

The Evaluation of Regional Water Balances Using Different Hydrometeorological Datasets

by

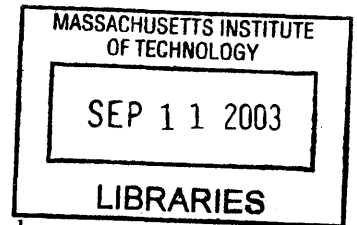
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B.S. Civil and Environmental Engineering  
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SUBMITTED TO THE DEPARTMENT OF CIVIL AND ENVIRONMENTAL  
ENGINEERING IN PARTIAL FULLFILMENT OF THE REQUIREMENTS FOR THE  
DEGREE OF

MASTER OF SCIENCE IN CIVIL AND ENVIRONMENTAL ENGINEERING  
AT THE  
MASSACHUSETTS INSTITUTE OF TECHNOLOGY

SEPTEMBER 2003



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**BARKER**

# The Evaluation of Regional Water Balances Using Different Hydrometeorological Datasets

by

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Submitted to the Department of Civil and Environmental Engineering on August 15, 2003 in  
Partial Fulfillment of the Requirements for the Degree of Master of Science in Civil and  
Environmental Engineering

## Abstract

This study attempts to use available hydrometeorological datasets to compute water balances over various regions of North and South America. The NCEP Reanalysis-1 and NCEP Reanalysis-2 are used as the primary sources of data, in addition to secondary datasets of precipitation and runoff. Time series of precipitation, evaporation, moisture flux, runoff, change in precipitable water and change in surface storage are generated from these sources and used in equations for the atmospheric and surface water balances.

Several different schemes are studied in order to utilize the available hydrometeorological data most effectively. These include various methods of calculating moisture flux, runoff, and change in surface storage. For both the atmospheric and surface balances various sizes and shapes of control volumes are studied including control volumes which are boxes of 7.5 and 15 degrees on a side, and basin sized control volumes resolved at 1 degree resolution varying in size from 318,000 km<sup>2</sup> to 4,620,000 km<sup>2</sup>. In between 60° N and 60° S the entire globe is broken into control volumes of 30, 15, and 7.5 degrees on a side, and the atmospheric water balance is evaluated for each control volume.

A general lack of ability to close regional water balances over land is found, specifically over a box of 7.5 degrees on a side. The integrity of the reanalysis data is simply not good enough to complete such a balance reliably on either the atmospheric or surface levels. Particularly unreliable are the moisture flux and runoff variables. There is some success in calculating the moisture flux by known methodologies, but this is largely in 15 and 30 degree control volumes. The runoff methodologies developed in this study were limited by the resolution of the data, and not terribly successful. A major recommendation of this work is that global hydrometeorological datasets such as the reanalysis should take into account the water balance at finer resolutions when deriving its product.

In addition, this work uses the Wang et al. (1999) methodology to compute global ground heat flux. This is part of a larger effort to create independent global energy flux maps that can refine our knowledge of the global energy balance. An original method for the calculation of the thermal inertia parameter necessary for this calculation is presented along with a new global ground heat flux product.

Thesis supervisor: Rafael L. Bras

Title: Bacardi and Stockholm Water Foundations Professor

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# 1. Introduction

Large-scale land changes have occurred all over the world in the last few centuries. One example is the deforestation of the South American continent, particularly in the Amazon River Basin where rain forest is being changed into pasture and grazing land. Such changes have large-scale environmental impacts, including impacts on the water environment. Key to understanding the water environment is the quantification of the water balance over areas of interest. For the purposes of this work, the water balance is defined as a time series of the different water fluxes in and out of a control volume, which make up the hydrological cycle for the control volume. These fluxes include precipitation, evaporation, atmospheric moisture flux, and runoff, as well as the changes in atmospheric and surface water storage.

All of these fluxes are or available or can be derived from the NCEP Reanalysis in order to compute the water balance (Kalnay et al., 1996). These datasets are based on most available historical data, and therefore are more reliable over areas with dense hydrometeorological monitoring (such as the United States and Europe), and not as accurate over remote areas like South America. Additional datasets pertaining to other water balance fluxes are also available.

This study attempts to use available hydrometeorological datasets to compute water balances over various regions of North and South America. The NCEP Reanalysis-1 and NCEP Reanalysis-2 are used as the primary sources of data, in addition to secondary datasets of precipitation and runoff. Several different schemes are studied in order to utilize the available hydrometeorological data most effectively.

The first chapter of this study will discuss the background and relevant previous studies. Chapter 2 will discuss the different types of data used and Chapter 3 will discuss the procedure followed in analyzing data in this study. Chapters 4, 5, and 6 will deal with the data analysis of water balance variables while Chapter 7 will deal with the data analysis of the ground heat flux. Concluding remarks will be made in Chapter 8.

## 1.1. Background on the Water Balance

The equations and the schematic of the water balance are given below:

$$\frac{dW}{dt} = -P + E + Q \quad \text{Eq. (1-1)}$$

$$\frac{dS}{dt} = P - E - R \quad \text{Eq. (1-2)}$$

**Control Volumes for Water Balance**

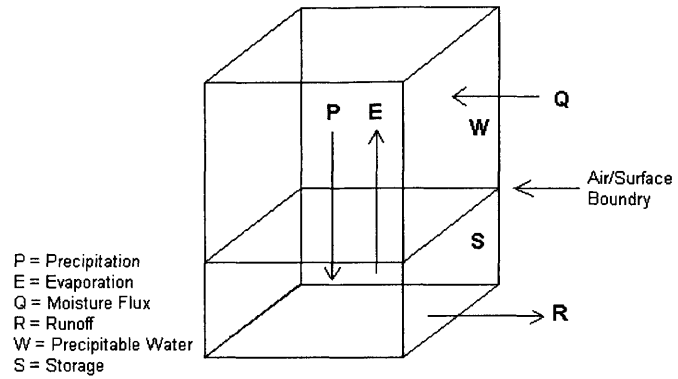


Figure 1-1: The Water Balance

For a given area of land, two control volumes are constructed. The atmospheric control volume contains all of the air above of a given area. The surface control volume contains everything below the air/surface boundary. The amount of water in each control volume at a given time is defined by the quantities W and S, which stand for atmospheric water vapor and surface storage respectively. The change of W and S with time must be equal to the net fluxes through the control volume. Equations 1-1 and 1-2 are the atmospheric and surface water balances respectively. The atmospheric water balance has inputs of Evaporation (E) and Moisture Flux (Q) and the output of Precipitation (P). Moisture flux is the moisture which moves into or out of an area of air over a land mass over a time period. In the surface water balance precipitation is an input while evaporation and runoff are outputs.

In this work we use a different time series of (P, E, Q, R,  $dW/dt$ ,  $dS/dt$ ) to evaluate the water balance over selected areas of the world. Monthly or yearly time periods are used. The time period (dt) over which we conduct these tests is either a month or a year. With six different variables in the water balance and a number of ways to arrive at several of the variables values, there are many alternative ways of computing the water balance.

## **1.2. Previous Relevant Studies**

### **1.2.1. Rasmussen**

One of the original and most complete water balance studies was undertaken by Eugene Rasmussen in the late 1960's in a series of three papers (Rasmussen 1967, 1968, 1971). These references will be mentioned extensively throughout our study, as they are our study's foundation. Rasmussen uses precipitation, evaporation, runoff, surface storage, and pressure level wind and specific humidity data. Pressure level, specific humidity and wind data are used to calculate moisture flux.

The purpose of Rasmussen's study was to compute the two water balance variables for which no measurements typically exist: evaporation, and change in surface storage. Large control volumes were created over major hydrological regions of North America (such as the Mississippi River Basin), and atmospheric sounding data and observational datasets from the early 1960's were used. The resolution of the 2.5 degree data in Rasmussen's study is the same used in the modern Reanalysis. Rasmussen's basins covered larger areas of North America than the areas used in our balances.

### **1.2.2. Model Based Studies**

Model derived water balances have as much of a history as data based balances starting with Baumgartner and Reichel (1975) and Willmott (1985). More recently Nicholson et al. (1996) used a model to compute the surface water balance over Africa using precipitation as a data input and land based parameters applied to a model to derive evaporation, storage, and runoff at 1 degree resolution. This data product is similar to that generated by Willmott (1985), a product which provides a fifty-year dataset of surface water balance variables on a 0.5 degree grid. Betts (1998) also used a model to produce a climatology, inputting precipitation for control volumes which are each part of the Mississippi River basin. A set of reanalysis data is used which yields good results on monthly time scales.

Lenters et al. (2000) used the NCEP-Reanalysis model in order to create a surface water balance climatology over the continental United States. The climatology is based on data from 1963-1995. The study utilized precipitation as well as other near surface variables and inputs. The study mentioned that the reanalysis precipitation is unfortunately "weakly constrained" by physical data. The NCEP Reanalysis surface water balance is classified as poor and the seasonal cycles for all variables are not consistent.

Oki et al. (2001) used modeling to generate a limited climatology for the Amazon River Basin. He took precipitation and broke it down into its surface balance counterparts (runoff, evaporation, and deep drainage). Several interesting concepts are used here including the use of an original simplified one degree runoff routing model called the TRIP pathways. Biases were analyzed and monthly runoff peaks were found to be shifted on the order of two to four months from precipitation peaks.

### 1.2.3. Data Based Studies

Recent work on regional water balance studies includes Higgins et al. (1996). An early version of the Reanalysis dataset was used to compute a moisture budget climatology for the Central United States, using data from 1985-1989. The study used Rasmussen's techniques of calculating moisture flux, and found biases in the water balance that were on the same order of magnitude as moisture flux. These problems were attributed to the poor quality of the reanalysis over mountainous regions.

Mo et al. (1996) studied the atmospheric water balance on a regional basis using two preliminary reanalysis datasets, again to form a climatology for a number of years (1982-1994). Large discrepancies in tropical moisture transport were found, which are attributed to uncertainties in the divergence of winds.

Gutowaki et al. (1997) also used the Rasmussen moisture flux methodology, and as before used the Reanalysis for a number of years to create a water balance climatology of relevant variables. Here the surface water balance was integrated into the data analysis, while in the previous two studies only the atmospheric water balance was considered. This allows a comparison of the total water balance, or a comparison between moisture convergence and runoff, when storage terms can be reduced and eliminated over longer periods of time. Their study discussed the requirements for a control volume. Basin control volumes are used which must be large enough to have many grid points (23 to 25 in this case) and moisture convergence is found over these volumes. The Moisture convergence is then compared to runoff, for which there is only one point, the discharge of the river basin.

Also discussed in Gutowaki is the idea of lag time as there is an observed difference between the peak cycles in moisture convergence and runoff in the absence of storage adjustments (which are eliminated for the climatology). Lag times are observed on the order of 30 to 60 days. Variability of the moisture convergence was found to be lower in areas such as the United States where the data quality is better.

More recently, Chen et al (2001) used Rasmussen's methodology of moisture flux calculation to produce a time series of water balance variables over a rectangular control volume in South America (5 N 15 S, 70 W 50 W) in order to study the circulation patterns associated with Amazon deforestation. Specifically the moisture flux is calculated by Rasmussen's method and plotted along with precipitation and evaporation thus in effect yielding an atmospheric water balance. The study found an increasing trend in both moisture flux and precipitation over the control volume from 1948 to present.

Roads et al (2002) provided one of the most extensive water balance studies involving a reanalysis, in this case the new Reanalysis-2. Both surface and atmospheric water balances are computed as well as energy balances. A variety of different control volumes are investigated in different parts of the world each with unique land features. Like in many studies balance nudging terms are used to force the balance into compliance. In terms of magnitude the greatest residuals are found over the Amazon Basin, however the variables of the water balance (precipitation, evaporation, etc.) are highest here.

#### **1.2.4. Precipitation Data Analysis Studies**

Precipitation is a very unique quantity as it is, along with evaporation, a link between the two distinct atmospheric and surface water balances. Likely non-coincidentally, precipitation datasets of the type used in our study are common and numerous. Although precipitation is only one variable in the water balance it is likely the most important because it is an exchange between the atmospheric and surface control volumes. Evaporation follows a fairly regular seasonal cycle and remains fairly even over a number of years, while precipitation varies.

Many of the existing precipitation datasets will be utilized in our study and discussed in detail in Chapter 2. Xie et al (1994) was one of the first studies involved in comparing globally resolved precipitation datasets on a monthly timescale. The study compared satellite and gauge based global precipitation datasets and observed many differences between different datasets. It is observed that gauges have a tendency to underestimate precipitation when aggregated over a larger area (gauges are point measurements), thus gauge datasets present a systematic bias. It was also determined that 5 gauges must be present in a 2.5 degree box in order to estimate precipitation within 10 % for that box.

Janowiak et al (1997), analyzed the Global Precipitation Climatology Project (GPCP), comparing it to the precipitation dataset derived from the NCEP Reanalysis. This task relies heavily on the data assimilation of the NCEP Reanalysis which contains a precipitation field. The Reanalysis precipitation shows a very regular pattern, while the GPCP shows a more realistic and random pattern. Observations made by Janowiak include very poor temporal agreement over South America and Africa (remote areas), but good over the United States, Europe, and Australia, a clear byproduct of poor data in remote areas with few observations. There is global precipitation agreement in these datasets, despite the regional differences.

Rudolf (2001) discussed a 1 degree precipitation global monthly precipitation dataset, derived from 2.5 degree information, and the importance of satellites in this process. Satellite studies such as the Tropical Rainfall Measuring Mission (current) and Global Precipitation Mission (proposed) are encouraged and satellite data is said to be especially important over the oceans.

### **1.3. Motivation for Study and Discussion**

Although there have been many water balances studies using various hydrometeorological datasets, these studies have had various limitations. Some of these studies use a time series of data of multiple years, but condense this data into a climatology only yielding a water balance for a typical year. Other studies calculate one or several of the water balance variables by solving the balance for that variable or deriving the variable from a model. Thus, the balance is forced. Most recently, Roads (2002) uses data for all variables in the water balance equation, and then adds an additional nudging term as needed to correct for imbalance. None of these studies evaluate a mutli-year monthly or yearly water balance using only data for each variable, and evaluate the imbalance of the water balance instead of correcting it.

At first water balances were tested over remote control volumes in South America in which long term climatological changes in the water balance were suspected. The NCEP Reanalysis could provide all of the data fields necessary to produce such a balance, but the various balance tests themselves, on both the monthly and yearly time scales yielded results far from correct. Errors were on the order of magnitude of the variables going into the balance. These findings begged the question of whether it was possible to complete a regionally sized water balance over a remote location. The idea of creating a water balance study for any region on the planet, inhabited or not, requires using remote sensing technologies to obtain fields where land based measurements are not available.

The NCEP Reanalysis-1 and NCEP Reanalysis-2 are integral parts of our study because they contain all necessary water balance data for each grid point globally. These datasets are based on all available hydrometeorological information, which may not be plentiful in remote areas. Therefore, in addition to choosing study areas over remote areas of South America it is also important to include areas where land based measurements are more prevalent. The United States is such a place, and contains areas that are large enough so that regional water balances over the United States can be compared to those in South America.

The flaw in the design of the water balance is that it is a measure of consistency in data and not accuracy. As long as the data used in the water balance provides a balance at each time step, the water balance is solved. However, if two or more water balance variables are incorrect the equation may still yield good results, if the errors are equal and opposite. In the case of an imbalance, it is also difficult to determine which water balance variable is off in the equation, or which is off by the most. In some cases this is easy to detect by common sense. Attention must be paid to false balances, but one reassuring thought is that if it is so hard to compute a truly balanced water balance, it is likely harder to compute a falsely balanced water balance. The key is to complete the water balance so as to insure the integrity of the data used.

#### **1.4. Relation of Energy Balance to Water Balance (Ground Heat Flux)**

The energy balance and water balance are linked by the evaporation term and as shown in the Roads' (2002) study. Similar methodologies can be used to balance the energy and water balance.

Consider here the surface energy balance:

$$R_{\text{net}} = G + SH + LH$$

The NCEP-Reanalysis provides each of these fields. Traditional methods of obtaining energy fluxes are complex, as measurements at several levels are generally necessary to estimate fluxes at a point for a given time. Wang et al. (1998, 1999) have derived ways to obtain ground, sensible, and latent heat fluxes from a time series of the relevant variables (generally temperature) at a point. In the case of ground heat flux, only ground surface temperature or skin temperature is need. In the case of sensible heat flux, only air temperature is needed, and in the case of latent heat flux only air temperature and humidity are needed. Additional information



based on land type, which is typically time invariant, is also utilized in these methods. The Net Radiation term can be computed or directly measured using well known methods.

This work uses the Wang et al. (1999) methodology to compute global ground heat flux. This is part of a larger effort to create independent global energy flux maps that can refine our knowledge of the global energy balance. Chapter 7 presents these results.

## 2. Datasets

The primary source of data for our study is the NCEP Reanalysis, which provides many useful hydrometeorological variables. All variables used in the water balance except precipitation and runoff, are provided or derived from the reanalysis. There are several large efforts to compile long term gridded and station oriented precipitation data sets. Precipitation is a very important component of the hydrological process, yet it is found that differing datasets provide very different estimates of mean monthly and mean annual precipitation for the study areas chosen. Independent precipitation estimates are all the more important because the reanalysis precipitation is solely model derived.

Our study considered monthly data through December 2002. Almost all of the datasets used are gridded for either part of or the entire globe. The size of the grid varies from 0.5 degree to 2.5 degree spacing depending on dataset. The distortion of the curvature of the earth is taken into consideration in all calculations. Its effect is very small in tropical regions.

### 2.1. NCEP Reanalysis

The NCEP Reanalysis is derived from a combination of land surface, ship, rawinsonde, aircraft, and satellite data, as well as other sources, and uses state of the art data assimilation and modeling techniques (Kalnay, 1996). The dataset is available monthly from January 1948 to present. Each variable is assigned a classification based on the methodology through which it is derived. A list of these classifications for relevant fields is given in Table 2-1.

Field	Class	Global Resolution	Application
U wind at 17 pressure levels	A	73 x 144	Q
V wind at 17 pressure levels	A	73 x 144	Q
Specific Humidity at 8 levels	B	73 x 144	Q
Precipitation	C	94 x 192	P
Evaporation (Latent Heat Flux)	C	94 x 192	E
Runoff	C	94 x 192	R
Precipitable Water	C	94 x 192	dW/dt
Snow Cover	C	94 x 192	dS/dt
Soil Moisture	C	94 x 192	dS/dt
Skin Temperature	B	94 x 192	Method GHF
Ground Heat Flux	C	94 x 192	Reanalysis GHF
Data Class	Comments		
A	Strongly influenced by observed data, most reliable class		
B	Influenced by both model and observed data		
C	Derived solely from the model		
D	Fixed from climatological models, does not depend on model		

Table 2-1: Reanalysis-1 Data Classification (Kalnay, 1996)

The humidity and wind fields, which are used to calculate the moisture flux, are classified as fairly reliable (Class A and B). It stands to reason that this analysis would be more accurate over the United States than over more remote areas where there is less observed data.

While alternative global precipitation datasets are easy to find, alternative evaporation data sets are generally unavailable. It is argued that for the purposes of our study the evaporation estimates for the reanalysis are fairly reliable because of the nature of evaporation. Evaporation data exhibits a fairly seasonal cycle with only small variations. Thus changes in the water balance are more likely to be the result of the other terms in the atmospheric water balance, like the precipitation for which there are many sources, and the moisture flux which is supposedly derived by fairly accurate data.

The data to be used for the surface water balance includes only variables derived from the model, class C, including precipitation, evaporation, runoff, soil moisture, and snow cover. The reliability of information on water balance data for water moving through and along the land surface is small. This work uses various alternatives to quantify runoff in order to improve the ability to close a surface water balance with reanalysis data.

The global resolution of the datasets refers to total amount of points provided by the dataset. The wind velocity and specific humidity fields are gridded every 2.5 degrees, while fields that are model derived are gridded approximately every 1.9 degrees in both latitude and longitude. The resolution of the data necessary for the moisture flux calculations is therefore very sparse. The reanalysis provides both wind and humidity data at 8 atmospheric levels over the entire globe, and this data is necessary for calculating moisture flux.

## **2.2. NCEP Reanalysis 2**

Recently, a second reanalysis data set has been released entitled the NCEP/DOE Reanalysis-2. This dataset has all the features of the original reanalysis that are useful for the water balance. There are many improvements made to the data assimilation techniques including: fixing cloud tuning and snow properties, implementing new simpler rainfall assimilation for soil wetness, updating the precipitation parameterization, fixing of the snow cover analysis, and removing the nudging of deep layer soil moisture (Kanamitsu et al., 2002).

The altering of the soil moisture nudging term is of particular interest due to the problems that have been observed with change in surface storage. The Reanalysis-1 used a 60 day climatology nudging term which caused a very high seasonal cycle, and this is improved in the Reanalysis-2. Also actual observed precipitation measurements are used in this dataset when the ground is not frozen.

There were improvements in several important fields including the soil wetness fields, winter time precipitation and tropical precipitation, and equatorial divergent winds (used for calculation of moisture flux). The tropical precipitation is improved but still classified as problematic. This dataset is only available for 23 years from 1979 to 2001. However, because data collection over remote areas has been more common in recent years, the overall water balance using Reanalysis-

2 data should be much better. Future plans include extending this dataset back to the 1950's thus allowing for a full comparison with the first reanalysis.

## **2.3. The Global Precipitation Climatology Project (GPCP)**

The Global Precipitation Climatology Project is one of the largest efforts directed at developing a global precipitation dataset. Data is incorporated from a variety of sources including low-orbit-satellite microwave data, geosynchronous-orbit satellite infrared data, and rain gauge observations. The original GPCP (version 1) was gridded on a 2.5 degree resolution thus its original component datasets are gridded at this resolution for consistency. The dataset originally spanned the time period of July 1987 through December 1995. The newer version of the GPCP has increased its resolution to 1 degree. The dataset now starts in January 1979 (around the time of the first remote sensing), and runs through the end of 2001. Since the dataset is on a 1-degree by 1-degree grid, each month of the dataset is composed of a 180 x 360 grid of the globe. This dataset classifies satellite data as useful, but it is improved a great deal using rain gauge data in its algorithms (Huffman et al., 1997).

### **2.3.1. Components**

In the derivation of the global precipitation climatology project several different precipitation datasets were created from independent data sources. It was the hypothesis of the GPCP that the combination of these data sources would yield the most accurate global precipitation dataset. However, the use of these products independently for a water balance can highlight the usefulness of each dataset. The GPCP states that each dataset is useful in a unique way. For instance one dataset may accurately predict the month to month variations in a water balance, while another may be a very strong annual predictor but have seasonal deficiencies. As stated above there are three major components of the GPCP: gauge based data, geostationary IR data, and microwave data. Each of these sources has different temporal and spatial variations, but each result in a gridded monthly dataset. In the following we discuss the gauge component and the geostationary IR component. Because the microwave estimates are based on SSM/I data, an initially independent precipitation dataset, it is discussed in detail in later sections.

#### **2.3.1.1. GPCC Gauge**

The GPCC is the land based gauge component of GPCP. It is a gridded 2.5-degree dataset of precipitation based on the interpolation of 6700 gauge stations onto a lat-long grid using SPHEREMAP developed by Wilmott et al. (1985). Because it is based on rain gauges this dataset has the potential for high accuracy, however it uses point sources of data, which are highly variable from point to point. The time coverage of this dataset is currently Jan 1986 through December 1996, which allows it to be used in the water balances with both reanalysis products, for all land based control volumes. Quality control technique with the GPCC involves first comparing data to regional means, and pulling out anomalies, and then studying each of these anomalies considering known relief and catastrophic events. This dataset is as a very important component of the GPCP; however it is also important to consider this dataset as compared to other gauge based datasets, especially the precipitation reconstruction dataset discussed later (Huffman et al., 1997).

### **2.3.1.2. Global Precipitation Index**

The global precipitation index is a subset dataset of the GPCP, which is derived almost exclusively from geostationary satellites specifically the Geostationary Operational Environmental Satellites (GOES) from the United States, the Geostationary Meteorological Satellite (GMS) from Japan, and the Meteosat from the European community. The satellites collect infrared imagery on a 3 hour basis which is integrated into a monthly time scale. When geostationary data is not available, NOAA polar orbiting satellites are used instead. The resolution is a 2.5 degree grid, consistent with GPCP, and the data is only available in the tropics from 40 degrees north to 40 degrees south. Thus, for some regions GPI data isn't available exclusively from geostationary satellites. The GPI is used in deriving one of the TRMM products discussed later. The time coverage of this dataset is January 1986 through the present. The major pitfall discussed pertaining to the GPI is that in cases where the satellite nadir point is far from the region of interest rainfall is overestimated (Huffman et al., 1997).

## **2.4. Precipitation Reconstruction (PREC)**

The precipitation reconstruction dataset is likely the most complete gridded monthly precipitation dataset based completely on rain gauges. Two major gauge sources are included, the Global Historical Climatology Network (GHCN) and the Climate Anomaly Monitoring System (CAMS). The result is information from over 17,000 stations, far more than the GPCC. The dataset is gridded on a 2.5 degree grid and runs from January 1948 to November 2002 for all land points. As is the case with other gauge datasets, the gathering of data often takes a few months and therefore at the time of data gathering for our water balance study the last month of 2002 was not available. Several different data assimilation techniques are evaluated in the process of preparing this dataset. Also included with this dataset is the number of gauges used for each grid point. Thus, for each control volume the amount of gauges can be tracked using knowledge of how many gauges per unit area are available for a given control volume (Chen et al, 2002). A graph of the number of precipitation gauges per area in some of the selected control volumes used in this work is given in Figure 2-1. The big differences occur between points in Brazil and the United States especially early in the study period. Recent rain gauge data is severely reduced due to the amount of time required to process rain gauge data and the recent switching to remote sensing technologies.

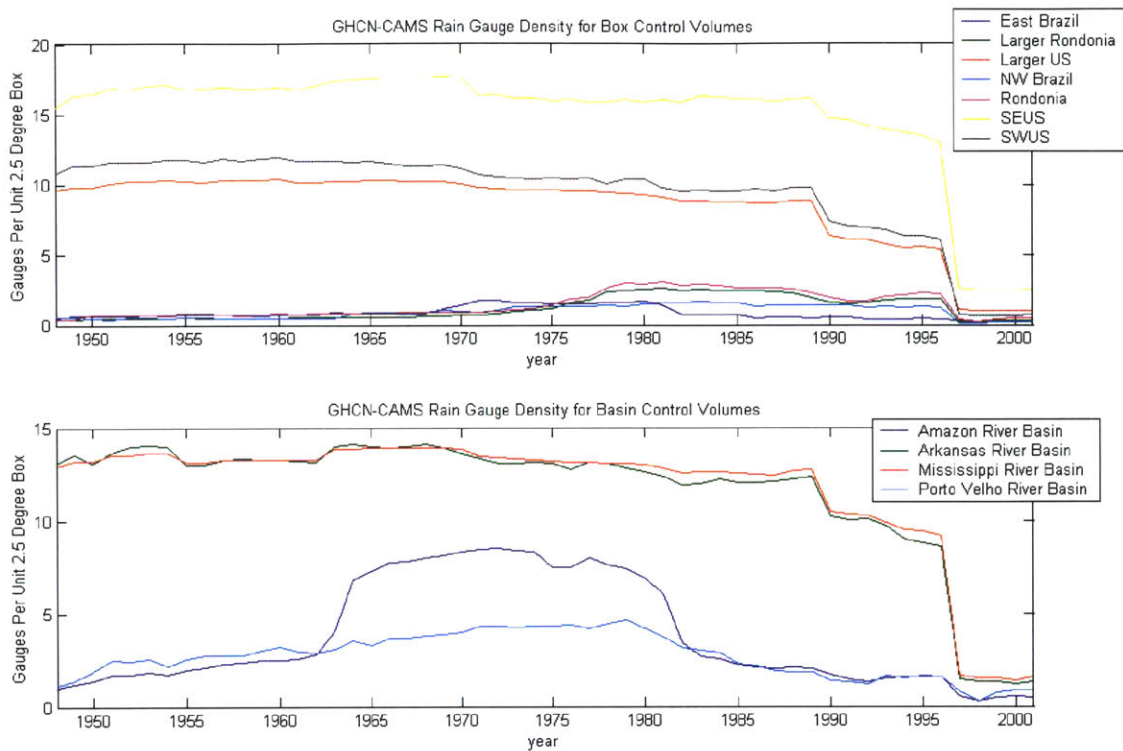


Figure 2-1: Rain Gauge Density in Selected Areas of Interest

## 2.5. CPC Merged Analysis of Precipitation (CMAP)

The CPC Merged Analysis of Precipitation combines multiple sources of data including gauge data and several different types of satellite data (the Global Precipitation index (GPI), the OLR Precipitation Index, two types of SSM/I, and a microwave sounding unit, MSU). All of these datasets with the exception of the microwave sounding data are analyzed as independent datasets, which are available individually. The CMAP dataset is similar to the GPCP in that it combines many sources of information. In addition, a second dataset is derived combining all sources and also using a model. The two products are referred to hereafter as CMAP and CMAP2 (model). Again the data is gridded on a 2.5 degree basis. An observation made in the documentation of the CMAP is that modeled precipitation is often calculated poorly in the tropical regions (Xie et al., 1997).

## 2.6. CAMS/OPI

CAMS/OPI is another global precipitation product gridded on a 2.5 degree basis. This dataset's goal is to produce a monthly precipitation that can be generated in real time. Products such as the GPCP and CMAP typically take a few months in order to collect the gauge data and combine it with other data sources to provide monthly means. This dataset combines a gauge and satellite dataset that are available shortly after the conclusion of a month. Rain gauge totals come from the Climate Anomaly Monitoring System (CAMS), and satellite estimates are included that are

derived from outgoing longwave radiation (OLR) for the OLR Precipitation Index (OPI) (Janowiak et al., 1999).

Two data subsets from the CAMS/OPI product are used in this work, the independent OPI estimates (OPI2), and the combined CAMS-OPI dataset (OPI3). The combined CAMS-OPI dataset is based on a CMAP climatology from 1987 to 1995 and therefore may be biased towards the CMAP dataset. The actual CAMS dataset is a gauge dataset similar and probably less accurate than other gauge based datasets and therefore is not included. The producers of this dataset warn about its use in studies such as the water balance study being conducted here and recommend the use of the GPCP or CMAP datasets. However, this dataset is unique in that it can be derived in real time and its effectiveness in water balance studies should be evaluated.

## **2.7. SSM/I**

The special sensor microwave imager is a satellite based remote sensing device. It has been used from July 1987 to the present and one of its products is a global precipitation dataset. The dataset is 1 degree, gridded globally. Two separate procedures are used to retrieve data over land and ocean. When processing the SSM/I time gaps were discovered. Some of the record prior to 1992 is incomplete which results in missing data in both the monthly and yearly resolutions of the water balances computed. There are several other sources of precipitation to apply during the time period of SSM/I. In addition one of the three products of the Tropical Rainfall Measuring Mission discussed in the next section is a 1 degree gridded SSM/I global dataset, which is likely more reliable than it's predecessor if for any reason, because it's data is more recent. The absolute magnitude of the SSM/I rainfall is found to be too high by the datasets producers, a hypothesis that can be tested by water balance analysis (Ferraro et al., 1996).

## **2.8. The Tropical Rainfall Measuring Mission (TRMM)**

The Tropical Rainfall Measuring Mission, is a project undertaken by NASA, to measure precipitation in the tropical regions of the globe (between 40 degrees north and 40 degrees south), for use in hydrologic research projects such as this water balance study. The need for tropical precipitation measurements is considered important because of its magnitude and relative importance to the global water cycle. The mission also allows for a much closer look at precipitation in the Amazon rain forests where primary sources of data are not available. The TRMM products are available from January 1998 to present.

Our study does not attempt to discuss the actual process and retrieval algorithms of the TRMM satellite or any other precipitation satellite. It rather discusses the difference of processes used in producing datasets and evaluates these processes through the water balance methodologies. Unfortunately it is impossible to apply the TRMM precipitation datasets to a basin control volume because runoff data for the basins studied is only readily available through 1996. Many other precipitation sources such as those described earlier are included in the TRMM analysis, most of which are incorporated in dataset 3B43 which is of primary interest for application to water balances. The following three datasets were derived as products of the TRMM mission and are used in this work.

### **2.8.1. Data Product 3A46**

The 3A46 TRMM data product is the SSM/I monthly tropical rainfall estimates from the TRMM project, gridded on a 1 degree by 1-degree basis providing a monthly grid of data 180 x 360. The record extends from January 1998 to August 2001, and is no longer being updated actively. The entire series of available monthly rainfall data is analyzed here.

### **2.8.2. GPCC\_TRMM**

An additional component of the TRMM precipitation dataset is the continuation of the gauge based analysis from the Global Precipitation Climate Center. This dataset is given as an independent product of TRMM and numbered 3A45B. It is in essence an extension of the GPCC Gauge precipitation dataset from January 1998 to the present, however the dataset at processing was only available (as PREC) until November 2002. The resolution of the data is increased to 1 degree to be consistent with other TRMM products.

### **2.8.3. Data Product 3B43**

The 3B43 TRMM data product takes the rainfall information from the TRMM satellite and merges it with other satellite and gauge data to produce a TRMM data product on a 1 degree grid. This dataset uses the actual TRMM data in combination with 3A46 and additional gauge based data including 3A45B, and is likely the best estimate of precipitation of a monthly basis from the tropical rainfall measuring mission, and the best estimate for gridded monthly precipitation over it's time period (1998-2002), and it's spatial coverage (40 N to 40 S).

## **2.9. USGS Runoff**

The USGS website contains extensive historical monthly runoff data for almost all of the significant rivers in the United States, especially those that would be useful in conducting water balances at a regional scale (<http://www.usgs.gov>). Two stations are used from this resource, one near the end of the Arkansas River and one near the end of the Mississippi River. The corresponding site numbers for these data sets are 0725800 and 0728600 respectively. Unfortunately from this source the runoff data isn't updated frequently, although a comprehensive dataset is available from (1948 to 1996) for the Mississippi River and from 1948 to 1993 for the Arkansas River. In order to better compare water balances, the Arkansas River runoff is extended to 1996 based on the given historical data.

## **2.10. RivDIS Runoff Data**

A very useful dataset is the LBA-HydroNET version 2, (available at <http://lba-hydronet.sr.unh.edu/database.html>). This dataset includes a comprehensive river gauge network of South American Rivers, with good temporal resolution. This data allows for the study of basin water balance studies on two scales, specifically the Madeira River which flows through Rondonia, Brazil, and the Amazon River, (Oki 1998, Chen 2001). Recently this dataset has been updated over South America giving runoff data for major rivers on the continent for about thirty years in most cases, while the original extent for these two stations was ten years. This new data



resource is very valuable in conducting a regional water balance analysis. The sites used are site id 13022 for the Madeira River and 13065 for the Amazon River. Data is available from 1970 to 1996 for both stations.

## **2.11. Total Runoff Integrating Pathways (TRIP)**

This dataset is an ASCII grid of runoff directions for all land points on the globe on a 1 degree and .5 degree grid. It was derived through two major processes. The first involved the use of a digital elevation model converted to a coarser resolution in order to find the lowest neighbor and steepest slope between grid points. This methodology then had to be modified subjectively by adjusting the data with an atlas. Basins are eventually derived for major rivers across the globe (those with a significant enough drainage basins that can resolved on a 1 degree or 0.5 degree grid, thus the 0.5 degree grid picks up more rivers.

The primary purpose for creating this dataset is to help in global circulation models and water balances studies such as this one. The documentation states that other water balance variables such as precipitation, evaporation, and soil moisture are now becoming available on a global basis, thus necessitating an accurate way of routing runoff across land through such models. The dataset also provides a method of finding runoff per unit area of a control volume or a basin, which is the form in which runoff is evaluated at in our water balances. Both the 0.5 degree resolution and 1 degree resolution are used in separate applications. The 0.5 degree data is used with a gridded runoff database, while the 1 degree data is used to derive the shape of the basin control volume.

## **2.12. Missing Precipitation Data**

One the major advantages of the reanalysis precipitation is it is seamless temporally and spatially. Unfortunately most of the precipitation datasets used are missing data either spatially or temporally. In most cases these missing values don't affect the ability to conduct water balances for the selected regions; however in some regions certain precipitation datasets can't be used. Table 2-2 outlines these errors in terms of the regions being studied. When the individual control volumes are analyzed separately in Sections 4.2, precipitation availability will be looked into in more detail.

It should be noted that although temporal gaps exist in the SSM/I dataset it is possible to use the SSM/I data for a monthly water balance. The Spatial coverage of SSM/I is flawless.

Precipitation Dataset	Start Year	Finish Year	Control Volume Errors	Basin Errors	Global Errors 60N-60S
Reanalysis – 1	1948	2002	None	None	None
Reanalysis – 2	1979	2001	None	None	None
GPCP	1979	2001	None	None	None
GHCN	1948	2001	East Coast	None	Not Applicable
GPCC	1986	1996	East Coast	None	Not Applicable
GPCC TRMM	1998	2001	East Coast	None	Not Applicable
GPI	86	2002	SWUS/Larger US/East Coast	Mississippi	Holes in Set early, good for global tropics, 30N/30S,
3A46	98	2002	SEUS/SWUS/Larger US/East Coast	Mississippi/Arkansas	Various errors in all but extreme tropical regions, useless for global evaluation
3B43	98	2002	SWUS/ Larger US/East Coast	Mississippi	Good for the 30 N/30 S
SSM/I	1988	1989	None	None	None
	1992	2002	None	None	None
OPI – Longwave	1979	2001	None	None	None
OPI – Combined	1979	2001	None	None	None
CMAP	1979	2001	None	None	None
CMAP + model	1979	2001	None	None	52.5 to 60 S contains

Table 2-2 Temporal Resolution of Precipitation Data

### 3. Procedure

The object of this study is applying the datasets described previously to a simple water balance model, and evaluating which datasets produce the best water balances. The exercise relies heavily on the NCEP Reanalysis, which contains all necessary fields. Only precipitation and runoff will be obtained from non reanalysis sources. Evaluating the water balance derived from the NCEP Reanalysis data, while using different methods and sources of precipitation and runoff will elucidate the value of information.

In this chapter the principles of a water balance are described and discussed in more detail than in Chapter 1, and the methodologies by which each variable in the water balance is obtained is discussed. Lastly, the data analysis tools used to evaluate the various water balances are described.

#### 3.1. A Water Balance

The general principle of a water balance is very simple and was outlined by Rasmussen in the late 1960's (Rasmussen 1967, 1968). These early investigations came up with a methodology for the United States where data for such a balance was first available. Two storage equations can be set up, one for the storage of atmospheric water vapor (the precipitable water), and another for the storage of surface, soil moisture, and groundwater. These equations are given below:

$$\frac{dW}{dt} = -P + E + Q \quad \text{Eq. (3-1)}$$

$$\frac{dS}{dt} = P - E - R \quad \text{Eq. (3-2)}$$

**Control Volumes for Water Balance**

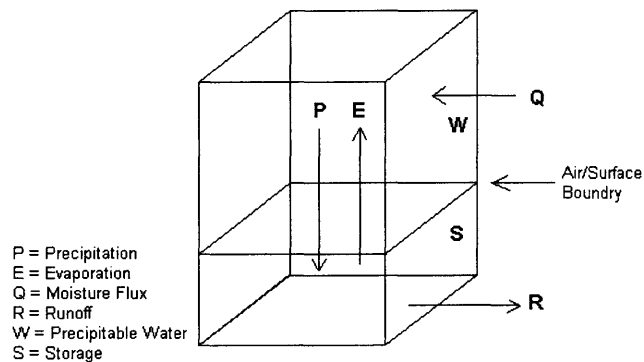


Figure 3-1: The Water Balance

A schematic of the control volume is given in Figure 3-1. The first equation, or the atmospheric water balance, equates the water coming in and out of the atmosphere, and refers to the top portion of the control volume drawn in Figure 3-1. The second equation, or the surface water balance, equates the water coming in and out of the ground surface, and refers to the bottom portion of the control volume drawn in Figure 3-1. All terms in these equations are time averaged values either monthly or yearly in this study.

In water balance calculations we use two types of data, spatially averaged data and boundary flux data. This has an effect on the manner which certain fields are calculated over the control volume. In this study the data is primarily gridded at a resolution less than that of the control volume. Spatially averaged terms include precipitation, evaporation, change in precipitable water, and change in storage. All these terms can be found by simply averaging the gridded data. Because the data is gridded by latitude and longitude, distortion effects are taken into account as a function of latitude.

Precipitation and evaporation can then be directly applied to the water balance equations. Change in surface and atmospheric storage can also be found by spatial averaging, with differences of values approximating time derivatives. Because runoff and moisture flux come and go through the actual boundaries of the control volume, different methodologies must be analyzed to determine the best way to utilize gridded datasets to obtain these values. Such methodologies are outlined later.

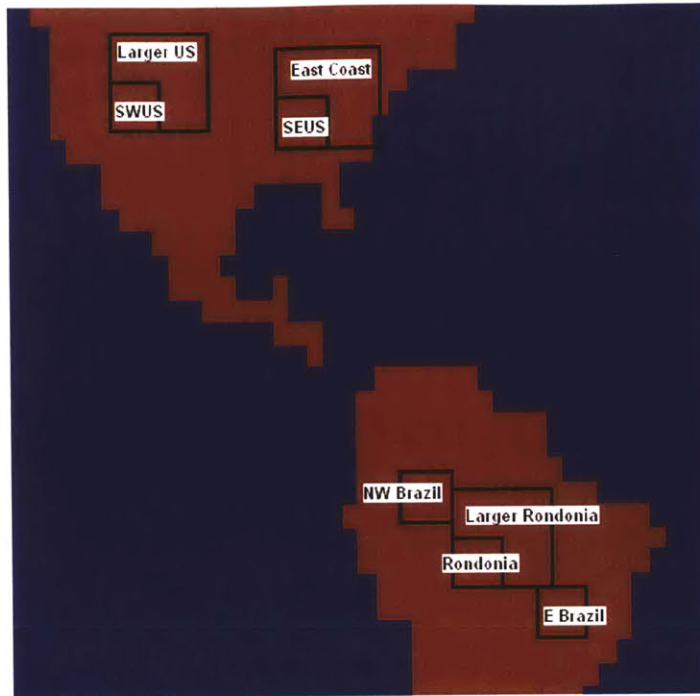
### **3.1.1. Box control volume**

The reanalysis as well as a number of monthly global precipitation datasets are presented in gridded latitude longitude boxes of varying sizes. In order to utilize these datasets in a regional scale water balance, the datasets are spatially averaged over a control area, which in its basic form is a rectangle with four coordinates. The programming developed finds the average value for a variable inside the coordinates.

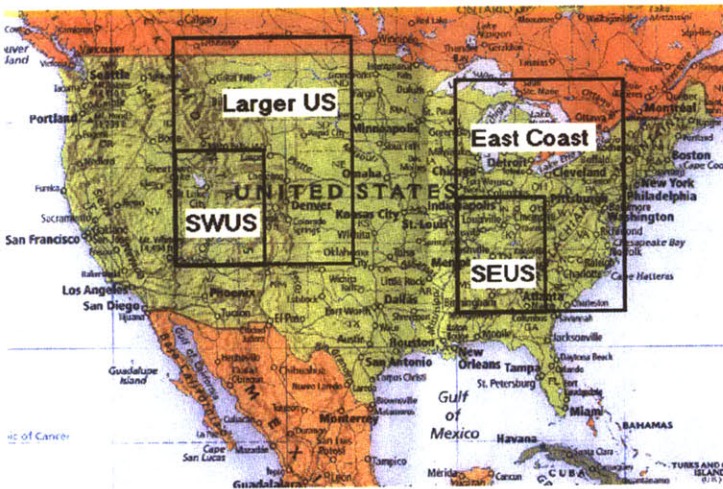
In the case of moisture flux calculations one point is used to estimate the flux through a boundary that is approximately 250 km wide. In order for the flux through the side of a control volume to be accurately computed it is important to use more than one data point. Taking three points on each side of the control volume as a minimum yields a box 7.5 degrees on a side. This allows for a meaningful average flux to be calculated across that boundary. It is also a small enough region over which significant land surface change can be observed in the past fifty years. Additional balances were performed for two cases using 15 degrees latitude and longitude on each side, as a comparison tool. Theoretically a better balance can be achieved over a larger control volume due to greater spatially averaging. It is also possible, however, that a larger control volume will cause already error prone data to become increasingly problematic.

Figure 3-2 shows specifically the eight box control volumes which were selected for this study. First, three control volumes were chosen of the 7.5 degree size. They represent three different stages of deforestation in the Amazon rain forest. The Eastern Brazil control volume is an example of already deforested land, the Rondonia control volume is an example of land currently being deforested, and the Norwest Brazil control volume is an example of land not yet deforested. In order to make comparisons, a control volume was chosen over the Southeast United States as a control. It is an area which in theory should have a good water balance since there is good observation data in this area. Also included at this smaller size is a control volume over the Southwest United States, a dry arid region, which we would like to contrast against the more tropical control volumes.

The Southwest United States and Rondonia control volumes are enlarged to form the Larger US and Larger Rondonia control volumes which are 15 degrees on a side. The comparison is again over an area in tropical remote South America, and a drier more detailed North America. Only two distinct control volumes can be formed over these continents as land masses of 15 degrees on a side are hard to find especially in the Western Hemisphere. An additional control volume over the East Coast of the United States was originally included to try and compute water balances over a land and sea area. This later proved to be an impossible task, so this control volume is included only in discussing the moisture flux methodology, nowhere else.



(a)



(b)



(c)

Figure 3-2: (a) Selected Box Control Volumes of the Reanalysis Land/Sea Grid, (b) North American Box Control Volumes, (c) South American Box Control Volumes

Table 3-1 denotes the different land areas of the box control volumes selected. The total vertical dimension of the control volume is any height through which water passes horizontally assuming no water enters or exits the top or bottom of the control volume. The total control volume is then broken into two control volumes at the air surface boundary, creating the atmospheric and surface control volumes. Although two standard control volume sizes are used (7.5 and 15 degrees on a side), the land areas of these control volumes are slightly different because of

latitude distortion. Rasmussen states that moisture flux calculations should be used only on a scale appropriate to the data given. In Rasmussen's last study in 1971 volumes were used between 500000 km<sup>2</sup> and 4200000 km<sup>2</sup> (similar to the scale used in this study). Specific dimensions of the box control volumes are not given here because there will be two slightly different versions of each of the box sized control volumes, as will be discussed in the Section 3.2.4, Moisture Flux Methodology Coordinates.

Control Volume	Land Area (km <sup>2</sup> )	Control Volume	Land Area (km <sup>2</sup> )
East Coast	2,106,000 km <sup>2</sup>	Northwest Brazil	692,000 km <sup>2</sup>
Eastern Brazil	655,000 km <sup>2</sup>	Rondonia	678,000 km <sup>2</sup>
Larger Rondonia	2,725,000 km <sup>2</sup>	SEUS	558,000 km <sup>2</sup>
Larger Untied States	2,027,000 km <sup>2</sup>	SWUS	539,000 km <sup>2</sup>

Table 3-1: Box Sized Control Volume Land Areas

**3.1.2. Basin Control Volumes**

In order to use an alternative source of runoff data for this study, a method was devised to create a control volume in the shape of a basin. An example of such a control volume is shown below for the Arkansas River Basin. The process by which these control volumes are derived and used is discussed in Section 3.3.2.2 Basin Type Control Volumes.

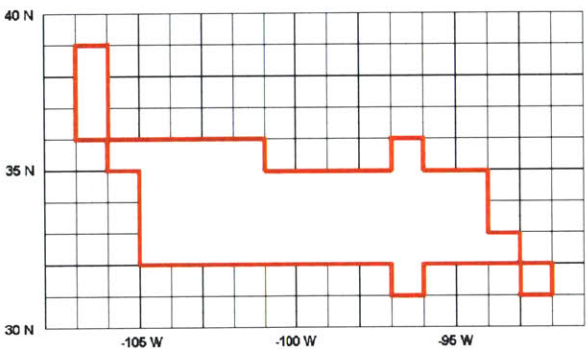


Figure 3-3: Arkansas Basin Sized Control Volume

**3.1.3. Global Analysis**

The programming devised for this study provides a time series of different water balance variables from different datasets based on four input boundaries: north, south, east, and west. Because datasets of a resolution applicable to this study are generally global in coverage, it is possible to input any four coordinates into the program and find the applicable water balance statistics. Although some precipitation and obviously all runoff datasets are limited only to land masses, full atmospheric water balances can be analyzed and calculated globally.

**3.1.4. Units**

The units that will be used for the remainder of this study are kg/m<sup>2</sup>-month. Yearly data will therefore be presented as average monthly data, in order to make the magnitude of the monthly

and yearly water balances comparable. Thus, water balance variables will be spatially averaged over the entire control volume. Normalized quantities will contain these units in both the numerator and denominator and therefore will be unitless.

### 3.2. Atmospheric Water Balance

The governing equation for the atmospheric water balance is:

$$\frac{dW}{dt} = -P + E + Q \quad \text{Eq. (3-1)}$$

Spatially averaged precipitation from a variety of sources and evaporation from both reanalyses are included in various analysis of the atmospheric water balance. The calculation of the time series of moisture flux and change in precipitable water require additional calculations of the reanalysis data.

#### 3.2.1. Atmospheric Change in Storage

As described before, the NCEP Reanalysis contains monthly precipitable water. The control volume in an atmospheric water balance is the air mass above a designated plot of land. The amount of water stored in this control volume is called the precipitable water or the total amount of water vapor present which could potentially be rained out of the sky. The change in precipitable water ( $dW/dt$ ) can generally be assumed to be zero on a year to year basis, although monthly changes may be significant.

The equation used in this research for change in precipitable water is the average change in precipitable water between a time period before and a time period after, where  $\Delta T$  is a month or a year:

$$\frac{dW_t}{dt} = \frac{W_{t+1} - W_{t-1}}{2\Delta T} \quad \text{Eq. (3-3)}$$

#### 3.2.2. Moisture Flux Calculation

Rasmussen devised a method of calculating the moisture flux through the boundary of a control volume. This method requires wind and specific humidity data at different levels, which is provided on a 2.5-degree grid by the Reanalysis-1 and Reanalysis-2. The moisture flux can be divided into an x and y component by:

$$Q_u = \frac{1}{g} \int_{p_u}^{p_s} \bar{q} \bar{u} dp = \frac{1}{9.8m/s^2} \int_{300mb}^{1000mb} \bar{q} \bar{u} dp \quad \text{Eq. (3-4)}$$

$$Q_v = \frac{1}{g} \int_{p_u}^{p_s} \bar{q} \bar{v} dp = \frac{1}{9.8m/s^2} \int_{300mb}^{1000mb} \bar{q} \bar{v} dp \quad \text{Eq. (3-5)}$$



The moisture flux terms are given by the integral of specific humidity times wind velocity from ground level atmospheric pressure to an atmospheric pressure where water vapor transport is negligible. The NCEP Reanalysis provides specific humidity u-wind velocity, and v-wind velocity data points at 8 pressure levels: 1000, 925, 850, 700, 600, 500, 400, and 300 mb. Above 300 mb specific humidity is very low and therefore its horizontal movement through the atmosphere is negligible in these calculations.

The flux at each point that is relevant to the control volume is found by a numerical integration technique. The pressure difference between two points is multiplied by the average value of the wind multiplied by the specific humidity for the two points being considered. The result of this numerical integration is a mass flux of water vapor through a point in kg/s per m of length through which the flux passes.

Regardless of methodology, considering a rectangular control volume, the moisture flux will enter the cube on each of its four faces, north, south, east, and west. The north and south fluxes are found by the equation for  $Q_v$ .  $Q_u$  gives the east and west fluxes. The total moisture flux for the control volume is the sum of the contributing moisture fluxes of each border.

$$Q = Q_W - Q_E + Q_S - Q_N \quad \text{Eq. (3-6)}$$

$Q$  is positive when moisture is flowing into the control volume. The units of this calculation are then multiplied by the length of the perimeter to obtain the total kg flux per time, then converted to monthly units and divided by the area of the control volume. This results in the units of  $\text{kg/m}^2\text{-month}$ , the standard units used for the other aspects of the water balance.

### 3.2.3. Moisture Flux Methodologies

Two methods for defining the moisture flux of a box control volume are investigated in this study (see Figure 3-4). The data points given by the reanalysis are gridded at a resolution of every 2.5 degrees. One method chooses the border so it aligns with this grid and thus the control volume is aligned on points as shown on the right side of Figure 3-4. The accuracy of the measurement therefore is dependant on using actual points. However, using this method a border that is 7.5 degrees long (approximately 834 km) is represented by only 3 data points.

### Moisture Flux Methodologies for Box Control Volumes

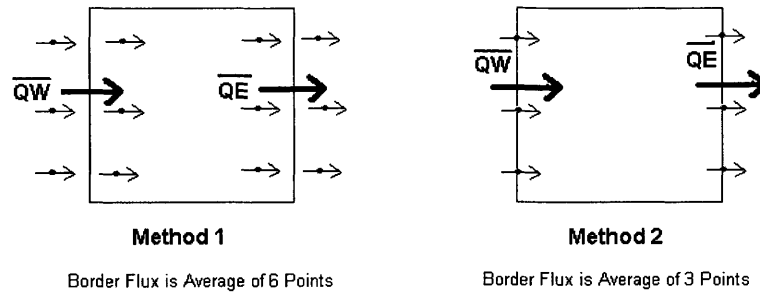


Figure 3-4: Moisture Flux Methodologies

Alternatively, if the control volume is selected between the points on the data grid, the flux can be calculated as the average of the two closest points to the border and for the same length of control volume border 6 data points are used, however each data point is a significant distance from the actual border. This method is shown on the left of Figure 3-4. Both methods are used for each control volume.

#### 3.2.4. Moisture Flux Methodology Coordinates

The control volumes that were selected for this study coincided with the gridding of the pressure level specific humidity and wind velocity in one of two ways. Data points for the gridding scheme in the reanalysis followed a 2.5 degree grid starting at 0 degrees East longitude and 90 degrees North latitude. Table 3.2 provides the coordinates of each of the box control volumes used in this study. Each control volume has two sets of boundaries for moisture flux methods 1 and 2. The control volume is shifted slightly by 1.25 degrees between methods. In the interests of keeping these control volumes over land boundaries so as to be able to calculate a surface and an atmospheric water balance, it wasn't possible to move all control volumes in the same direction.

Control Volume	Method 1 Control Volumes				Method 2 Control Volumes			
	North	South	West	East	North	South	West	East
Rondonia, Brazil	6.25 S	13.75 S	66.25 W	58.75 W	7.5 S	15 S	65 W	57.5 W
Southeast US	38.75 N	31.25 N	88.75 W	81.25 W	40 N	32.5 N	90 W	82.5 W
Southwest US	33.75 N	41.25 N	116.25 W	108.75 W	42.5 N	35 N	115 W	107.5 W
Northwest Brazil	1.25 N	6.25 S	71.25 W	63.75 W	2.5 N	5 S	72.5 W	65 W
Eastern Brazil	13.75 S	21.25 S	51.25 W	43.75 W	15 S	22.5 S	52.5 W	45 W
Larger Rondonia	1.25 N	13.75 S	66.25 W	51.25 W	0	15 S	65 W	50 W
Larger US	48.75 N	33.75 N	116.25 W	101.25 W	50N	35 N	115 W	100 W
East Coast	46.25 N	31.25 N	88.75 W	73.75 W	47.5 N	32.5 N	90 W	75 W

Table 3-2: Coordinates of box control volumes for the two different moisture flux methodologies

### 3.3. Surface Water Balance

The general surface water balance equation is given as:

$$\frac{dS}{dt} = P - E - R \quad \text{Eq. (3-2)}$$

Again, spatially averaged precipitation from a number of sources and evaporation from the reanalysis is used with this equation for each time step. For all control volumes both runoff and change in surface storage can be derived from the reanalysis, however in the case of a basin sized control volume there is a much more accurate alternative source of runoff, actual stream gauge data.

#### 3.3.1. Surface Storage Term

Rasmussen (1967) defined the storage term as the change in surface water, soil moisture, and groundwater. It is also stated that in a land based control volume, with a land area of at least 15000 km<sup>2</sup>, that the change in surface water will be minimal over a month or year unless there is the presence of a lake of similar scale, or part of the control volume is over the ocean. However, for all regional water balance analyzes in this study, change in surface water is seen as a factor, perhaps because in the reanalysis surface storage is such an important balancing term.

Soil moisture and change in snow cover are taken to be the primary cause in change of surface water storage, which is derived from the shallow soil moisture (0-10 cm) and deeper soil moisture (10-200 cm) data points of the NCEP Reanalysis, as well as snow cover. Soil moisture fields in the Reanalysis are given as a water level including the height of water in the soil. Water can only reside in the pores, which are assumed to be 30 % of the total space. Therefore the height of water in a “reservoir” (the surface and subsurface) is arrived at by multiplying the soil moisture by .3 and the snow height by .1. Snow is assumed to be 90 % air and 10 % water. The height of surface storage in cm can be converted to kg/m<sup>2</sup>.

The change in storage for a month was found by using one unit future and one unit past data.

$$\frac{dS}{dt} = \frac{S_{t-1} - S_{t+1}}{2\Delta T} \quad \text{Eq. (3-7)}$$

#### 3.3.2. Runoff Methodologies

The NCEP Reanalysis provides a gridded runoff dataset with the same units as the precipitation and evaporation. A gridded runoff dataset is counterintuitive as runoff is typically measured at a point. A gridded runoff dataset will likely include many points of runoff leaving the grid point in many directions, if it is of any significant size. However, given the constraint of a box control volume it is impossible to identify all point runoff fluxes which flow through the control volume boundaries, as these points are numerous and for the most part not gauged. Therefore the

gridded runoff dataset given by the reanalysis is useful, but methods must be devised to determine the routing of the runoff. These methods include the use of Oki's TRIP pathways.

### 3.3.2.1. Uses of TRIP Pathways in combination with Reanalysis Runoff

In order to determine the runoff flux through the selected control volumes the reanalysis runoff and the Oki TRIP pathways are combined. Figure 3-3 (a-e) shows a series of diagrams outlining this combination.

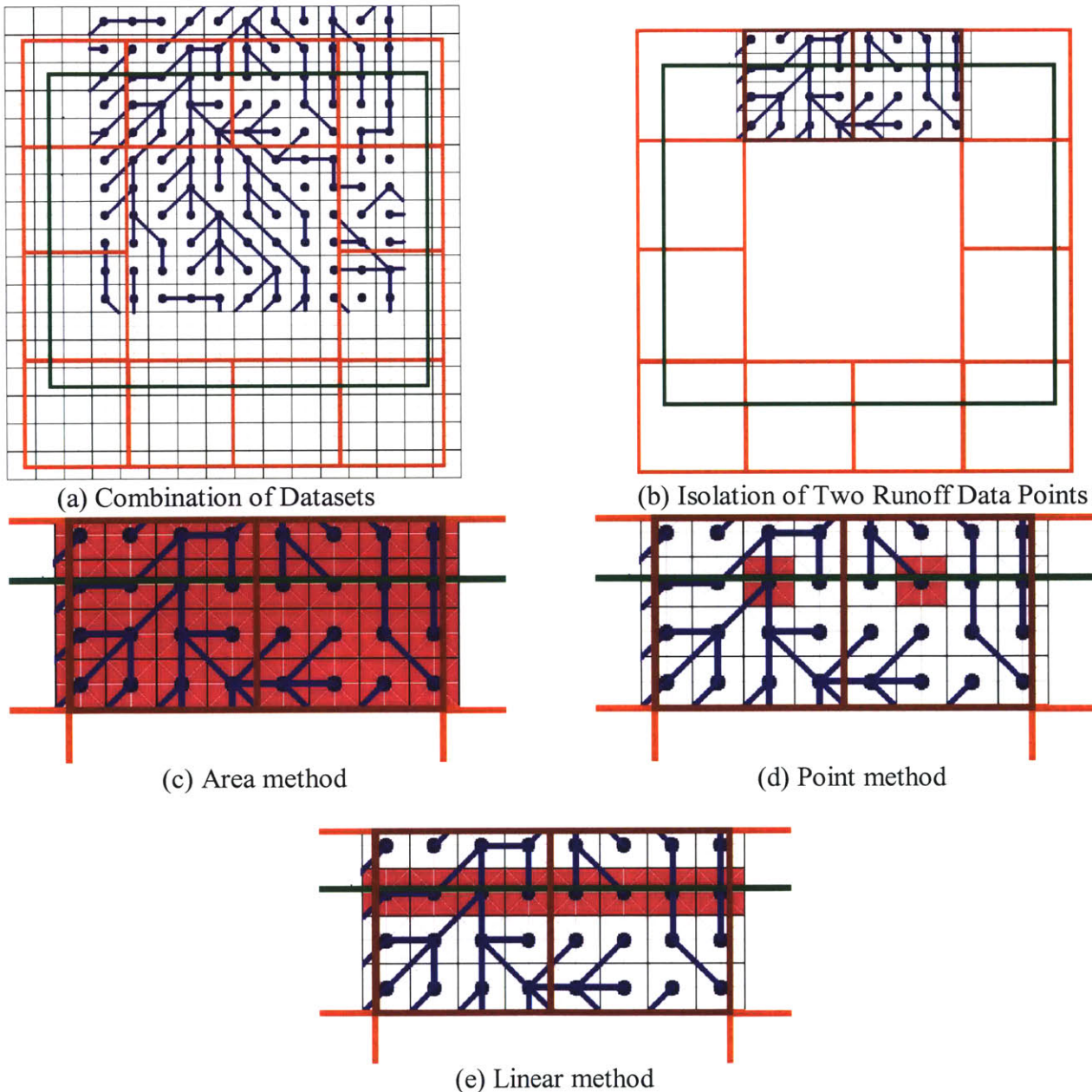


Figure 3-5: Derivation of the Runoff Routing Methodology For the Boxed Control Volume

Figure 3-5 (a) shows the integration of the datasets involved. The larger green box indicates a hypothetical control volume for which the water balance is being taken. The red boxes indicate the gridded NCEP Reanalysis runoff values, which have a resolution of slightly less than 2 degrees on a side. Each red box through which the green control volume boundary passes through represents one runoff value in the Reanalysis. The center of the control volume is not gridded because these runoff values are unnecessary in finding the flux of runoff through the control volume's sides. The blue lines represent the TRIP pathways network at .5-degree resolution. In the dataset, each node represents a direction over which runoff flows from the node (as in a river network), and these directions are used to route the gridded Reanalysis runoff (red), through the boundary of the control volume (green).

To better illustrate the procedure used to apply the runoff data to the control volume, two reanalysis runoff grid boxes are isolated in Figure 3-5 (b). As can be seen here, each Reanalysis runoff grid has several TRIP directions in it, which can be used to determine in which direction this runoff value is traveling in the larger box. If all of these directions indicate that runoff is flowing out of the control volume for a certain box, then the full runoff mass is added to the runoff total of the control volume. If the runoff directions indicate half of the water in the area is flowing in and half is flowing out, then there is no mass of runoff through the control volume boundary. The reanalysis runoff value for the red gridded boxes is given as an average over an area, and with the grid area known, this value can be converted into kg/month. This mass can then be multiplied by the fraction of this runoff, which is traveling outwards from the control volume based on the TRIP pathways. This fraction may be negative if all directions point inward. Summing the masses of water exiting the control volume and subtracting the mass of the water entering the control volume yields the total runoff mass per month, which can then be spatially averaged over the control volume so that it's units are the same as the other properties in the water balance.

Because the runoff data available is gridded at a much coarser resolution than the routing data available, this study evaluates three separate methodologies for the use of the routing data, which are shown in Figure 3-5 (c-e). It is impossible to route runoff data at a coarser resolution than that which is used here because too many "rivers" are lost at such a coarse resolution. Each method involves using a certain number of the available routing directions within the runoff data grid in order to determine how much of the runoff runs through the control volume boundary. Runoff directions, which run perpendicular and outward from the control volume, are weighted with a value of 1. Runoff in the opposite direction is weighted with a value of -1. Runoff directions which are parallel to the control volume boundary are weighted by zeros, since they don't run through the control volume boundary, and directions which run diagonal to the boundary are weighted either .5 or -.5 depending on whether they are oriented partially outward or partially inward to the control volume respectively.

The difference between methodologies lies in the directions that are used to determine how much runoff is passing through the control volume boundary. Diagram (c) indicates the first method, the area method. By this method the directions of runoff at all of the nodes in the area of the gridded runoff piece of data are used to determine the routing of that runoff value. There are usually about sixteen directions in such cases. Diagram (d) indicates the second method the point method. In this method only one point, the point in the runoff data box which is closest to

the center of the runoff grid in one direction and closest to the control volume boundary in the other direction. Diagram (e) indicates the third method the linear method. In this method only routing points through which the control volume boundary passes are used.

It is proposed that the linear method is the most accurate way to route the runoff through a control volume boundary. By this method the TRIP pathways which are used are the ones which are closest to the control volume boundary. This is really the only place we care how much runoff is passing through. Although the runoff data itself is coarse, the routing of the data can be accomplished in a more precise manner around the boundary of the control volume, and it is the directions around this boundary which we are most interested in. The point method doesn't use all the data available, and the area method uses runoff routing that can be up to 100 km away from the control volume boundary. The point and area method, however may be more effective in certain cases and therefore should at least be considered.

### 3.3.2.2. Basin type control volumes

In order to better utilize the alternative source of runoff data an alternative method was derived for the creation of a control volume in the shape of a basin. Four of these control volumes were created for the Mississippi River Basin, the Arkansas River Basin, the Amazon River Basin, and the Madeira River Basin. The Madeira River Basin is gauged at, and from here on referred to, as the Porto Velho Basin. The Porto Velho and Arkansas River Basins were chosen because of their comparable sized to the 7.5 degree on a side box control volumes, and the Mississippi and Amazon River basins were chosen as being comparable in size to 15 degree on a side control volumes. Additionally we have a contrast of data rich and data poor areas as well as a contrast between tropical and non-tropical areas. Table 3-3 gives the approximate land areas of these control volumes based on the runoff data used. The basins are derived from a one degree grid and are drawn out in later Figures 3-6, 3-7, and 3-8, as their derivation is described.

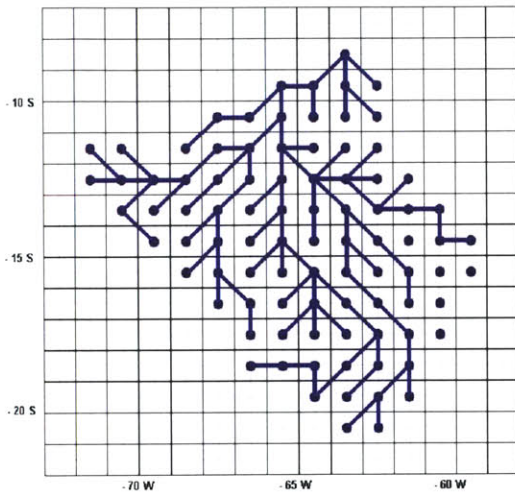
Control Volume	Land Area (km <sup>2</sup> )
Amazon River Basin	4,620,000 km <sup>2</sup>
Arkansas River Basin	318,000 km <sup>2</sup>
Mississippi River Basin	2,350,000 km <sup>2</sup>
Porto Velho River Basin	954,000 km <sup>2</sup>

Table 3-3: Basin Sized Control Volume Land Areas

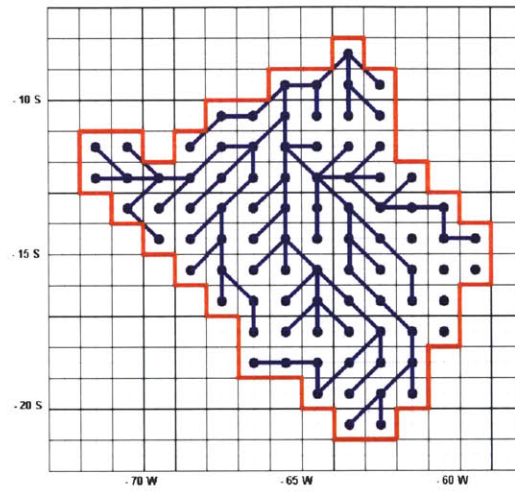
Figure 3-6 shows the evolution of the basin sized control volume. A number of different datasets are combined here to compute the water balance over an area that in theory has one runoff outlet, a gauged point on a large river. The reanalysis is used again for evaporation and moisture flux calculations as well as all relevant storage calculations. Precipitation is altered as before over a dozen alternative precipitation datasets. Runoff can also be calculated for this control volume by the reanalysis, however given the coarser resolution of the runoff grid and the finer resolution of the border of the basin, the reanalysis calculated runoff may not be as accurate.

The procedure for the derivation of a basin control volume is as follows. First a river basin must be selected that can be resolved reasonably well over the one-degree resolution TRIP pathways. The one-degree resolution of the TRIP pathways is used in this case in order to make the

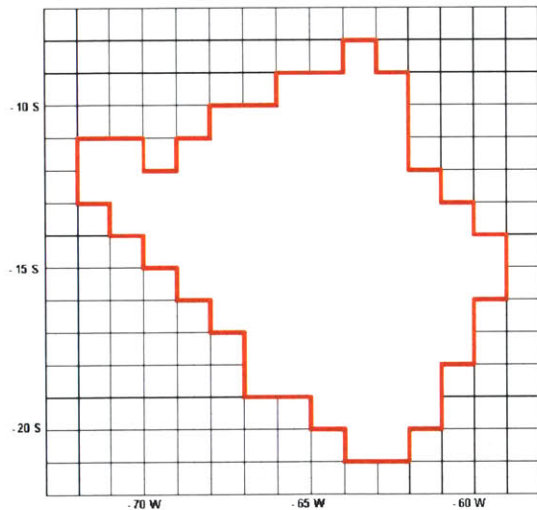
boundaries of the control volume a reasonable size for the gridded 2.5-degree moisture flux calculations, and is acceptable for river routing because the rivers which are being studied are large enough so that they can be resolved by a one degree grid. The TRIP pathways, which converge to this point, are then found. For the Madeira River basin, which is gauged at Porto Velho, Brazil in the RivDis dataset this layout is given in Figure 3-6(a). The basin is then drawn in Figure 3-6 (b) including all the grid points that lead to the outlet where runoff is known. The pathways are removed in the diagram so the actual shape of the control volume can be seen in Figure 3-6(c). In Figure 3-6 (d) the control volume is split into boxes, the coordinates of which can be input into the programming developed for this study, which receives four boundaries, a north, south, east, and west face.



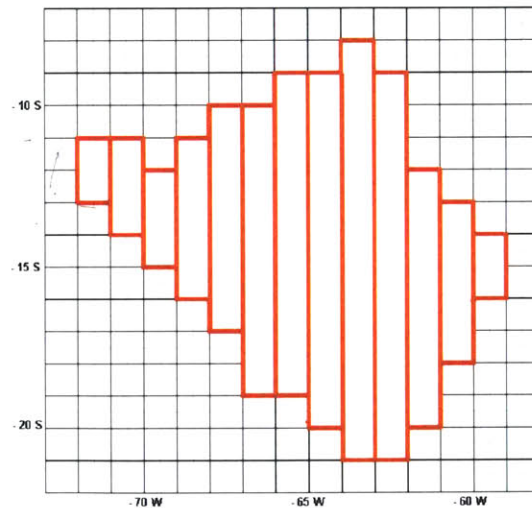
(a) One Degree Runoff Trip Grid



(b) Runoff Grid Used to Outline Basin



(c) Outline of the Basin Control Volume

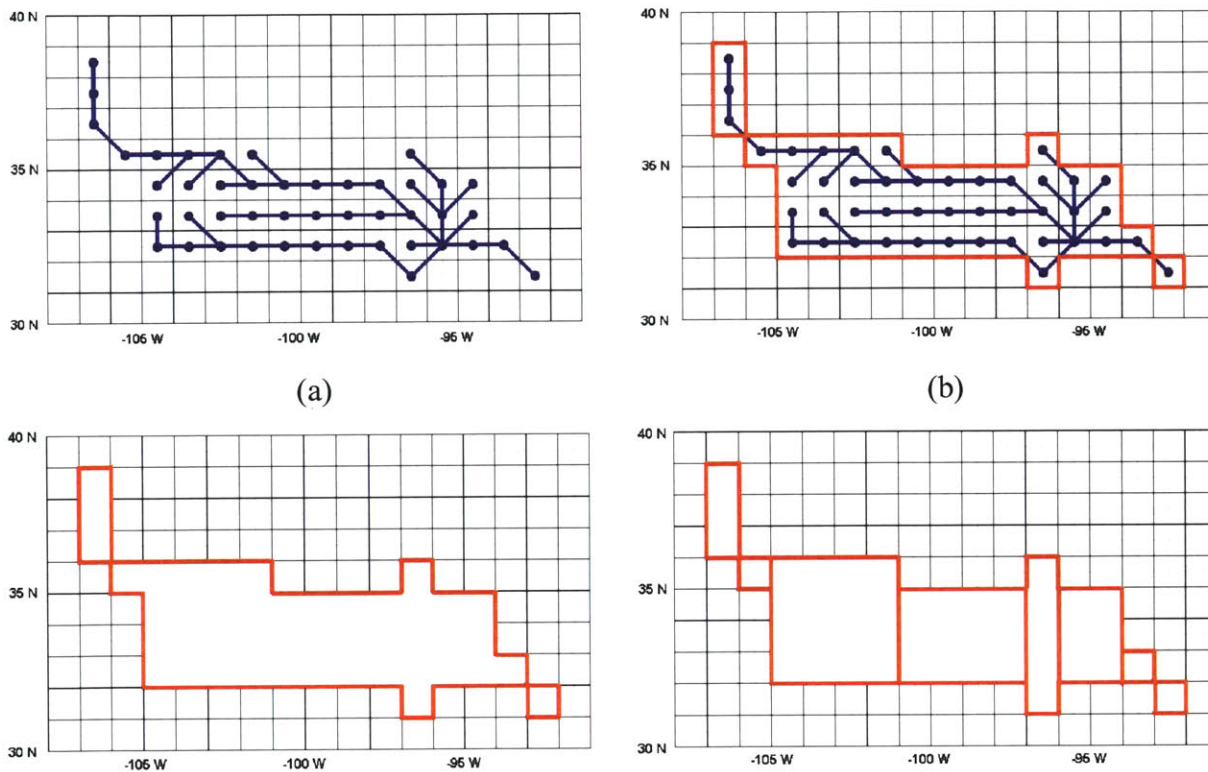


(d) Component Rectangular Control Volumes

Figure 3-6: Derivation of the Basin Sized Control Volume for Water Balance Calculations

Madeira River through Porto Velho, Brazil

The moisture flux and runoff in and out of each control volume is summed and then divided by the area of the basin. Moisture flux calculations for the smaller control volumes that make up basin control volumes often have coordinates that are not consistent with either Method 1 or Method 2 of the moisture flux methodologies outlined earlier. In other words the border of the control volume contains points neither on points of the 2.5 degree grid or directly between points of the 2.5 degree grid. In this case the nearest points to a control volume border are used to calculate the moisture flux across that border. Because the basin control volumes are gridded on a one-degree basis, a situation does not occur in which the boundary of such a control volume is equidistant between two points.



(c) Outline of the Basin Control Volume      (d) Component Rectangular Control Volumes  
 Figure 3-7: Derivation of the Basin Sized Control Volume For Water Balance Calculations  
 Arkansas River

Figure 3-7 shows the same evolution for the Arkansas River Basin. This basin was chosen because it was similar in size to the Porto Velho River Basin. Similar control volumes were found for the Mississippi and Amazon River basins by using the trip pathways, however these basins are too large to draw out in such detail, and are presented from graphics available at the TRIP website (Figure 3-8). A more detailed coordinate schematic was derived by hand. Note that the actual basins used are smaller because the runoff data used for the balances comes from the point indicated by the black dot in Figure 3-8. In the case of the Amazon basin, the actual basin used does not include the drainage directions for much of the far eastern part of the basin.



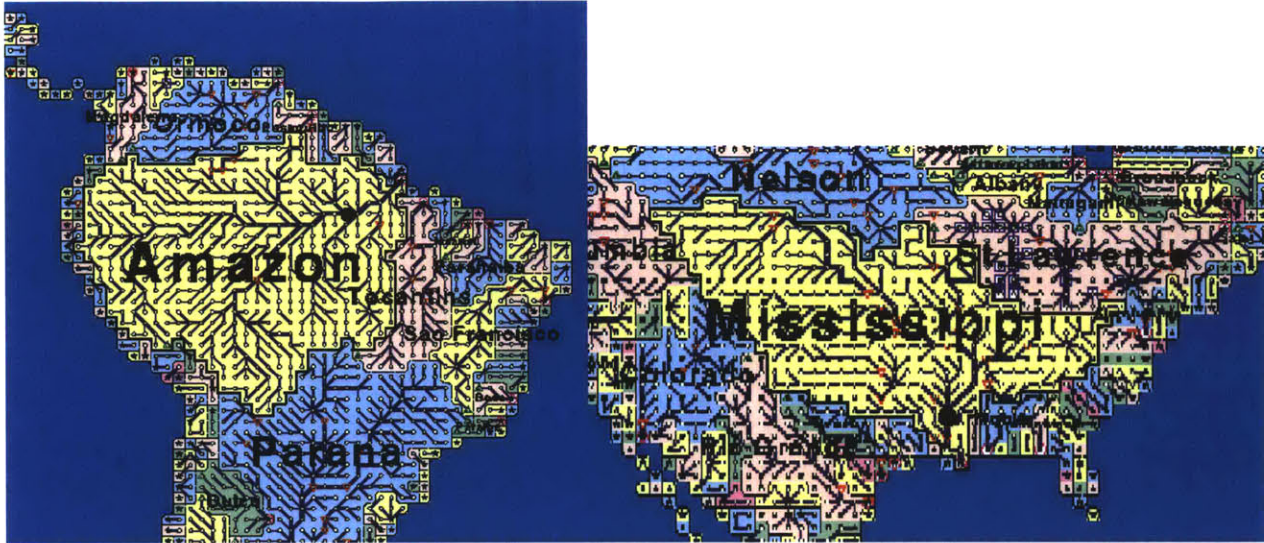


Figure 3-8: Basin Sized Control Volumes for the Amazon and Mississippi River (Oki, 1998)

### 3.3.2.3. Potential Adjustments in Storage and Lag Time

While water vapor moves through a regional control volume fairly quickly over the course of a month, water in the surface and subsurface likely moves more slowly. Therefore it is conceivable that in a month precipitation water does not completely travel through the control volume and produce corresponding runoff. The process is lagged by a number of days or months.

In an attempt to improve the surface water balance, the change in storage term was removed and instead lag times of 0, 1, 2, 3, and 4 months were tested to satisfy the surface water balance, in addition to the analysis with the change in storage term. Given  $L$  = lag time in months, and  $t$  as an arbitrary time the mass balance equation becomes:

$$-P_{t-L} + E_{t-L} + R_t = 0 \quad \text{Eq. (3-8)}$$

## 3.4 Data Analysis

### 3.4.1. Residual Analysis

Residuals are the resultant of a water balance calculation that theoretically should be equal to zero. Solving the atmospheric and surface water balance:

$$P - E - Q + \frac{dW}{dt} = residual \quad \text{Eq. (3-9)}$$

$$-P + E + R + \frac{dS}{dt} = residual \quad \text{Eq. (3-10)}$$

Smaller residuals are desired for a better balance. Residuals are calculated for all surface and atmospheric water balances in this work and analyzed in two ways, bias and absolute error.

### 3.4.1.1. Absolute Error

For the atmospheric balance the mean absolute error is:

$$AE_{atmos.} = \frac{\sum_{t=1}^n \left| P_t - E_t - Q_t + \frac{dW_t}{dt} \right|}{n} \quad \text{Eq. (3-11)}$$

For the surface balance the absolute error equation is:

$$AE_{surf.} = \frac{\sum_{t=1}^n \left| -P_t + E_t + R_t + \frac{dS_t}{dt} \right|}{n} \quad \text{Eq. (3-12)}$$

The variable n stands for the length of the dataset, and t stands for the time index. The absolute error preserves the magnitude of the error calculation throughout the time period of the water balance. This is especially important on the monthly time scale where seasonal departures from the zero balance in opposite directions may give the illusion of a good balance in the bias calculation.

### 3.4.1.2. Bias

Bias refers to the long-term average of the residuals. The equation used for this term is as follows:

For the atmospheric balance:

$$Bias_{atmos.} = \frac{\sum_{t=1}^n (P_t - E_t - Q_t + \frac{dW_t}{dt})}{n} \quad \text{Eq. (3-13)}$$

For the surface balance:

$$Bias_{surf.} = \frac{\sum_{t=1}^n (-P_t + E_t + Q_t + \frac{dS_t}{dt})}{n} \quad \text{Eq. (3-14)}$$

The bias is a measure of the agreement of the water balance averaged over the entire time period. Because this study relies primarily on the length of precipitation and runoff datasets, the bias is a good evaluation of how these datasets fit into a water balance for the time period for which they are being collected.

### 3.4.1.3. Normalization

In this work residuals are found for a number of different datasets and a number of different hydrological conditions around the globe. In order to compare residuals for different water balances, the absolute error and bias statistics, as well as the root mean square error of the correlations terms are normalized by the average precipitation of the given water balance. Precipitation is taken as a normalization factor because it is the most basic component of the water balance. For this normalization the statistic is simply divided by the average precipitation of all points used for the particular balance. The equations for the normalized residual statistics are shown below.

For the atmospheric water balance:

$$\text{Normalized } AE_{atmos.} = \frac{\sum_{t=1}^n \left| P_t - E_t - Q_t + \frac{dW_t}{dt} \right|}{n \times \bar{P}} \quad \text{Eq. (3-15)}$$

$$\text{Normalized } Bias_{atmos.} = \frac{\sum_{t=1}^n (P_t - E_t - Q_t + \frac{dW_t}{dt})}{n \times \bar{P}} \quad \text{Eq. (3-16)}$$

For the surface water balance:

$$\text{Normalized } AE_{surf.} = \frac{\sum_{t=1}^n \left| -P_t + E_t + R_t + \frac{dS_t}{dt} \right|}{n \times \bar{P}} \quad \text{Eq. (3-17)}$$

$$\text{Normalized } Bias_{surf.} = \frac{\sum_{t=1}^n (-P_t + E_t + \nabla Q_t + \frac{dS_t}{dt})}{n \times \bar{P}} \quad \text{Eq. (3-18)}$$

In cases when only one control volume is being analyzed, the normalization is not necessary and regular bias and absolute error terms can be used.

### 3.4.2. Correlation Analysis

Solving for P-E in the atmosphere and land surface balances equations (3.1 and 3.2) results in the three elements of Eq. 3-19.

$$P - E = Q - \frac{\partial W}{\partial t} = R + \frac{\partial S}{\partial t} \quad \text{Eq. (3-19)}$$

(1)      (2)      (3)

Although each of the three comparison terms are actually comprised of two terms each, the change in atmospheric storage and surface storage terms are known to be smaller in nearly all cases as compared to moisture flux and runoff respectively. Likewise, the evaporation term follows a more regular seasonal and yearly cycle than precipitation. In each of the comparison terms one variable is clearly more dependent (precipitation, moisture flux, and runoff) at the time index and particular data set used than the other variable (evaporation, change in precipitable water, and change in storage).

In theory each of the three quantities outlined above should equate to each other for each time step in the water balance. Taking any two time series and plotting the quantities against each other a 1:1 scatter plot should result. Of course, this theoretical plot is never obtained, and we must use comparative statistics in order to evaluate how well the theoretical relation is met. Three comparisons are possible, an atmospheric balance comparison (terms (1) and (2)), a surface balance comparison (terms (2) and (3)), and a total balance comparison (terms (1) and (3)). The comparison can be plugged into a linear regression, and additionally a correlation coefficient and the root mean square error can be calculated for all three comparisons. Such correlations can be hereafter referred to as the atmospheric regression or correlation, the surface regression or correlation, and the total regression or correlation.

### 3.4.2.1. Linear Regression

A linear regression of a two sided water balance can be conducted to produce a slope, intercept,  $R^2$ , and  $R^2$  1:1. The lesser-known predictor is graphed on the y-axis and regressed against the actual value on the x-axis. Unfortunately, in this work there is no actual known value, and we can only assume one set of data is more reliable than the other. This study makes this classification based on the ease by which sides of the balance are calculated. This classification is altered based on whether a box sized control volume or a basin sized control volume is being used.

For the box control volume water balance the classification is as follows:

- 1) P-E
- 2)  $Q - dW/dt$
- 3)  $R + dS/dt$

Thus, P-E is always taken as the actual value, and the moisture flux term is taken as the actual value when the total regression is performed. Although precipitation and evaporation are considered class C data products in the reanalysis, they are derived by simple spatial averaging. The moisture flux is derived by a less direct method, which has been documented, and therefore its term is ranked second. The runoff is derived by a methodology original to this work, and therefore is likely the most experimental and variable term, thus is ranked last.

However, when the basin sized control volume is used the rankings are changed. In this case, runoff data is considered the most reliable because actual river runoff data is being used, with spatial averaging only in the change in storage term. Conversely, the moisture flux is considered to be the least accurate measure due to the fact that it is derived from the combination of a number

of smaller control volumes, where it is inferred that less points on the boundary of a control volume produce a higher probability of variability.

For the basin control volume water balance the classification is as follows:

- 1)  $R + dS/dt$
- 2)  $P - E$
- 3)  $Q - dW/dt$

Linear regressions are performed for the three comparisons (atmospheric, surface, and total), and from these regressions, a slope, intercept and  $R^2$  value corresponding to the slope and intercept are found as well as an  $R^2$  1:1 value, which is an  $R^2$  measure fitting the data to a 1:1 line. The slope of the regression is theoretically one and on the monthly scale. The slope should be a good indicator of the seasonal strength of signal of one side of the water balance compared to the other. The term signal refers to the amplitude of various peaks and troughs of the time series of each side of the water balance being considered. A slope greater than one indicates the y-axis property has a stronger signal, while a slope less than one indicates the y-axis property has a weaker signal. Signals also exist on the yearly scale, to a lesser degree.

The intercept of the regression is another check of bias, and is useful in checking regressions for large errors and inaccuracies. The  $R^2$  value is a similar analysis tool. It measures the tightness of fit to the regressed line, but is a good measure only if the slope and intercept of the regression make sense. The perfect  $R^2$  value is 1, and at zero it becomes meaningless. The  $R^2$  1:1 value is in theory a very good metric for this study because it tests the fit of the regression to a 1:1 line. Unfortunately, the regressions in this study are rarely close to 1:1, causing this metric to become very largely negative in most cases. However, in some cases the statistic is still useful.

### 3.4.2.2. Correlation Coefficient

The correlation coefficient is a useful metric like the RMSE because it allows for the comparison of two independent variables, either of which may be faulty. The coefficient is the off diagonal term of the covariance matrix obtained by comparing a time series of two water balance variables (i.e. A, B) and is found by the formula:

$$CorrCoef = \frac{\sum_{t=1}^n ((A_t - \bar{A}) \times (B_t - \bar{B}))}{\sqrt{\sum_{t=1}^n ((A_t - \bar{A})^2 \times (B_t - \bar{B})^2)}} \quad \text{Eq. (3-20)}$$

The correlation coefficient is a normalized relationship of the linear relationship strength between variables independent of slope or intercept. The correlation coefficient will vary from -1 to 1, with 1 representing a perfect linear correlation, 0 representing no correlation, and -1 representing a perfect inverse correlation. The linear regression analysis finds the slope and intercept of the relation to find the differences in the regression from the predicted 1 to 1 regression. Unfortunately in this study good linear correlations, and correlations to the 1:1 line

are rare, and it is more useful to look for any correlation possible as evaluation tool for the potential faulty data. Because this coefficient is normalized between -1 and 1, the evaluation statistic is more under control for some of the potentially volatile data that will be looked at in this study. In some it is important to simply see if a correlation exists at all. The normalization in this case is performed based on the actual data rather than the precipitation magnitude, which is an added advantage for the use of this statistic.

### 3.4.2.3. RMSE

The root mean square error gives an error term to the two independent data sources:

$$RMSE = \sqrt{\frac{\sum_{t=1}^n (A_t - B_t)^2}{n}} \quad \text{Eq. (3-21)}$$

The root mean square error is a standard comparative statistic that gives a good representation of the magnitude of the error between two sides of a water balance. Because the resulting statistic has the same units as the water balance variables themselves, the term can be normalized, so different control volumes may be compared. The measure is most like the absolute error and bias measurements used before in that it gives a magnitude of error, therefore it is normalized by precipitation.

$$Normalized\ RMSE = \frac{\sqrt{\frac{\sum_{t=1}^n (A_t - B_t)^2}{n}}}{P} \quad \text{Eq. (3-22)}$$

## 4. Data Analysis of Water Balance Methods

In this Chapter, differences in water balance methodology will be discussed for all of the control volumes. This initial analysis will indicate which methodologies are more effective when computing regional water balances. Therefore, where applicable, normalized data analysis tools will be utilized. In order to analyze which methods are best suited for the computing of water balances a set of default characteristics must be given, so that they may be held constant while other characteristics are altered. Table 4-1 lists the characteristics which will be studied, the possible alternatives, and the default conditions.

Water Balance Characteristics	Alternatives	Default
Moisture Flux Methodology	Method 1, Method 2	Method 2
Runoff Methodology	Point, Area, Line	Line
Surface Storage Method	dS/dt, 0-4 month lags	dS/dt
Reanalysis	1 and 2	Reanalysis 2
Precipitation	13 Types	Reanalysis 2

Table 4-1: Water Balance Methodologies

For moisture flux methodology Method 2 is assumed as the default method. This is the aligned method, which is also used for the global analysis, and is the more conventional method using the actual grid of the reanalysis. The default runoff methodology used is the line methodology. As discussed before, this is theoretically the best way to calculate runoff. The default surface storage methodology uses the dS/dt term in the equation and no time lag. The Reanalysis-2 is used as the default primary source of data for the water balances while studying methodology. Also used as the default precipitation dataset is the Reanalysis-2 precipitation. As will be shown in the precipitation analysis data section the Reanalysis-2 provides generally the best closure for water balances over its duration. Unfortunately, this precipitation dataset is known to have its problems. Alternative precipitation datasets will be looked at very closely to find their advantages.

There will be two additional sections to analyze the time scale used in the data analysis contrasting the monthly and yearly time steps with all default conditions in place and the size and shape of control volumes used. There are four distinct types of control volumes used: 7.5 degree box, 15 degree box, small basin, and large basin, and each of the types of control volume are represented in North and South America. The final section will give a sense of how the balances compare to each other and lead into the following Section 4.2, where each control volume will be studied independently.

### 4.1. Moisture Flux Methodology

As discussed in the procedure section two methods of obtaining moisture flux through a boundary are investigated for a box control volume. Because of the coarse resolution of the data it is hypothesized that it may be more effective to compute a water balance in between grid points (Method 1), as opposed to aligned with grid points (Method 2).

Figure 4-1 shows monthly and yearly normalized absolute error and bias contrasting the two moisture flux methods. A normalized error of one indicates error on the order of magnitude of precipitation and therefore a magnitude of error on the order of 6 or 7 would be completely useless. Problems develop in the atmospheric water balance computations of the Southwest US and in the Larger US control volumes, which contain similar areas. The mountainous nature of these control volumes may contribute to poor balance.

For the first three larger (15 degree) control volumes the aligned Method 2 has a larger absolute error over both time scales. The bias for Method 2 is also slightly larger (more non-zero) for these three control volumes. Mixed results are found for smaller control volumes in terms of absolute error. Bias calculations are consistent across monthly and yearly time scales as the bar graphs on the bottom left and the bottom right are almost identical. The bias of the water balance should be the same regardless of time scale because it is measure of the aggregate water balance, and the yearly time scale is simply the aggregate of the monthly time scale. Bias results are in general agreement with absolute error observations.

Figures 4-2 and 4-3 show the regression statistics and other statistics for the comparison of the flux through the bottom of the atmospheric control volume (P-E) with the flux through the sides of the control volumes (Q-dW/dt). All of the statistics computed in the study are shown for this analysis for both monthly and yearly time scales. For the majority of control volumes (excluding the SWUS and Larger US control volumes), the regressions are fairly good and therefore all regression statistics are useful. No clear conclusion can be made from these plots as to which flux computation methodology works better.

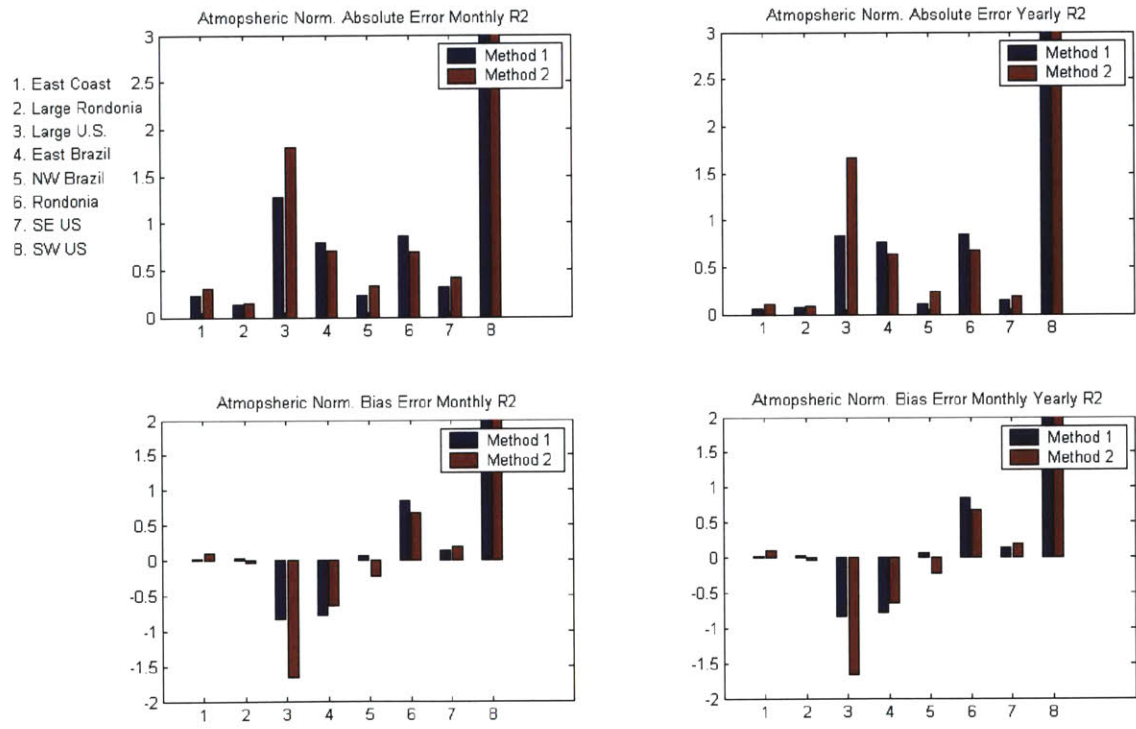


Figure 4-1: Moisture Flux Method Comparison: Absolute Error and Bias



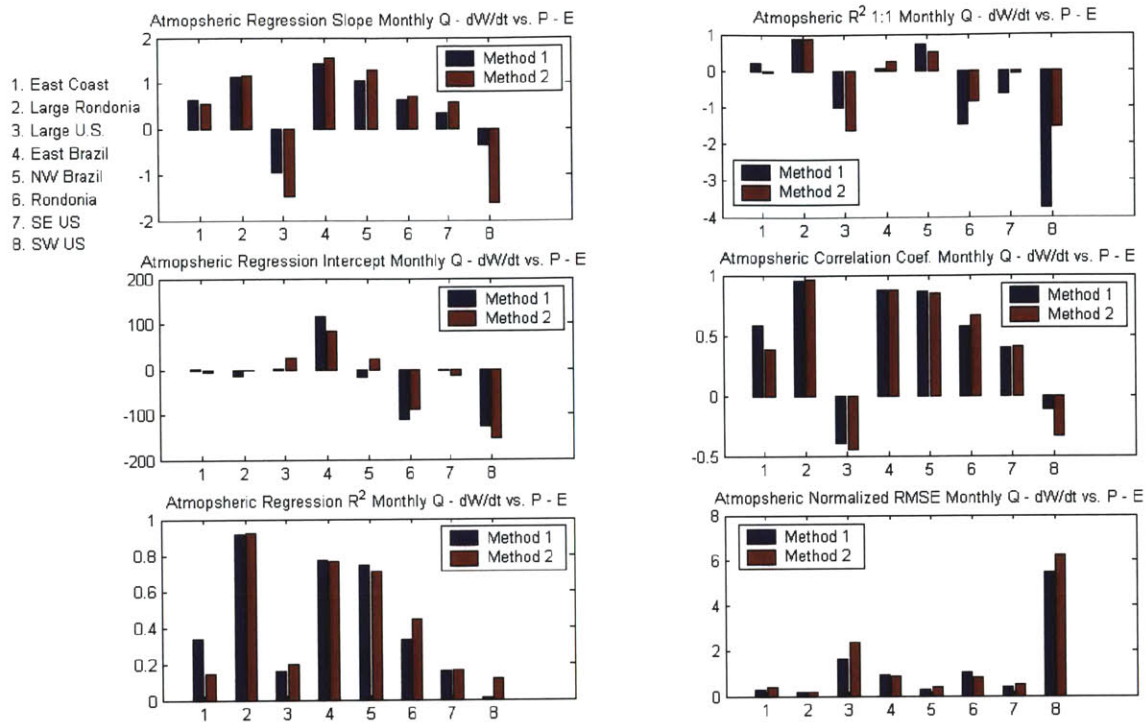


Figure 4-2: Moisture Flux Method Comparison: Monthly Correlation Statistics

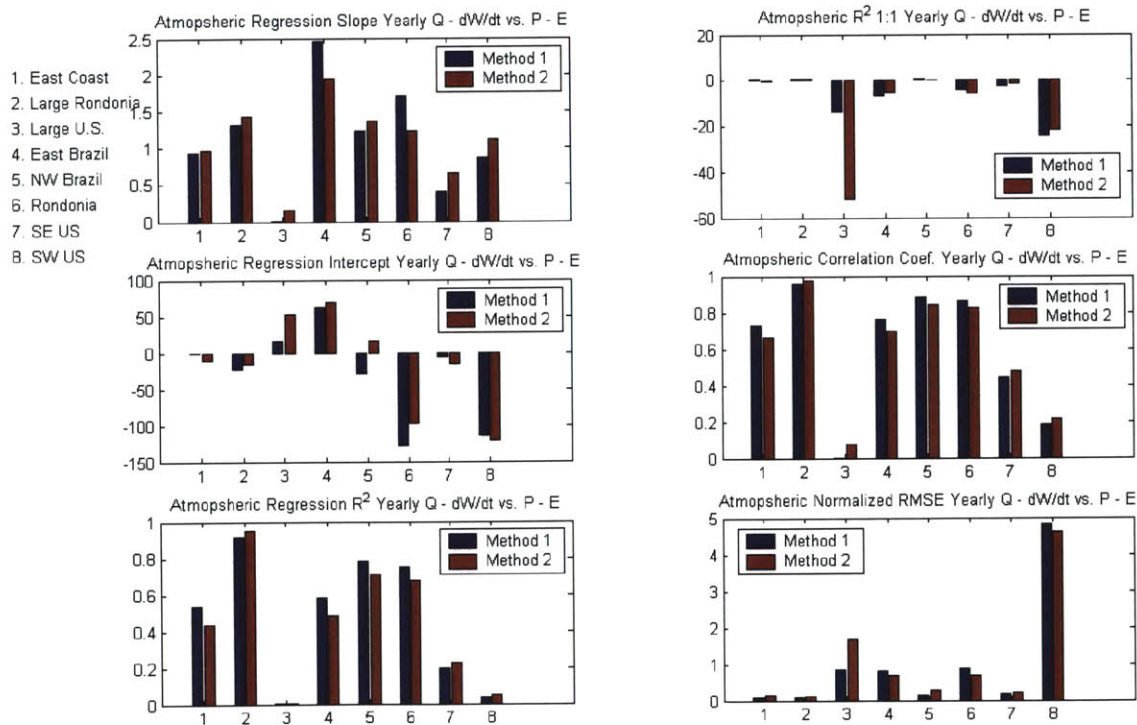


Figure 4-3: Moisture Flux Method Comparison: Yearly Correlation Statistics

Figure 4-4 shows the regression of monthly water balance data for Rondonia, the properties of which appear as the sixth column of each of the figures in 4-2. The bottom portion of Figure 4-4 shows an example time series of the two correlated properties plotted against each other. As is the case with most control volumes there is little difference between methods. A difference can be seen between time series plots of  $Q + dW/dt$  in the bottom portion of the Figure 4-4 for the given four year period. The P-E term will be the same for each case. These correlations are rather good in both cases. As shown in Figure 4-1, a positive bias can be seen for this control volume, as  $Q$  is in general underestimated and therefore the average residual is positive.

Figure 4-5 gives a yearly contrast of moisture flux methodology regressions for the Larger Rondonia control volume. The regression is better for a number of reasons. The control volume area is increased by a factor of 4, and balance is found each year not each month. There is again little contrast between methods. In general, the accuracy of this water balances are good by both methods in most cases, and it therefore could be said that both methods work. This is surprising considering the uncertainty of using only three data points on each side of the control volume.

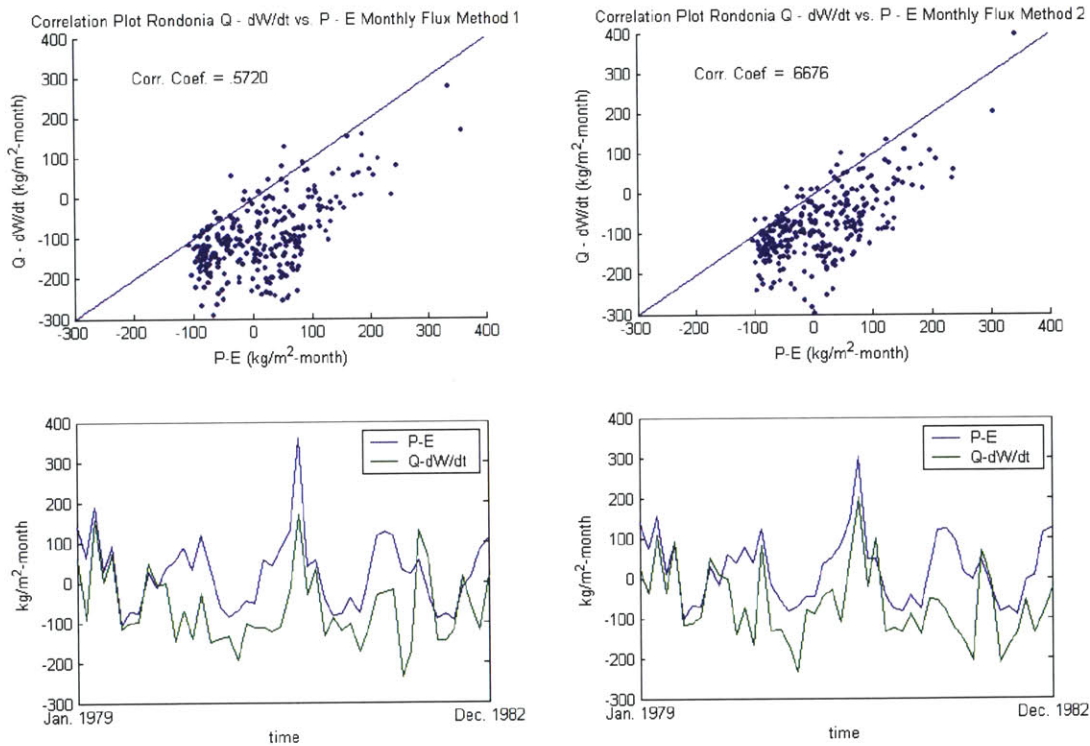


Figure 4-4: Correlation Plots for Rondonia – Different Moisture Flux Methods, Monthly

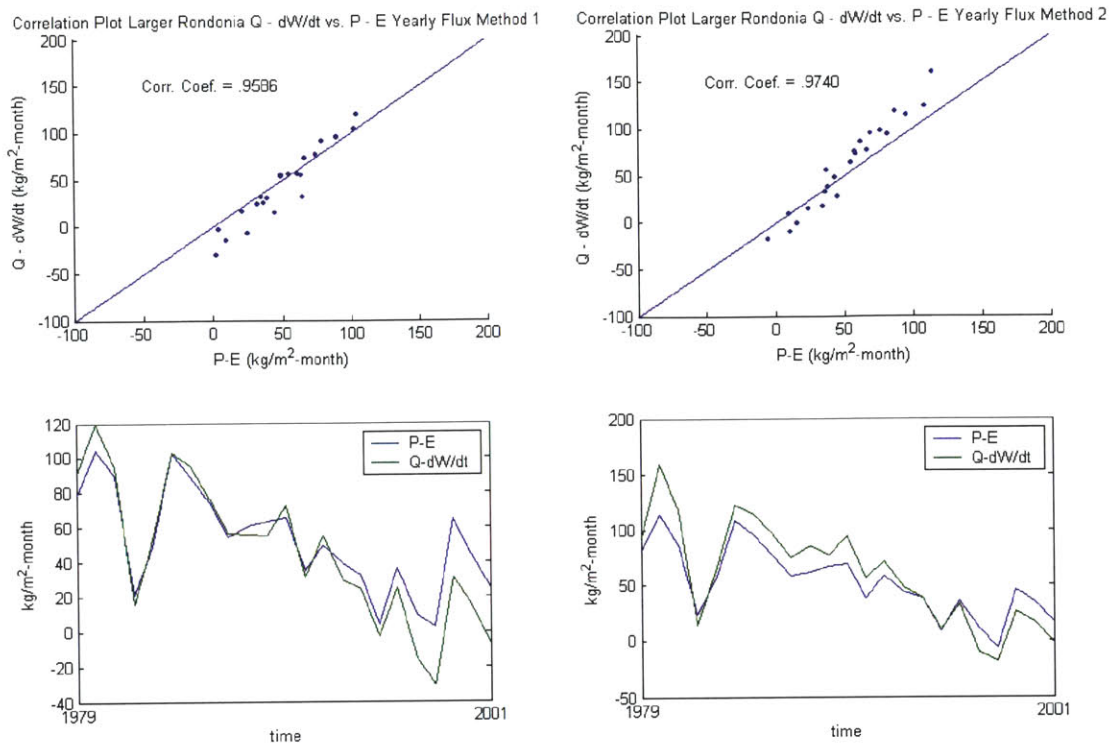


Figure 4-5: Correlation Plots for L. Rondonia – Different Moisture Flux Methods, Yearly

## 4.2. Runoff Methodology

Three methodologies have been presented for calculating the over ground runoff through the sides of a box control volume: point, area, and linear. Only the land surface water balance is relevant here. No discernable pattern is seen in the absolute error and bias calculations between these methods for calculating the surface balance, shown in Figure 4-6.

Figures 4-7 and 4-8 show the results of a regression of the surface balance for each scenario. In this case P-E is regressed against the runoff and the change in storage. As we look at methodologies to compute moisture flux in the previous section and runoff in this section we can get a sense of the ability to close the relevant atmospheric and surface balances respectively. While for the majority of the atmospheric balances relevant to moisture flux methodology we were able to find desirable regression properties, such properties may not be found for the surface water balance which is relevant to this section. Different runoff methodologies work the best over different control volumes with no systematic pattern. Looking at the correlation coefficient the point methodology is consistently the best method. In general over all evaluation the point methodology shows consistently more desirable results.

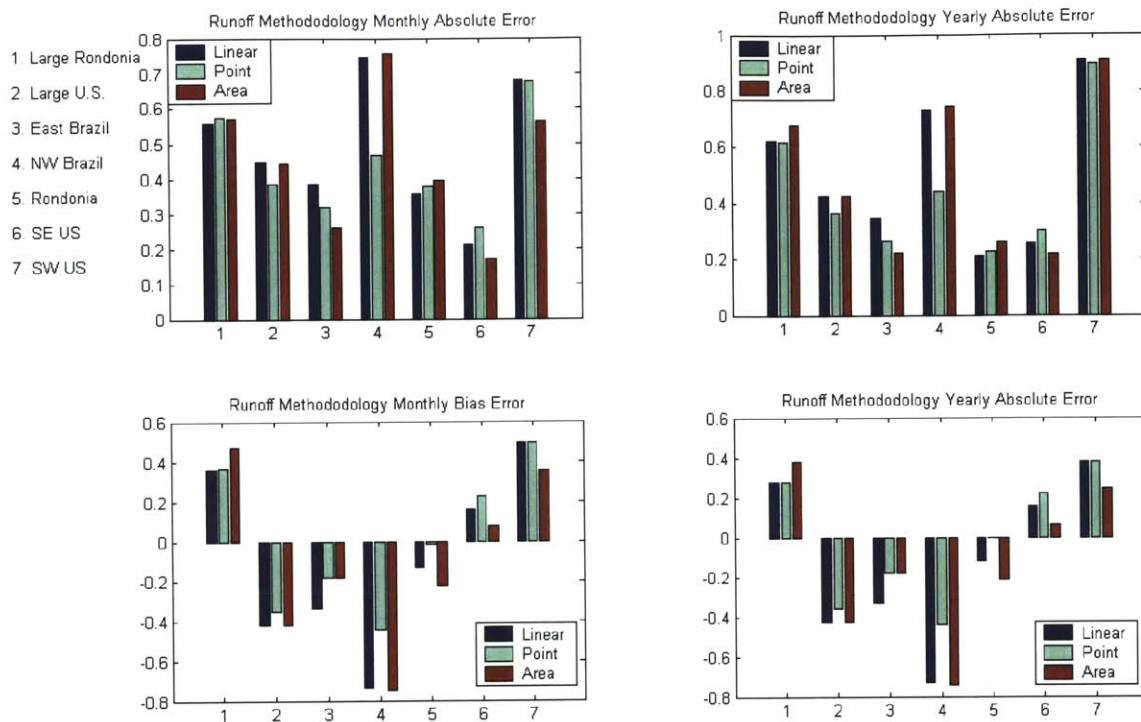


Figure 4-6: Runoff Method Comparison - Absolute Error and Bias

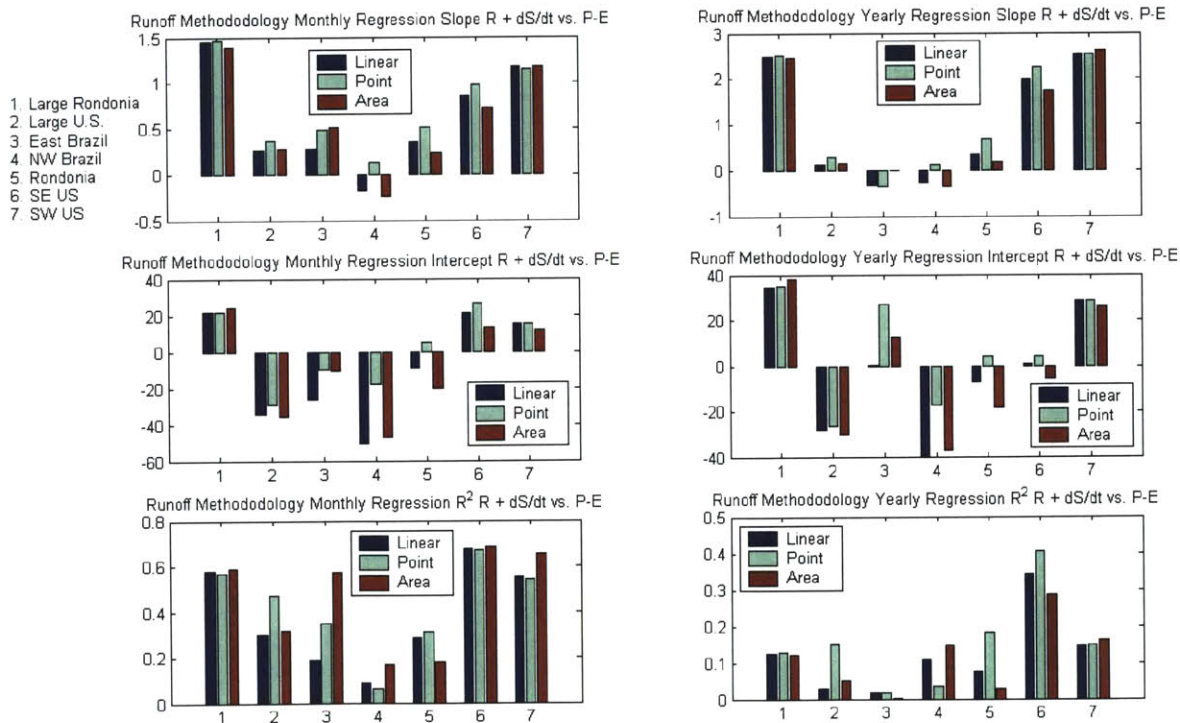


Figure 4-7: Runoff Method Comparison – Correlation Statistics 1

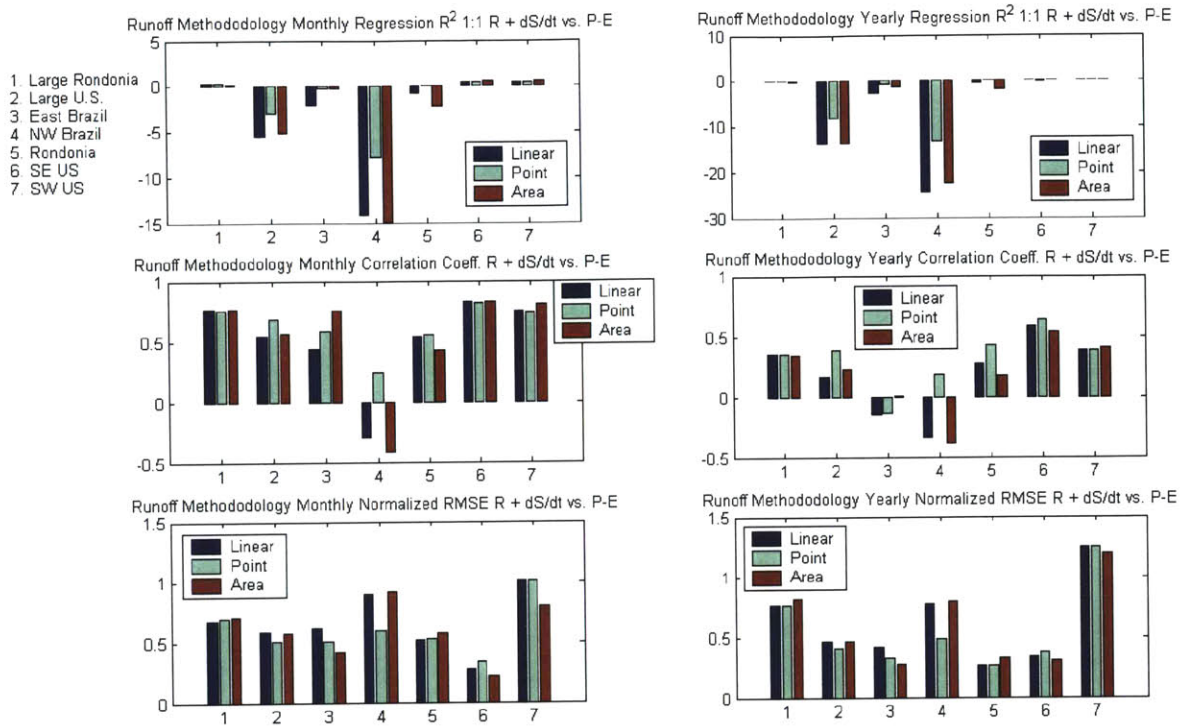


Figure 4-8: Runoff Method Comparison – Correlation Statistics 2

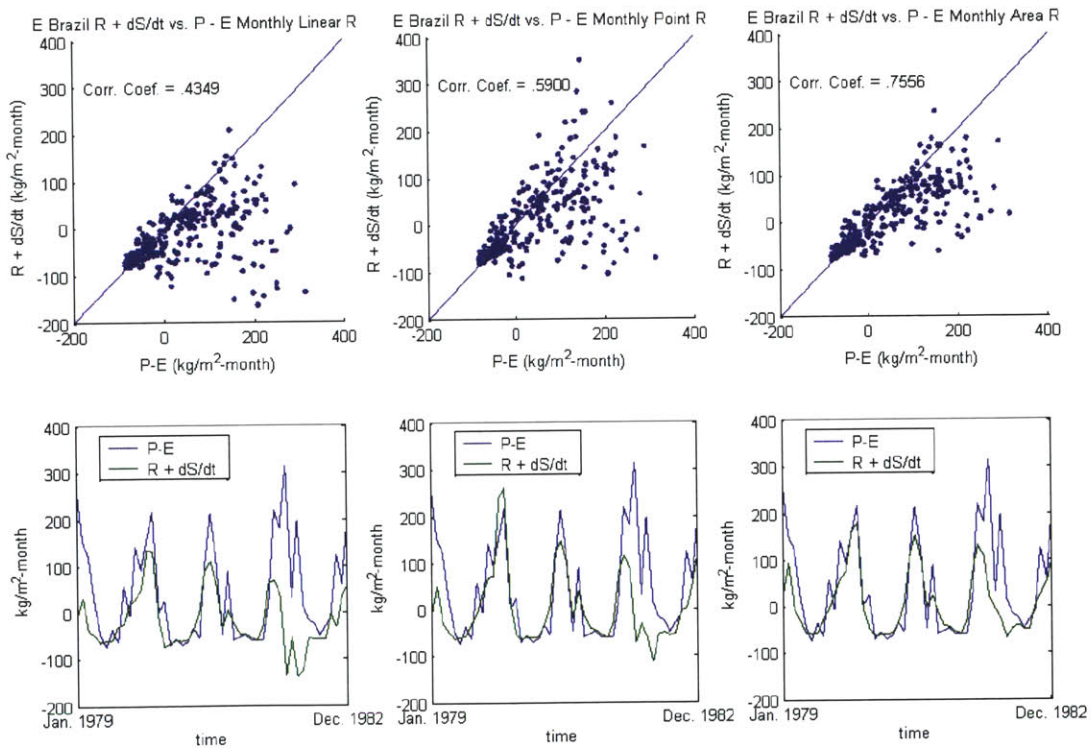


Figure 4-9: Correlation Plots for E Brazil – Different Runoff Methods, Monthly

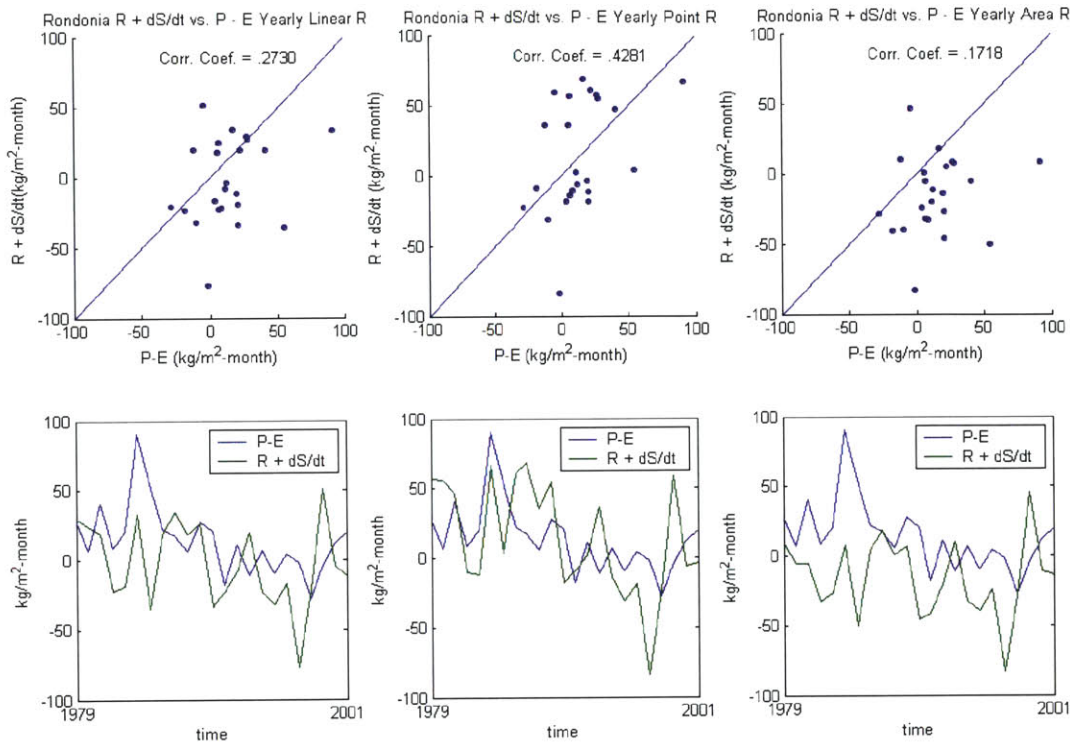


Figure 4-10: Correlation Plots for Rondonia – Different Runoff Methods, Yearly

Figures 4-9 and 4-10 give examples of the regressions whose statistics are tabulated in Figures 4-7 and 4-8. For the monthly case we refer to Figure 4-9 which shows the monthly correlation for the Eastern Brazil control volume for the three different runoff methodologies. While the P-E quantity remains constant, the runoff is altered and combined with the also constant change in storage. The overall trend of the seasonal cycle is preserved for each of the methods and the water balances well. The slopes of these regressions are all below one indicating the seasonal cycle of all of the runoff methods isn't as strong as the corresponding precipitation cycle. The yearly contrast is given as an example in Figure 4-10 for the Rondonia control volume. In this case there is no regular seasonal cycle therefore the slope statistic isn't as useful a tool, and what is more important is the correlation of the water balance as well as the bias and absolute error. Each of the three runoff methods provides a different level runoff that matches the precipitation signal reasonably well. It can be seen here that the point methodology for the Rondonia control volume, and in general, results in a more accurate overall magnitude of runoff.

In the analysis of runoff methodologies it is useful to combine the seven control volumes for the various statistics outlined, so that the methods can be contrasted directly. The seven control volumes represent a sample of the possible control volumes in the northern and southern hemisphere at varying resolutions (7.5 and 15 degrees), and have comparable statistics across the board. In other words there isn't one control volume with particularly large or small errors that would bias the averaging of the properties across each methodology.

Figure 4-11 shows the monthly and yearly atmospheric error and bias averaged over the seven control volumes. Both atmospheric error and bias error are shown on a monthly and yearly scale. The differences between the different runoff methodologies are very small for both statistics. The bias discrepancy looks misleadingly large, because the departure from zero of bias is very small for all methods. This result is encouraging because it indicates that given the sample of control volumes combined, the resulting bias is nearly zero. High and low biases cancel each other out over all control volumes selected.

Figures 4-12 and 4-13 show the regression and correlation statistics for the three runoff methodologies, averaged over all seven control volumes for both monthly and yearly time steps. Again the slopes and intercepts are consistent with what we are looking for with slopes around one and intercepts around zero. The point method is slightly better on the monthly time scale in terms of regression statistics, but not on the yearly time scale. The  $R^2$  1:1 statistic in this case is uninformative as it is negative for both time scales and all methods. However the correlation coefficient indicates good agreement between the two sides of the water balance, with an advantage given to the point methodology. The NRMSE is just about the same across each method, with errors on the magnitude of 50 % of the precipitation values.

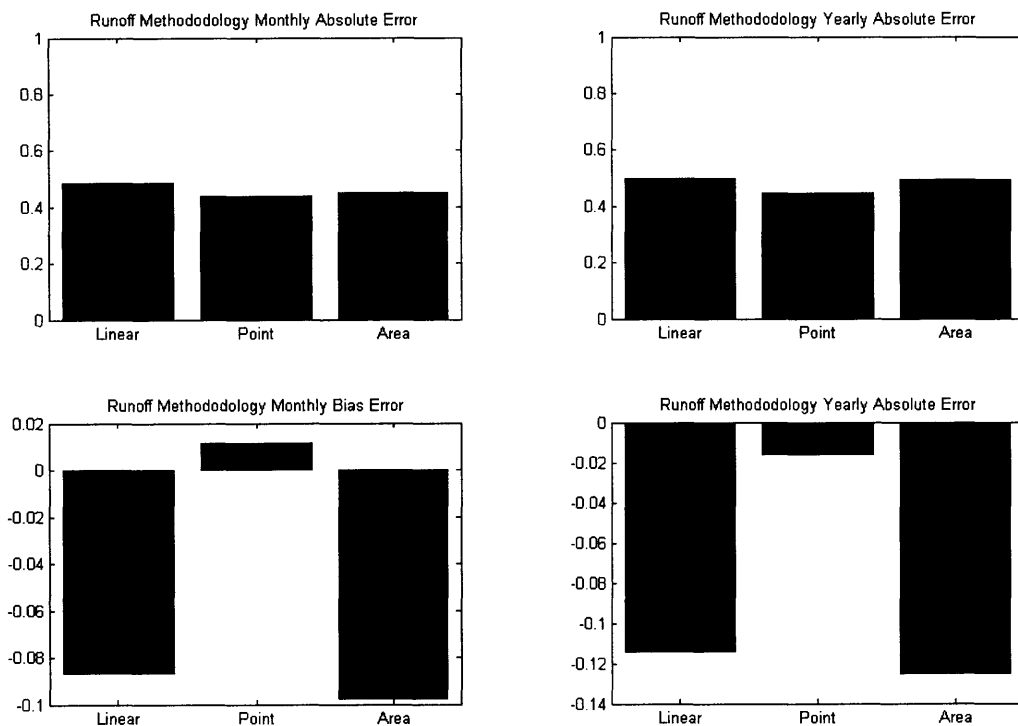


Figure 4-11: Runoff Method Comparisons - Absolute Error and Bias Avg. Across All CV's

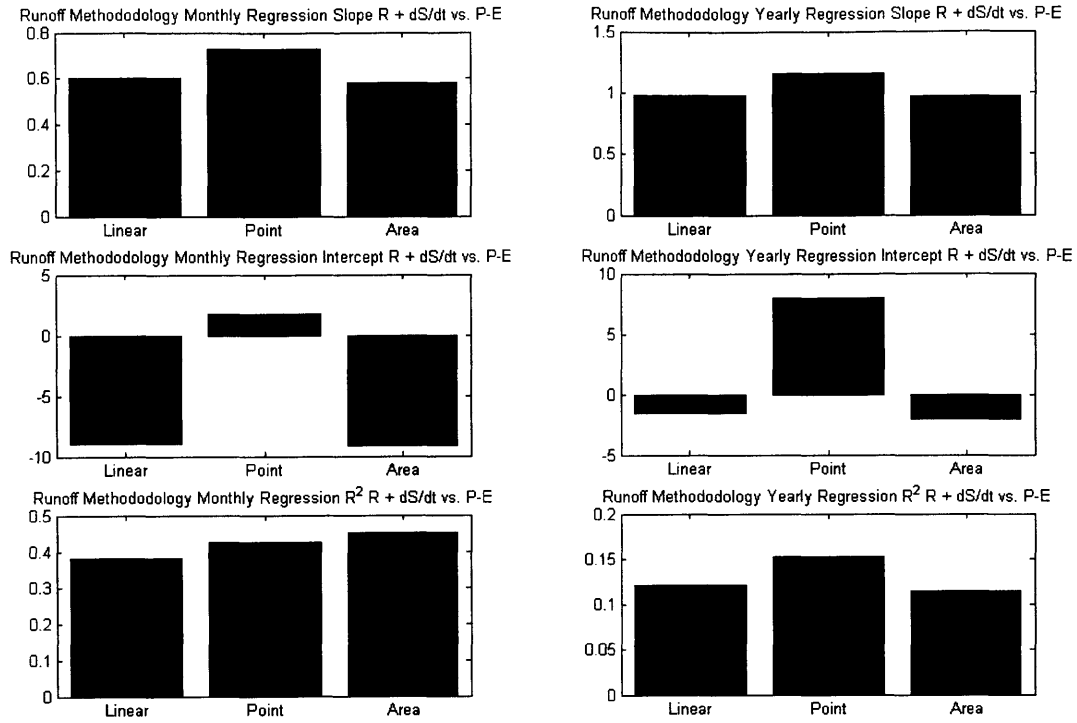


Figure 4-12: Runoff Method Comparison Regression Statistics Averaged Across All CV's

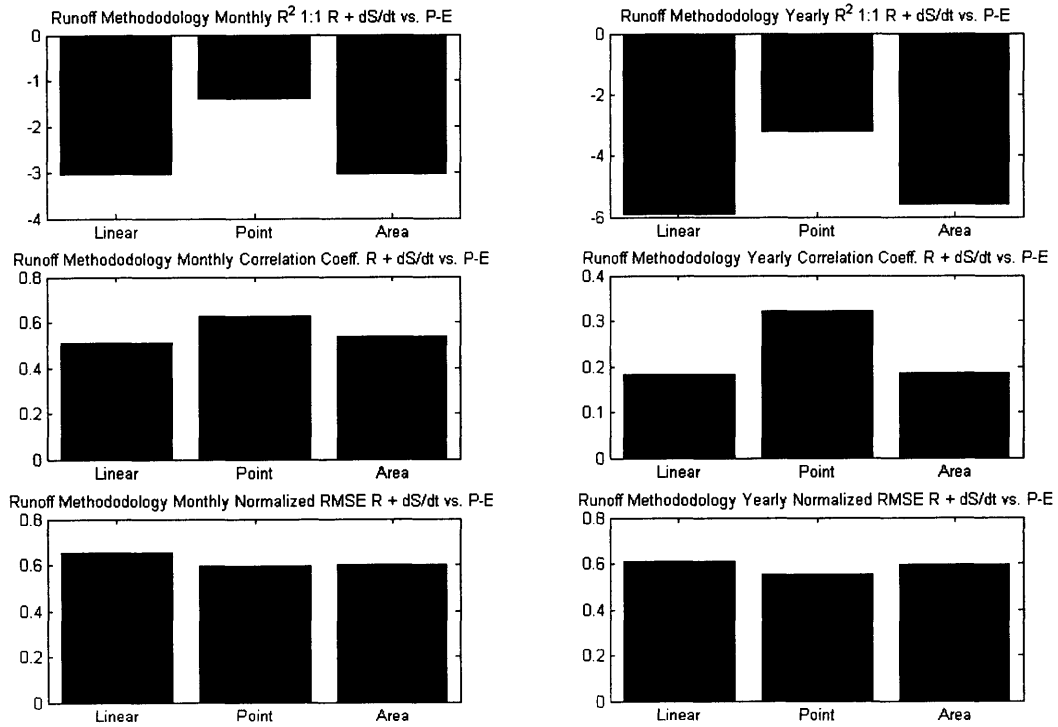


Figure 4-13: Runoff Method Comparison Correlation Statistics Averaged Across All CV's



No strong conclusion favoring a runoff estimation procedure can be reached that is applicable to all control volumes. The point methodology has a slight advantage based on this evaluation, but this conclusion does not have a strong theoretical backing. Hence, throughout the rest of this section the linear method will be used as a default because its theoretical basis is sounder.

### 4.3. Reanalysis

Alternative sources to the reanalysis are unfortunately limited to a number of precipitation and runoff datasets. However, there are two separate reanalyses, whose differences are discussed in detail in the data section. While the Reanalysis-2 is used in most of the methodology studies because it is assumed (and by most evidence is) more accurate, the Reanalysis-1 covers a much larger period of time, which is more valuable in conducting climate studies. In this section the components of the water balance are extracted as a time series for each reanalysis, and the adherence to balance is compared between the reanalyses. The duration of each dataset is tested in full, thus 55 years of the Reanalysis-1 are compared to 23 years of the Reanalysis-2.

Figure 4-14 shows the absolute error for the monthly and yearly time scales for both reanalysis datasets. For both time scales in the atmospheric balance about half of the control volumes favor each reanalysis. For the surface balance on the monthly time scale the Reanalysis-1 yields generally lower absolute error, but for the yearly time scale the Reanalysis-2 yields lower absolute error. The Reanalysis-2 has a consistently higher surface balance bias (see Figure 4-15) for most control volumes, indicating differences in the total amounts of surface water balance variables (runoff and change in storage) between the two reanalyses. The Larger US control volume is the only one which does not show the higher bias for the surface in the Reanalysis-2.

Examining Figures 4-16 and 4-17 for the monthly time scale, statistics tend to favor the Reanalysis-2. Correlations are clearly better for the Reanalysis-2, for both surface and atmospheric balances, as shown by the slope,  $R^2$ , and correlation coefficient statistics. The NRMSE looks very even for both atmospheric and surface balances. Yearly properties are presented in Figure 4-18 and 4-19. For both reanalysis products in both the atmospheric and surface balance the correction relationships are not found for the SWUS and the Larger US control volumes. Each of these control volumes contain high relief regions not present in any other control volume which may interfere with moisture flux calculations. Precipitation is also sparser and extreme over these regions, but it is odd these problems persist in the Reanalysis-2. There are no general statements for the reanalysis contrast based on the yearly atmospheric water balance. The correlations are better with the Reanalysis-2 for the yearly surface water balance but only for a limited number of control volumes.

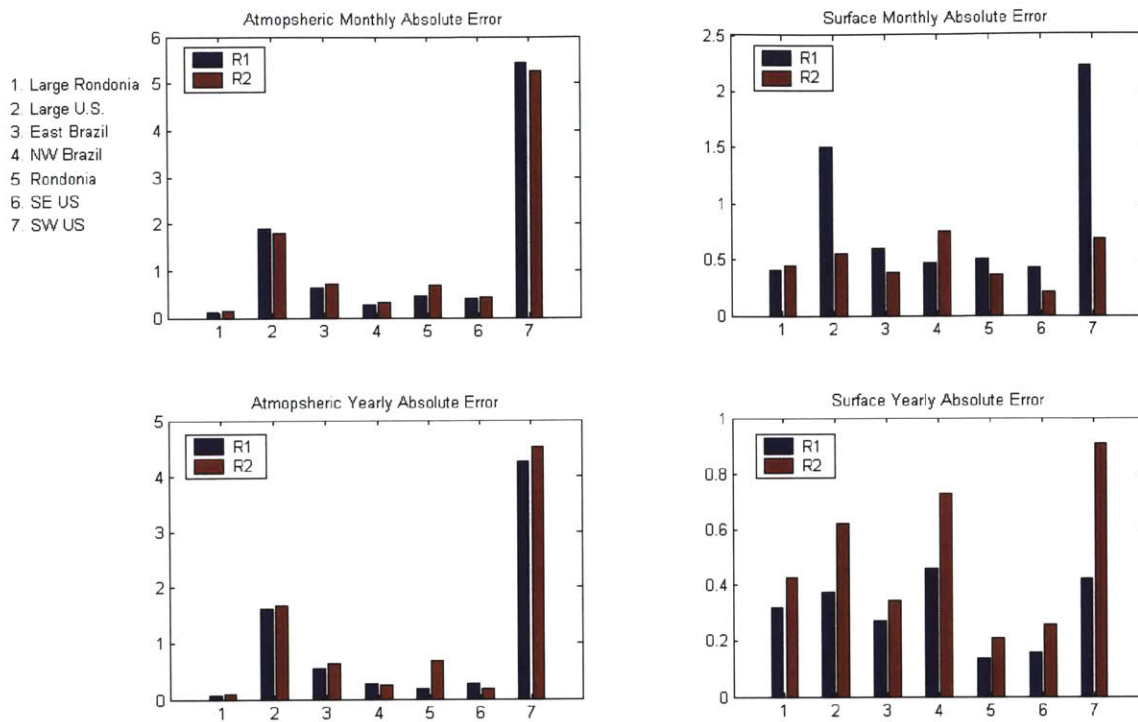


Figure 4-14: Reanalysis Comparison – Absolute Error Monthly and Yearly

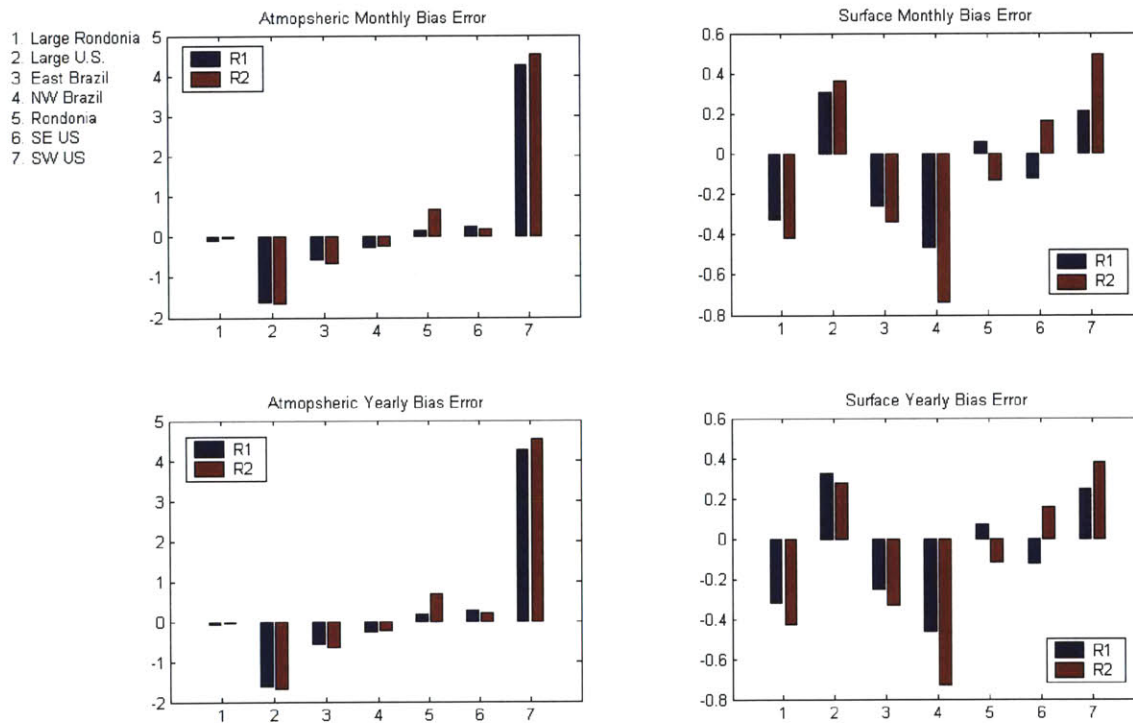


Figure 4-15: Reanalysis Comparison – Bias Error Monthly and Yearly

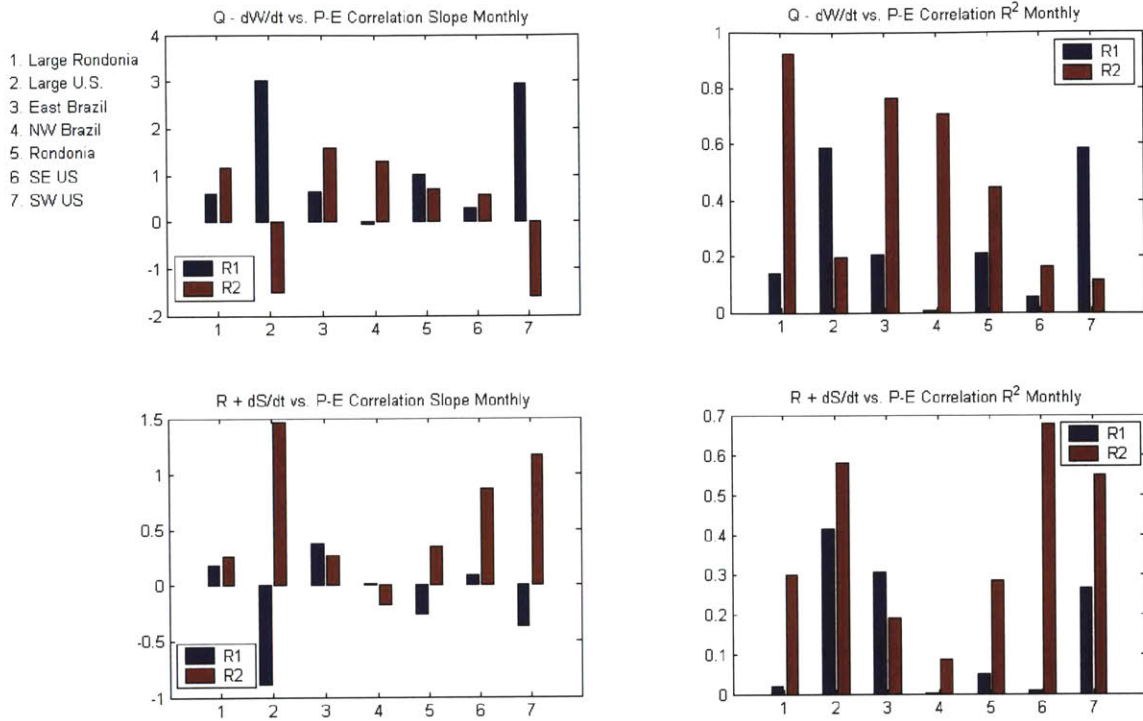


Figure 4-16: Reanalysis Comparison – Monthly Regression Statistics

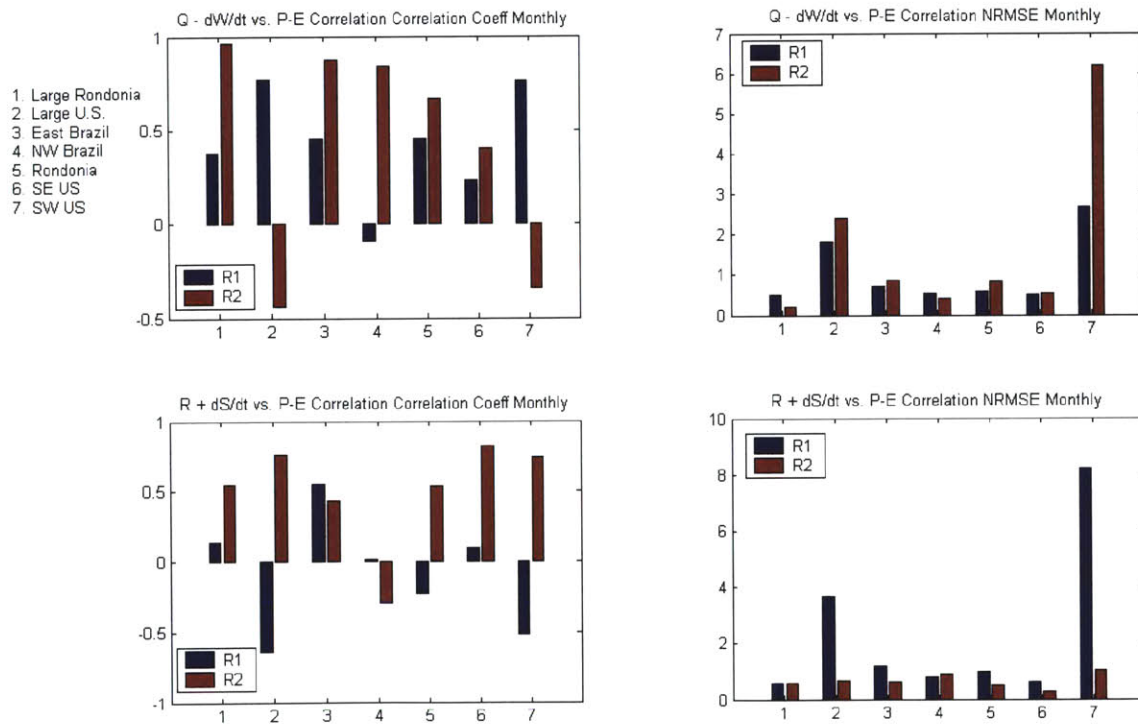


Figure 4-17: Reanalysis Comparison – Monthly Correlation Statistics

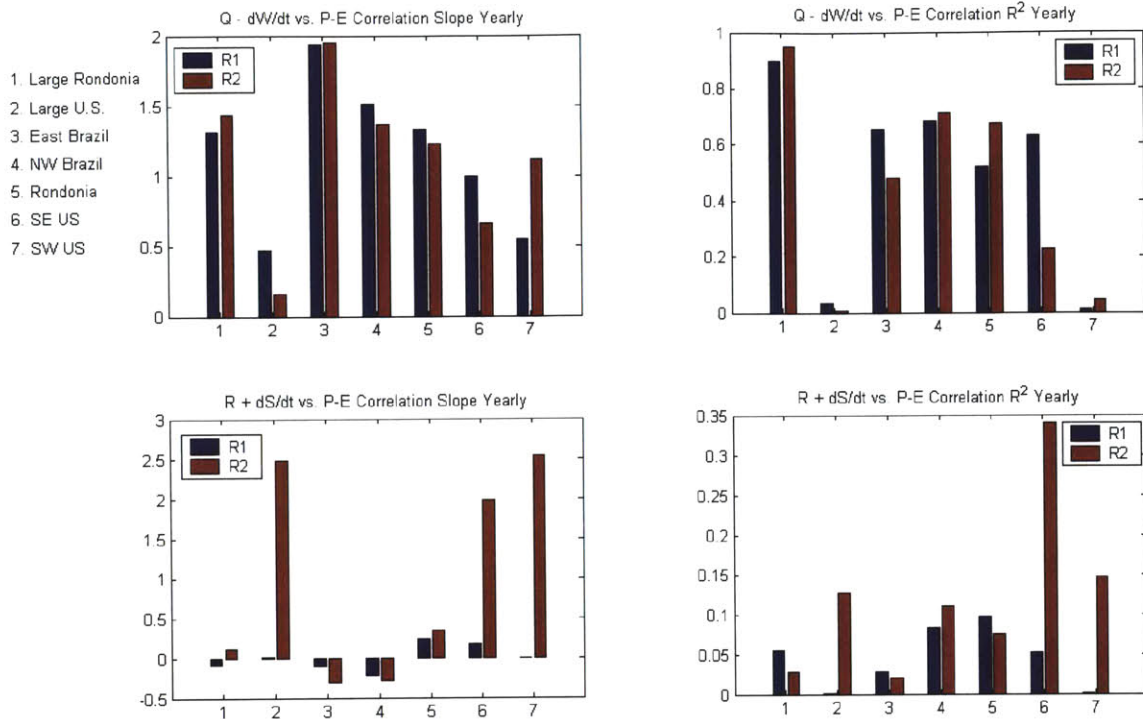


Figure 4-18: Reanalysis Comparison – Yearly Regression Statistics

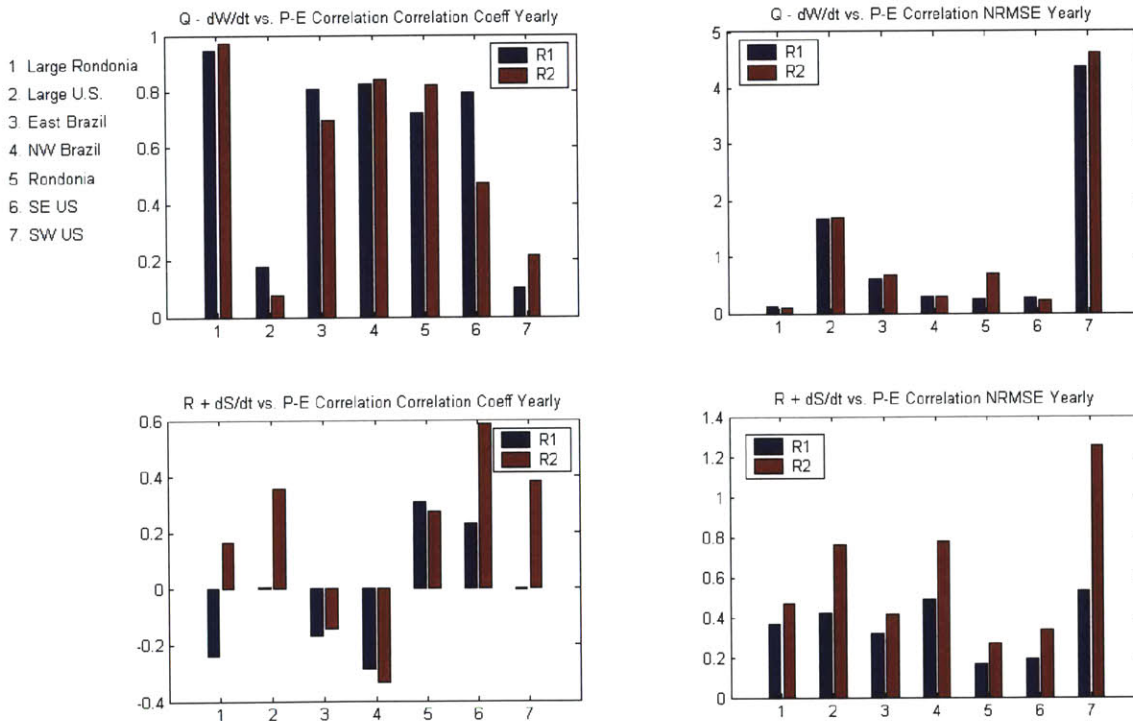


Figure 4-19: Reanalysis Comparison – Yearly Correlation Statistics

A more detailed look at the reanalysis contrast for specific control volumes is shown in Figures 4-20 through 4-23. In Figure 4-20 the East Brazil control volume is shown for the atmospheric and surface monthly time scale. For the atmospheric balance there is little change in the general shape of the regression. However, there is increased correlation and signal pick up in the case of the Reanalysis-2, as not just the yearly cycle is picked up, but smaller month to month peaks are reproduced in the time interval shown in the bottom portion of the figure. In the surface balance, the Reanalysis-2 appears to be a slightly worse balance as the seasonal signals are not as resolved.

Figure 4-21 shows the NW Brazil control volume's yearly atmospheric and surface balances. For the atmospheric balance, the yearly signals have good consistency, though a large portion of the Reanalysis-1 balance is off in the middle of the yearly time series, a time period not covered well by the Reanalysis-2. The Reanalysis-2 atmospheric balance looks better and is more tightly correlated. For the surface balance, both reanalyzes fail to close the water balance well. Since precipitation and evaporation lead to a well balanced atmospheric cycle, the conclusion is that the runoff and surface storage data is flawed. NW Brazil is a fairly remote area for which there is little hydrometeorological data, which may account for these problems.

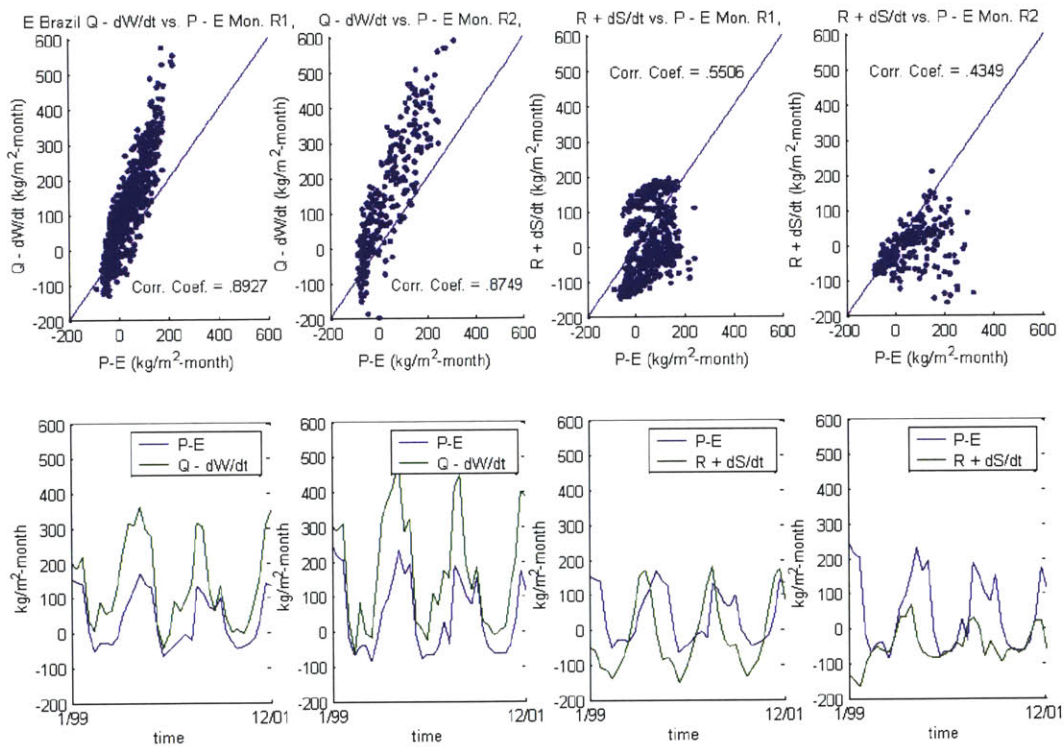


Figure 4-20: Correlation Plots for Eastern Brazil – Reanalysis Comparison Monthly

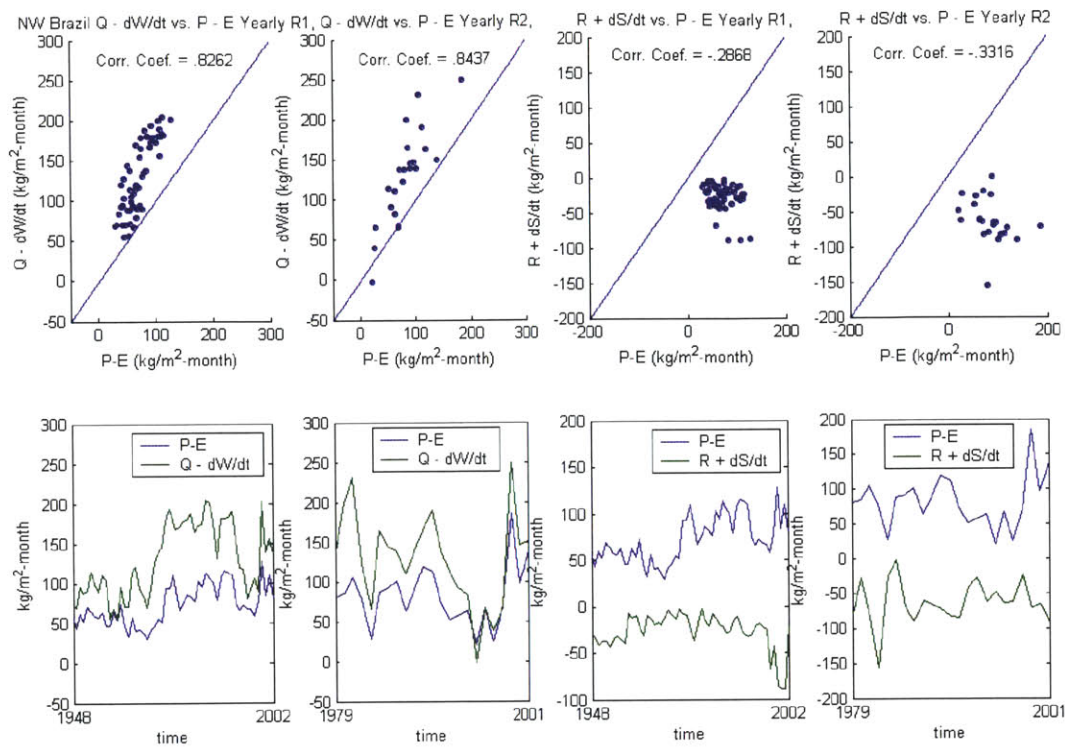


Figure 4-21: Correlation Plots for NW Brazil – Reanalysis Comparison Yearly

Figures 4-22 and 4-23 look closely at the Rondonia control volume on the monthly and yearly time scales, respectively. On the monthly time scale the observations made for the Eastern Brazil control volume still hold. The poor correlation seen in the Reanalysis-1 surface balance is due to points which aren't visible in the example plot. One important observation about the original Reanalysis-1 water balance over Rondonia is that the runoff data was vastly improved over the last five years of the balance, but wrong everywhere else. This leads to good behavior in the last five years as shown in the bottom panels of Figure 4-22, but a fairly bad regression overall.

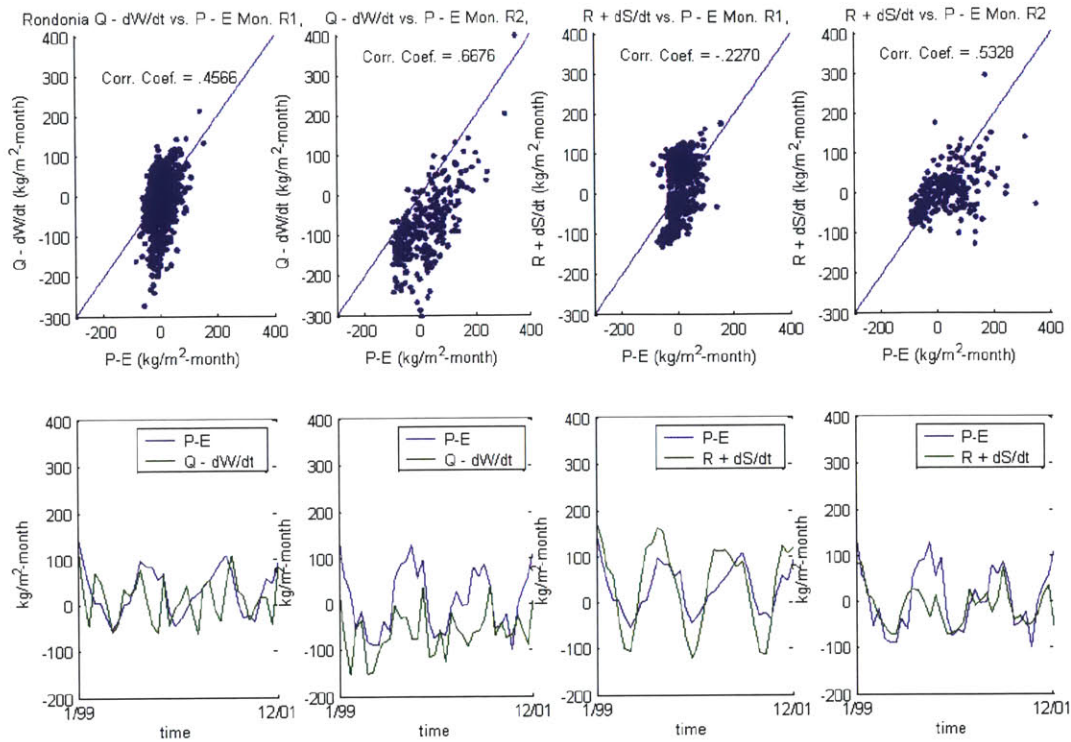


Figure 4-22: Correlation Plots for Rondonia – Reanalysis Comparison Monthly

Interestingly in Figure 4-23, on the left hand side, the atmospheric water balance for the Reanalysis-1 has a small overall bias when compared to the Reanalysis-2. However, though badly biased, the Reanalysis-2 water balance correlates better as the signals in the water balance are very well aligned (the peaks and troughs). In the surface balance a weaker overall correlation is found than in the atmospheric balance with smaller differences on the yearly scale than on the monthly scale.

As expected the Reanalysis-2 tends to close water balances slightly better than the Reanalysis-1 in general for the sample box control volumes. Imbalances arise in the surface balance for the default conditions in both reanalyzes much more than in the atmospheric balance. Many of these imbalances are control volume dependant and the usefulness of the reanalysis for each box control volume will be looked into in great detail in Chapter 5.

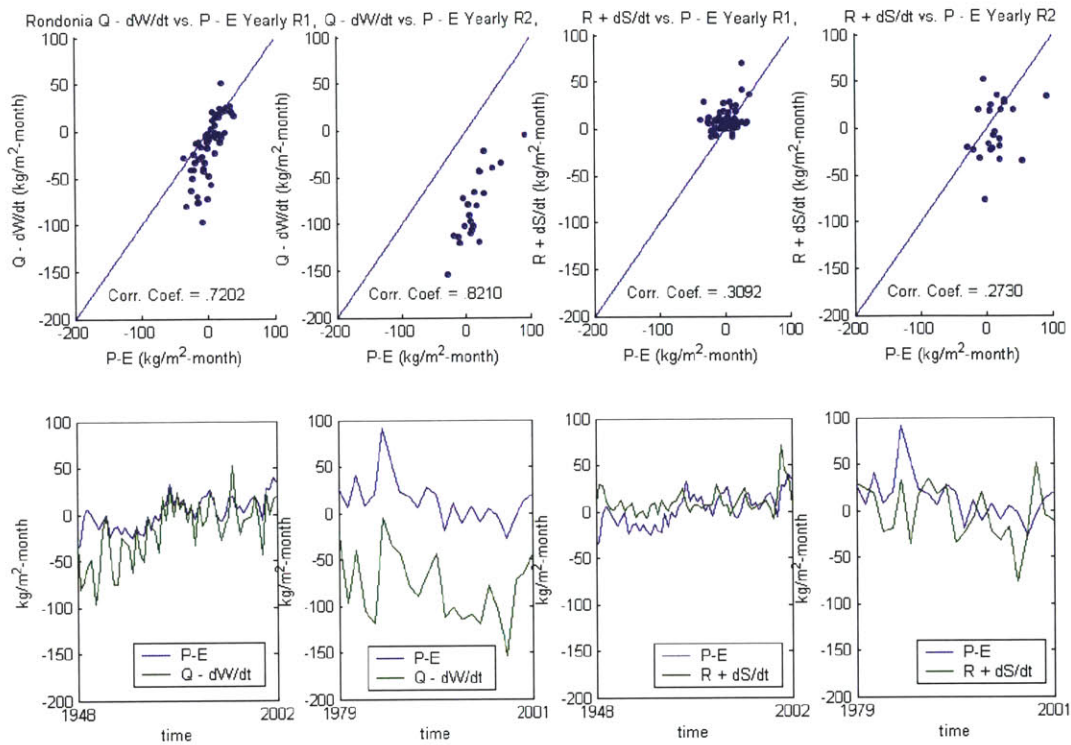


Figure 4-23: Correlation Plots for Rondonia – Reanalysis Comparison Yearly

#### 4.4. Precipitation Datasets

The majority of the alternative data used in this study is in the form of various precipitation datasets. Twelve different precipitation datasets are used with the Reanalysis-2 water balance in this section, in addition to the given Reanalysis-2 precipitation. Both the atmospheric and surface water balances are computed. Each water balance is evaluated for the time period over which the precipitation dataset coincides with the Reanalysis-2, therefore this period varies by precipitation dataset.

We will first look at all of these datasets at the same time for the seven box control volumes. Unfortunately, it is difficult to derive meaningful conclusions about these datasets while looking at them all at once, therefore several important datasets will be pulled out and broken up into groups. Of special interest are datasets which cover the same historical period as the Reanalysis-2 (79-01). These datasets include the GPCP, the GHCN-CAMS gridded dataset, and the best estimates from the OPI and CMAP datasets, which all can replace every value of the Reanalysis-2 precipitation. Also of particular interest in this study are the TRMM datasets, which are only available since January of 1998. Because the Reanalysis-2 is only available through 2001, the Reanalysis-1, which runs through 2002, is used for this precipitation analysis.



#### 4.4.1. All Precipitation Datasets

Figures 4-24 through 4-32 contrast precipitation datasets over all of the atmospheric and surface water balances on monthly and yearly time scales. Figures 4-24 through 4-26 deal with the residuals created by the water balances in terms of absolute error and bias. Figures 4-27 through 4-29 show the slope, correlation coefficient, and NRMSE for the atmospheric balance. Figures 4-30 through 4-32 show the same statistics for the surface water balance. As stated previously because of the large amounts of different precipitation data it is hard to make specific statements from this analysis. More particular observations can be made when a closer look is given to particular subsets of the data, and when some of these datasets are applied to a global type analysis in Section 4.3. For a few control volumes there are no results for certain precipitation datasets because they have errors or do not cover the region (see Section 2.12). They pertain mostly to tropical precipitation datasets which cannot be applied to control volumes which extend past 40 degrees north latitude.

Looking at the atmospheric absolute error in Figure 4-24 the differing precipitation datasets seem to yield very similar results over the monthly and yearly time scales for all precipitation datasets used in the Larger Rondonia, NW Brazil, Rondonia, and SEUS control volumes. Over East Brazil the reanalysis clearly yields the best balance, however, better balances are also seen with remotely sensed technologies. Again the situations for the Larger US and the SWUS control volumes for the atmospheric balance are not good. Because of the large errors in the SWUS atmospheric balance, regardless of precipitation dataset, the data can be disregarded. There are larger differences in surface absolute error shown in Figure 4-25 over precipitation datasets, specifically for the monthly case where the Larger Rondonia, NW Brazil, and SEUS control volumes show significant improvement with the OPI and CMAP precipitation datasets.

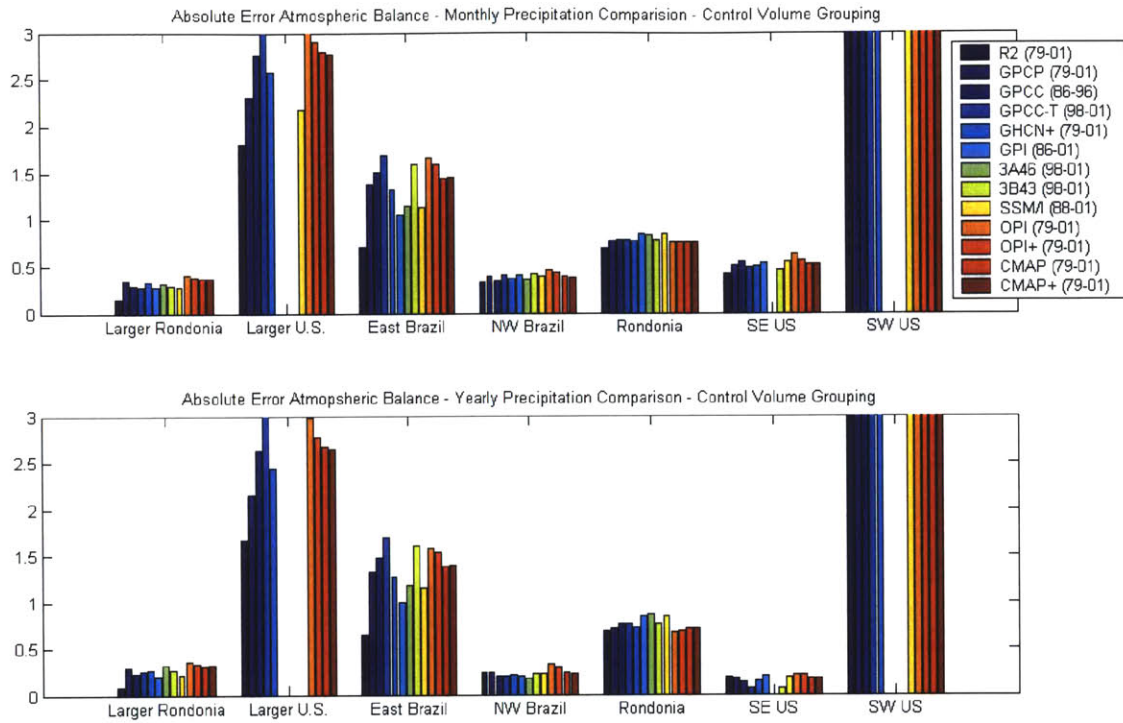


Figure 4-24: Precipitation Comparison – Atmospheric Absolute Error Monthly and Yearly

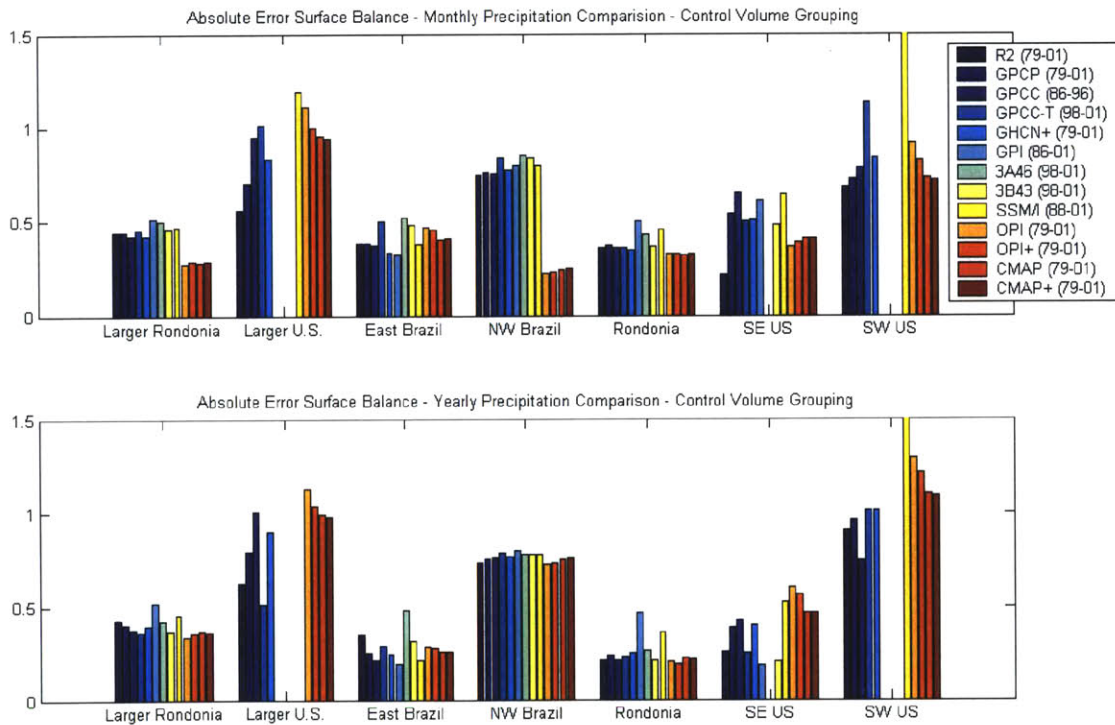


Figure 4-25: Precipitation Comparison – Surface Absolute Error Monthly and Yearly

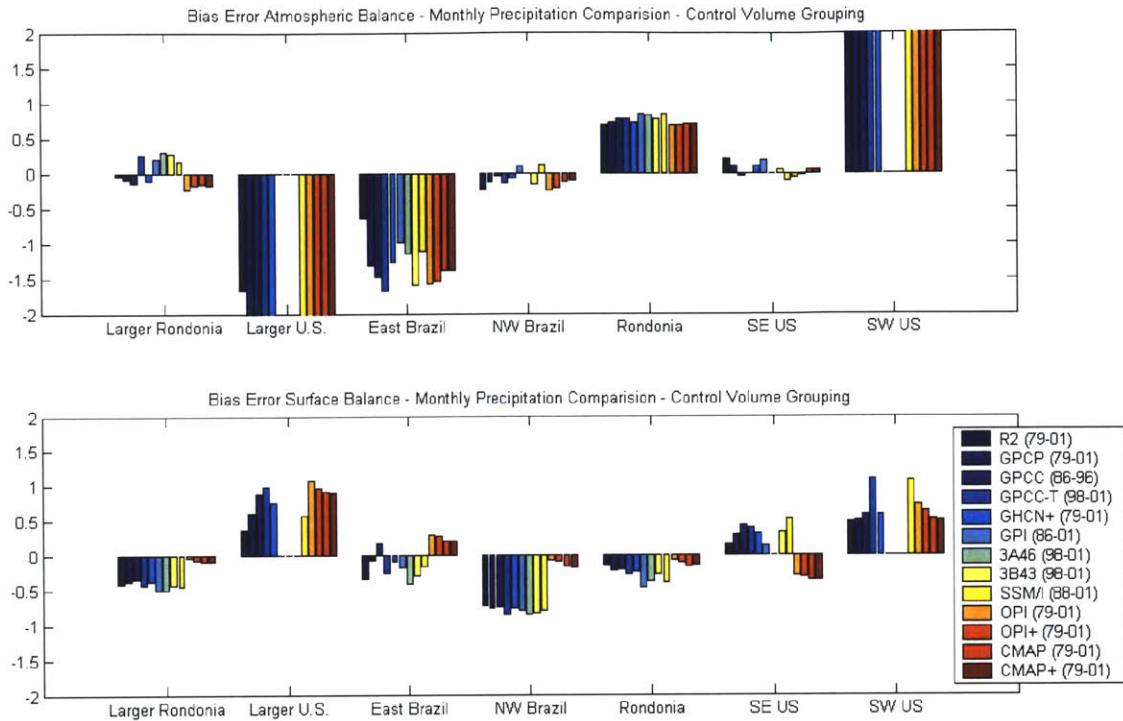


Figure 4-26: Precipitation Comparison – Atmospheric and Surface Bias Error Monthly

Monthly bias is shown for the atmospheric and surface balances in Figure 4-26. Yearly values are similar but not shown. The figure highlights the difference in total precipitation magnitude as applied to the different control volumes. The precipitation datasets can be evaluated for whether they are too low or too high relative to other components of the water balance. Biases are generally consistent with absolute errors in the atmospheric balance, but differ in the surface balance. The complete closure of the water balance does not seem to rely on precipitation datasets nearly as much as it depends on the control volume being considered.

Figures 4-27 through 4-29 show relevant regression and correlation statistics for the atmospheric monthly and yearly balances. As far as the monthly case goes the results are very consistent in terms of closure of the water balance to observations made for the residual statistics, with the exception of the NW Brazil control volume. For the atmospheric balance this control volume will be examined more closely later. For the yearly atmospheric balances, large discrepancies start to arise in terms of slope on the bottom of Figure 4-27. For all control volumes there are either very good correlations for a particular precipitation data set, or very bad correlations. The general problems with the Larger US and SWUS seem to be a problem with the monthly cycle across different precipitation datasets. These control volumes behave similarly to the other control volumes on a yearly basis. There is little consistency between control volumes on the yearly time scale based on precipitation dataset other than that the Reanalysis-2 consistently balances well in all cases, as it should.

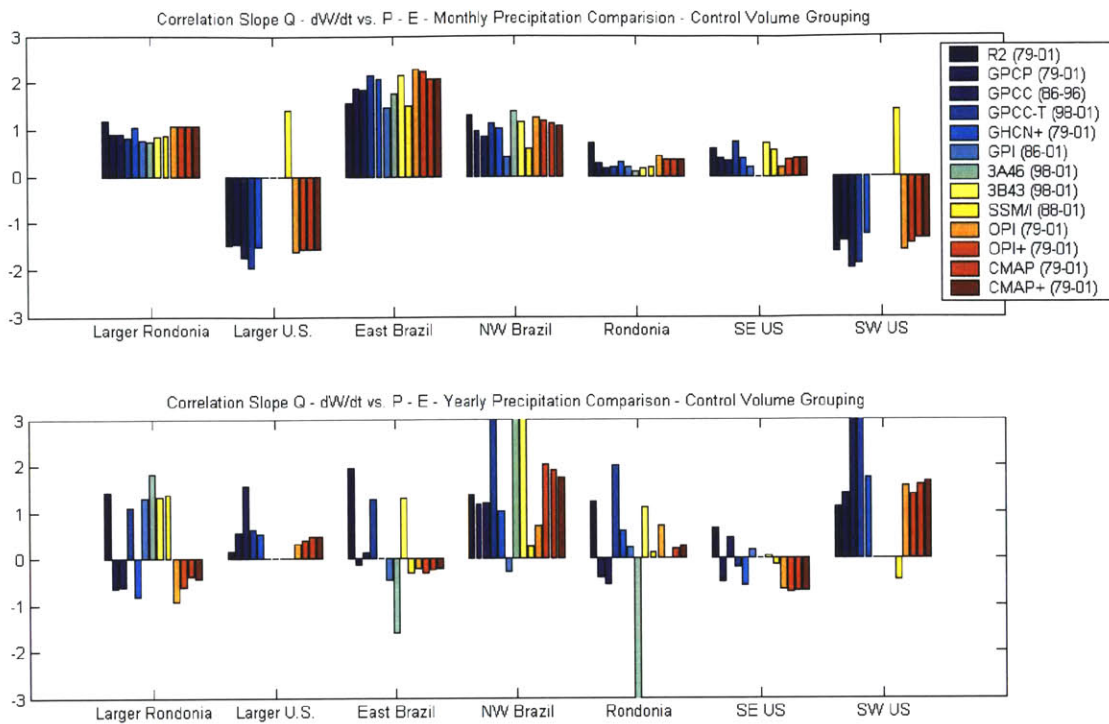


Figure 4-27: Precipitation Comparison – Atmospheric Regression Slope Monthly and Yearly

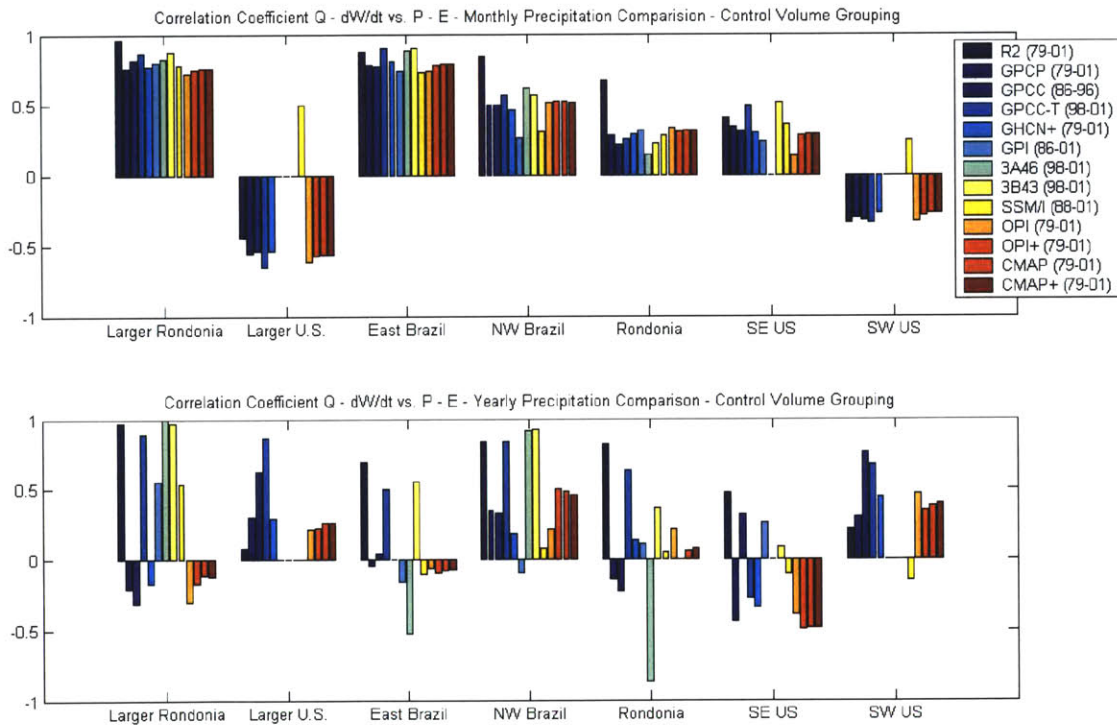


Figure 4-28: Precipitation Comparison – Atmospheric Correlation Coeff Monthly and Yearly

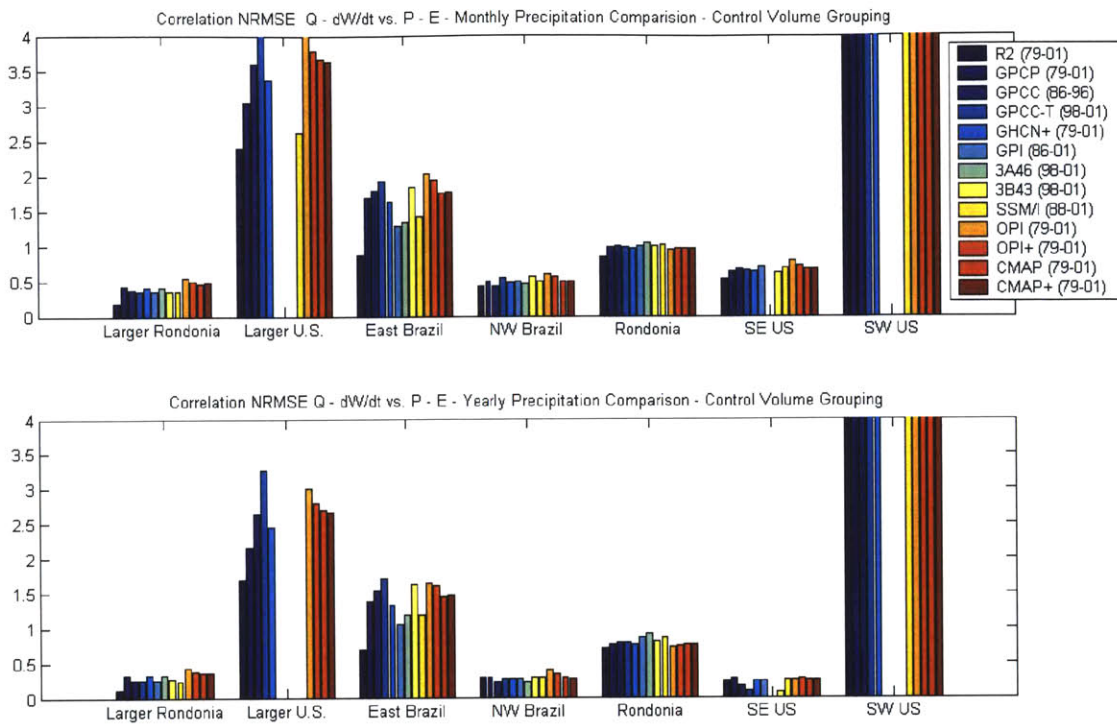


Figure 4-29: Precipitation Comparison – Atmospheric NRMSE Monthly and Yearly

Figures 4-30 through 4-32 show the relevant regression and correlation statistics for the surface balance contrasting precipitation sources. There are differences between the alternate precipitation datasets for most of the control volumes. Generally there is reasonable consistency in the seasonal cycle as the precipitation datasets allow for (P-E) to balance the surface runoff and change in storage. Slopes are consistently zero or negative for all precipitation datasets applied to the NW Brazil control volume indicating poor balances. Differences in slope are more evident on the yearly time scale, particularly in the case of the SEUS control volume.

The monthly surface correlation coefficients are good for most precipitation datasets over most control volumes. An interesting situation again arises over the SEUS where the reanalysis precipitation provides a strong surface correlation, while all other alternatives do not. Correlations tend to be not as great for the surface balance for the OPI and CMAP datasets. These precipitation datasets tend to consistently follow each other in most of these analyses as they are of similar origin. Consistency however is rarely seen for the gauge based precipitation datasets (Columns 3 through 5) and remotely sensed precipitation datasets (Columns 6 through 9). The GPI and SSM/I precipitation datasets have consistently high NRMSE for both the monthly and yearly time scales. Overestimation was a factor in the derivation of both of these datasets.

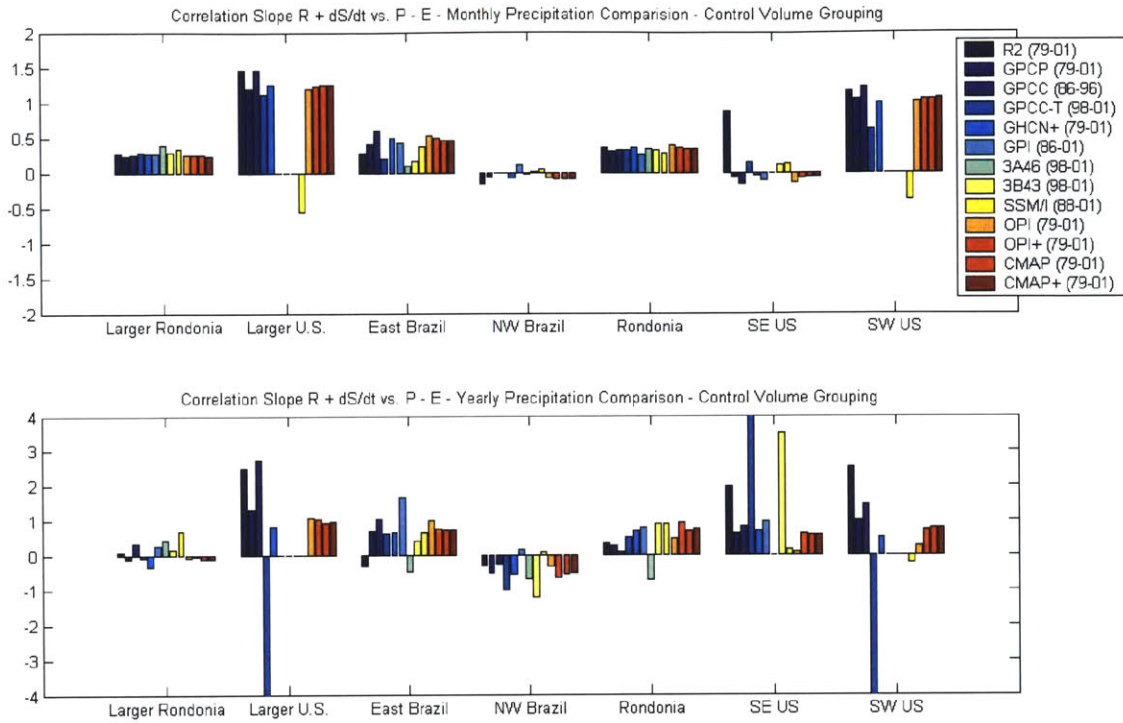


Figure 4-30: Precipitation Comparison – Surface Regression Slope Monthly and Yearly

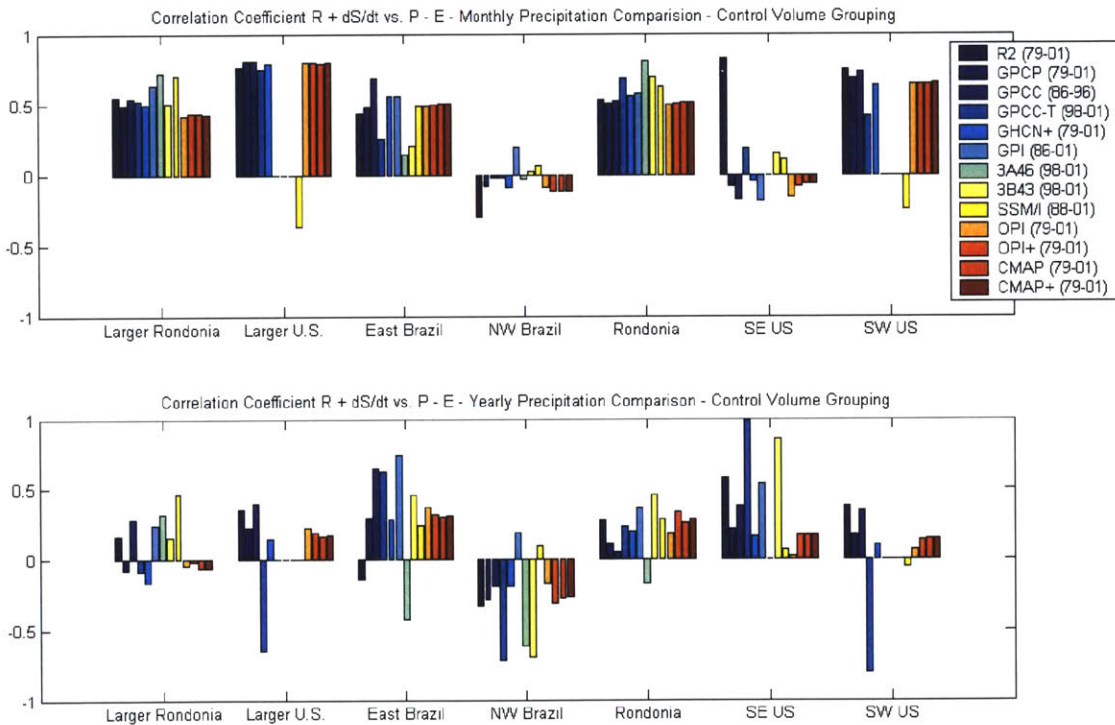


Figure 4-31: Precipitation Comparison – Surface Correlation Coefficient Monthly and Yearly

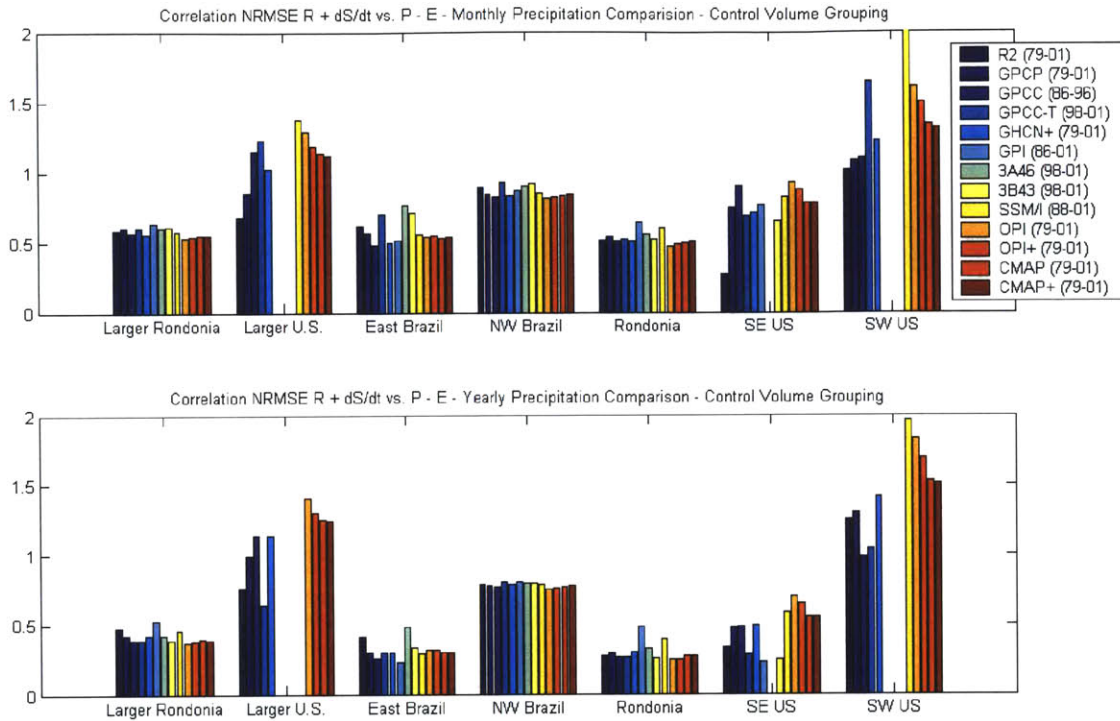


Figure 4-32: Precipitation Comparison – Surface NRMSE Monthly and Surface

As in previous sections we now select some sample precipitation regressions to elucidate the bar graphs above. Here we will select three precipitation datasets which represent the different types of precipitation data available. First shown is the reanalysis precipitation, which is the default precipitation set in this work and commonly provides the best balance. Next, shown is the GPI, a fairly long (1986-2001) tropical satellite based precipitation dataset, to represent satellite based data. Finally, the GHCN-CAMS dataset is used to represent the gauged datasets.

First shown is the NW Brazil control volume as an example for the atmospheric monthly and yearly water balances in Figures 4-33 and 4-34. For the monthly time scale the satellite precipitation is generally underestimated as can be seen by the positive slope of the regression line for this control volume. For the yearly case in Figure 4-34, the reanalysis balance looks good, while the addition of the GPI precipitation degrades the balance, especially in the 1990's. The gauge correlation is better than the GPI correlation, but shows a weak signal compared to the runoff flux term, which results in a larger positive slope. It is apparent from these figures that on the monthly time scale many of the precipitation datasets agree with the reanalysis balance. Nevertheless, the water balances exhibit differences between datasets.

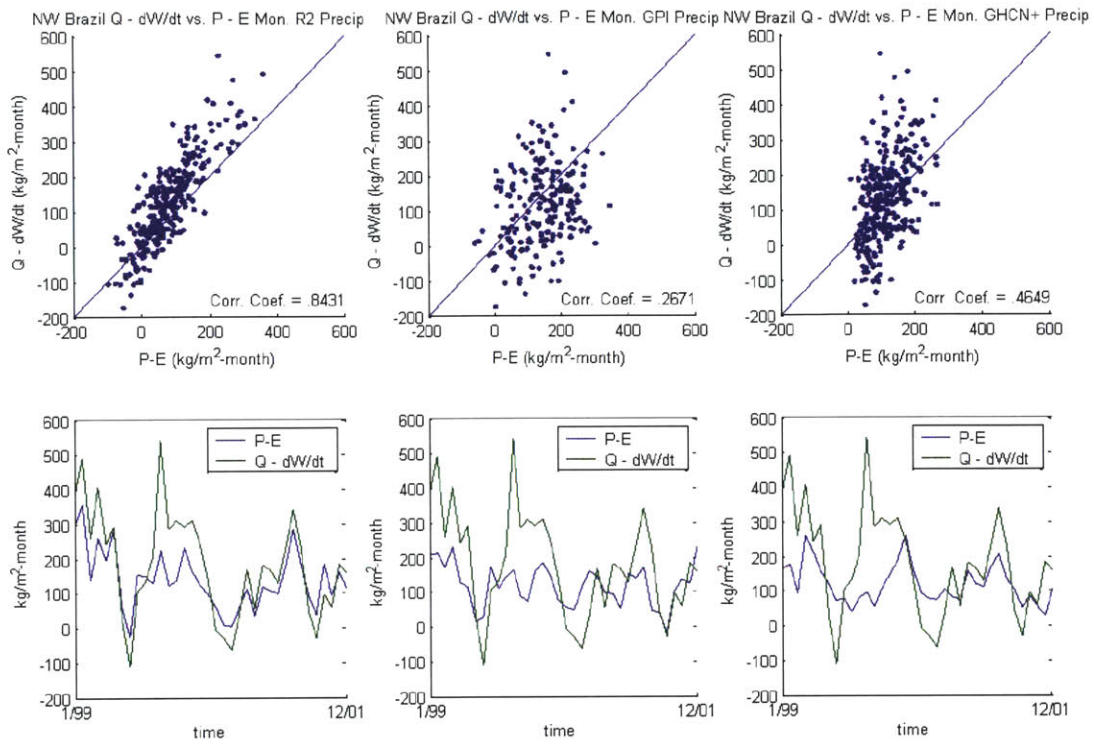


Figure 4-33: Correlation Plots for NW Brazil – Precipitation Comparison Monthly Atmospheric

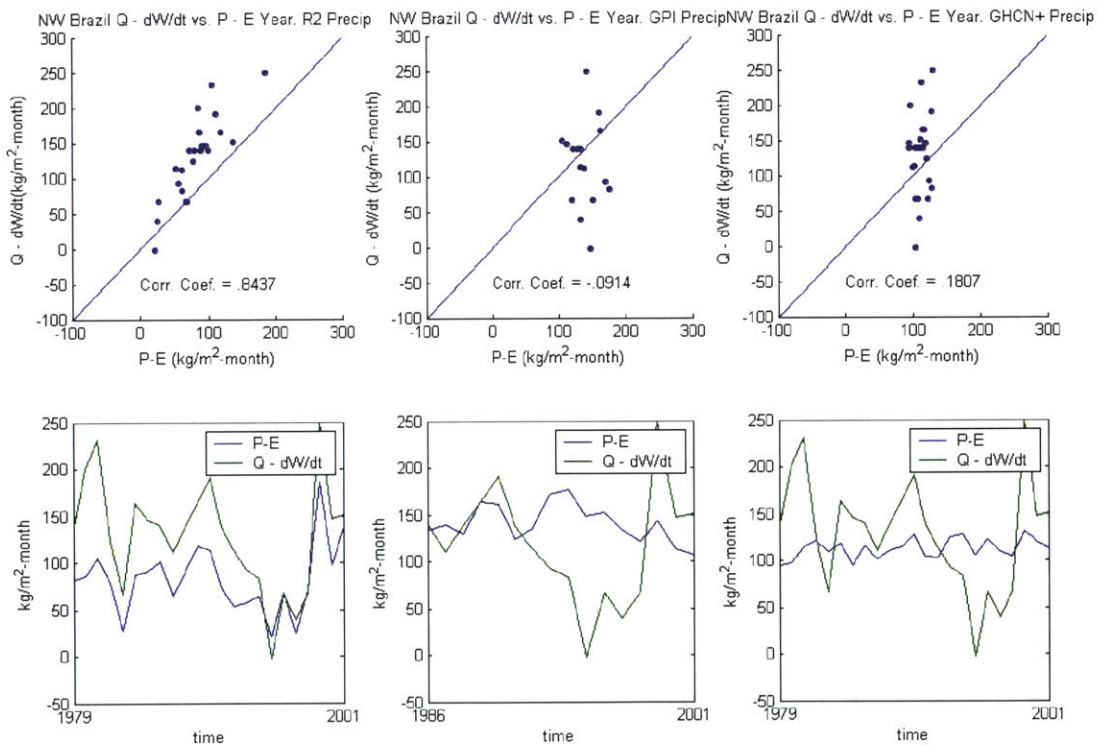


Figure 4-34: Correlation Plots for NW Brazil – Precipitation Comparison Yearly Atmospheric



Figures 4-35 and 4-36 show example correlation plots for the SEUS control volume. This control volume was selected for closer inspection as there are sharp contrasts between water balance properties between the precipitation datasets as they apply to the surface water balance. In Figure 4-35, the reanalysis surface monthly water balance shows a very strong correlation and is very well balanced. However, when reanalysis precipitation is replaced by gauge and satellite data some points correlate well while others correlate poorly. The runoff and change in storage data remains the same as can be seen by the green line represented in the bottom panels, however, the P-E quantity changes dramatically with precipitation dataset. One explanation is that the model derived reanalysis precipitation for this region was made to fit the surface water balance of the reanalysis.

For the yearly time scale in Figure 4-36, the goodness of balance steadily decreases from left to right, with some signal pick up in the Reanalysis-2, but no signal pickup the when using gauge precipitation. This is interesting because the SEUS is a heavily gauged precipitation area, and therefore it would seem the reanalysis disregards the rain gauge data completely.

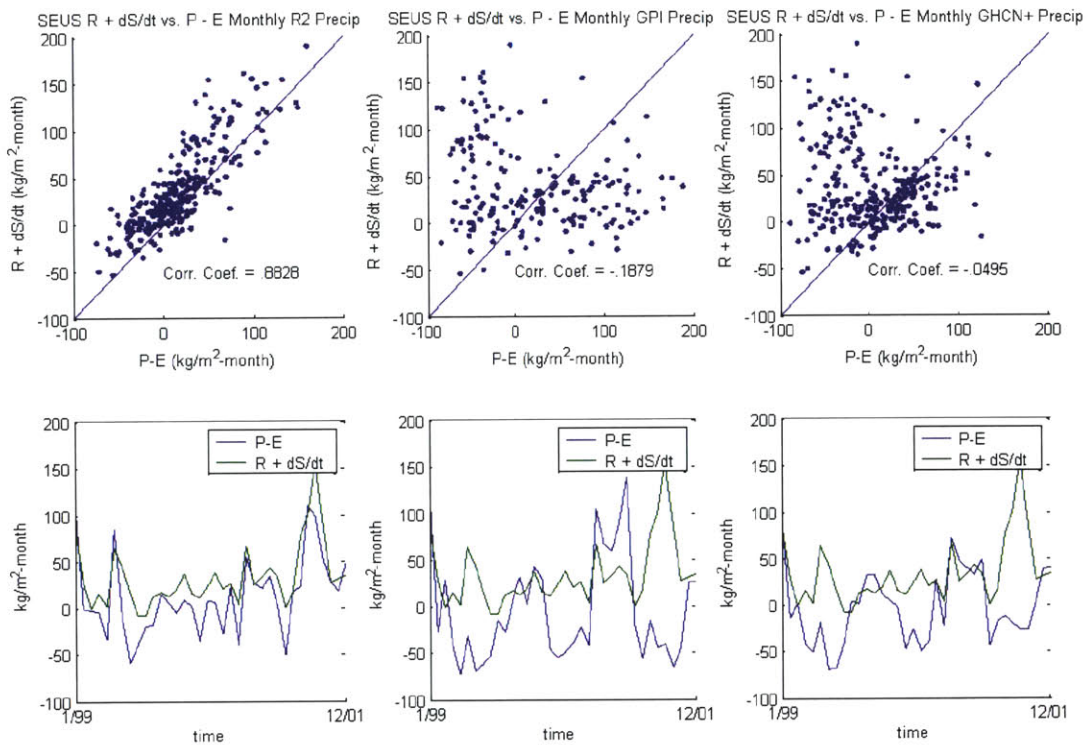


Figure 4-35: Correlation Plots for SEUS – Precipitation Comparison Monthly Surface

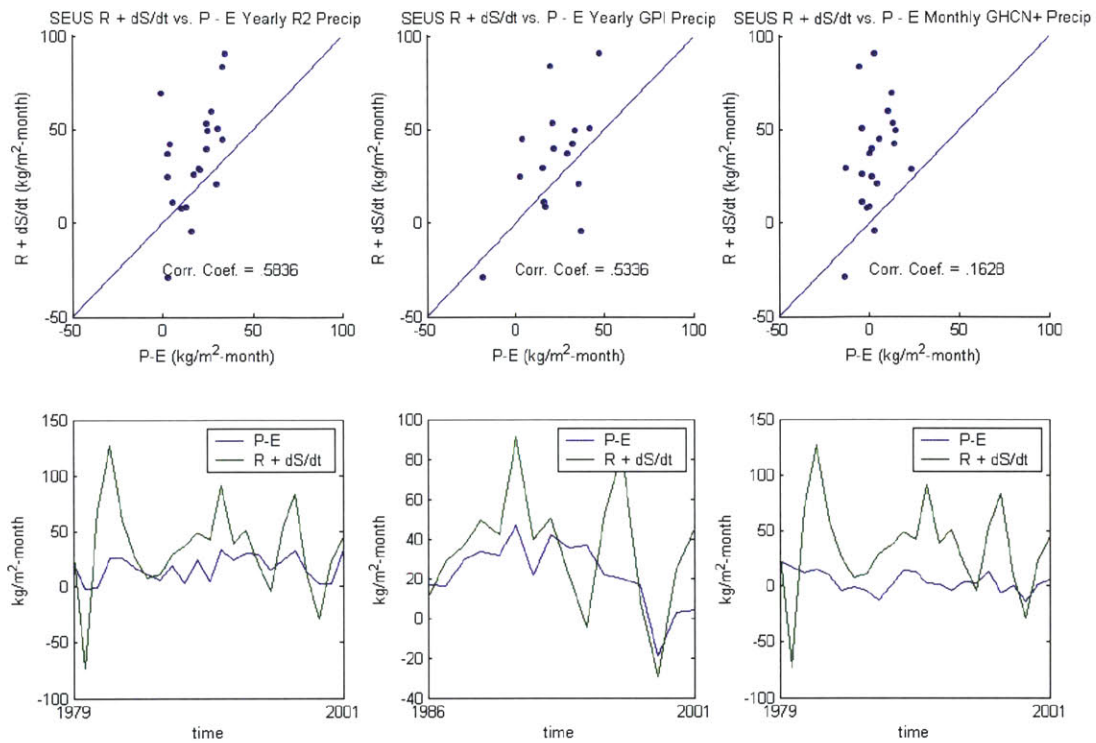


Figure 4-36: Correlation Plots for SEUS – Precipitation Comparison Yearly Surface

#### 4.4.2. Longer Precipitation Datasets

We now look at precipitation datasets which can completely replace the Reanalysis-2 precipitation (1979-2001) in the water balance and thus determine if an alternative source of precipitation is better at closing the water balance using the reanalysis for all other fields. Datasets used are the GPCP, CMAP, OPI, and GHCN-CAMS products. For the first time here we will consider the basin sized control volumes. Because of the nature of these balances, the defaults will be considered slightly different. Surface runoff is considered default in basin sized control volumes. Thus for the surface balance of the basin sized control volumes, both precipitation and runoff may differ from the reanalysis.

Figures 4-37 through 4-40 display the effect of change in precipitation source on the absolute error and bias calculations for the monthly/yearly and atmospheric/surface conditions. In the atmospheric balance no matter what the precipitation is, the Mississippi and Arkansas water balances have higher errors, as was the case with the SWUS and Larger United States water balances before. The reanalysis consistently provides a better balance for all control volumes. Very few anomalies arise in the atmospheric balances, as all the alternate precipitation datasets do slightly worse in atmospheric absolute error and bias, and absolute errors decrease on the yearly basis. For the surface balance (Figures 4-39 and 4-40) the alternative precipitation datasets result in a better balance than the reanalysis for certain control volumes. In Rondonia, the OPI+ dataset works well, while over the Amazon the gauge network seems to work well.

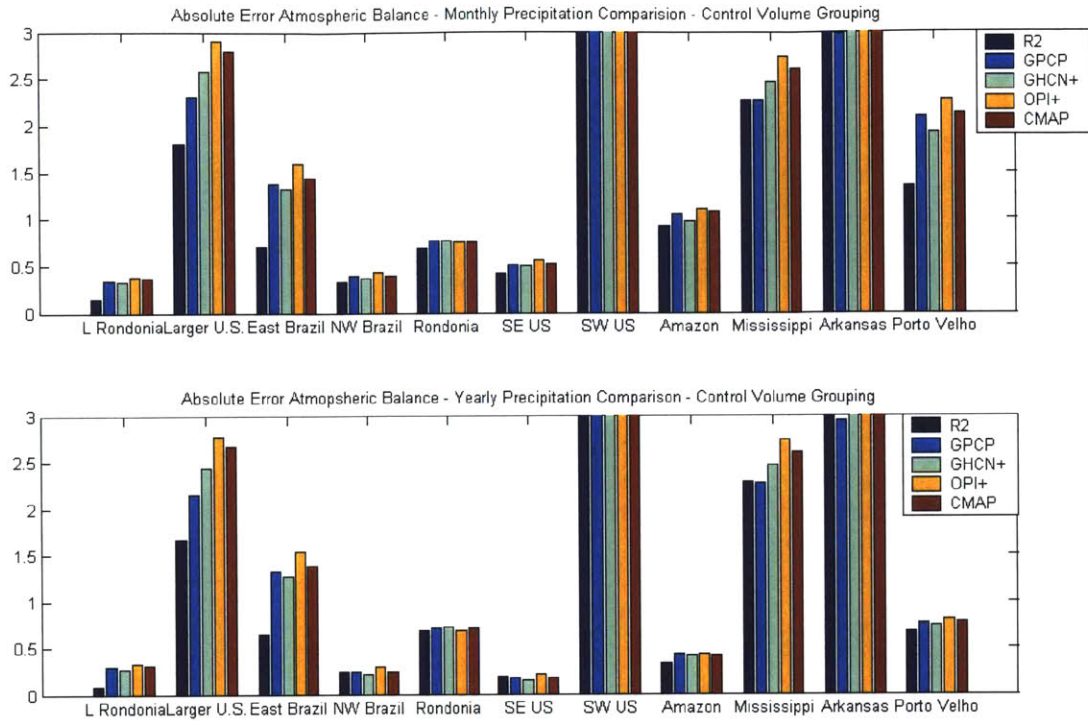


Figure 4-37: Large Precipitation Comparison – Atmospheric Absolute Error Monthly and Yearly

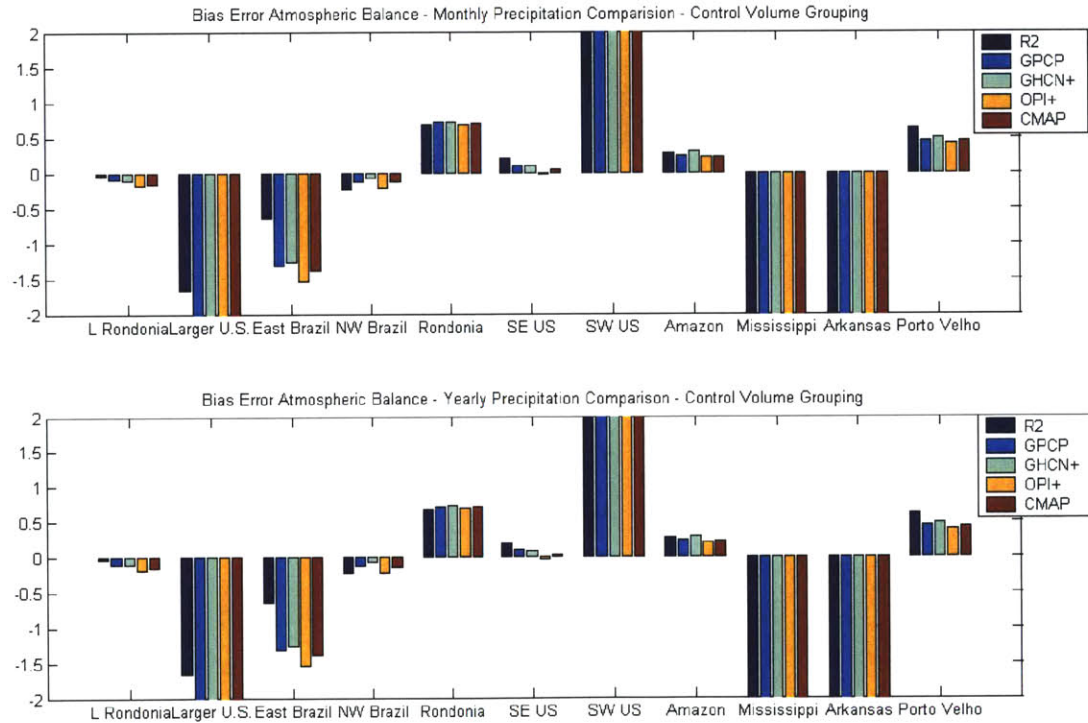


Figure 4-38: Large Precipitation Comparison - Atmospheric Bias Error Monthly and Yearly

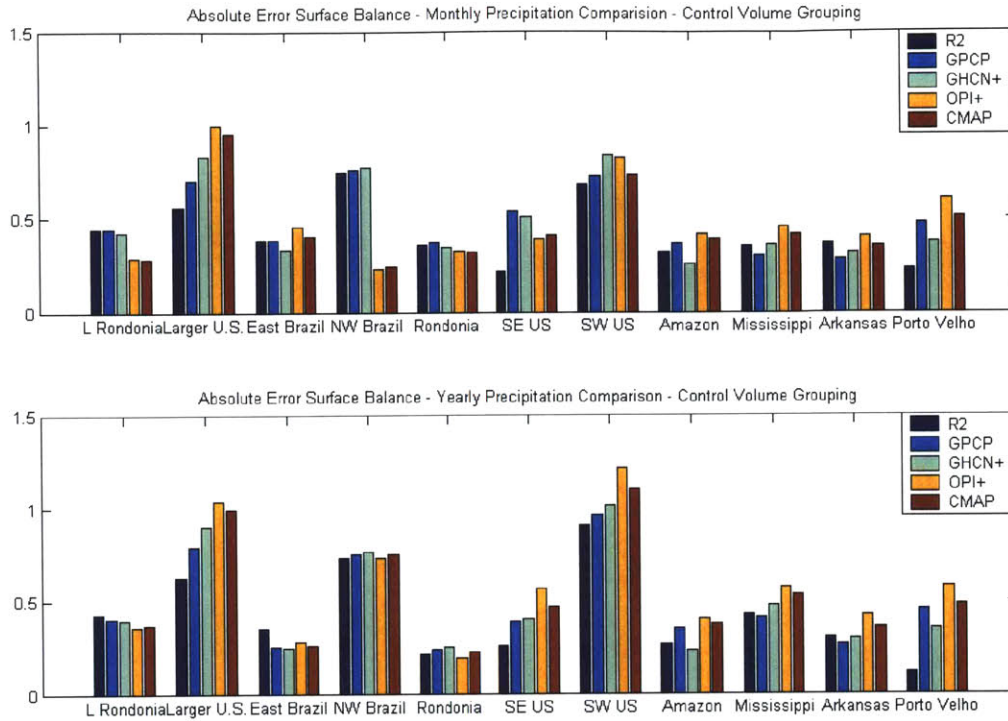


Figure 4-39: Large Precipitation Comparison – Atmospheric Absolute Error Monthly and Yearly

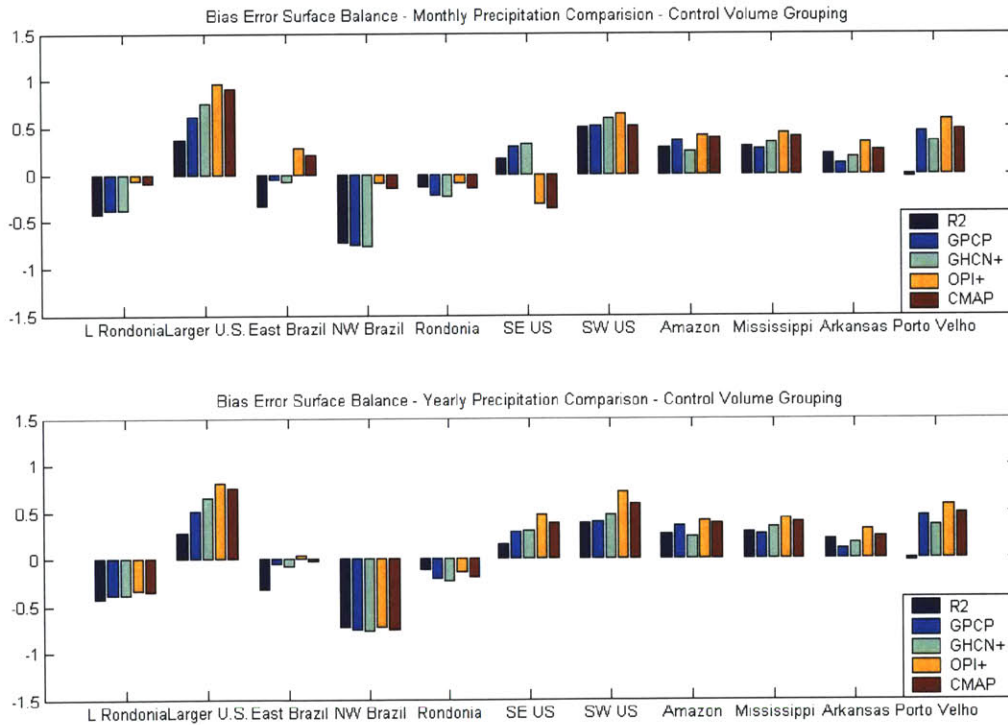


Figure 4-40: Large Precipitation Comparison – Atmospheric Bias Error Monthly and Yearly

Figures 4-41 through 4-43 explore the atmospheric balance correlation. In Figure 4-41, in almost all cases, the reanalysis has the closest slope to one. Some of the yearly problems are cleared up in the basin control volumes, specifically over the Mississippi and Arkansas basins. Figure 4-42 paints a similar picture of the atmospheric correlation, where on the yearly time scale sharply higher correlations are found using default reanalysis precipitation. Figure 4-43 shows the NRMSE for the atmospheric water balance. The reanalysis performs the best here, while the OPI+ dataset consistently has the worst NRMSE. This is consistent with what is known about the OPI+ dataset as it is not recommended for studies of this nature.

Figures 4-44 through 4-46 show the land surface balance correlation statistics for the larger precipitation datasets. In the surface correlation the reanalysis is not clearly the best precipitation dataset in terms of either slope or correlation coefficient. In fact for the Arkansas and Mississippi control volumes it is the worst. What we have done here is use a dataset which is not part of the reanalysis to close the surface balance (runoff). Over the United States alternative precipitation datasets likely match better with these runoff datasets. On the monthly time scale, looking at the SEUS, very poor correlations are found for all precipitation datasets but the reanalysis. This is a box control volume and not a basin control volume and the runoff is coming from an entirely different source. For the NRMSE in Figure 4-46, errors are found to be largest generally for the OPI+ dataset again and lowest for the reanalysis.

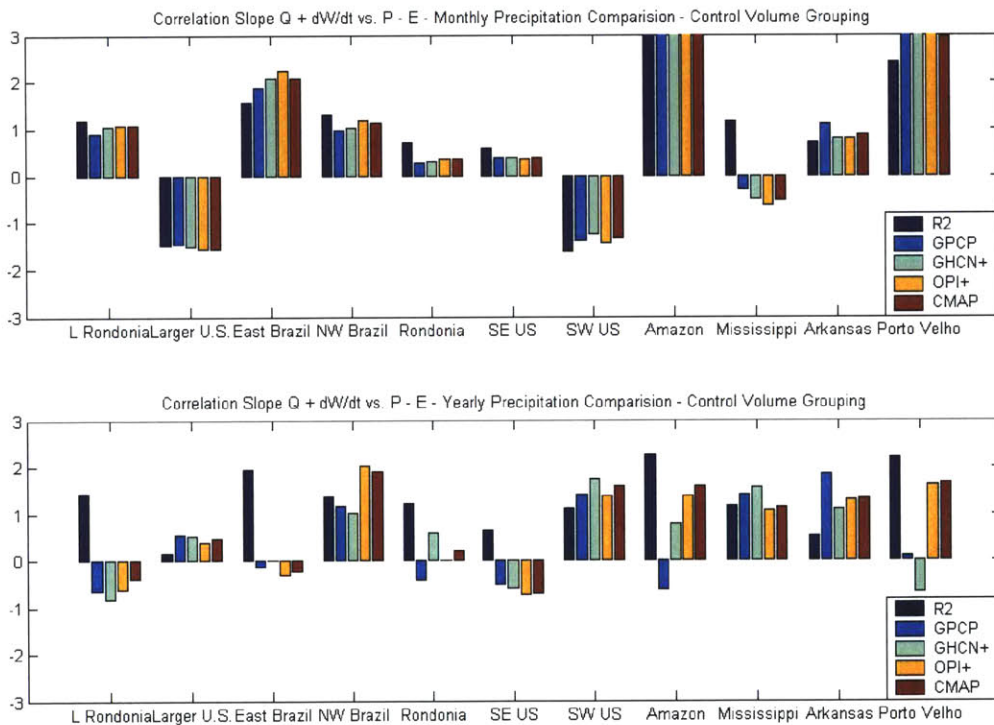


Figure 4-41: L. Precipitation Comparison – Atmospheric Regression Slope Monthly and Yearly

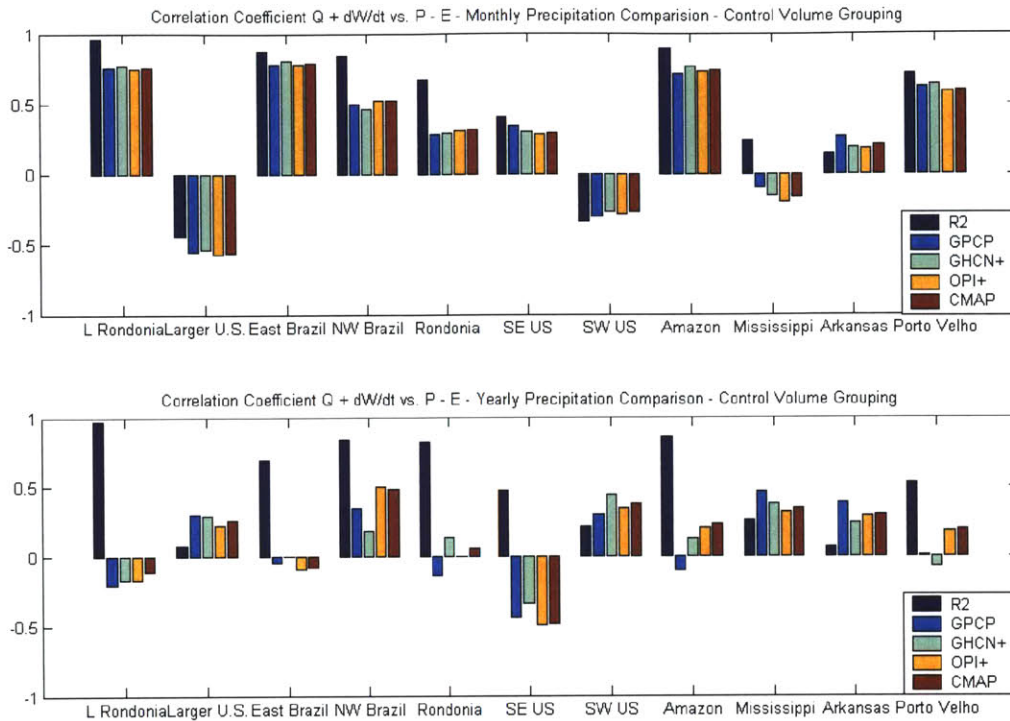


Figure 4-42: L. Precipitation Comparison – Atmospheric Correlation Coeff Monthly and Yearly

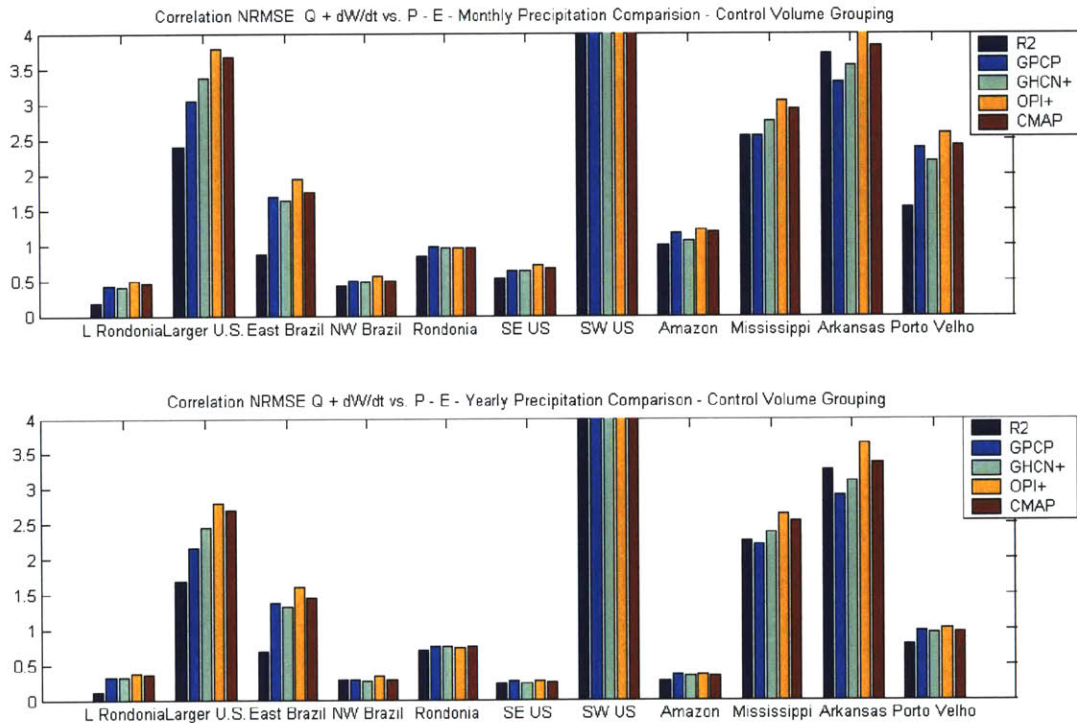


Figure 4-43: Large Precipitation Comparison – Atmospheric NRMSE Monthly and Yearly

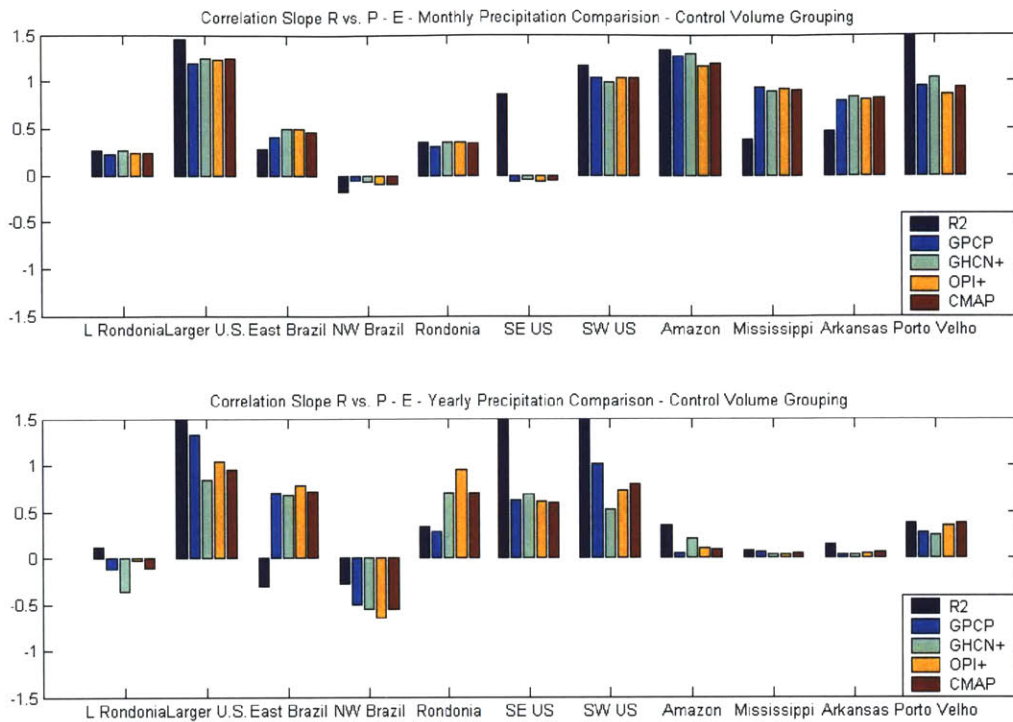


Figure 4-44: Large Precipitation Comparison – Surface Regression Slope Monthly and Yearly

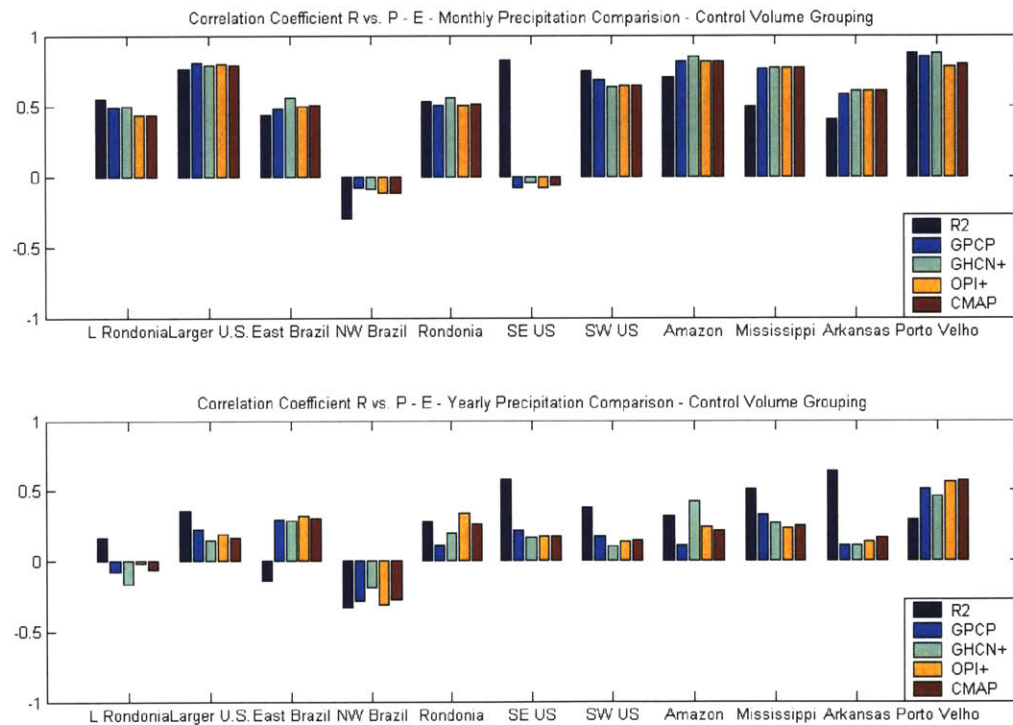


Figure 4-45: L. Precipitation Comparison – Surface Correlation Coefficient Monthly and Yearly

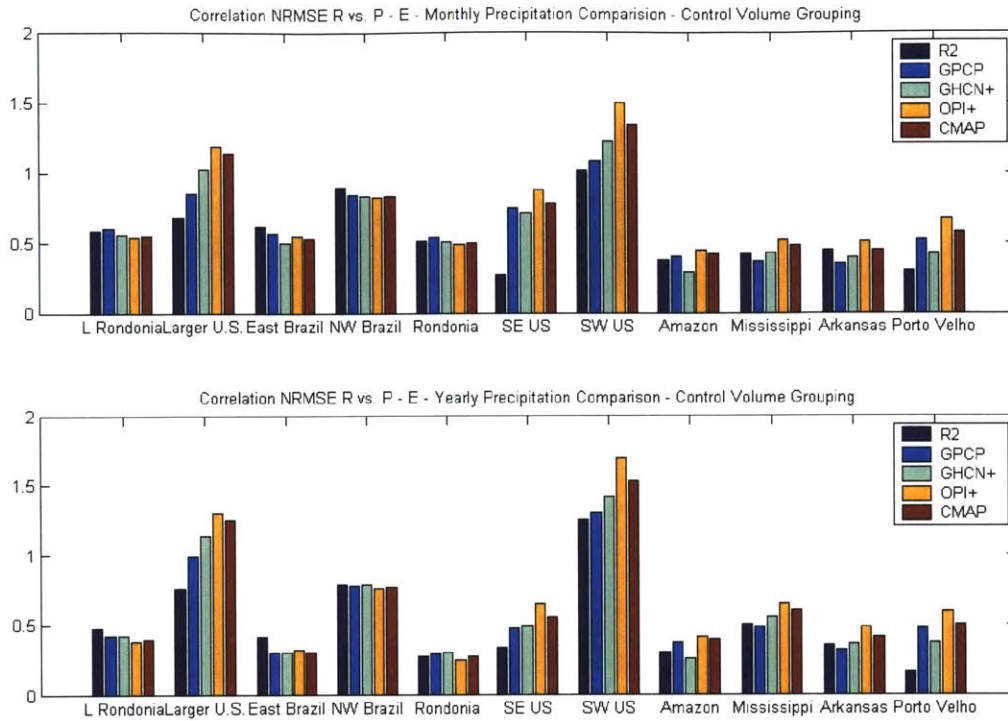


Figure 4-46: Large Precipitation Comparison – Surface NRMSE Monthly and Yearly

Example correlation plots for the above figures are provided in Figures 4-47 through 4-49. First we will look at the Northwest Brazil control volume for the atmospheric balance and yearly time scale. The three alternative precipitation data sets show very similar annual trends with P-E underestimating moisture flux. The reanalysis does far better than any other precipitation dataset.

Figures 4-48 and 4-49 explore the surface water balance on the yearly basis. These balances are of particular interest because alternative precipitation seemed to improve the water balances for the Arkansas and Mississippi control volumes. The amplitudes of the two sides of the surface balance don't match up very well, but the signals do. For the Mississippi Basin the same observations can be made. The Reanalysis correlation is strong, but the scatter plots seem to be better for the alternative precipitation datasets. The balances in general are not great in these two cases.



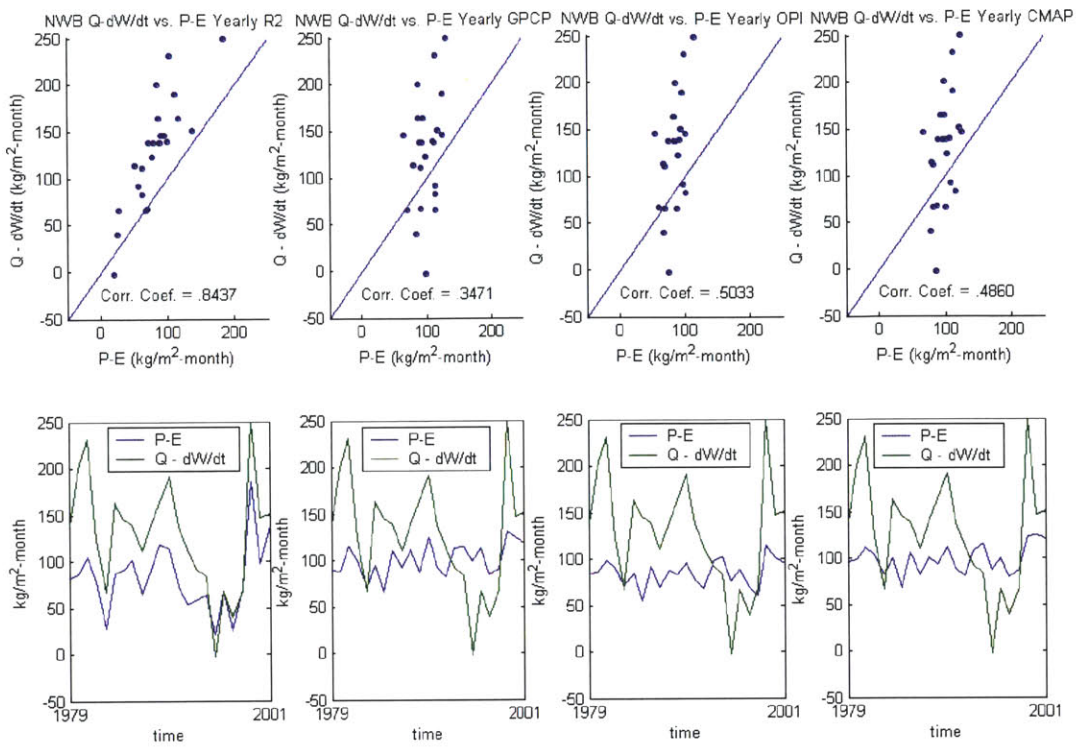


Figure 4-47: Correlation Plots for NW Brazil – Large Precip Comparison Yearly Atmospheric

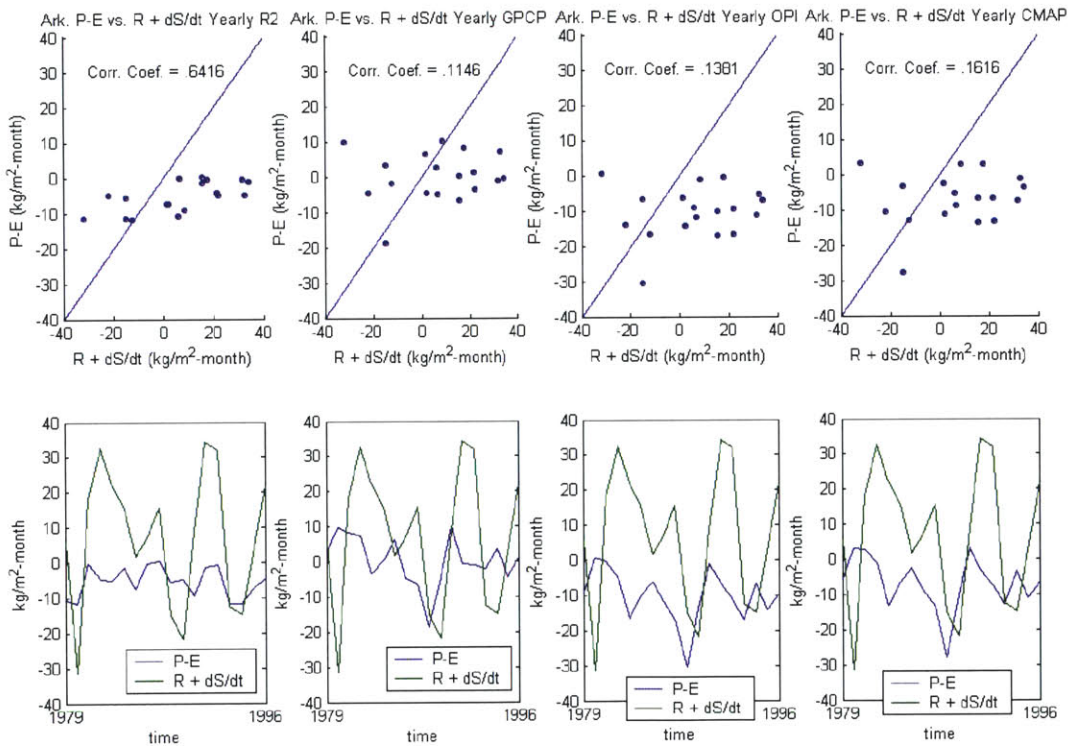


Figure 4-48: Correlation Plots for Arkansas Basin – Large Precip Comparison Yearly Surface

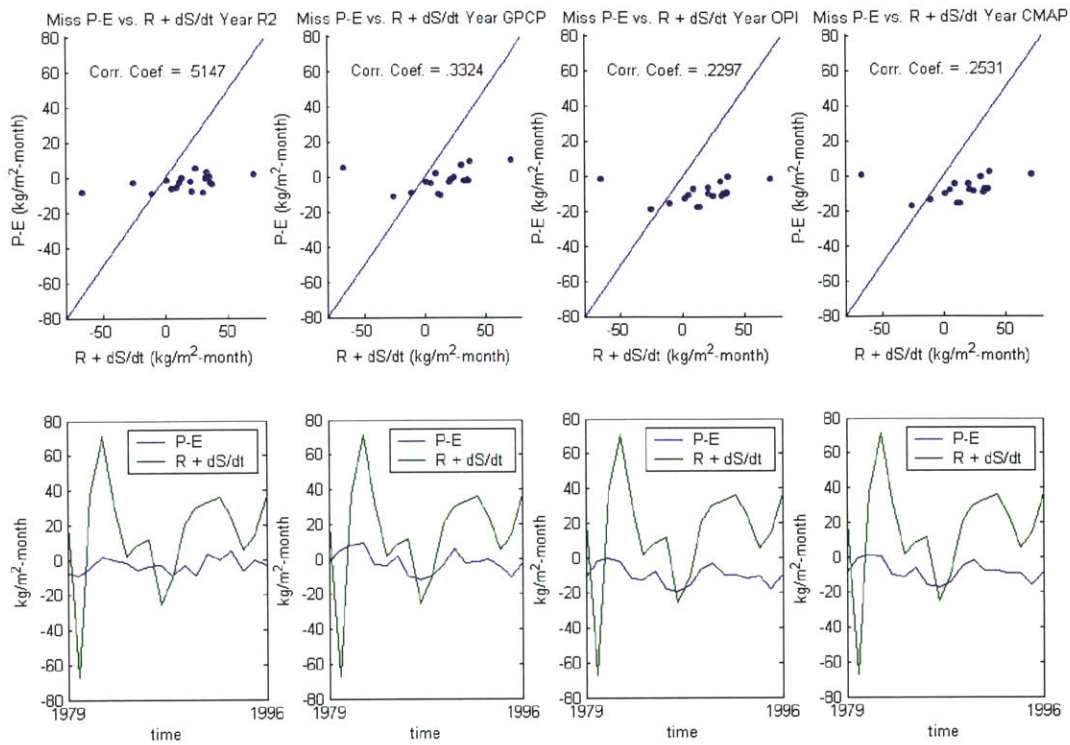


Figure 4-49: Correlation Plots for Mississippi Basin – Large Precip Comparison Yearly Surface

#### 4.4.3. TRMM Precipitation Datasets

The TRMM precipitation datasets represent the latest of current precipitation sensing technologies. Three monthly precipitation products are provided which can be conveniently used in this study, the gauge based TRMM GPCP dataset, the SSM/I based TRMM 3A46 dataset, and the actual TRMM precipitation product the TRMM 3B43 dataset. Since TRMM focuses on tropical precipitation between 40 degrees north and 40 degrees south latitude, only some of the control volumes of this work can be used with the dataset. These control volumes are: Larger Rondonia, East Brazil, Northwest Brazil, Rondonia, Amazon Basin, and Port Velho Basin.

In Figure 4-50 the atmospheric residuals are analyzed for the TRMM products in terms of absolute error and bias. In general, the TRMM 3A46 dataset results in the lowest absolute error and bias in the different water balances. Figure 4-51 shows the residual statistics when using TRMM products to obtain the surface water balance. Here the errors are slightly higher for the TRMM 3A46 dataset. This is unfortunate because it is hard to determine whether the atmospheric or surface balance is a more reliable evaluation of the precipitation products. By the classifications developed earlier for the box control volume we would tend to believe the atmospheric balance is more reliable while for the basin sized volume we would tend to think the surface balances were more reliable.

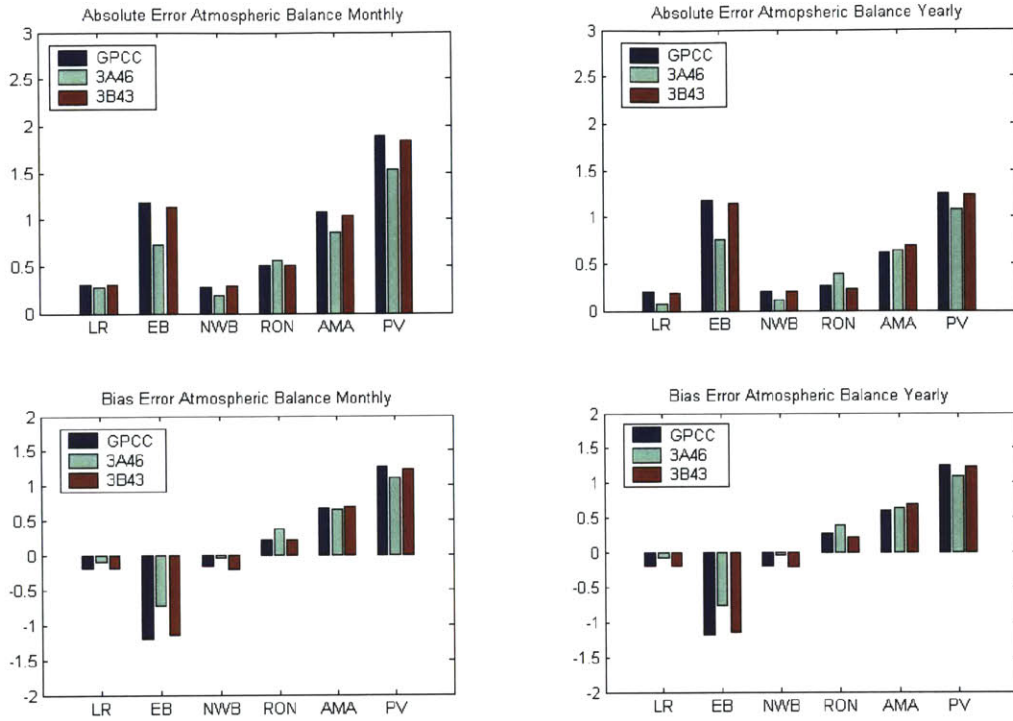


Figure 4-50: TRMM Precipitation Comparison – Atmospheric AE and Bias Monthly and Yearly

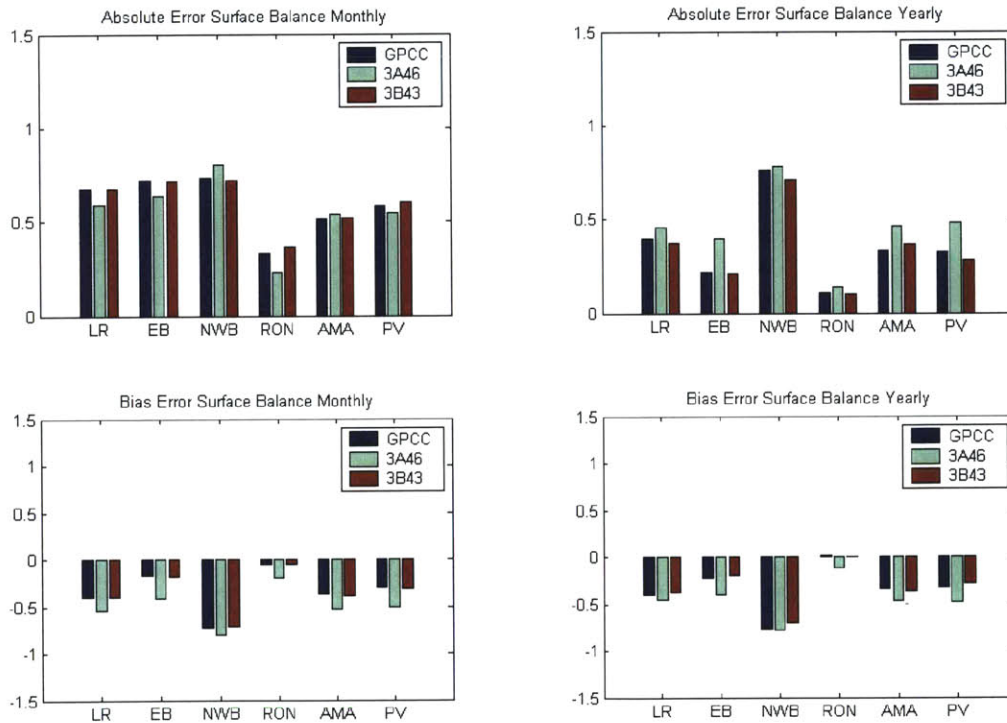


Figure 4-51: TRMM Precipitation Comparison – Surface AE and Bias Monthly and Yearly

Figures 4-52 and 4-53 show the relevant correlation statistics for the atmospheric and surface balances respectively. It should be noted here that these regressions are based on only 43 to 60 points on the monthly basis, and only 3 to 5 points on the yearly basis. For the monthly time scale the TRMM 3A46 in some cases has a slope closer to one, a correlation coefficient closer to one, and a lower NRMSE, but not as consistently as in the residual analysis.

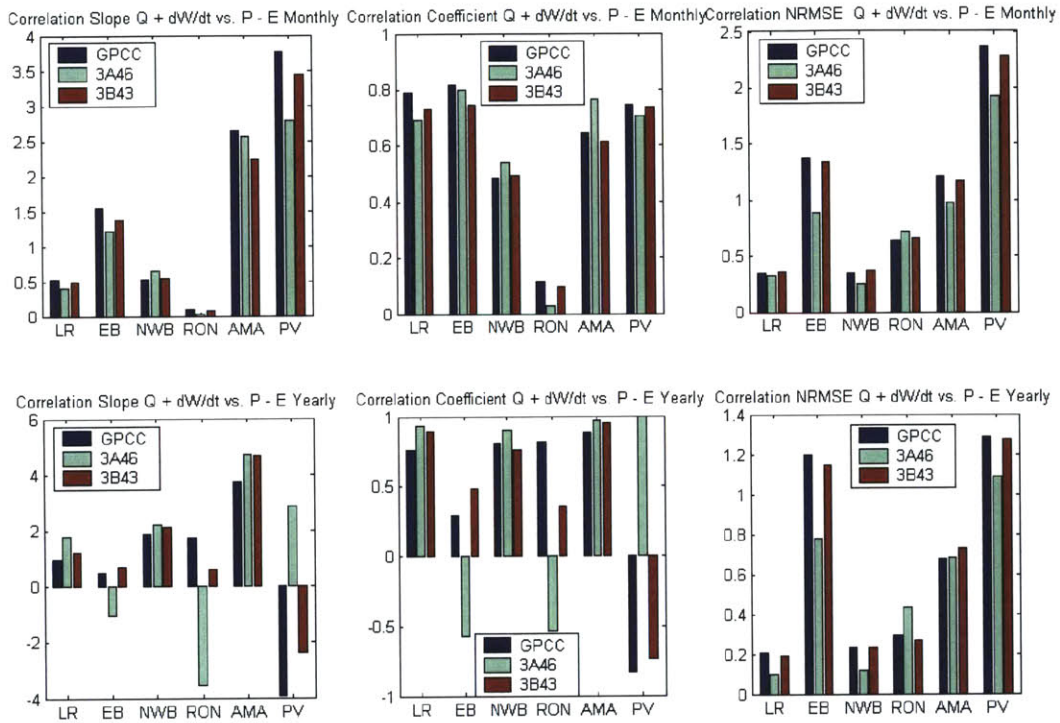


Figure 4-52: TRMM Precipitation Comparison – Atmos. Regression Stats Monthly and Yearly

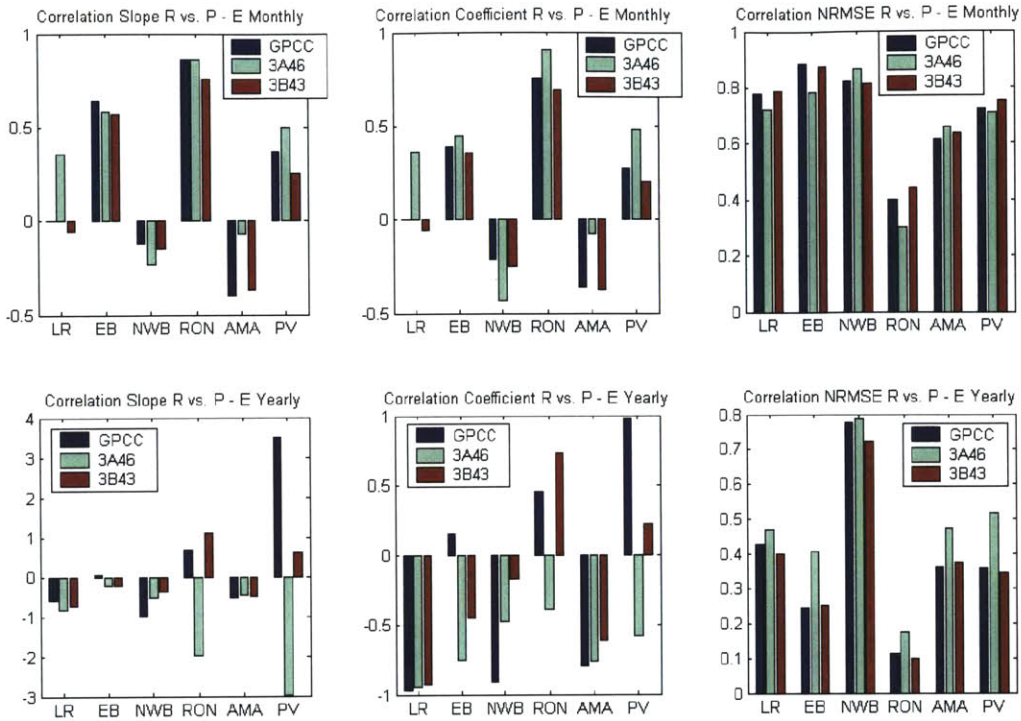


Figure 4-53: TRMM Precipitation Comparison – Surface Regression Stats Monthly and Yearly

Figures 4-54 and 4-55 give example correlation plots for the TRMM precipitation analysis. Of specific interest in this study is the application of these products to the Rondonia control volume. The monthly time step is looked at in detail as the yearly time scale would only contain a maximum of five points. In Figure 4-54 the atmospheric case is shown and though it seems poor correlations exist, looking at the bottom figures, the monthly and seasonal signals seem to be reproduced well by all three TRMM datasets, with the TRMM GPCC maybe having a slight edge. For the surface case the TRMM precipitation datasets form a tight correlation for the years they are available. The 3A46 dataset performs the best here, but it is clear that for these balances all of the TRMM datasets could be applied well and the TRMM datasets are valuable for the evaluation of the water balance in Rondonia over the last five years.

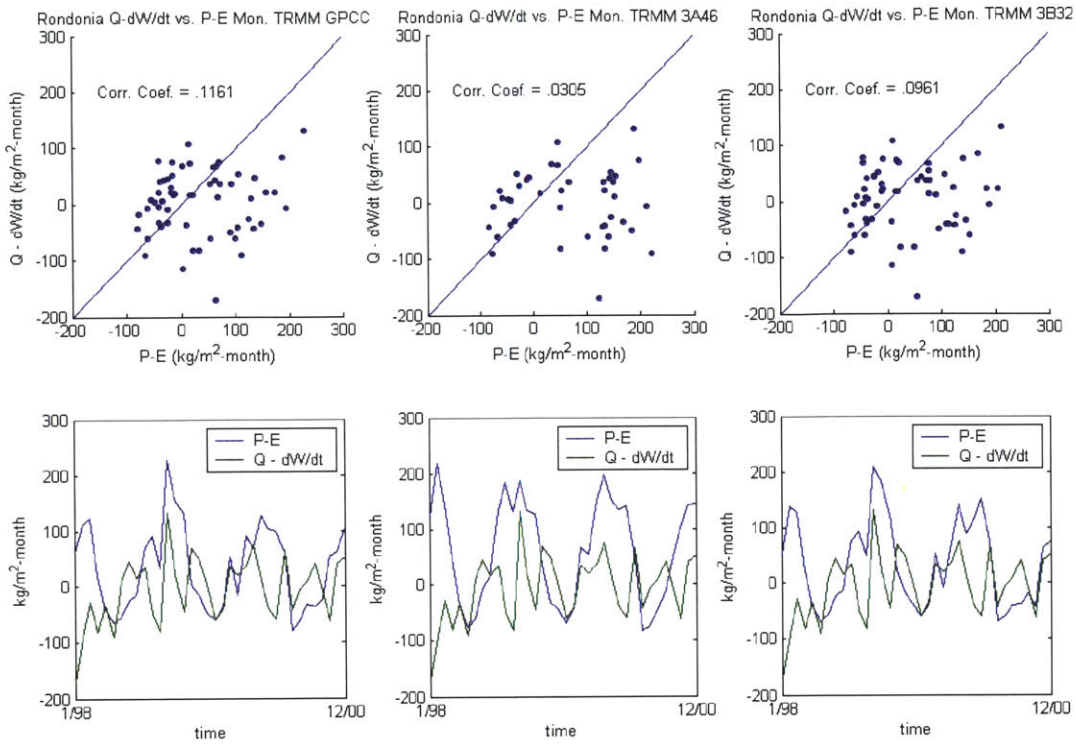


Figure 4-54: Correlation Plots for Rondonia – TRMM Precipitation Comparison Monthly Atmos.

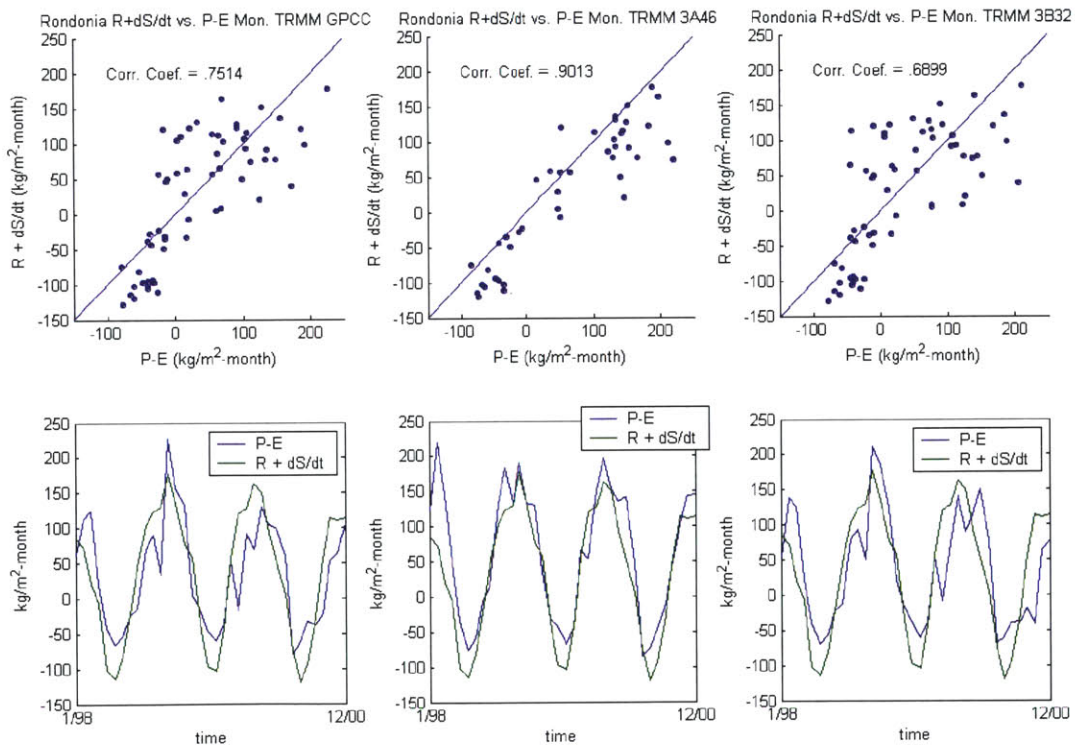


Figure 4-55: Correlation Plots for Rondonia – TRMM Precipitation Comparison Monthly Surf.

## 4.5. Time Scale

The time scale of the water balance has been taken into account in the study of each water balance situation presented. On a monthly basis a water balance typically follows a seasonal cycle, and thus even if error exists in some fields, a balance can still be established because the water balance variables are for example all high in the summer and low in the winter. There is, however, more potential for variability on a monthly time scale because the time averaging of the water balance is less than that of the yearly cycle. A yearly water balance alternatively contains no regular seasonal cycle, but likely is composed of less variable and better averaged data. These properties can be seen using our analysis tools and all default conditions.

Figure 4-56 shows the residual statistic analysis for the default conditions for each of the box control volumes, and contrasts the statistics by time scale of the water balance. For the monthly water balance, the absolute error is consistently slightly higher. However, for the surface balance there is no discernable pattern between the two time scales. There is barely any difference between the bias on the monthly and yearly time scales, and small differences in bias are likely due to rounding errors.

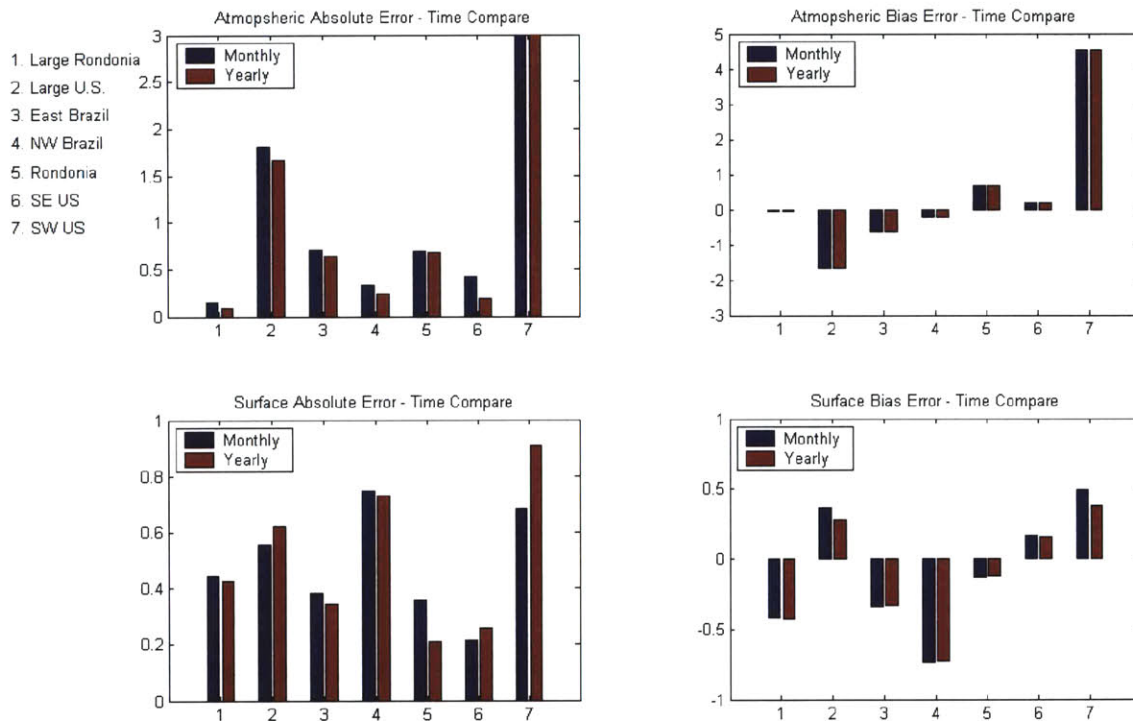


Figure 4-56: Time Scale Comparison – Atmospheric and Surface Absolute Error and Bias Error

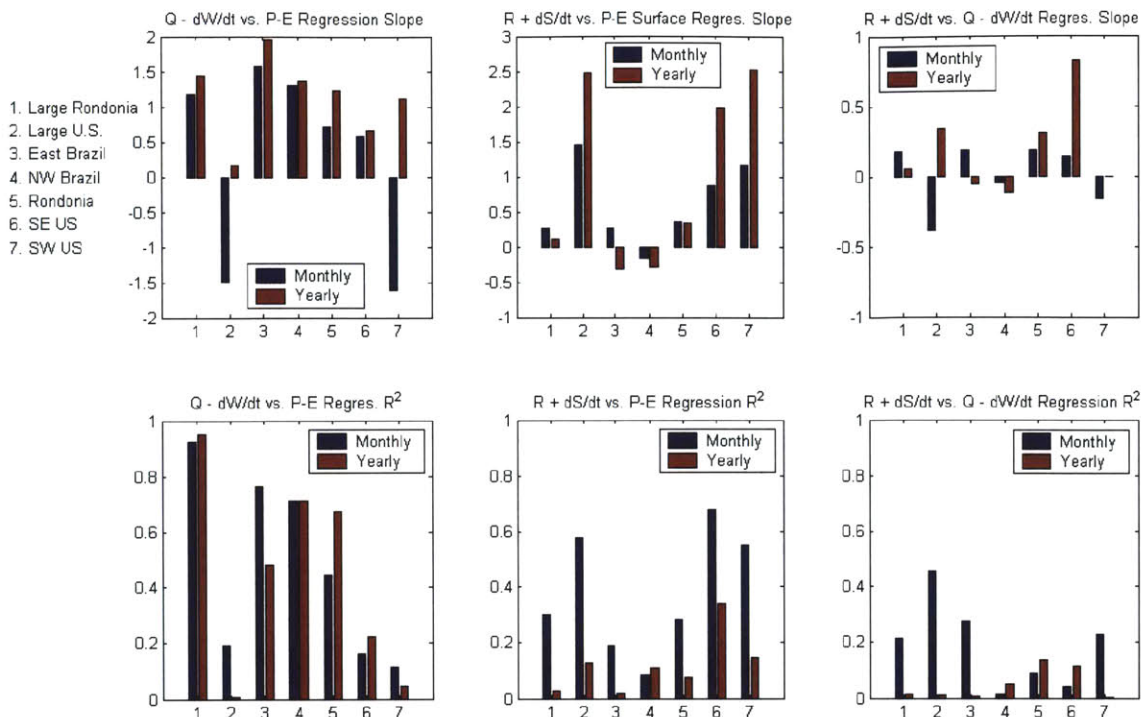


Figure 4-57: Time Scale Comparison – Atmospheric, Surface, and Total Regressed Slope and  $R^2$

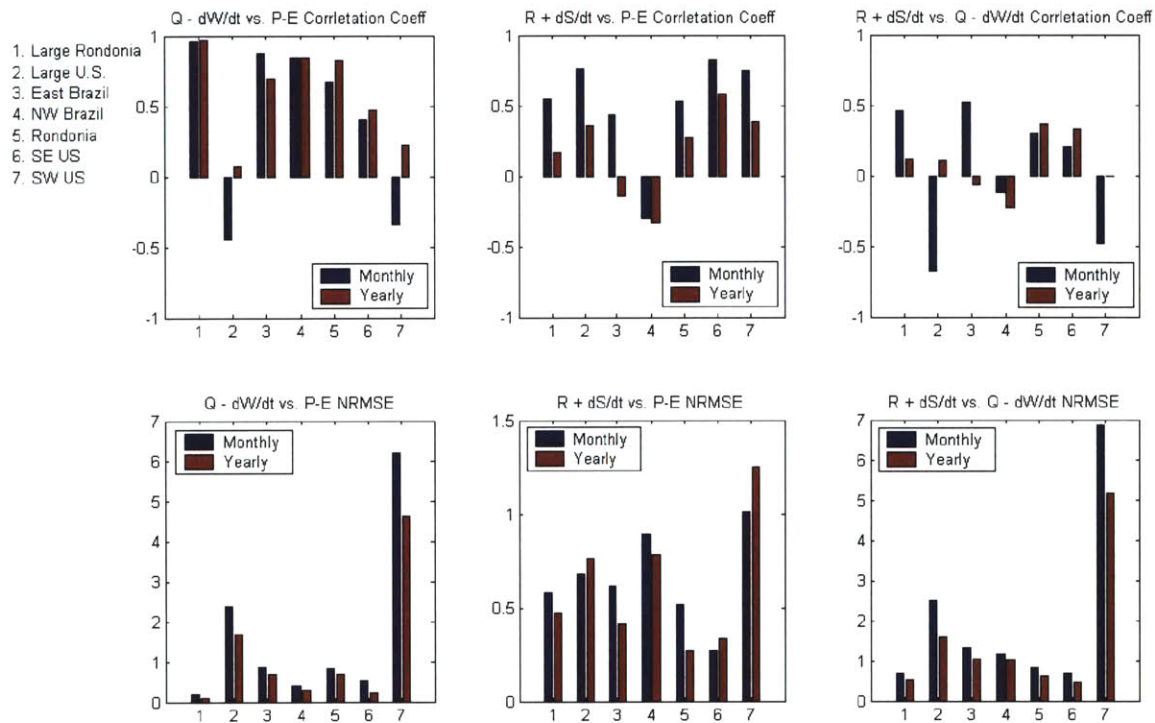


Figure 4-58: Time Scale Comparison – Atmos, Surface, and Total Corr. Coeff. and NRMSE



Figures 4-57 and 4-58 show the relevant correlation statistics for all three of the possible regressions, atmospheric, surface and total, so they can be contrasted by time. For the atmospheric regression the slopes are consistently closer to one on a monthly basis, while the  $R^2$  and correlation coefficients are mostly a draw. The NRMSE is lower for the yearly balance on a consistent basis. However, the correlation between the two sides of the balance is not greatly improved on the monthly basis as shown by the  $R^2$  and correlation statistics.

Three control volumes of interest are presented for the time scale analysis, as we explore all three types of correlations, atmospheric, surface, and total. In both of the first two examples, Figures 4-59 and 4-60, we see how differences can arise between the monthly and yearly time scales. For the atmospheric case in East Brazil, both the monthly and yearly time scales provide a well correlated atmospheric balance. On the monthly scale the peaks match up very well because of the seasonal cycle. However, when the yearly average is taken, the moisture flux values are clearly higher than the precipitation – evaporation term at all points. The same observation is made for the SWUS in Figure 4-60 to an even more extreme state (and replacing moisture flux with runoff). This proves that just because a monthly water balance is good, it doesn't mean when this data is averaged annually the balance will still be good, because the monthly water balance is highly dependent on a seasonal pattern.

In Figure 4-61 the total water balance correlation is observed for Rondonia. The signals are correlated well from month to month and year to year, however the balance is still flawed. There is a large bias that can be clearly seen by the residual analysis.

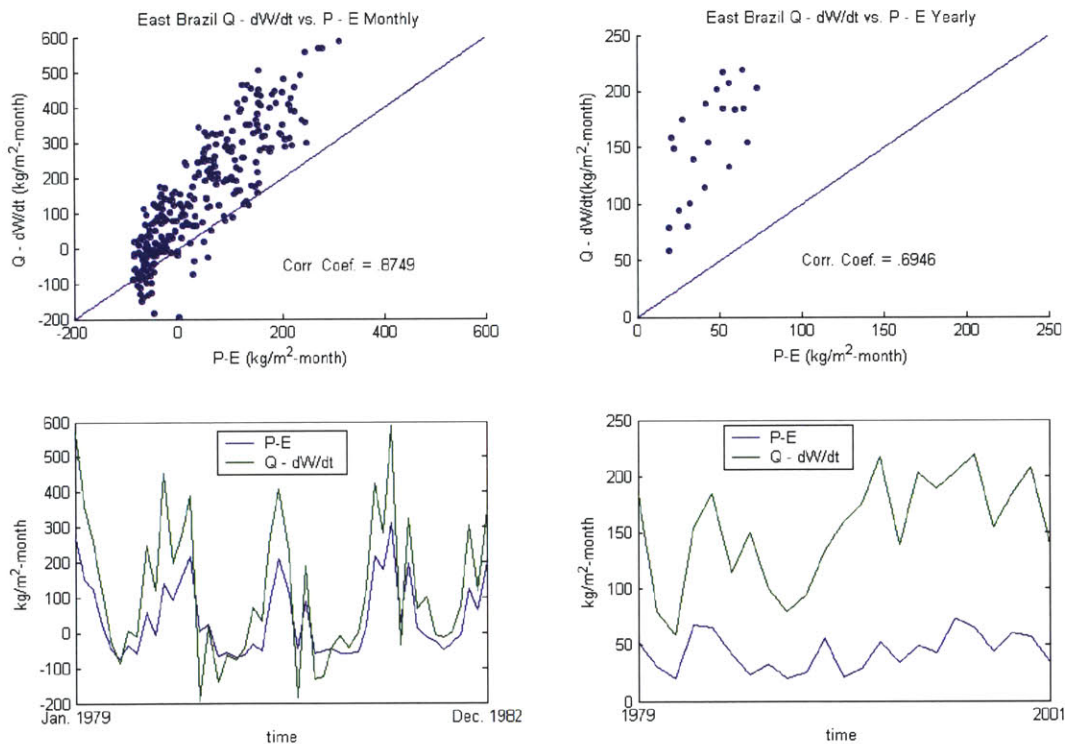


Figure 4-59: Correlation Plots for East Brazil – Time Scale Comparison Atmospheric

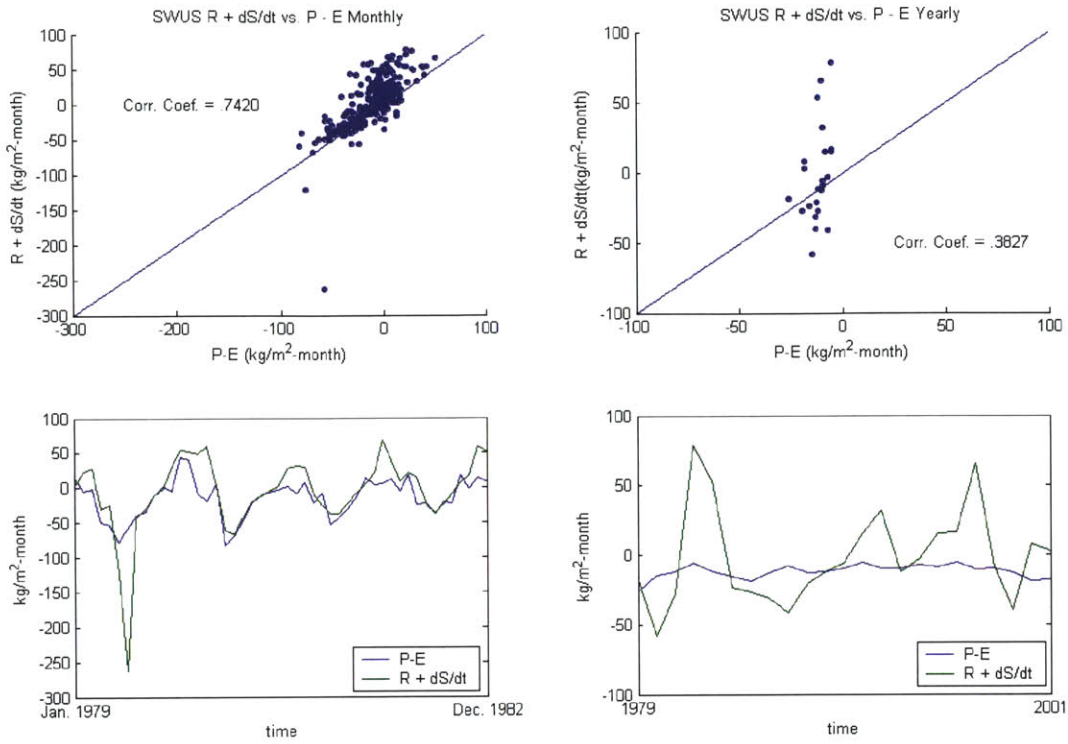


Figure 4-60: Correlation Plots for SWUS – Time Scale Comparison Surface

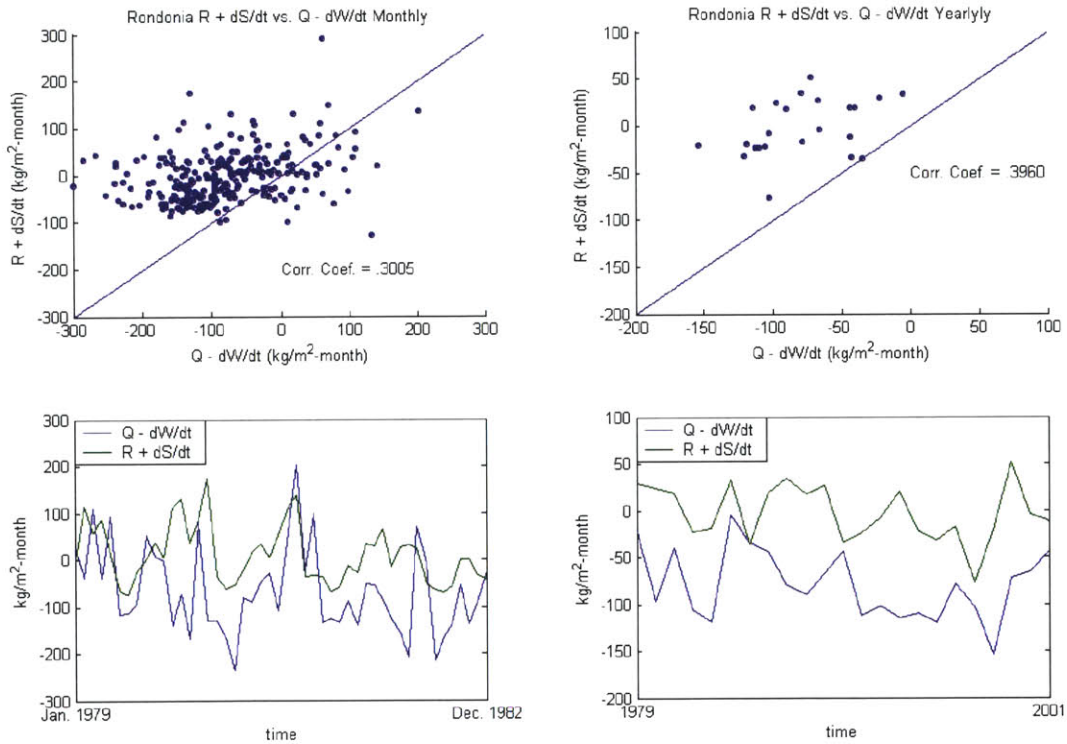


Figure 4-61: Correlation Plots for Rondonia – Time Scale Comparison Total Balance

## 4.6. Storage Considerations

Early data processing indicated poor surface water balances in most areas studied, which necessitated the search for an alternative method for closing the surface water balance. This, as discussed in Section 3.3.2.3 involves the removal of the storage term ( $dS/dt$ ) in the surface balance, and compensating for this deletion by lagging the runoff time series by a period of months when applying it to the water balance. In this analysis only the surface water balance need be considered on the monthly time scale. First the box control volumes will be analyzed in this manner, followed by the basin control volumes. For the basin control volumes the reanalysis runoff will be used in addition to the surface runoff data, as the application of the lag time runoff methods may be useful to both sources of runoff.

In Figure 4-62 the residual statistics for the box control volumes are shown for all of the storage methodologies. In theory the absolute error should be lowest in the case where change in storage is included. Then, in the series of monthly lags, a minimum error should become prevalent for one of the months indicating the best correlation of precipitation and runoff signals without storage. In all but one case (Larger US) the change in surface storage scenario has the lowest absolute error, as it should have. No best lag can be identified. The bias should be about the same for all conditions including when change in storage is included, because the sum of all of the change in storage terms should be zero over the duration of the balance. This behavior is observed.

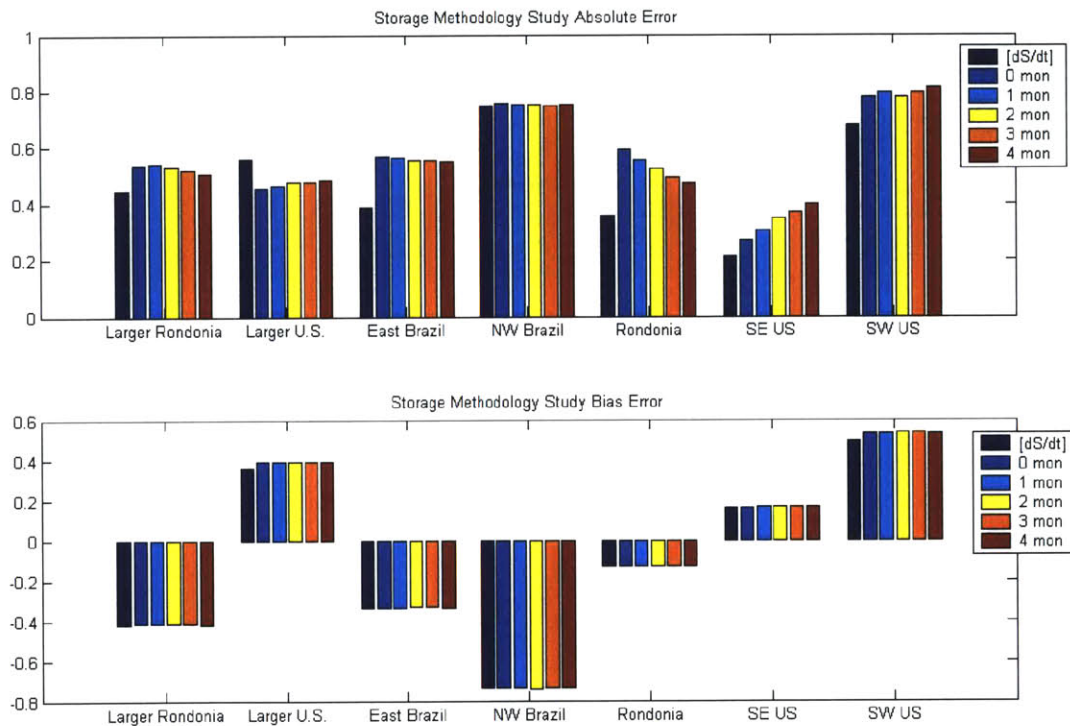


Figure 4-62: Storage Method Comparison – Absolute Error and Bias, Box Control Volumes

In Figures 4-63 through 4-66 the surface correlation is observed between our different surface flux term and both the precipitation – evaporation term and the moisture flux term. Here the total correlation is useful as in the case of the total water balance we are trying to match the storage methodology to the selected moisture flux methodology. Figure 4-63 shows the first attempt at regression with P – E. What is seen here are good correlations when the storage is included, but no regressions are found with near correct properties for any control volumes except the SEUS when looking at the lag time trials. The SEUS control volume displays a lag time of zero months, which is likely the case since it is a small control volume and fairly active in a hydrological sense. Regression and correlation statistics are consistently off for all other lag situations in Figure 4-63 and 4-64.

In Figures 4-65 and 4-66 we test the total correlation. Results are nearly the same. In both cases the correlation and  $R^2$  in NW Brazil appear to get worse as lag time increases, however the slopes on these regressions are negative. The primary purpose of these figures is to show the impossibility of matching the moisture flux cycle to the runoff cycle without surface storage for the box control volume. This just simply can't be done given the coarseness of the data and the small size of these control volumes.

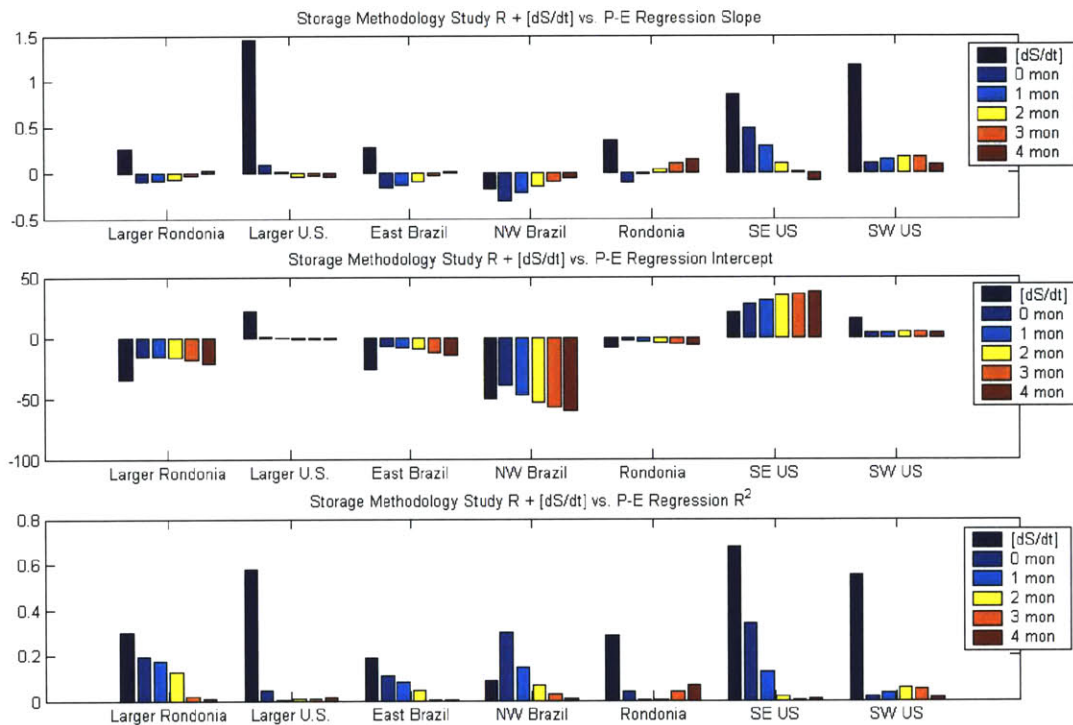


Figure 4-63: Storage Method Comparison – Surface Regression Stats, Box Control Volumes

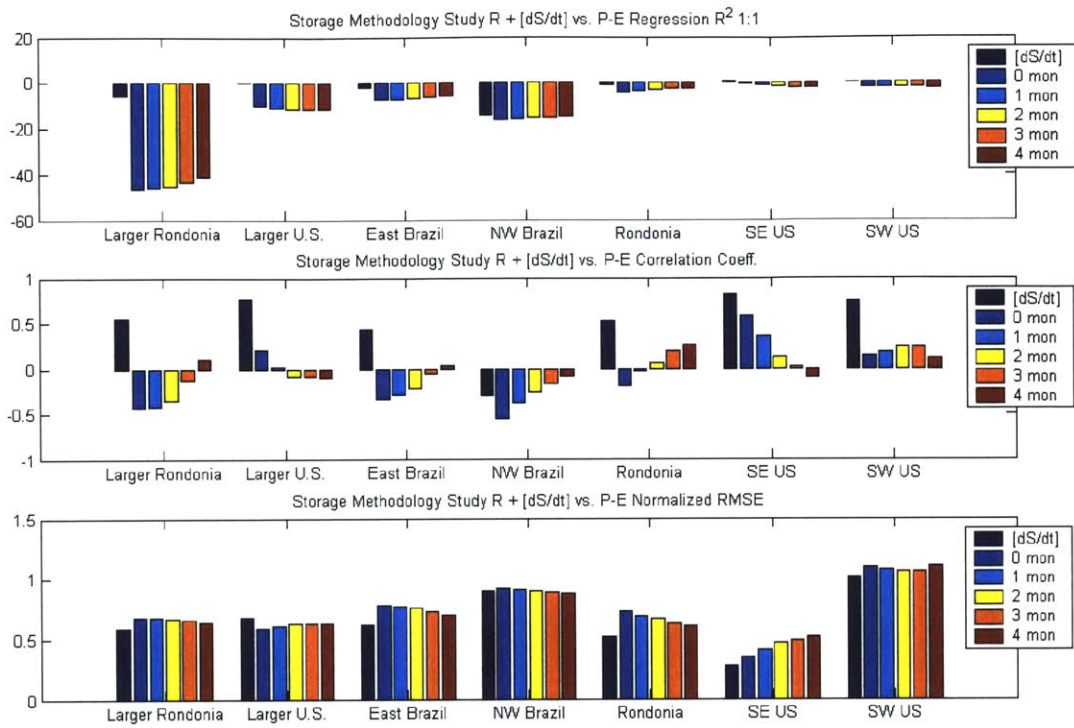


Figure 4-64: Storage Method Comparison – Surface Correlation Stats, Box Control Volumes

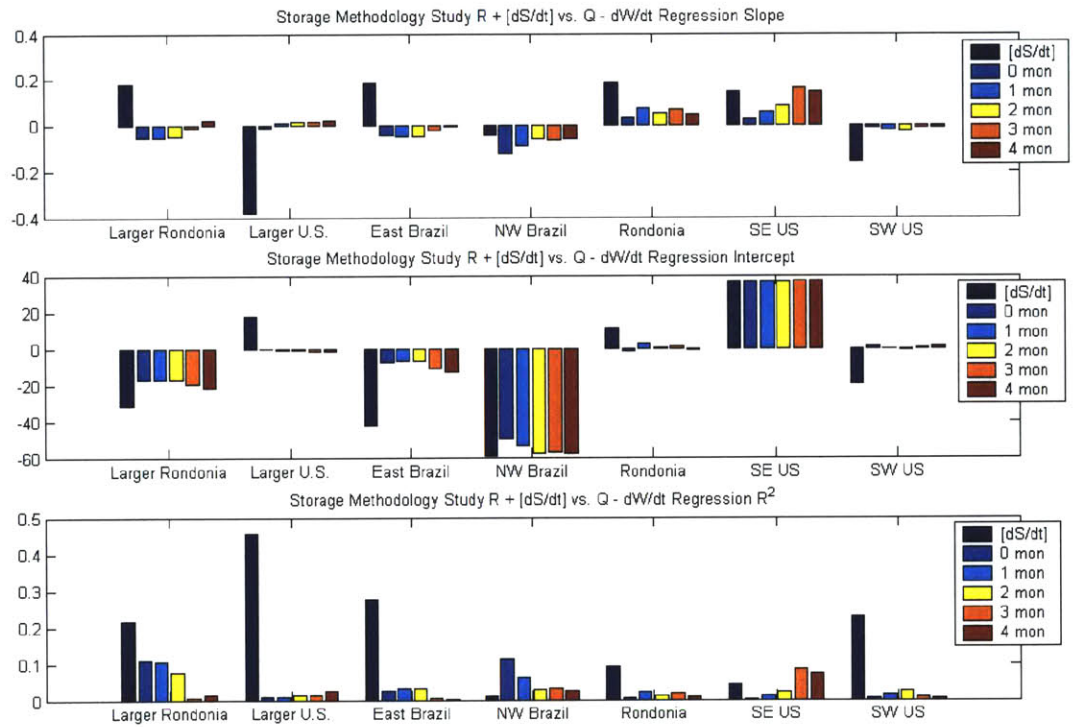


Figure 4-65: Storage Method Comparison – Total Regression Stats, Box Control Volumes

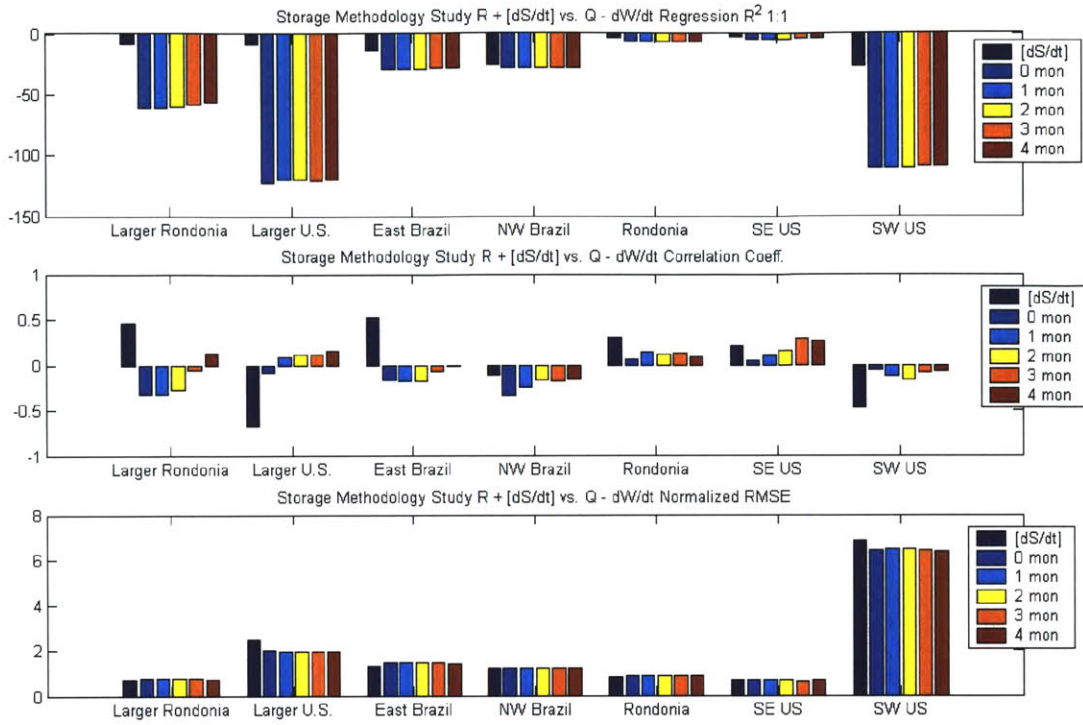


Figure 4-66: Storage Method Comparison – Total Correlation Stats, Box Control Volumes

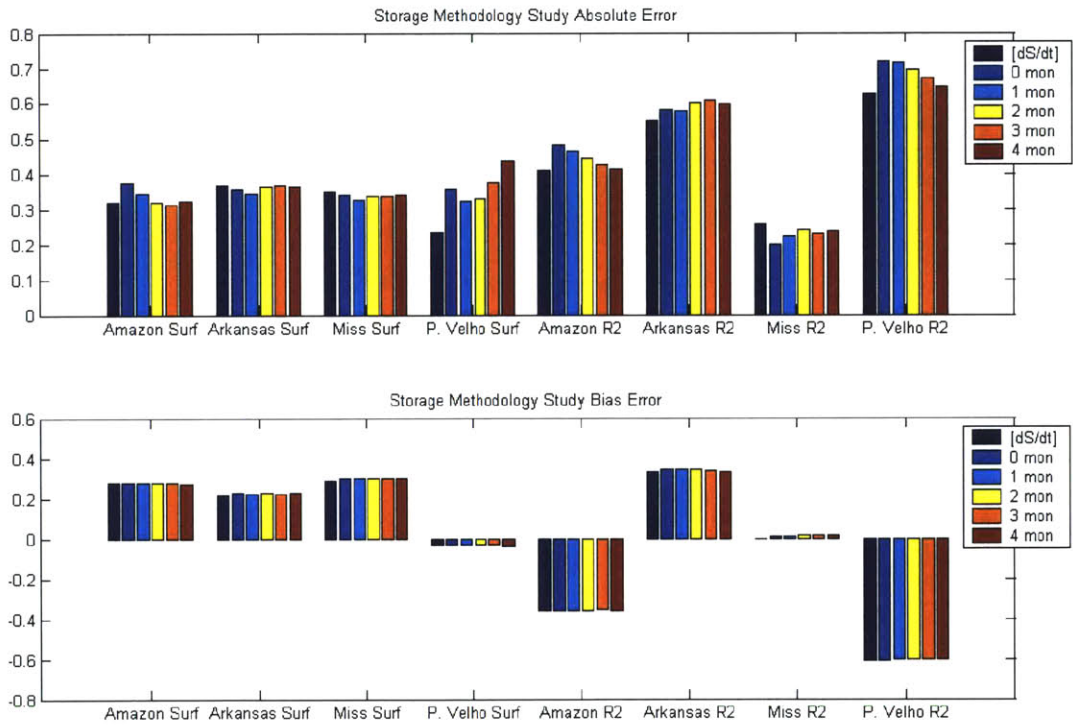


Figure 4-67: Storage Method Comparison – Absolute Error and Bias, Basin Control Volumes

In Figure 4-67 we apply the different storage conditions to the basin control volumes. All different storage and runoff combinations are analyzed for the default conditions. For the surface runoff we begin to see effective lag analysis in terms of absolute error. For the Amazon control volume we see a minimum absolute error at three months, and for the other three basin control volumes a lag time of one month is observed. We do not see these tendencies when using the reanalysis runoff, which is derived by the linear (default) method. In all cases but the Mississippi Basin control volume, the surface runoff improves the balance over the reanalysis runoff. The bias is an ineffective analysis tool for long term averages. What can be observed, however, is the bias of the long term balance. The Porto Velho surface balance is very good when surface runoff is used, while the Mississippi River basin is balanced very well when the reanalysis runoff is used over the long haul.

In Figures 4-68 through 4-71 we again look for a correlation between the various fluxes. Slopes are found to be the best for the lag conditions at one month for the Amazon, Arkansas, and Mississippi control volumes which is realistic. However, when the reanalysis runoff is applied to the basin sized control volume no correlation at all exists between the runoff and precipitation signals. Looking at the correlation statistics is more revealing because what we are looking for in the lag situation is not necessarily a balance, but a correlation between the two sides of the balance. In the bottom section of Figure 4-68, for the surface runoff, the  $R^2$  of the regression is almost always highest for the inclusion of storage, and the lag times match those that we found in the absolute error analysis. The only balance to make any use of the reanalysis runoff is the Mississippi Basin control volume which for storage conditions shows a promising regression and correlation. We will explore this correlation further.

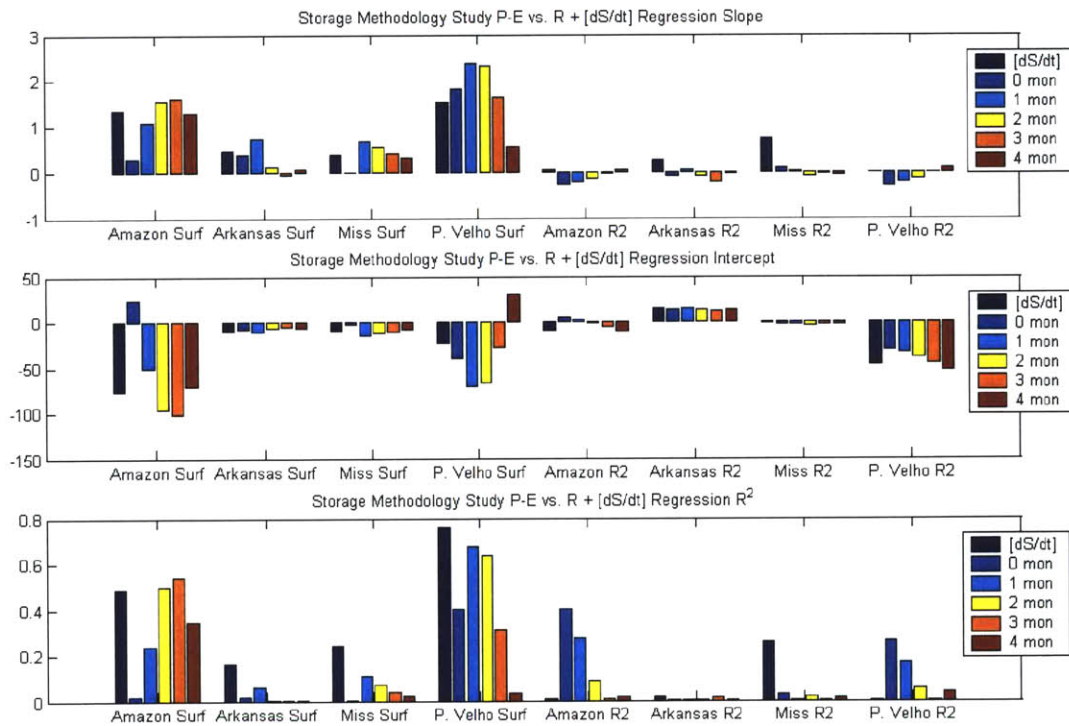


Figure 4-68: Storage Method Comparison – Surface Regression Stats, Basin Control Volumes

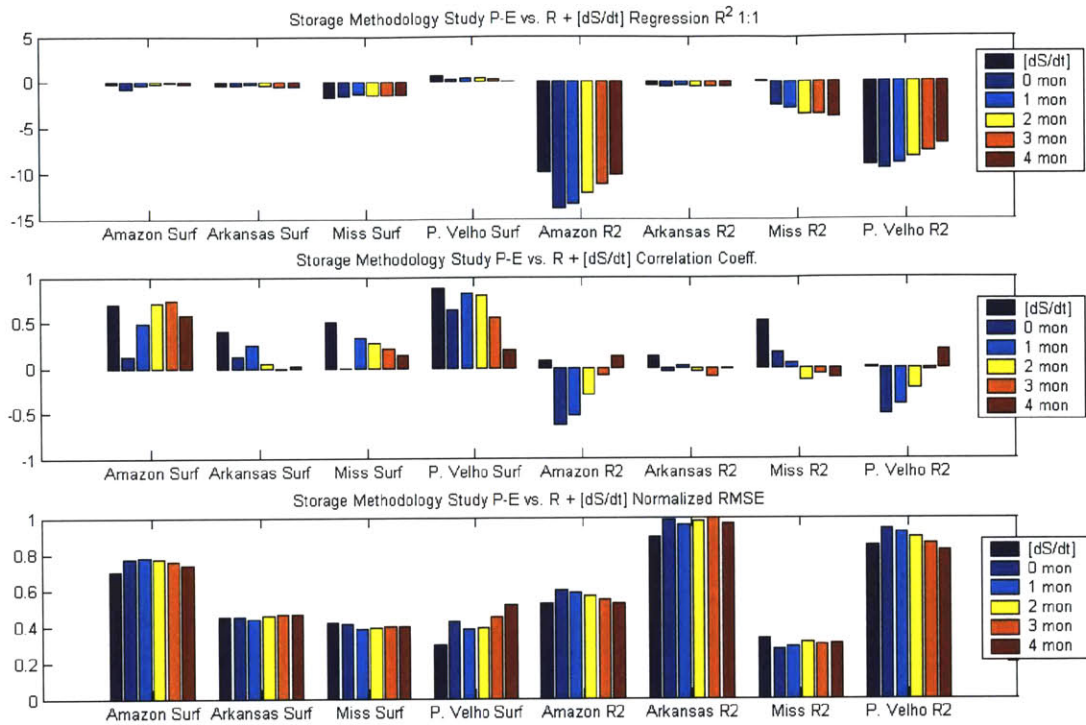


Figure 4-69: Storage Method Comparison – Surface Correlation Stats, Basin Control Volumes

Because of the success at using the surface balance to find lag times, we attempted to find the same patterns by performing the total water balance regression and correlation in Figures 4-70 and 4-71. Nevertheless, there are difficulties finding the moisture flux of a basin sized control volume, and the regression yields wild results, especially in the case of using the reanalysis runoff. Looking at the correlation coefficients in Figure 4-71, we see the same peaks for the lag times that were derived earlier for surface runoff. Being able to find these lag times for a basin is very important to a water balance study because it allows for precipitation and surface runoff to be compared directly without change in storage and thus allows for the completion of a rough water balance without knowledge of soil moisture in a basin.



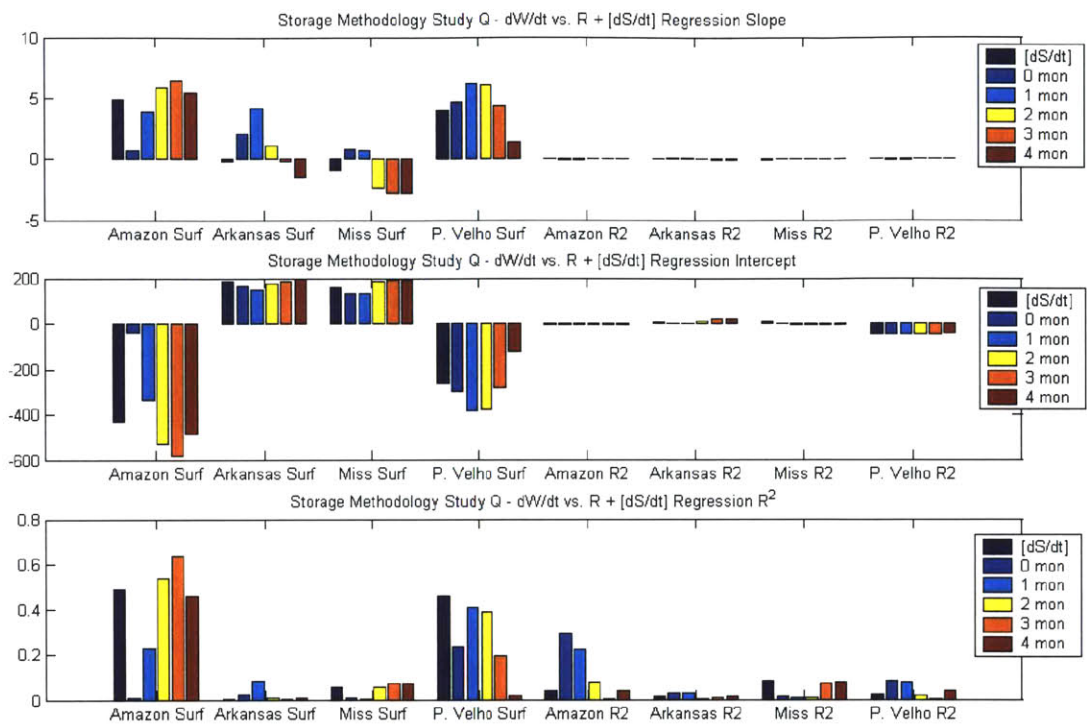


Figure 4-70: Storage Method Comparison – Total Regression Stats, Basin Control Volumes

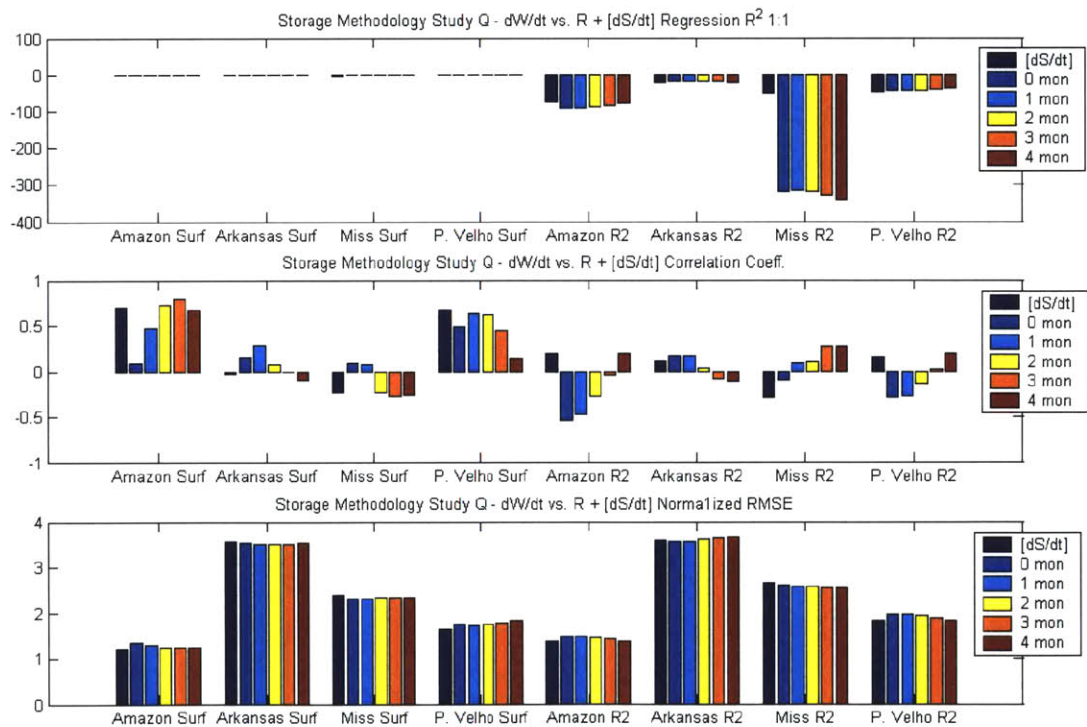


Figure 4-71: Storage Method Comparison – Total Correlation Stats, Basin Control Volumes

Returning to the box control volume, we present the example correlation plots for the SEUS control volume. On the leftmost side of Figure 4-72 a strong correlation and balance is observed for default conditions with change in storage. However, as storage change is removed and lag time is increased the balance and correlation breaks down rapidly. It can be concluded then rather definitively for this control volume, that there is no lag time on a monthly scale between the runoff and the precipitation cycles of the water balance. In Figure 4-73 an even less encouraging example of the storage analysis is provided with the Larger US control volume. With change in storage this water balance does well, however as soon as change in storage is removed the correlation of the two sided balance disappears.

In Figures 4-74 through 4-77 the basin sized control volumes of the Mississippi and Porto Velho basins are analyzed with example correlation plots. The first two figures use surface runoff data, while the second two use the reanalysis runoff. In Figure 4-74 and 4-75 the leftmost balances, the ones which include the surface storage, are the best. However, for the one month lag case in each figure (3<sup>rd</sup> from the left), the runoff is well correlated with the precipitation – evaporation term, which can be better seen in the lower panel snapshot of the data for a few selected years. The signal pickups are subtle in the Mississippi Basin, but not in the case of the Porto Velho Basin, where the pronounced seasonal cycle is picked up on a one month lag. It is interesting that in both of these cases the lag time between the runoff and the precipitation signals are one month, while the size of the basins is quite different.

In Figures 4-76 and 4-77 reanalysis runoff is applied to the same control volumes. The purpose of Figure 4-76 is to show the usefulness of the reanalysis runoff in closing the water balance for the Mississippi Basin control volume, which is balanced well when storage is included. Because of the size of the Mississippi Basin control volume, it could be possible that a consistent lag time can not be found for the control volume because there are such a wide variety of basin response times and hydrological features contained in the basin.

In Figure 4-77 the reanalysis runoff term does not complete the balance well for any of the scenarios presented. The runoff methodology seems to route the runoff in the wrong direction, meaning there is a problem applying the runoff methodology to the basin control volume. These results indicate the reanalysis storage is much more accurate than the runoff term for the basin control volumes. This makes sense because the basins are large control volumes, and the storage change is spatially averaged to fit the precipitation minus evaporation cycle.

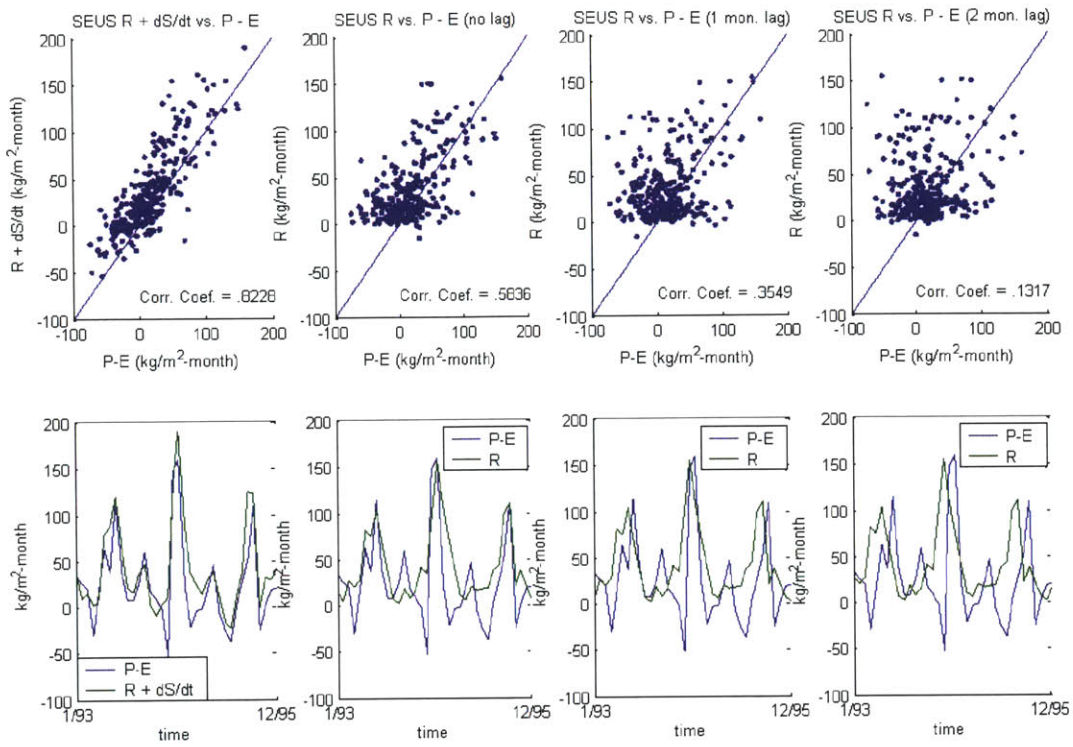


Figure 4-72: Correlation Plots for SEUS – Storage Method Comparison Surface

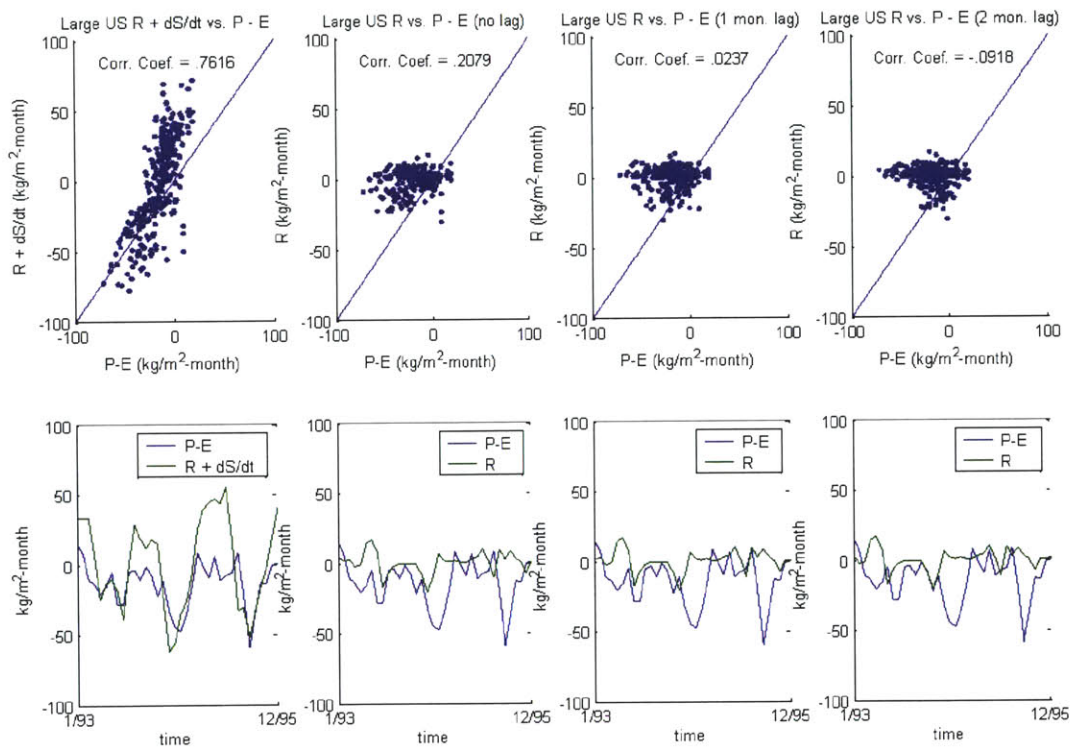


Figure 4-73: Correlation Plots for Larger US – Storage Method Comparison Surface

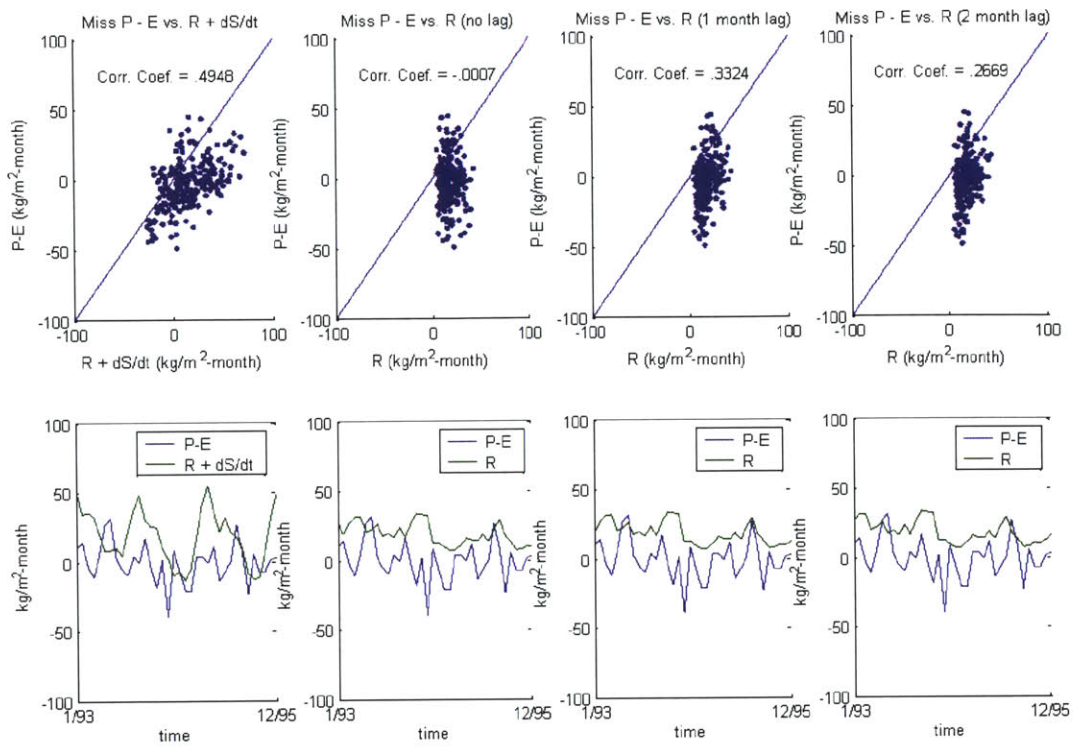


Figure 4-74: Correlation Plots for Miss. Basin – Storage Method Comparison Surface Runoff

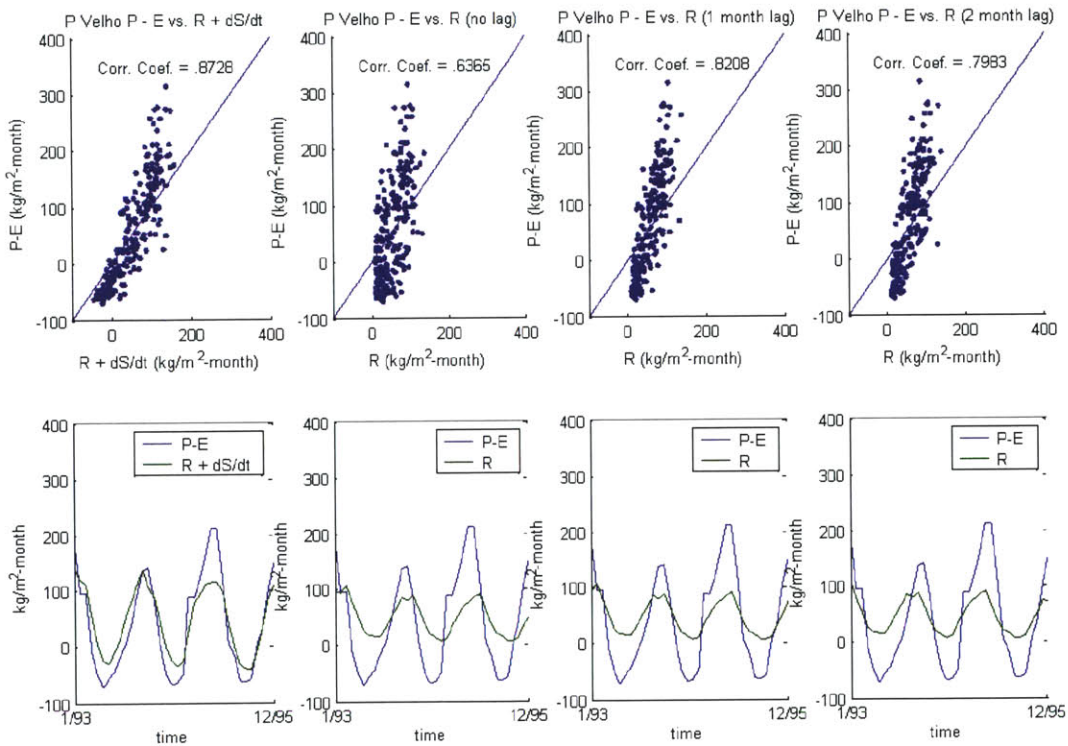


Figure 4-75: Correlation Plots for P. Velho Basin – Storage Method Comparison Surface Runoff

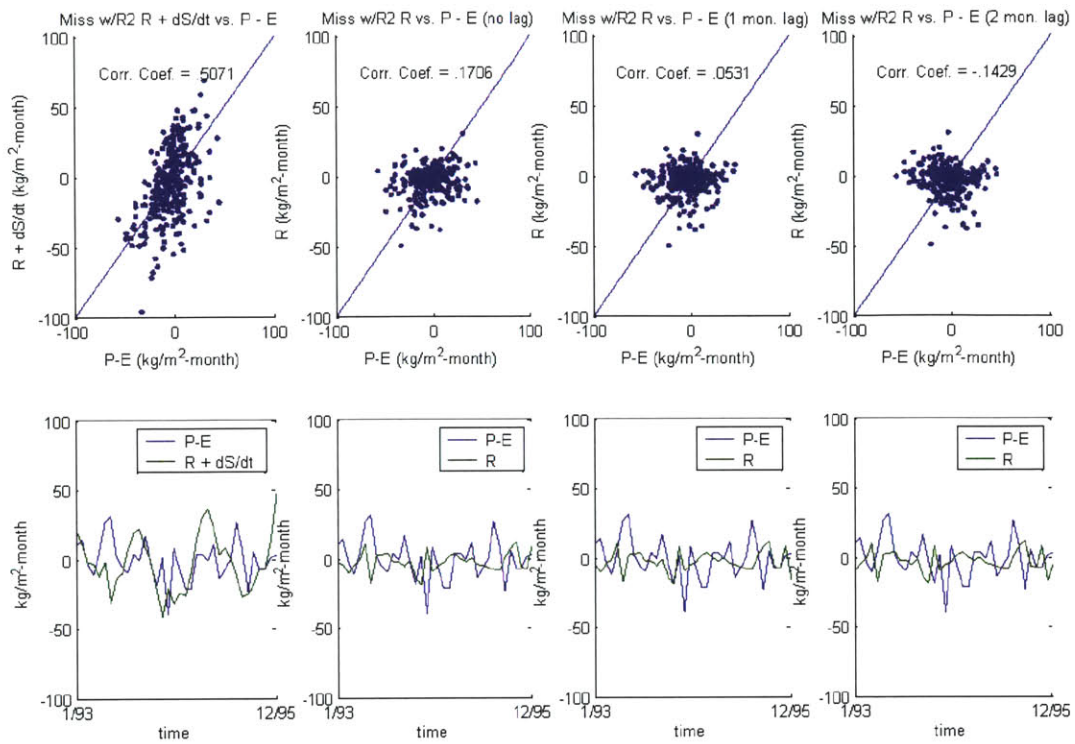


Figure 4-76: Correlation Plots for Miss. Basin – Storage Method Comparison Reanalysis Runoff

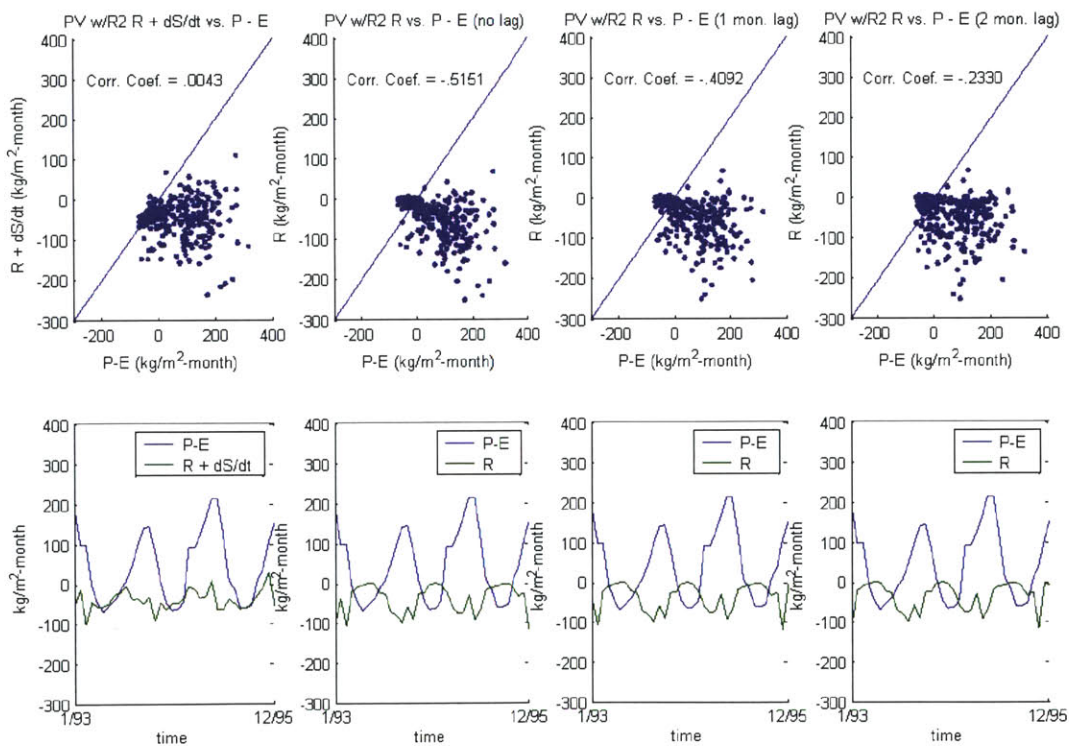


Figure 4-77: Correlation Plots for P. Velho Basin – Storage Method Comparison Re. Runoff

In summary, the lag time balance method is effective only for basin sized control volumes with surface runoff, where meaningful results can be obtained. This can be seen especially for the Port Velho Basin in Figure 4-75. In most other cases, the inclusion of the surface storage term in the water balance is paramount.

## 4.7. Control Volume Type

Although much of this information has been presented before, it is necessary to look here at all 11 control volumes at default conditions so as to directly look at the properties of these balances when compared to one another. By this analysis we can get a sense of how different control volumes of size and shape balance.

Figure 4-78 shows the residual statistics for all 11 of the control volumes for both the atmospheric and surface balances, default conditions. As has been observed before, the Larger US and SWUS control volumes for the atmospheric balance display larger absolute errors and biases. The closure of the box control volume atmospheric water balance relies not only on size, but also on the location of the control volume, with errors common over particularly arid or mountainous land. The Large Rondonia control volume has a better balance than the other four smaller tropical, control volumes. A more complete classification of different sized box control volumes for atmospheric water balances will be considered in the global water balance in Section 4.3. For the basin control volumes and the atmospheric balance the same general trends exist with size and location of control volume. The Amazon Basin and Mississippi Basin balances are better than their smaller sub basins. The tropical Amazon and Porto Velho basins are better balances than the Mississippi and Arkansas basins.

There are no normalized errors in the surface balance greater than one for any control volume. There is more variability associated with the moisture flux calculation as much more water is passing through the atmosphere than the surface in a given month. The trends that existed between groups of control volumes for the atmospheric balance can't be made for the surface balance. The errors are lower for the basin sized control volumes in terms of the absolute error, with comparable errors for the SEUS control volume.

Figures 4-79 and 4-80 show the regression and correlation statistics for all 11 control volumes under default conditions. For this comparison we will consider four statistics for all three of the correlation types, atmospheric, surface, and total. The total regression is appropriate for use here because we are using default conditions, but it is likely not very useful. This information has already been views in Section 4.5 for the box control volumes, but here the basin control volumes are added. In terms of slope of the atmospheric balances, the Larger Rondonia, Northwest Brazil, and Mississippi Basin control volumes show the best adherence to one for both monthly and yearly time scales. Good slopes are hard to come by on the yearly time scale for the surface balances, but a handful are good on the monthly basis, with the best in the SEUS. The total correlation only yields "decent" slopes for the SEUS and Amazon control volumes

The  $R^2$  and correlation coefficient statistics tell very similar stories about the tightness of correlation between the different sides of the water balance, in Figures 4-79 and over to Figure 4-

80. For the atmospheric balance the best correlations are found for the Larger Rondonia and Amazon control volumes while the SWUS and Larger US control volumes are the worst. Monthly correlation for most water balances is stronger than yearly correlation for the surface water balance with strong correlations in the SEUS and the Porto Velho control volumes. A lot of the poor total correlations occur in the basins over the United States and over the smaller control volumes. The basins over the United States (Mississippi and Arkansas) have weak correlations for all three balances (atmospheric, surface, and total). The box control volumes have varying correlations over different geographic areas. The use of the surface runoff data seems to have a pronounced effect on the correlation of the basins being good or bad, and the USGS Runoff data seems to be substandard when plugged into these regressions. In general as we go from atmospheric to surface to total correlations, the coefficients go down.

Continuing on to the NRMSE for the different control volumes on the bottom of Figure 4-80 we see the same atmospheric weaknesses, all in control volumes over the United States (box and basin with the exception of the SEUS). The magnitude of precipitation is smaller over these control volumes decreasing the normalization factor which likely drives the NRMSE up, and these effects are more prominent in the atmospheric balance. For the surface balance we have good conditions for all of the basin control volumes and the SEUS. The total NRMSE is one in nearly all cases, but is smallest for Larger Rondonia, Rondonia, and SEUS, with the Amazon Basin.

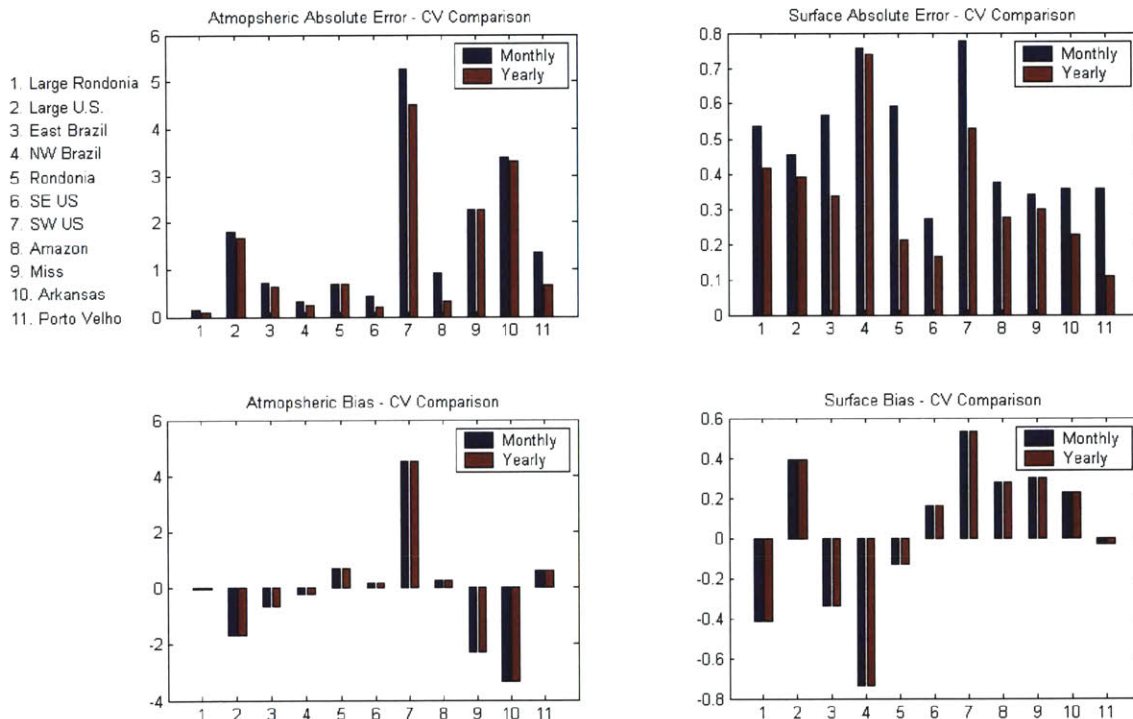


Figure 4-78: Control Volume Comparison – Atmos. and Surface Absolute Error and Bias Error

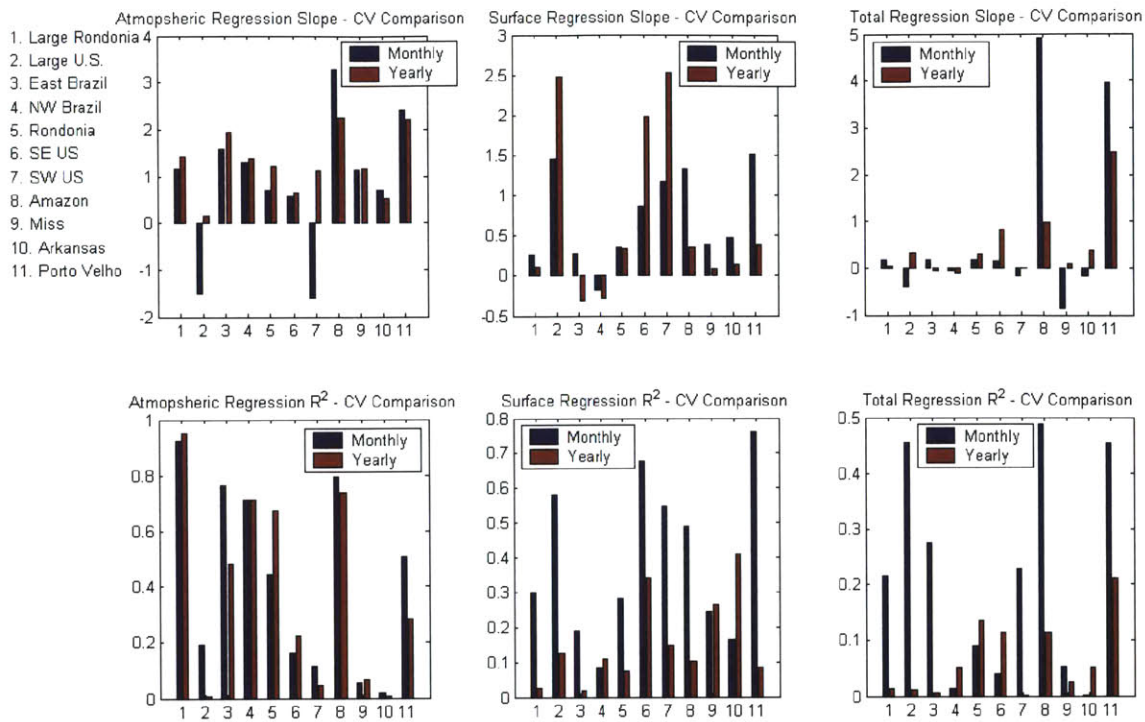


Figure 4-79: Control Volume Comparison - Atmos, Surface, and Total Regressed Slope and  $R^2$

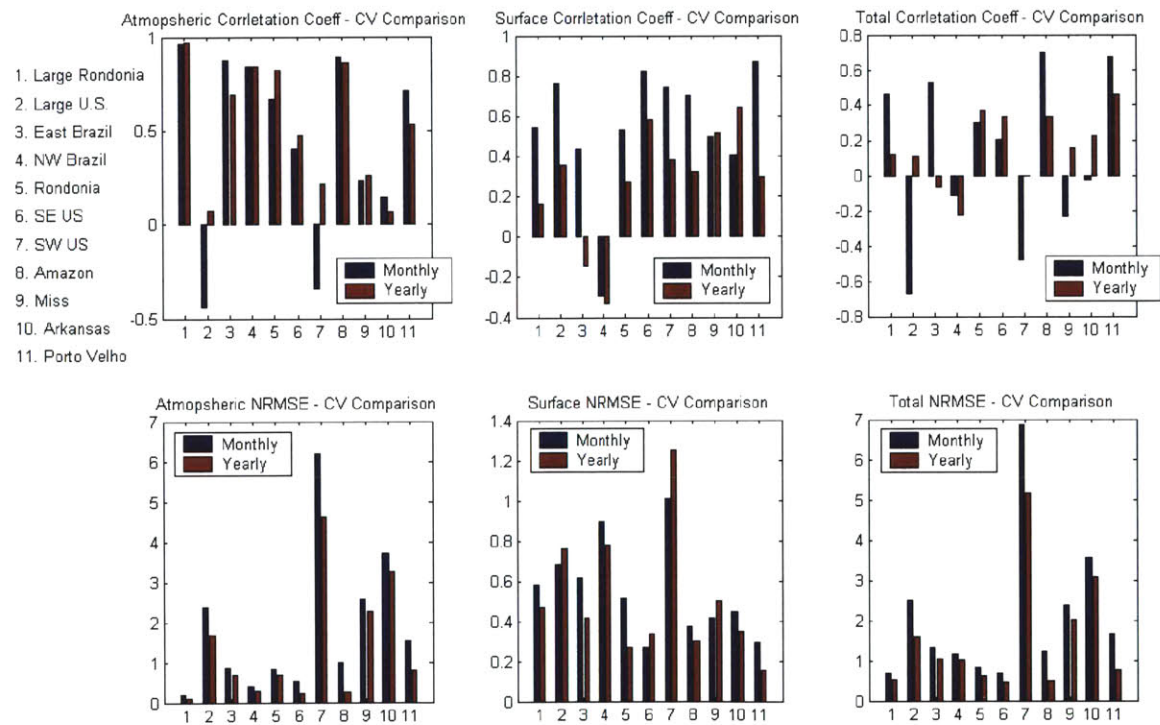


Figure 4-80: Control Volume Comparison - Atmos, Surface, and Total Corr. Coeff. and NRMSE



Four control volumes are selected for a more detailed look, each different in nature. The Rondonia and Larger Rondonia control volumes are selected as the small and large box control volumes, and the Amazon and Porto Velho control volumes are selected as the small and large basin control volumes. All of these control volumes overlap to some extent and represent the same general hydrological conditions. Figure 4-81 displays the residual statistics for only these four selected and different control volumes. The larger areas are better balanced on the atmospheric level, but only the Larger Rondonia control volume is well balanced. The surface balances are improved on the monthly scale for the basin control volumes, and biases are reduced.

In Figure 4-82 and 4-83 the regression and correlation statistics for all three types of correlation are considered. For the atmospheric balance the slope is much better aligned to one for the box control volumes. The Amazon basin slope is fairly high, which illustrates the difficulties of applying the moisture flux methodology to a basin control volume. The surface slopes are poor in all cases but the Porto Velho basin, and therefore because alignment does exist in the atmospheric balance it won't exist in the total balance. However, for the Amazon Basin control volume the alignment does occur for the total two sided balance because of the high slope of the atmospheric balance correlates to the low slope of the surface balance.

The atmospheric  $R^2$  and correlation coefficients are again similar in terms of trends, and for the atmospheric balance shows little difference among types of control volumes, all being fairly good, with the Porto Velho lagging a little behind. The only good correlations in the surface balance are in Porto Velho basin. While the total Amazon slope looked good the correlation is clearly not that strong. For the NRMSE on the bottom of Figure 4-83, the atmospheric balance gets worse from left to right as expected.

In Figure 4-84 the atmospheric correlation plots for the four control volumes of interest are shown on the yearly time scale. The Larger Rondonia control volume has a very tight atmospheric correlation, while the Rondonia control volume has a good signal correlation, but a clear bias. In Figure 4-85 the surface balance is considered. In this case the Larger Rondonia control volume is very biased, while the Rondonia control volume is well balanced. However, neither balance is very well correlated. The surface correlations look better especially in the Porto Velho control volume, as the numbers are realistic with positive side surface flux and precipitation – evaporation. The total correlations are given in Figure 4-86. The Rondonia control volume clearly results in the best balance out of the four balances presented.

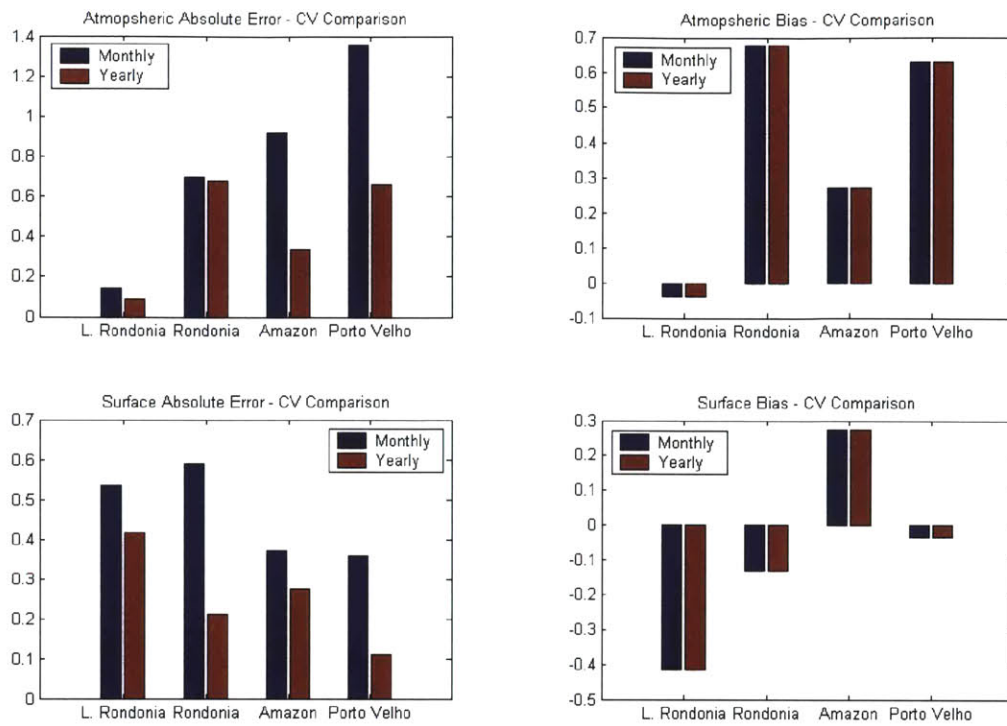


Figure 4-81: Tropical Volume Comparison – Atmos. and Surface Absolute Error and Bias Error

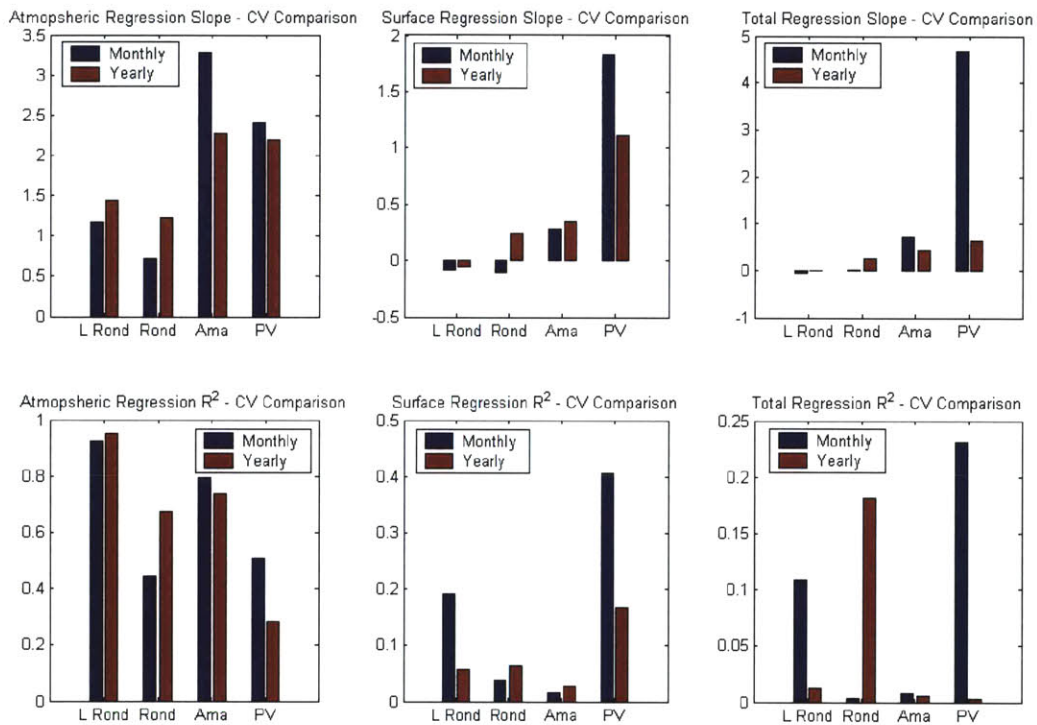


Figure 4-82: Tropical Volume Comparison - Atmos, Surface, and Total Regressed Slope and R<sup>2</sup>

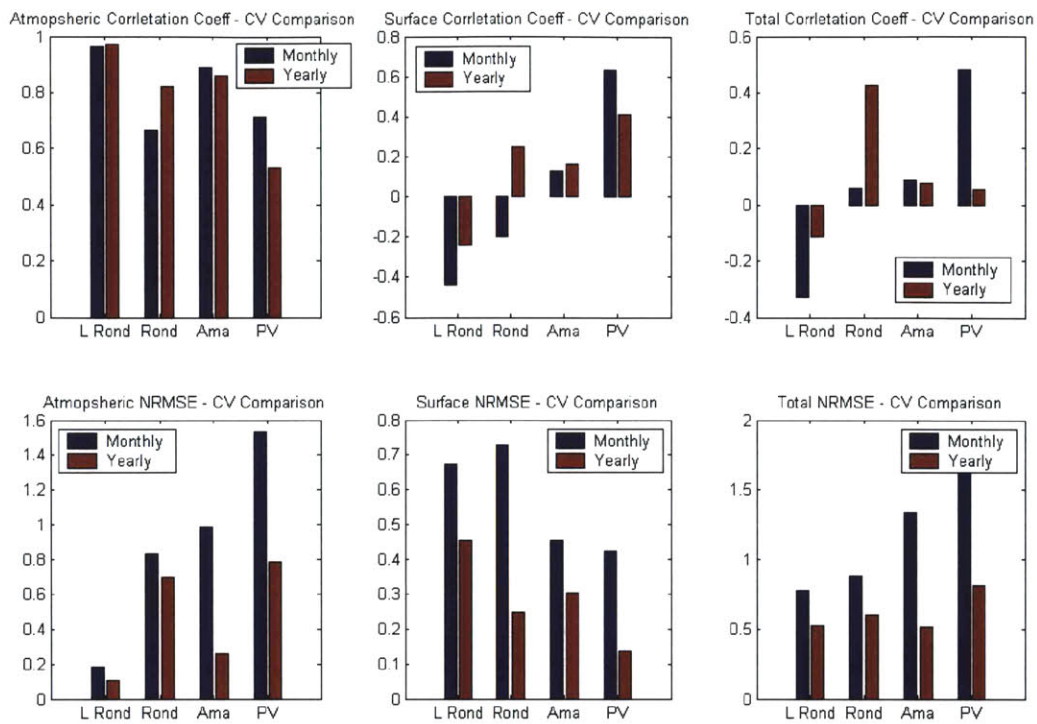


Figure 4-83: Tropical Volume Comparison - Atmos, Surf, and Total Corr. Coeff. and NRMSE

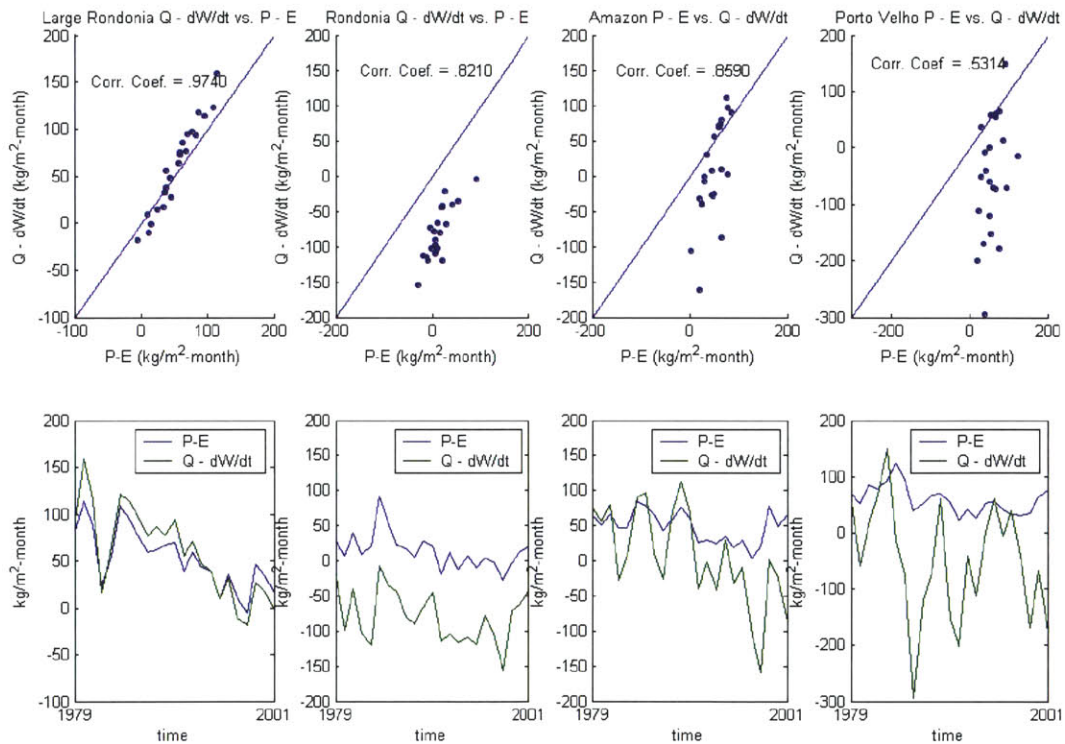


Figure 4-84: Tropical Correlation Plots – Atmospheric Regression

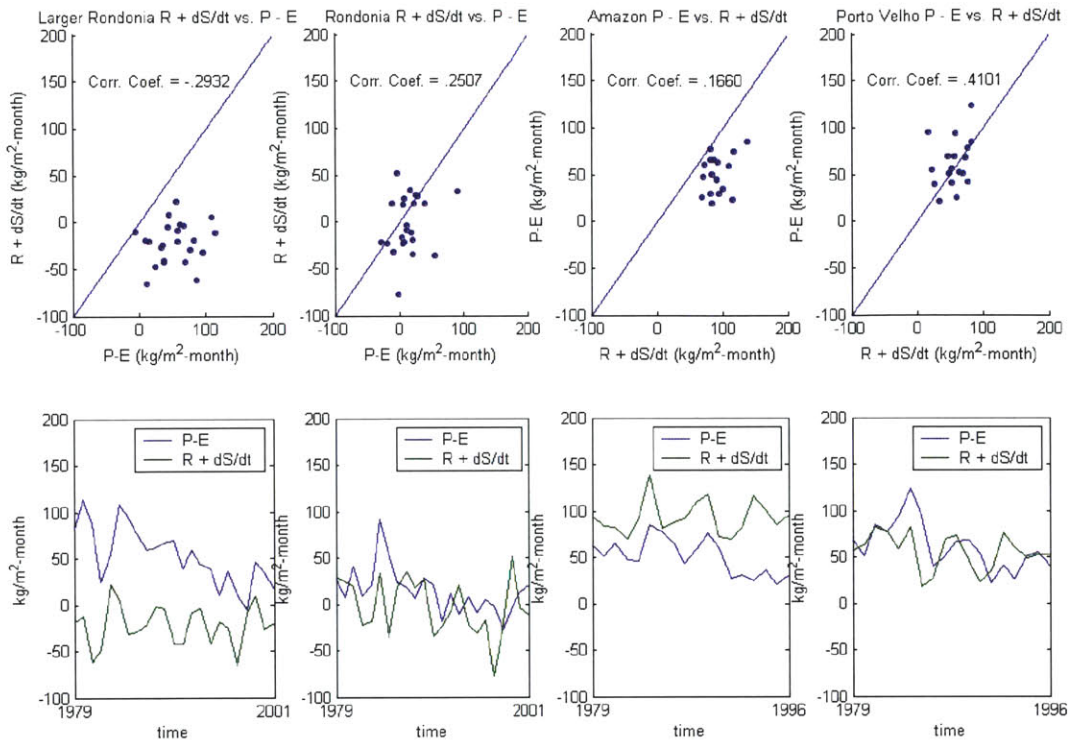


Figure 4-85: Tropical Correlation Plots – Surface Regression

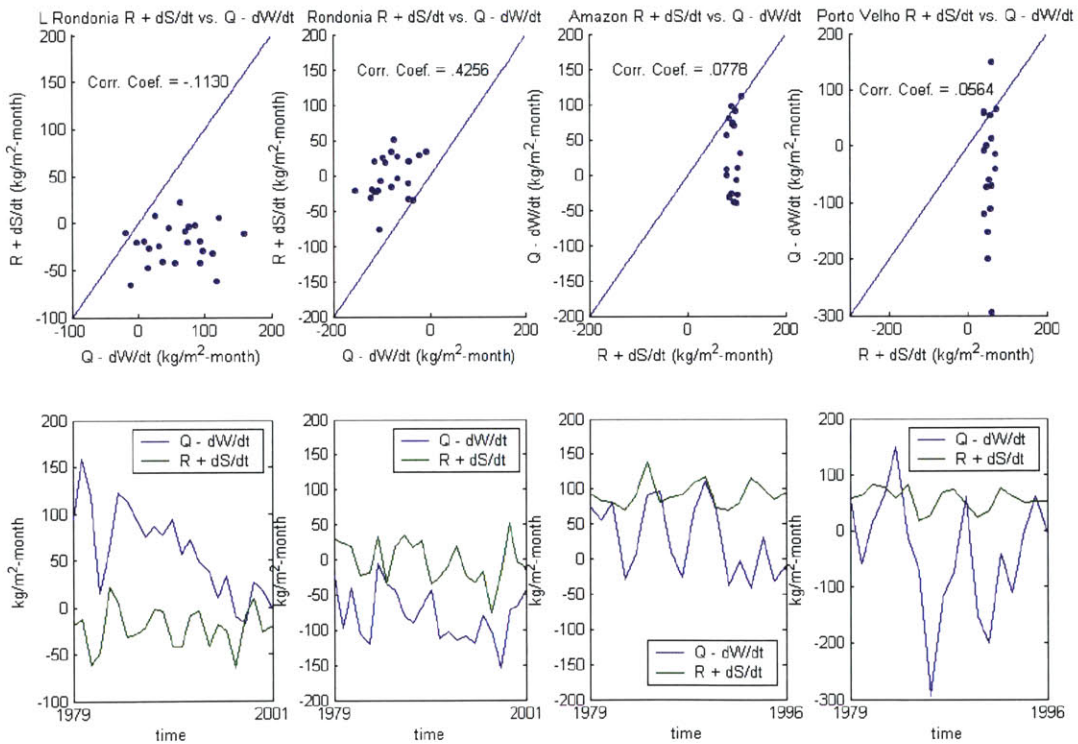


Figure 4-86: Tropical Correlation Plots – Total Regression

## 5. Data Analysis by Study Region

For each control volume there are four basic water balances that we wish to close for each study region: the monthly atmospheric water balance, the yearly atmospheric water balance, the monthly surface water balance, and the yearly surface water balance. Each of these water balances can be evaluated using the Reanalysis-1 and the Reanalysis-2 (55 years and 23 years) respectively. The primary purpose of this section will be to present the water balance over each region for both reanalyses. These water balances are presented with default conditions which include: moisture flux method 2, non-zero change in storage, and the linear runoff methodology routing methodology.

For each control volume there are a large number of permutations of use of characteristics which can be used to perform the water balances. Tables 5.1-5.7 show the characteristics which can be changed in each water balance over each control volume. Five evaluation criteria were used to evaluate the water balances: absolute error, bias error, regression slope, correlation coefficient, and RMSE. All of the normalized statistics were not used because there is no comparison of different control volumes in this evaluation. The  $R^2$  values were not used as most of the water balances yield very poor  $R^2$  results and the adherence of the balance to a linear relationship is evaluated in the correlation coefficient. Intercept is disregarded as it is accounted for in the bias residual statistic.

All possible water scenarios for each control volume and balance are placed in a spreadsheet and with the five evaluation criteria. The different scenarios are sorted and ranked by performance in each criterion. In the case of ranking bias and slope the departure from zero and one are ranked based on a modification to the original properties. Once each of the criteria is ranked, the rankings are summed for each scenario, and the best balances are then sorted and ranked from the lowest sum of ranks to the highest sum of ranks. The spreadsheets which were used to carry out this evaluation are given in Appendix A for reference purpose as the water balance statistics for every balance conducted over all eleven control volumes are presented. Once the evaluation was completed the data for each control volume and water balance was analyzed carefully to pick the best possible water balance, which in some cases was not the balance ranked the highest by the evaluation criterion. These selections will be discussed and displayed further, along with the default reanalysis balances on an area-by-area basis. Also included in each section will be the annual time series of precipitation for all control volumes on a yearly basis.

Control Volume	Moisture Flux	Runoff Method	Storage	Runoff Data	Reanalysis	Time	Precipitation	Time		
Monthly Atmospheric	Method 1	1	N/A	N/A	1	1/1948-12/2002	Reanalysis	N/A		
	Method 2				2	1/1979-12/2001	GPCP	1/1979-12/2001		
							GPCC	1/1986-12/1996		
							GPCC TRMM	1/1998-12/2002		
							GHCN/CAMS	1/1948-11/2002		
							GPI	1/1986-12/2002		
							TRMM 3A46	1/1998-8/2001		
							TRMM 3B43	1/1998-12/2002		
							SSM/I	7/1987-11/1987		
								1/1988-6/1990		
								1/1992-12/2002		
							OPI2	1/1979-8/2002		
							OPI3	1/1979-8/2002		
							CMAP	1/1979-6/2002		
							CMAP2	1/1979-6/2002	Total	
<b>Sum Variables</b>	<b>2</b>	<b>2</b>					<b>13</b>		<b>52</b>	
Control Volume	Moisture Flux	Runoff Method	Storage	Runoff Data	Reanalysis	Time	Precipitation	Time		
Yearly Atmospheric	Method 1	1	N/A	N/A	1	1948-2002	Reanalysis	N/A		
	Method 2	2			2	1979-2001	GPCP	1979-2001		
							GPCC	1986-1996		
							GPCC TRMM	1998-2002		
							GHCN/CAMS	1948-2002		
							GPI	1986-2002		
							TRMM 3A46	1998-2000		
							TRMM 3B43	1998-2002		
							SSM/I	1988-1989		
								1992-2002		
							OPI2	1979-2001		
							OPI3	1979-2001		
							CMAP	1979-2001		
							CMAP2	1979-2001	Total	
<b>Sum Variables</b>	<b>2</b>	<b>2</b>					<b>13</b>		<b>52</b>	
Control Volume	Moisture Flux	Runoff Method	Storage	Runoff Data	Reanalysis	Time	Precipitation	Time		
Monthly Surface	Method 1	Linear	Yes	Reanalysis	1	1/1948-12/2002	Reanalysis	N/A		
		Point	No		2	1/1979-12/2001	GPCP	1/1979-12/2001		
		Area	1 mon					GPCC	1/1986-12/1996	
			2 mon					GPCC TRMM	1/1998-12/2002	
		3 mon				GHCN/CAMS	1/1948-11/2002			
		4 mon				GPI	1/1986-12/2002			
						TRMM 3A46	1/1998-8/2001			
						TRMM 3B43	1/1998-12/2002			
						SSM/I	7/1987-11/1987			
								1/1988-6/1990		
								1/1992-12/2002		
							OPI2	1/1979-8/2002		
							OPI3	1/1979-8/2002		
							CMAP	1/1979-6/2002		
							CMAP2	1/1979-6/2002	Total	
<b>Sum Variables</b>	<b>1</b>	<b>3</b>	<b>6</b>	<b>1</b>	<b>2</b>		<b>13</b>		<b>468</b>	
Control Volume	Moisture Flux	Runoff Method	Storage	Runoff Data	Reanalysis	Time	Precipitation	Time		
Yearly Surface	Method 1	Linear	Yes	Reanalysis	1	1948-2002	Reanalysis	N/A		
		Point	No		2	1979-2001	GPCP	1979-2001		
		Area					GPCC	1986-1996		
							GPCC TRMM	1998-2002		
							GHCN/CAMS	1948-2002		
							GPI	1986-2002		
							TRMM 3A46	1998-2000		
							TRMM 3B43	1998-2002		
							SSM/I	1988-1989		
								1992-2002		
							OPI2	1979-2001		
							OPI3	1979-2001		
							CMAP	1979-2001		
							CMAP2	1979-2001	Total	
<b>Sum Variables</b>	<b>1</b>	<b>3</b>	<b>2</b>	<b>1</b>	<b>2</b>		<b>13</b>		<b>156</b>	

Table 5-1: Water Balance Combinations for Eastern Brazil, Northwest Brazil, Larger Rondonia, Rondonia

Control Volume	Moisture Flux	Runoff Method	Storage	Runoff Data	Reanalysis	Time	Precipitation	Time	
Monthly Atmospheric	Method 1	1	N/A	N/A	1	1/1948-12/2002	Reanalysis	N/A	
	Method 2				2	1/1979-12/2001	GPCP	1/1979-12/2001	
							GPCC	1/1986-12/1996	
							GPCC TRMM	1/1998-12/2002	
							GHCN/CAMS	1/1948-11/2002	
							GPI	1/1986-12/2002	
							TRMM 3B43	1/1998-12/2002	
							SSM/I	7/1987-11/1987	
								1/1988-6/1990	
								1/1992-12/2002	
							OPI2	1/1979-8/2002	
							OPI3	1/1979-8/2002	
							CMAP	1/1979-6/2002	
						CMAP2	1/1979-6/2002		
<b>Sum Variables</b>	<b>2</b>	<b>2</b>					<b>12</b>		<b>48</b>
Control Volume	Moisture Flux	Runoff Method	Storage	Runoff Data	Reanalysis	Time	Precipitation	Time	
Yearly Atmospheric	Method 1	1	N/A	N/A	1	1948-2002	Reanalysis	N/A	
	Method 2	2			2	1979-2001	GPCP	1979-2001	
							GPCC	1986-1996	
							GPCC TRMM	1998-2002	
							GHCN/CAMS	1948-2002	
							GPI	1986-2002	
							TRMM 3B43	1998-2002	
							SSM/I	1988-1989	
								1992-2002	
							OPI2	1979-2001	
							OPI3	1979-2001	
							CMAP	1979-2001	
							CMAP2	1979-2001	
<b>Sum Variables</b>	<b>2</b>	<b>2</b>					<b>12</b>		<b>48</b>
Control Volume	Moisture Flux	Runoff Method	Storage	Runoff Data	Reanalysis	Time	Precipitation	Time	
Monthly Surface	Method 1	Linear	Yes	Reanalysis	1	1/1948-12/2002	Reanalysis	N/A	
		Point	No		2	1/1979-12/2001	GPCP	1/1979-12/2001	
		Area	1 mon				GPCC	1/1986-12/1996	
			2 mon				GPCC TRMM	1/1998-12/2002	
			3 mon				GHCN/CAMS	1/1948-11/2002	
			4 mon				GPI	1/1986-12/2002	
							TRMM 3B43	1/1998-12/2002	
							SSM/I	7/1987-11/1987	
								1/1988-6/1990	
								1/1992-12/2002	
							OPI2	1/1979-8/2002	
							OPI3	1/1979-8/2002	
							CMAP	1/1979-6/2002	
						CMAP2	1/1979-6/2002		
<b>Sum Variables</b>	<b>1</b>	<b>3</b>	<b>6</b>	<b>1</b>	<b>2</b>		<b>12</b>		<b>432</b>
Control Volume	Moisture Flux	Runoff Method	Storage	Runoff Data	Reanalysis	Time	Precipitation	Time	
Yearly Surface	Method 1	Linear	Yes	Reanalysis	1	1948-2002	Reanalysis	N/A	
		Point	No		2	1979-2001	GPCP	1979-2001	
		Area					GPCC	1986-1996	
							GPCC TRMM	1998-2002	
							GHCN/CAMS	1948-2002	
							GPI	1986-2002	
							TRMM 3B43	1998-2002	
							SSM/I	1988-1989	
								1992-2002	
							OPI2	1979-2001	
							OPI3	1979-2001	
							CMAP	1979-2001	
							CMAP2	1979-2001	
<b>Sum Variables</b>	<b>1</b>	<b>3</b>	<b>2</b>	<b>1</b>	<b>2</b>		<b>12</b>		<b>144</b>

Table 5-2: Water Balance Combinations for SEUS

Control Volume	Moisture Flux	Runoff Method	Storage	Runoff Data	Reanalysis	Time	Precipitation	Time		
Monthly Atmospheric	Method 1	1	N/A	N/A	1	1/1948-12/2002	Reanalysis	N/A		
	Method 2				2	1/1979-12/2001	GPCP	1/1979-12/2001		
							GPCC	1/1986-12/1996		
							GPCC TRMM	1/1998-12/2002		
							GHCN/CAMS	1/1948-11/2002		
							SSM/I	7/1987-11/1987		
								1/1988-6/1990		
								1/1992-12/2002		
								OPI2	1/1979-8/2002	
								OPI3	1/1979-8/2002	
								CMAP	1/1979-6/2002	
							CMAP2	1/1979-6/2002		
<b>Sum Variables</b>	<b>2</b>	<b>2</b>					<b>10</b>		<b>40</b>	
Yearly Atmospheric	Method 1	1	N/A	N/A	1	1948-2002	Reanalysis	N/A		
	Method 2	2			2	1979-2001	GPCP	1979-2001		
							GPCC	1986-1996		
							GPCC TRMM	1998-2002		
							GHCN/CAMS	1948-2002		
							SSM/I	1988-1989		
								1992-2002		
								OPI2	1979-2001	
								OPI3	1979-2001	
								CMAP	1979-2001	
								CMAP2	1979-2001	
<b>Sum Variables</b>	<b>2</b>	<b>2</b>					<b>10</b>		<b>40</b>	
Monthly Surface	Method 1	Linear	Yes	Reanalysis	1	1/1948-12/2002	Reanalysis	N/A		
		Point	No		2	1/1979-12/2001	GPCP	1/1979-12/2001		
		Area	1 mon					GPCC	1/1986-12/1996	
			2 mon					GPCC TRMM	1/1998-12/2002	
			3 mon					GHCN/CAMS	1/1948-11/2002	
			4 mon					SSM/I	7/1987-11/1987	
								1/1988-6/1990		
								1/1992-12/2002		
								OPI2	1/1979-8/2002	
								OPI3	1/1979-8/2002	
								CMAP	1/1979-6/2002	
							CMAP2	1/1979-6/2002		
<b>Sum Variables</b>	<b>1</b>	<b>3</b>	<b>6</b>	<b>1</b>	<b>2</b>		<b>10</b>		<b>360</b>	
Yearly Surface	Method 1	Linear	Yes	Reanalysis	1	1948-2002	Reanalysis	N/A		
		Point	No		2	1979-2001	GPCP	1979-2001		
		Area						GPCC	1986-1996	
								GPCC TRMM	1998-2002	
								GHCN/CAMS	1948-2002	
								SSM/I	1988-1989	
								1992-2002		
								OPI2	1979-2001	
								OPI3	1979-2001	
								CMAP	1979-2001	
								CMAP2	1979-2001	
<b>Sum Variables</b>	<b>1</b>	<b>3</b>	<b>2</b>	<b>1</b>	<b>2</b>		<b>10</b>		<b>120</b>	

Table 5-3: Water Balance Combinations for SWUS



Control Volume	Moisture Flux	Runoff Method	Storage	Runoff Data	Reanalysis	Time	Precipitation	Time		
Monthly Atmospheric	Method 1	1	N/A	N/A	1	1/1948-12/2002	Reanalysis	N/A		
							Method 2	2	1/1979-12/2001	GPCP
								GPCC	1/1986-12/1996	
								GPCC TRMM	1/1998-12/2002	
								GHCN/CAMS	1/1948-11/2002	
								OPI2	1/1979-8/2002	
								OPI3	1/1979-8/2002	
								CMAF	1/1979-6/2002	
								CMAF2	1/1979-6/2002	
	<b>Sum Variables</b>	<b>2</b>	<b>2</b>					<b>9</b>		<b>36</b>
Control Volume	Moisture Flux	Runoff Method	Storage	Runoff Data	Reanalysis	Time	Precipitation	Time		
Yearly Atmospheric	Method 1	1	N/A	N/A	1	1948-2002	Reanalysis	N/A		
							Method 2	2	1979-2001	GPCP
								GPCC	1986-1996	
								GPCC TRMM	1998-2002	
								GHCN/CAMS	1948-2002	
								OPI2	1979-2001	
								OPI3	1979-2001	
								CMAF	1979-2001	
								CMAF2	1979-2001	
	<b>Sum Variables</b>	<b>2</b>	<b>2</b>					<b>9</b>		<b>36</b>
Control Volume	Moisture Flux	Runoff Method	Storage	Runoff Data	Reanalysis	Time	Precipitation	Time		
Monthly Surface	Method 1	Linear	Yes	Reanalysis	1	1/1948-12/2002	Reanalysis	N/A		
			No		2	1/1979-12/2001	GPCP	1/1979-12/2001		
		Area		1 mon				GPCC	1/1986-12/1996	
			2 mon				GPCC TRMM	1/1998-12/2002		
			3 mon				GHCN/CAMS	1/1948-11/2002		
			4 mon				OPI2	1/1979-8/2002		
								OPI3	1/1979-8/2002	
								CMAF	1/1979-6/2002	
								CMAF2	1/1979-6/2002	
	<b>Sum Variables</b>	<b>1</b>	<b>3</b>	<b>6</b>	<b>1</b>	<b>2</b>		<b>9</b>		<b>324</b>
Control Volume	Moisture Flux	Runoff Method	Storage	Runoff Data	Reanalysis	Time	Precipitation	Time		
Yearly Surface	Method 1	Linear	Yes	Reanalysis	1	1948-2002	Reanalysis	N/A		
			No		2	1979-2001	GPCP	1979-2001		
		Area						GPCC	1986-1996	
							GPCC TRMM	1998-2002		
							GHCN/CAMS	1948-2002		
							OPI2	1979-2001		
								OPI3	1979-2001	
								CMAF	1979-2001	
								CMAF2	1979-2001	
	<b>Sum Variables</b>	<b>1</b>	<b>3</b>	<b>2</b>	<b>1</b>	<b>2</b>		<b>9</b>		<b>108</b>

Table 5-4: Water Balance Combinations for Larger US

Control Volume	Moisture Flux	Storage	Runoff Data	Runoff Time	Reanalysis	Time	Precipitation	Time	
Monthly Atmospheric	N/A	N/A	N/A	N/A	1	1/1948-12/2002	Reanalysis	N/A	
					2	1/1979-12/2001	GPCP	1/1979-12/2001	
							GPCC	1/1986-12/1996	
							GPCC TRMM	1/1998-12/2002	
							GHCN-CAMS	1/1948-11/2002	
							GPI	1/1986-12/2002	
							TRMM 3A46	1/1998-8/2001	
							TRMM 3B43	1/1998-12/2002	
							SSM/I	7/1987-11/1987	
								1/1988-6/1990	
								1/1992-12/2002	
							OPI2	1/1979-8/2002	
							OPI3	1/1979-8/2002	
							CMAP	1/1979-6/2002	
							CMAP2	1/1979-6/2002	
<b>Sum Variables</b>	<b>1</b>	<b>1</b>	<b>1</b>		<b>2</b>		<b>13</b>		<b>Total</b>
									<b>26</b>
Control Volume	Moisture Flux	Storage	Runoff Data	Runoff Time	Reanalysis	Time	Precipitation	Time	
Yearly Atmospheric	N/A	N/A	N/A	N/A	1	1948-2002	Reanalysis	N/A	
		N/A			2	1979-2001	GPCP	1979-2001	
							GPCC	1986-1996	
							GPCC TRMM	1998-2002	
							GHCN-CAMS	1948-2002	
							GPI	1986-2002	
							TRMM 3A46	1998-2000	
							TRMM 3B43	1998-2002	
							SSM/I	1988-1989	
								1992-2002	
							OPI2	1979-2001	
							OPI3	1979-2001	
							CMAP	1979-2001	
							CMAP2	1979-2001	
<b>Sum Variables</b>	<b>1</b>	<b>1</b>	<b>1</b>		<b>2</b>		<b>13</b>		<b>Total</b>
									<b>26</b>
Control Volume	Moisture Flux	Storage	Runoff Data	Runoff Time	Reanalysis	Time	Precipitation	Time	
Monthly Surface	N/A	Yes	Surface	1/1970-12/1996	1	1/1948-12/2002	Reanalysis	N/A	
		No	Reanalysis	N/A	2	1/1979-12/2001	GPCP	1/1979-12/2001	
		1 mon					GPCC	1/1986-12/1996	
		2 mon					GPCC TRMM	1/1998-12/2002	
		3 mon					GHCN-CAMS	1/1948-11/2002	
		4 mon					GPI	1/1986-12/2002	
							TRMM 3A46	1/1998-8/2001	
							TRMM 3B43	1/1998-12/2002	
							SSM/I	7/1987-11/1987	
								1/1988-6/1990	
								1/1992-12/2002	
							OPI2	1/1979-8/2002	
							OPI3	1/1979-8/2002	
							CMAP	1/1979-6/2002	
							CMAP2	1/1979-6/2002	
<b>Sum Variables (SurfR)</b>	<b>1</b>	<b>6</b>	<b>1</b>		<b>2</b>		<b>10</b>		<b>Total</b>
<b>Sum Variables (ReR)</b>	<b>1</b>	<b>6</b>	<b>1</b>		<b>2</b>		<b>13</b>		<b>156</b>
<b>Sum Variables</b>									<b>276</b>
Control Volume	Moisture Flux	Storage	Runoff Data	Runoff Time	Reanalysis	Time	Precipitation	Time	
Yearly Surface	Method 1	Yes	Surface	1970-1996	1	1948-2002	Reanalysis	N/A	
		No	Reanalysis	N/A	2	1979-2001	GPCP	1979-2001	
							GPCC	1986-1996	
							GPCC TRMM	1998-2002	
							GHCN-CAMS	1948-2002	
							GPI	1986-2002	
							TRMM 3A46	1998-2000	
							TRMM 3B43	1998-2002	
							SSM/I	1988-1989	
								1992-2002	
							OPI2	1979-2001	
							OPI3	1979-2001	
							CMAP	1979-2001	
							CMAP2	1979-2001	
<b>Sum Variables (SurfR)</b>	<b>1</b>	<b>2</b>	<b>1</b>		<b>2</b>		<b>10</b>		<b>40</b>
<b>Sum Variables (ReR)</b>	<b>1</b>	<b>2</b>	<b>1</b>		<b>2</b>		<b>13</b>		<b>52</b>
<b>Sum Variables</b>									<b>92</b>

Table 5-5: Water Balance Combinations for Amazon and Port Velho Basins

Control Volume	Moisture Flux	Storage	Runoff Data	Runoff Time	Reanalysis	Time	Precipitation	Time	
Monthly Atmospheric	N/A	N/A	N/A	N/A	1	1/1948-12/2002	Reanalysis	N/A	
					2	1/1979-12/2001	GPCP	1/1979-12/2001	
							GPCP	1/1986-12/1996	
							GPCP TRMM	1/1998-12/2002	
							GHCN-CAMS	1/1948-11/2002	
							GPI	1/1986-12/2002	
							TRMM 3B43	1/1998-12/2002	
							SSM/I	7/1987-11/1987	
								1/1988-6/1990	
								1/1992-12/2002	
							OPI2	1/1979-8/2002	
							OPI3	1/1979-8/2002	
							CMAP	1/1979-6/2002	
							CMAP2	1/1979-6/2002	
<b>Sum Variables</b>	<b>1</b>	<b>1</b>	<b>1</b>		<b>2</b>		<b>12</b>		<b>Total</b>
									<b>24</b>
Control Volume	Moisture Flux	Storage	Runoff Data	Time	Reanalysis	Time	Precipitation	Time	
Yearly Atmospheric	N/A	N/A	N/A	N/A	1	1948-2002	Reanalysis	N/A	
		N/A			2	1979-2001	GPCP	1979-2001	
							GPCP	1986-1996	
							GPCP TRMM	1998-2002	
							GHCN-CAMS	1948-2002	
							GPI	1986-2002	
							TRMM 3B43	1998-2002	
							SSM/I	1988-1989	
								1992-2002	
							OPI2	1979-2001	
							OPI3	1979-2001	
							CMAP	1979-2001	
							CMAP2	1979-2001	
<b>Sum Variables</b>	<b>1</b>	<b>1</b>	<b>1</b>		<b>2</b>		<b>12</b>		<b>Total</b>
									<b>24</b>
Control Volume	Moisture Flux	Storage	Runoff Data	Runoff Time	Reanalysis	Time	Precipitation	Time	
Monthly Surface	N/A	Yes	Surface	1/1970-12/1996	1	1/1948-12/2002	Reanalysis	N/A	
		No	Reanalysis	N/A	2	1/1979-12/2001	GPCP	1/1979-12/2001	
		1 mon					GPCP	1/1986-12/1996	
		2 mon					GPCP TRMM	1/1998-12/2002	
		3 mon					GHCN-CAMS	1/1948-11/2002	
		4 mon					GPI	1/1986-12/2002	
							TRMM 3B43	1/1998-12/2002	
							SSM/I	7/1987-11/1987	
								1/1988-6/1990	
								1/1992-12/2002	
							OPI2	1/1979-8/2002	
							OPI3	1/1979-8/2002	
							CMAP	1/1979-6/2002	
							CMAP2	1/1979-6/2002	
<b>Sum Variables (SurfR)</b>	<b>1</b>	<b>6</b>	<b>1</b>		<b>2</b>		<b>10</b>		<b>Total</b>
<b>Sum Variables (ReR)</b>	<b>1</b>	<b>6</b>	<b>1</b>		<b>2</b>		<b>12</b>		<b>144</b>
<b>Sum Variables</b>									<b>264</b>
Control Volume	Moisture Flux	Storage	Runoff Data	Runoff Time	Reanalysis	Time	Precipitation	Time	
Yearly Surface	Method 1	Yes	Surface	1970-1996	1	1948-2002	Reanalysis	N/A	
		No	Reanalysis	N/A	2	1979-2001	GPCP	1979-2001	
							GPCP	1986-1996	
							GPCP TRMM	1998-2002	
							GHCN-CAMS	1948-2002	
							GPI	1986-2002	
							TRMM 3B43	1998-2002	
							SSM/I	1988-1989	
								1992-2002	
							OPI2	1979-2001	
							OPI3	1979-2001	
							CMAP	1979-2001	
							CMAP2	1979-2001	
<b>Sum Variables (SurfR)</b>	<b>1</b>	<b>2</b>	<b>1</b>		<b>2</b>		<b>10</b>		<b>40</b>
<b>Sum Variables (ReR)</b>	<b>1</b>	<b>2</b>	<b>1</b>		<b>2</b>		<b>12</b>		<b>48</b>
<b>Sum Variables</b>									<b>88</b>

Table 5-6: Water Balance Combinations for Arkansas Basins

Control Volume	Moisture Flux	Storage	Runoff Data	Runoff Time	Reanalysis	Time	Precipitation	Time	
Monthly Atmospheric	N/A	N/A	N/A	N/A	1	1/1948-12/2002	Reanalysis	N/A	
					2	1/1948-12/2001	GPCP	1/1979-12/2001	
							GPCP	1/1986-12/1996	
							GPCP TRMM	1/1998-12/2002	
							GHCN-CAMS	1/1948-11/2002	
							SSM/I	7/1987-11/1987	
								1/1988-6/1990	
								1/1992-12/2002	
							OPI2	1/1979-8/2002	
							OPI3	1/1979-8/2002	
							CMAP	1/1979-6/2002	
							CMAP2	1/1979-6/2002	
<b>Sum Variables</b>	<b>1</b>	<b>1</b>	<b>1</b>		<b>2</b>		<b>10</b>		<b>Total</b>
									<b>20</b>
Control Volume	Moisture Flux	Storage	Runoff Data	Runoff Time	Reanalysis	Time	Precipitation	Time	
Yearly Atmospheric	N/A	N/A	N/A	N/A	1	1948-2002	Reanalysis	N/A	
		N/A			2	1948-2001	GPCP	1979-2001	
							GPCP	1986-1996	
							GPCP TRMM	1998-2002	
							GHCN-CAMS	1948-2002	
							SSM/I	1988-1989	
								1992-2002	
							OPI2	1979-2001	
							OPI3	1979-2001	
							CMAP	1979-2001	
							CMAP2	1979-2001	
<b>Sum Variables</b>	<b>1</b>	<b>1</b>	<b>1</b>		<b>2</b>		<b>10</b>		<b>Total</b>
									<b>20</b>
Control Volume	Moisture Flux	Storage	Runoff Data	Runoff Time	Reanalysis	Time	Precipitation	Time	
Monthly Surface	N/A	Yes	Surface	1/1948-12/1996	1	1/1948-12/2002	Reanalysis	N/A	
		No	Reanalysis	N/A	2	1/1948-12/2001	GPCP	1/1979-12/2001	
		1 mon					GPCP	1/1986-12/1996	
		2 mon					GPCP TRMM	1/1998-12/2002	
		3 mon					GHCN-CAMS	1/1948-11/2002	
		4 mon					SSM/I	7/1987-11/1987	
								1/1988-6/1990	
								1/1992-12/2002	
							OPI2	1/1979-8/2002	
							OPI3	1/1979-8/2002	
							CMAP	1/1979-6/2002	
							CMAP2	1/1979-6/2002	
<b>Sum Variables (SurfR)</b>	<b>1</b>	<b>6</b>	<b>1</b>		<b>2</b>		<b>9</b>		<b>Total</b>
									<b>108</b>
<b>Sum Variables (ReR)</b>	<b>1</b>	<b>6</b>	<b>1</b>		<b>2</b>		<b>10</b>		<b>Total</b>
									<b>120</b>
<b>Sum Variables</b>									<b>228</b>
Control Volume	Moisture Flux	Storage	Runoff Data	Runoff Time	Reanalysis	Time	Precipitation	Time	
Yearly Surface	Method 1	Yes	Surface	1948-1996	1	1948-2002	Reanalysis	N/A	
		No	Reanalysis	N/A	2	1948-2001	GPCP	1979-2001	
							GPCP	1986-1996	
							GPCP TRMM	1998-2002	
							GHCN-CAMS	1948-2002	
							SSM/I	1988-1989	
								1992-2002	
							OPI2	1979-2001	
							OPI3	1979-2001	
							CMAP	1979-2001	
							CMAP2	1979-2001	
<b>Sum Variables (SurfR)</b>	<b>1</b>	<b>2</b>	<b>1</b>		<b>2</b>		<b>9</b>		<b>Total</b>
									<b>36</b>
<b>Sum Variables (ReR)</b>	<b>1</b>	<b>2</b>	<b>1</b>		<b>2</b>		<b>10</b>		<b>Total</b>
									<b>40</b>
<b>Sum Variables</b>									<b>88</b>

Table 5-7: Water Balance Combinations for Mississippi Basins

## 5.1. East Brazil

The East Brazil control volume was selected for this study as one of three small control volumes over the tropical portion of the South American continent. This control volume represents the more populated and developed portions of Brazil, and thus likely is a better observed region. Figure 5-1 shows the yearly Reanalysis-1 water balances for Eastern Brazil. In the atmospheric balance precipitation exceeds evaporation for much of the second half of the balance. This should result in a slightly positive moisture flux, unfortunately a very negative moisture flux is found. The moisture flux does increase slightly from 1950-1955 and in the later half of the balance which is consistent with the P-E difference. The result in general is a very biased residual at the beginning and the end of the balance of similar magnitude to the other water balance variables. In the surface balance residual is consistently negative. Similarities in signals between the P-E and storage change exist, but the methodology fails to pick up much of the runoff. A balance is achieved in the first half on the time period, but not in the second half. The magnitudes of the errors are high for these water balances.

To see if the second half of these balances can be improved we refer to Figure 5-2 which shows the Reanalysis-2 yearly balances. For the atmospheric balance there are few differences from the Reanalysis-1. For the surface balance runoff has a little more amplitude here from year to year however this does not lead to much improvement. While the entire time series is unbalanced in the Reanalysis-1 a balance is achieved in the middle portion of the Reanalysis-2 surface balance. Both the absolute error and bias increase from the Reanalysis-1 to Reanalysis-2 for both the atmospheric and surface water balances. Virtually no correlation exists between elements of the atmospheric and surface water balances from either dataset.

Figure 5-3 displays alternative precipitation datasets on an annual basis for the Eastern Brazil control volume, which yields some interesting results. Up until 1979 the Reanalysis-1 seems to be consistent with the alternative sources of precipitation, however, in 1979 the Reanalysis-1 and Reanalysis-2 jump and are significantly higher for the rest of the time period. This behavior was observed in the atmospheric balance and compensated for in the moisture flux, but in the surface balance it isn't compensated for until the Reanalysis-2. The alternative precipitation datasets are very consistent with each other especially in the large dataset category. The larger datasets are probably the correct ones; however these datasets will not necessarily be able to close this balance.

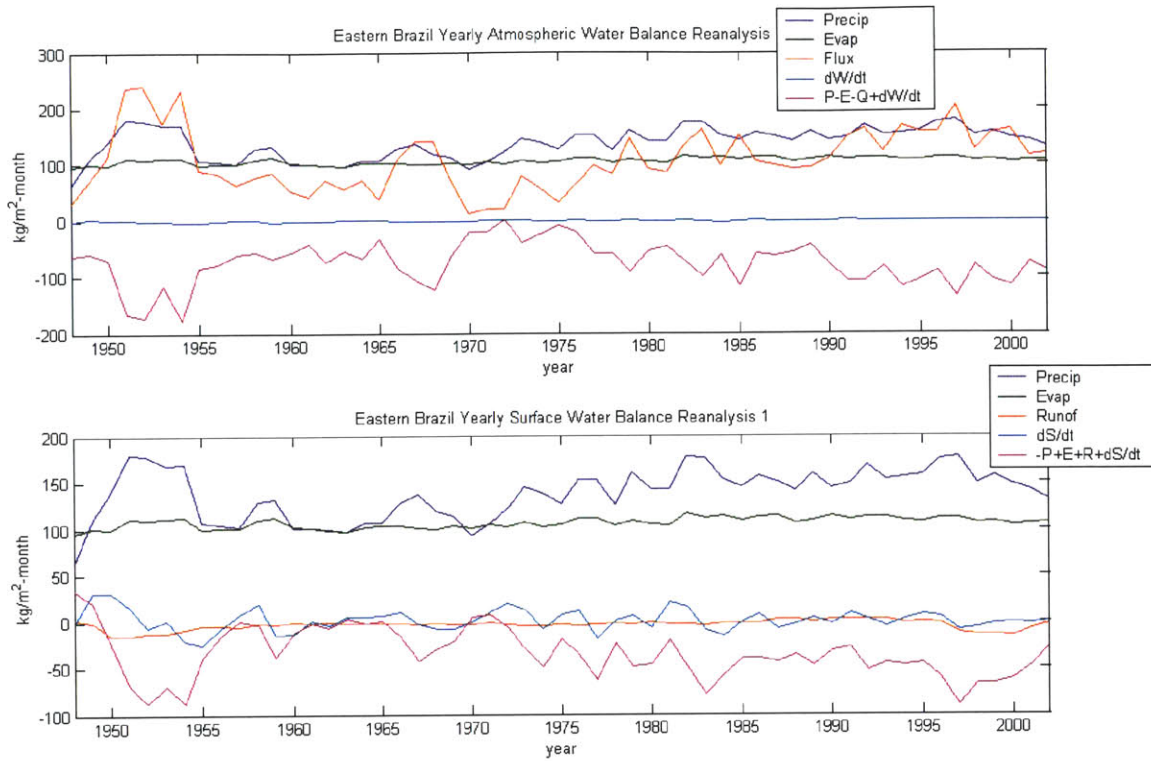


Figure 5-1: Eastern Brazil Reanalysis-1 Yearly Atmospheric and Surface Balances

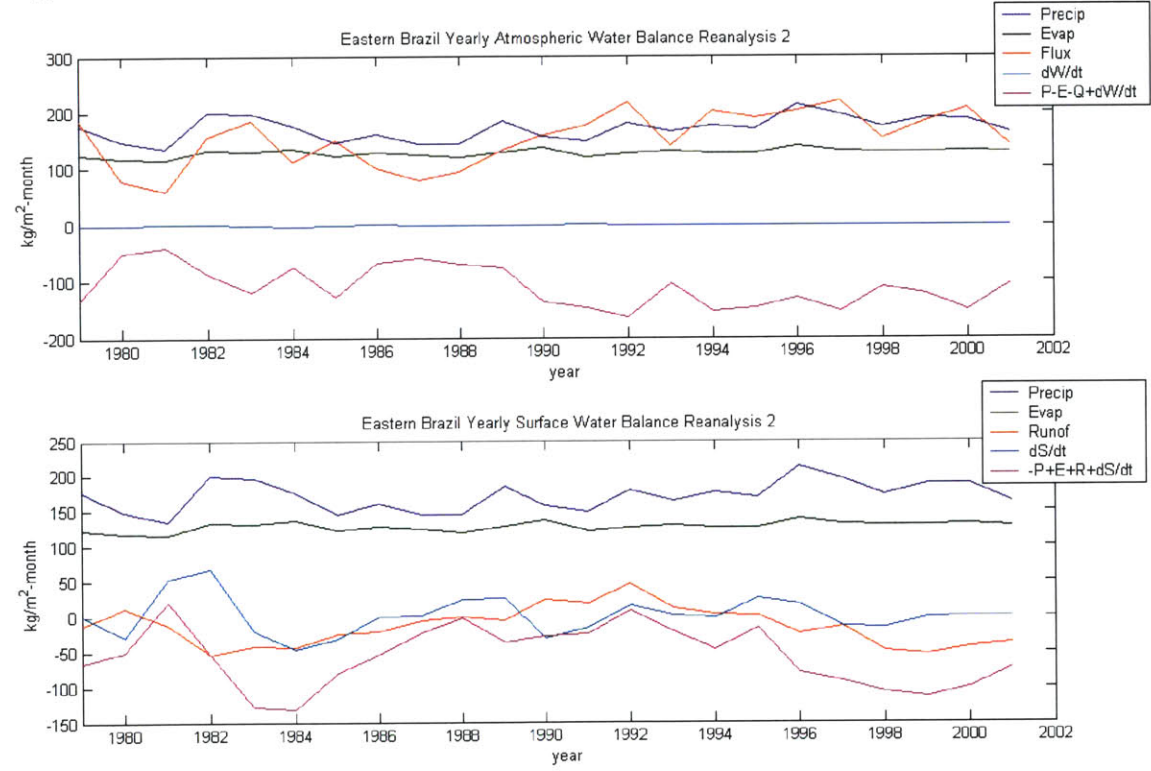


Figure 5-2: Eastern Brazil Reanalysis-2 Yearly Atmospheric and Surface Balances

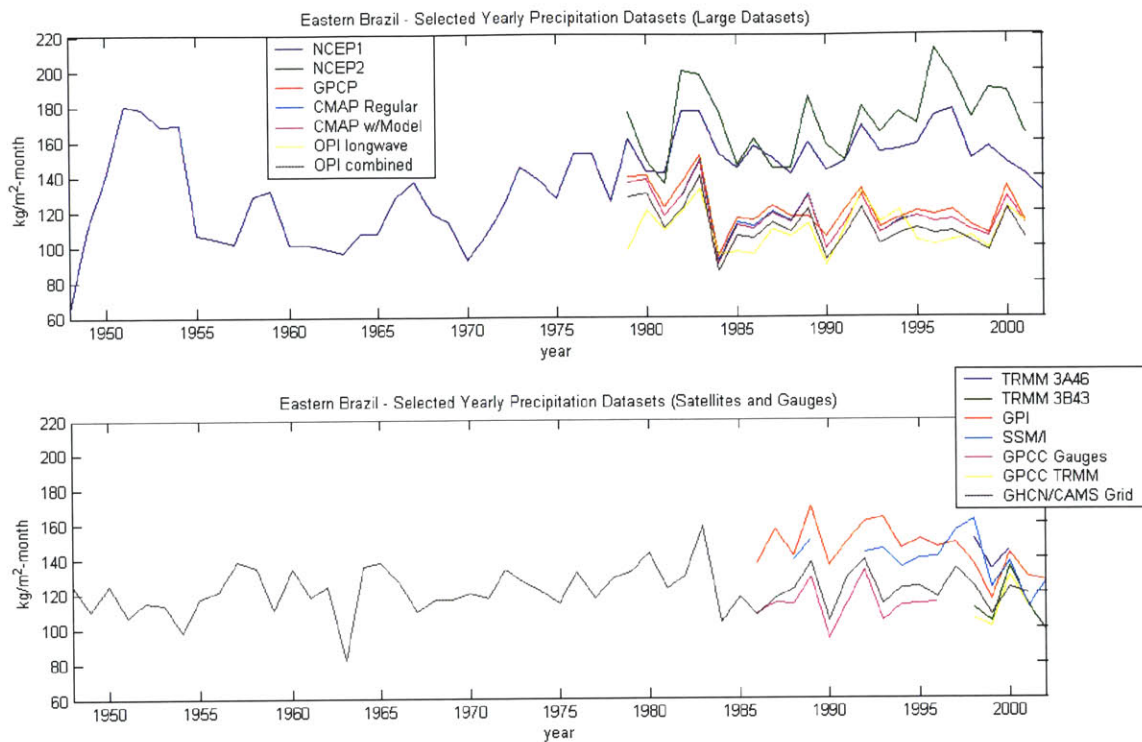


Figure 5-3: Eastern Brazil Different Sources of Yearly Precipitation

Turning to the monthly balance, we observe the Reanalysis-1 in Figure 5-4. For the atmospheric monthly balance we see a strong correlation between the two elements of the water balance. Moisture flux is clearly overestimated as was seen on the yearly time scale which results in a slope above one. But on a monthly basis there is a tight correlation, and the seasonal cycle can be seen well in the example plot. The correlation coefficient for this balance is .8927, while the slope of the regression is 1.83. For the surface balance we also see a good correlation between the two sides of the balance, however the runoff isn't well aligned with the precipitation which can be seen in the bottom right panel. Despite the fact that change in surface storage is included in the balance, there appears to be a lag time in the runoff. Unfortunately, lag time is only considered without storage in this study, because change in storage should account for any lag time. In the Reanalysis-1 for this control volume a one or two month lag time may be very useful. The correlation coefficient here is .4511 with a slope of .6511.

Turning to the Reanalysis-2 in Figure 5-5, we see a very similar atmospheric water balance. The slope of the regression is reduced to 1.57. The lag time problem is completely removed in the example plot of the Reanalysis-2 surface balance, however the balance here is deceptive as can be seen by the scatter plot. On the yearly basis for this control volume we observed a very good balance over the early 1990's. The monthly balance is very good here too, and the seasonal cycle is very well alligned. However, for other portions of the time series, there appears to be significant errors, so by our ranking system given in Appendix A, the Reanalysis-2 only provides for slight improvements in the surface water balance from the Reanalysis-1. The correlation coefficient for the Reanalysis-2, .4349, is almost exactly the same as in the Reanalysis-1.

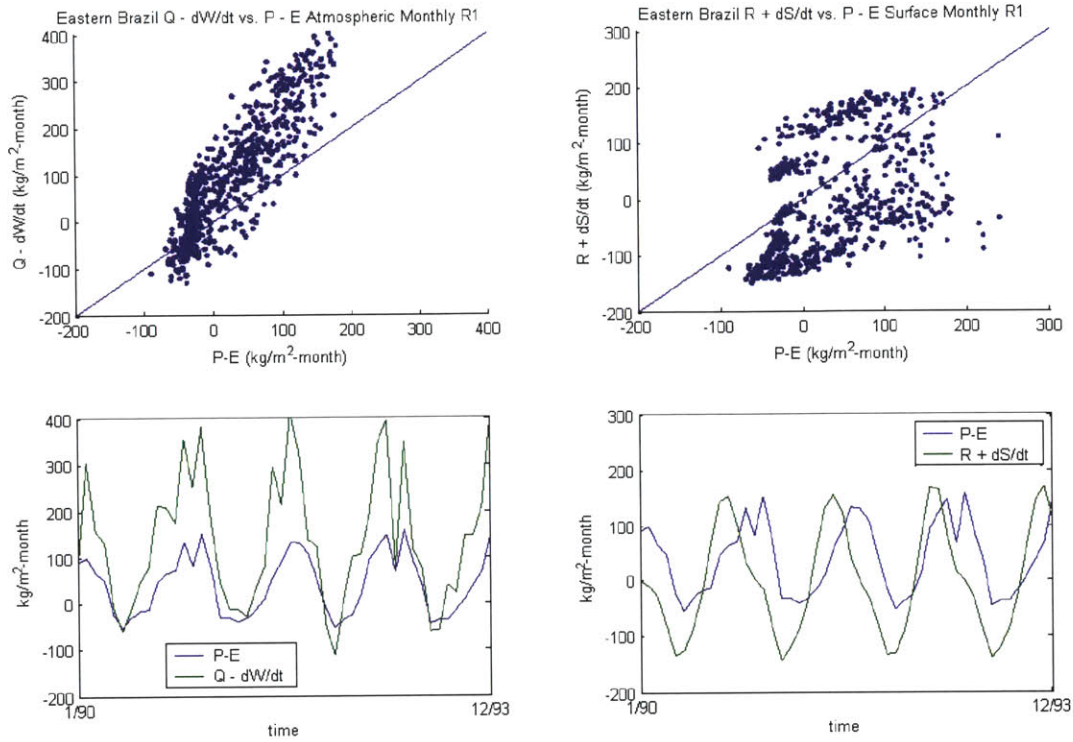


Figure 5-4: Eastern Brazil Reanalysis-1 Monthly Atmospheric and Surface Balances

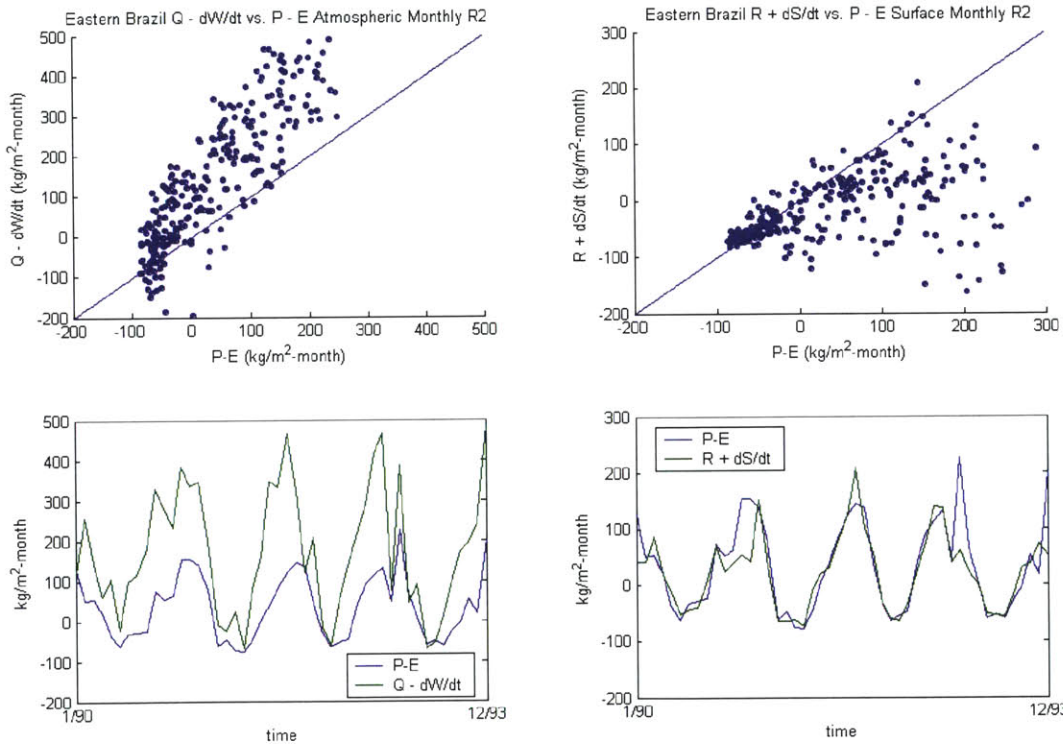


Figure 5-5: Eastern Brazil Reanalysis-2 Monthly Atmospheric and Surface Balances



Trying to improve on the Eastern Brazil water balance we refer to the ranking system devised and presented in Appendix A. Table 5-1 indicated that for this control volume, for the atmospheric yearly water balance, we can vary flux methodology (2), reanalysis (2), and precipitation dataset (13). For the yearly surface balance we can vary runoff methodology (2), reanalysis (2), storage method (2), and precipitation dataset (13).

For the East Brazil atmospheric yearly balance the default Reanalysis-1 and Reanalysis-2 balances are among the top 3 of all the balances evaluated. The Reanalysis-1 provides the best balance which is very encouraging in the context of use for climate studies. However, absolute errors and biases are high. Inputting the GPCP precipitation dataset into the later portion of the Reanalysis-1 results in a comparable balance to that of the reanalysis (shown in Figure 5-6). The properties of this balance are very similar to the default reanalysis balances. Of special note here is the TRMM 3B43 dataset which under certain conditions ranks 5<sup>th</sup>, but the temporal resolution is only 5 years which isn't significant on the yearly basis.

In the yearly surface balance the reanalysis balances are not as good. We replace the precipitation with GPI precipitation, change the runoff method to area, and eliminate the storage term. This balance is ranked second. However the best balance contains GPCC TRMM, and contains only four years of data. When the change in storage is removed and the precipitation replaced the balance improves a lot especially in its second half. The correlation coefficient is .6303 and the bias is only 2 kg/m<sup>2</sup>-month, indicating almost all water is accounted for over the duration of the balance.

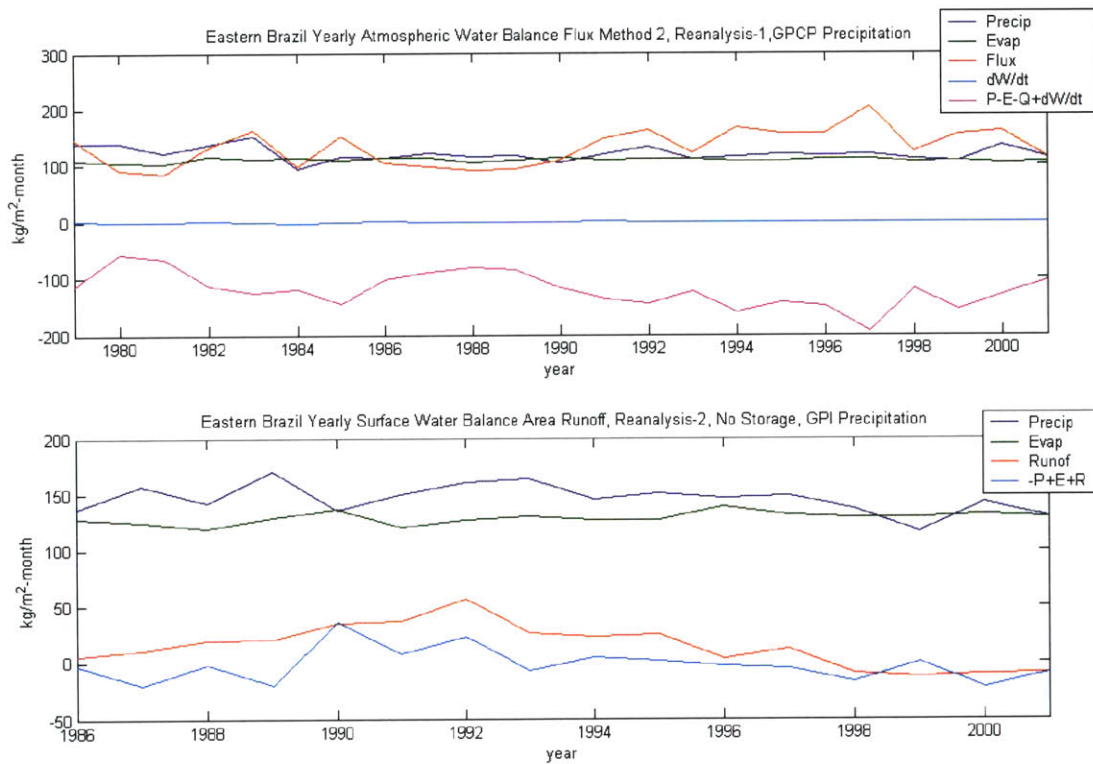


Figure 5-6: Eastern Brazil Selected Yearly Water Balances

For the monthly time scales the improved balances are shown in Figure 5-7. There are as many possibilities here as there were for the yearly time scale (52) for the atmospheric balance. Again for the atmospheric balance, not much improvement could be achieved from the reanalysis balance. The highest ranking balance is the TRMM 3A46 precipitation applied to the Reanalysis-1 with moisture flux Method 2. In general the Reanalysis-1 balances with various precipitation datasets tend to be better balanced than the Reanalysis-2 for the East Brazil control volume. There is little improvement though, and by this dataset moisture flux is too high for all data points.

For the monthly surface balance there are a great number of possible water balances which can be applied and improvement can be found. Alternatives are found in runoff method (3), reanalysis (2), storage method (6), and precipitation (13). Based on the evaluation criteria the GPI precipitation is again a good fit for the water balance with area runoff, and remaining default conditions. The balance ranks well among all evaluation criteria and the correlation factor is improved to nearly .8, almost double what it was before. Possible explanations include the elimination of data points early in the time series, and just a greater effectiveness of the GPI, combined with a better runoff model for this region. GPI precipitation is similar to the reanalysis, but slightly lower for this control volume as shown in Figure 5-3.

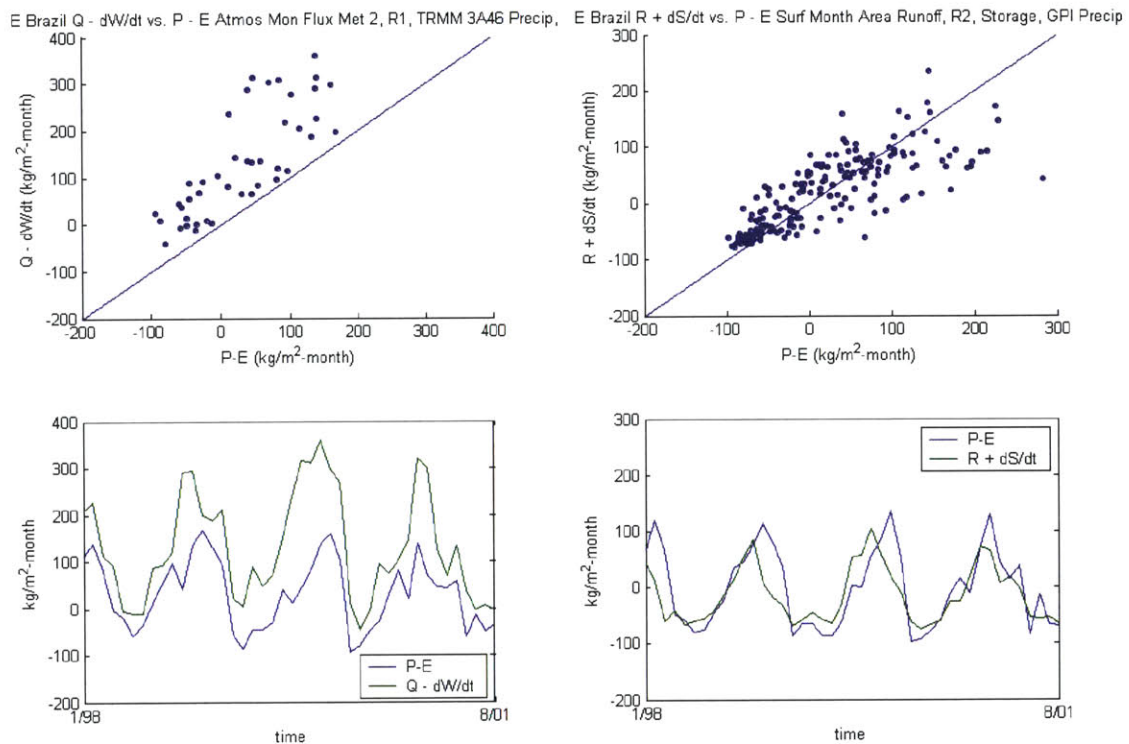


Figure 5-7: Eastern Brazil Selected Monthly Water Balances

## 5.2. Larger Rondonia

The Larger Rondonia control volume was selected for this work to compare to the smaller 7.5 degree box control volumes in the Amazon rain forest region, to see if there were substantial advantages to using a control volume with six moisture flux points on a side, and four times the area. The size of this water balance should allow for the anomalous flux and other fields to have a smaller affect on the balance, and should in general improve the balance because more data is being averaged. We have shown how this control volume clearly improves the ability to obtain a water balance in this region in the previous chapter, and we will now study it more closely.

Figure 5-8 shows the yearly water balance for the Reanalysis-1. The atmospheric water balance is very good and the results are very interesting. There is a marked increase in the magnitude of the precipitation and a corresponding increase in the magnitude of moisture flux over the second half of the water balance. The balance term dips slightly here with this increase. A similar situation was observed in the Eastern Brazil balance, and also by Chen (2001), as discussed in Chapter 1, whose control volume was near the same size and in the same area as this one. The atmospheric water balance closes well, with small absolute and bias error on the magnitude of  $10 \text{ kg/m}^2\text{-month}$  and a correlation of .9740. There is solid evidence of a change in the water balance over the Amazon Rain Forest here. We will continue to look for this pattern in other water balances in this geographic region. In the Reanalysis-1 surface balance the change in storage signal is weak and steady and the runoff is nearly flat. There is a -.2362 correlation in this balance with an absolute error and bias of about  $75 \text{ kg/m}^2\text{-month}$ .

In the Reanalysis-2 shown in Figure 5-9 the yearly atmospheric balance again is closed very well, though there are some differences in the data especially at the end of the time series. There are very small differences in evaluation criteria between reanalyzes, with the surface balance continuing to be a failure. The runoff is routed in the wrong direction and ends up negative, and the storage change follows a very regular pattern every few years. The correlation coefficient is at least above zero as positive changes in storage correspond to precipitation increases.

A look at the annual precipitation (Figure 5-10) for this control volume is very relevant, as there is a marked increase in precipitation in the second half of the Reanalysis-1 water balance and we wish to see if alternative precipitation datasets agree. The same trend is observed as in Eastern Brazil, the precipitation increases in the Reanalysis-1 over the second half of the 55 year period, but the alternative precipitation values are consistent with the reanalysis precipitation prior to this time. The Reanalysis-2 seems to improve the precipitation to make it more consistent with alternative sources.

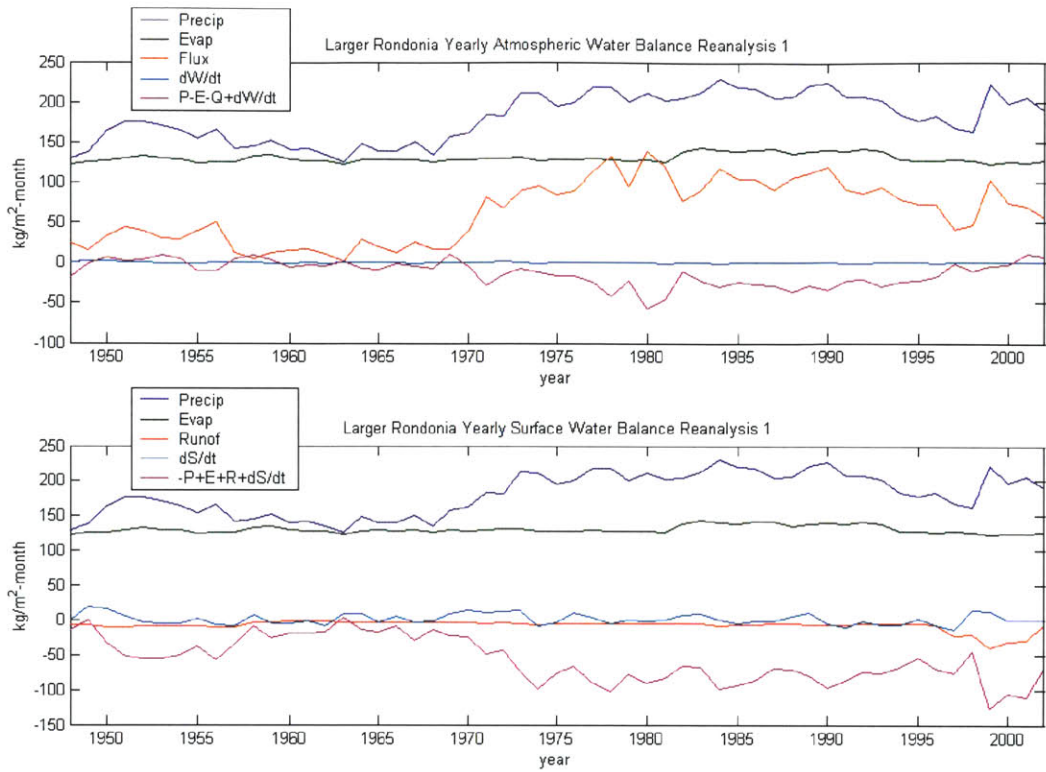


Figure 5-8: Larger Rondonia Reanalysis-1 Yearly Atmospheric and Surface Balances

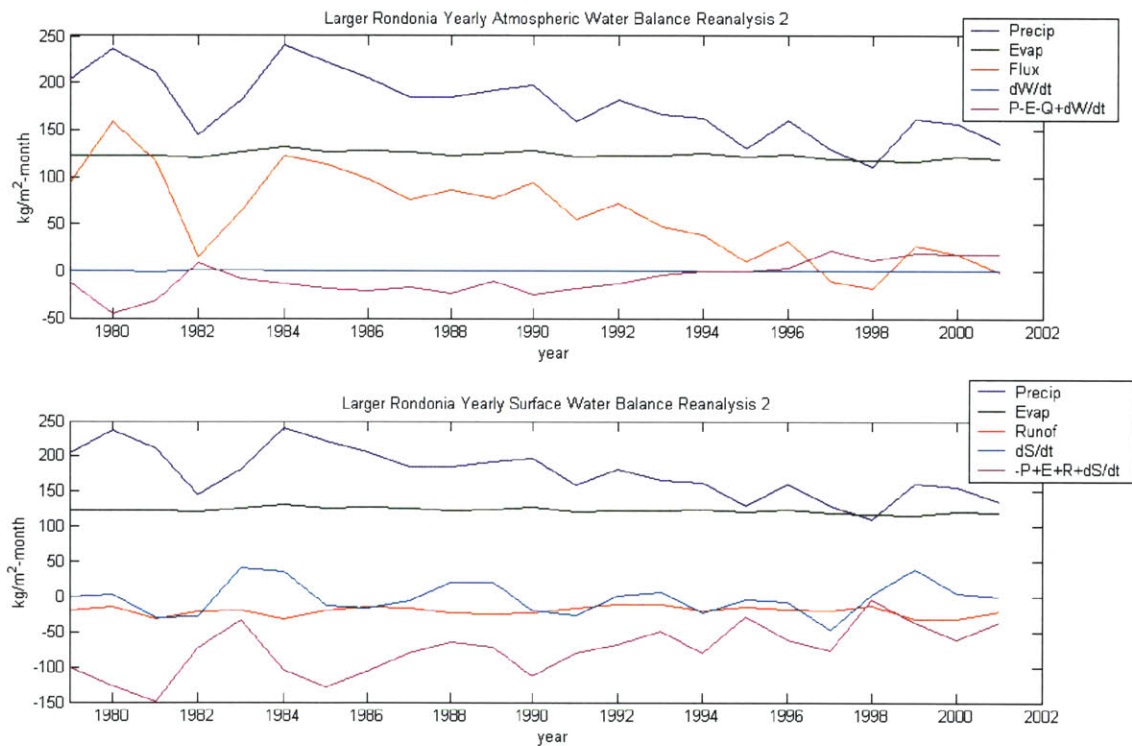


Figure 5-9: Larger Rondonia Reanalysis-2 Yearly Atmospheric and Surface Balances

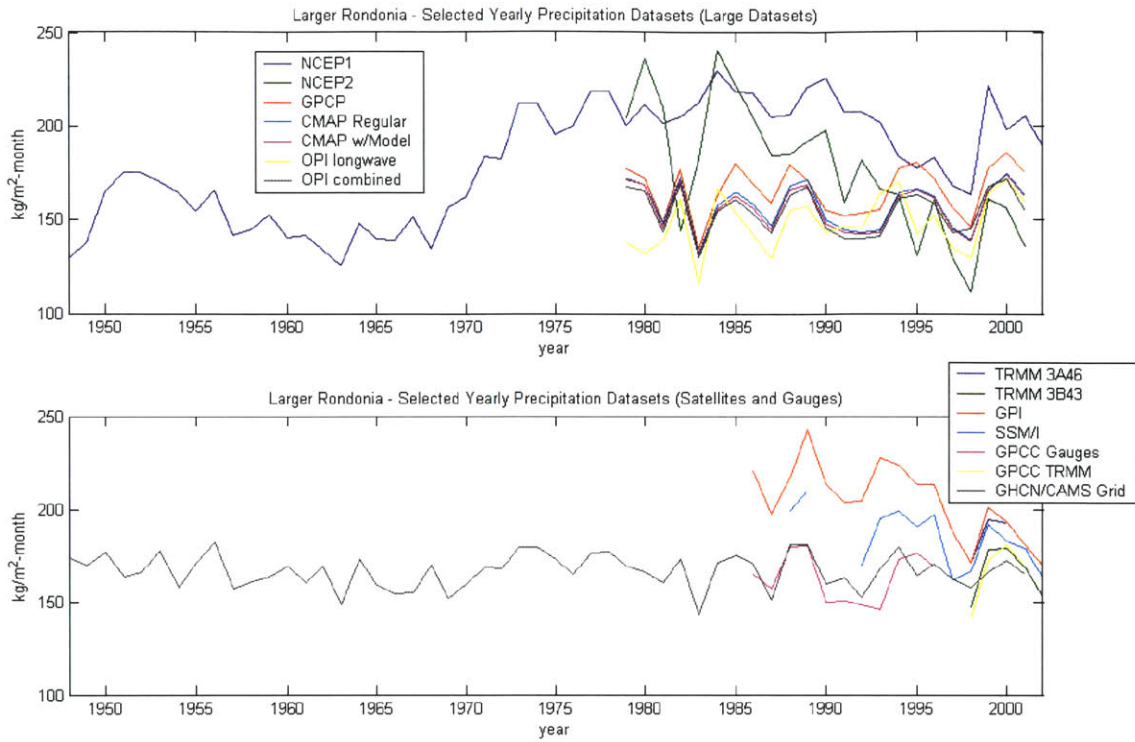


Figure 5-10: Larger Rondonia Different Sources of Yearly Precipitation

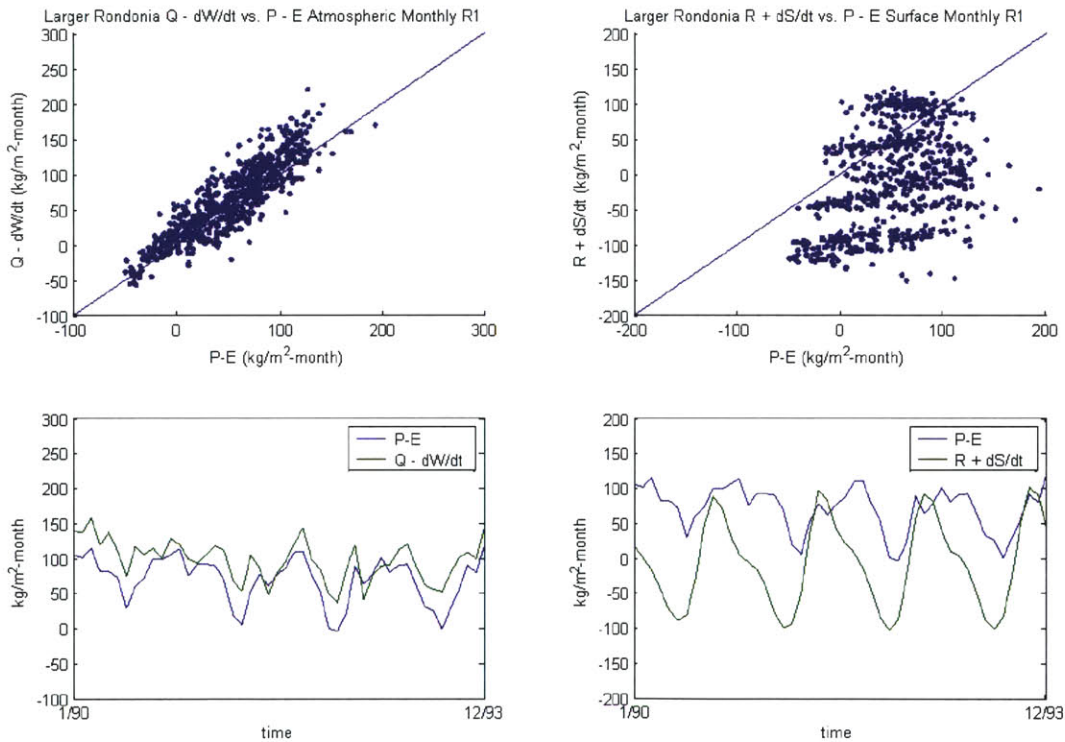


Figure 5-11: Larger Rondonia Reanalysis-1 Monthly Atmospheric and Surface Balances

Figure 5-11 shows the monthly balances for Larger Rondonia for the Reanalysis-1. The results look similar to those obtained in the Eastern Brazil control volume. Although these regions have similar hydrological properties it should be noted that they don't overlap (though they do share a partial boundary). The slope of the regression of the atmospheric balance is much improved (1.0332), almost exactly aligned with one. For the surface balance a very regular change in change in storage cycle seems to dominate the balance and it looks as if the runoff is missing or negative when it should be positive, as was sensed in the yearly results. The seasonal cycle and correlation looks poor for the surface balance. The surface balance has a correlation slope of .59 and a correlation coefficient of .3743.

The Reanalysis-2 is considered under the same circumstances in Figure 5-12. The atmospheric balance looks just as good as the Reanalysis-1 based on the evaluation criteria. The surface balance looks to be improved in terms of seasonal cycle. However the bias and absolute errors increase in this case because the negative runoff forces the runoff and change in storage term in the wrong direction. For this control volume the linear runoff method clearly fails to capture even the correct direction of the runoff as the change in storage is responsible for any type of correlation present (coefficient .5473). It is encouraging that these cycles do line up fairly well even with a complete failure in runoff. Perhaps a more sophisticated runoff routing system and better data could result in both a good surface balance for this control volume.

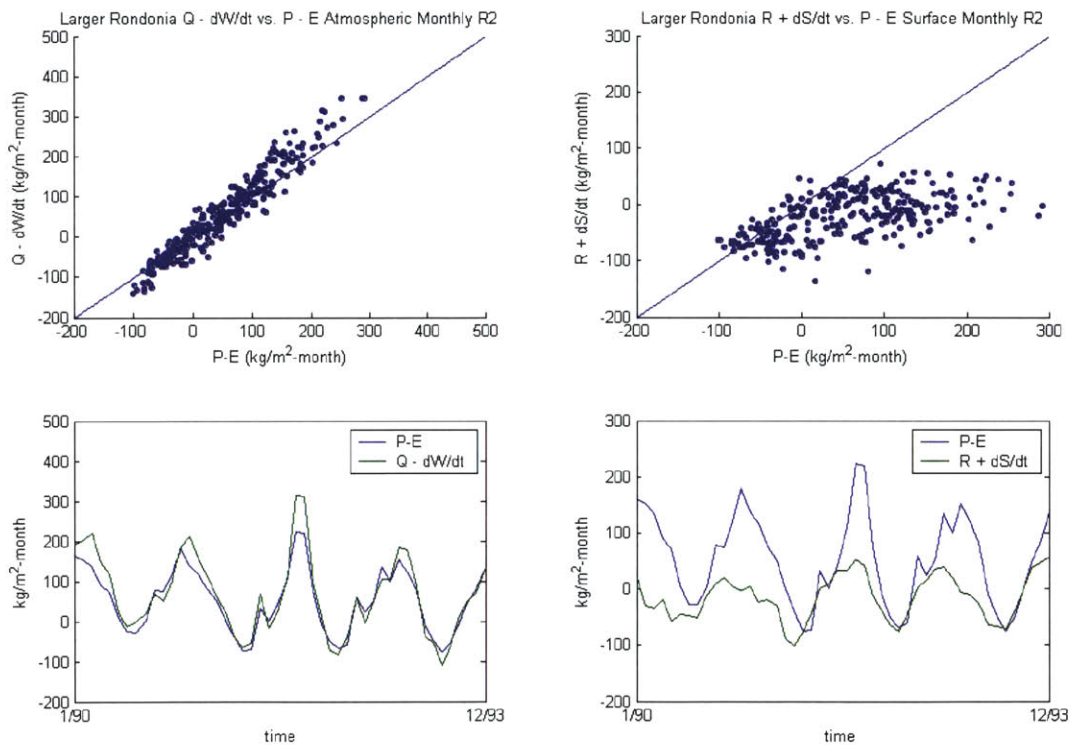


Figure 5-12: Larger Rondonia Reanalysis-2 Monthly Atmospheric and Surface Balances

The amount of scenarios for the Larger Rondonia control volume balances are the same as those for the Eastern Brazil control volume in Table 5-1-1, providing a large amount of possibilities for

improvement. As was the case with the previous control volume, it is difficult to improve on any of the atmospheric balances because of the good job the Reanalysis does fitting its own precipitation to the moisture flux. On the yearly basis the balance is slightly improved by changing the moisture flux methodology to 1, or changing the precipitation dataset to TRMM 3A46. However, in this case we will use an alternative precipitation dataset, the GPI. This precipitation dataset is high like that provided by the reanalysis thus it helps make the atmospheric balance better, as shown in Figure 5-13. To improve the yearly surface balance, the evaluation criteria are used and the best balance is arrived at by changing to the point runoff method, applied to the Reanalysis-1, and using GPCC presentation. This provides an absolute error and bias of 34.63 kg/m<sup>2</sup>-month, or in other words approximately the difference between the precipitation and evaporation. The application seems to simply lower the precipitation and nullify the negative pull of the runoff. Several scenarios with the GPCC precipitation work well for this control volume, likely because this precipitation provides for the lowest bias. It is of special note though that gauges may underestimate precipitation in this control volume (a general note about gauges). What we really need here is better surface flux data (especially runoff).

In Figure 5-14 improvements are made to the monthly water balances. There aren't any improvements for the atmospheric balance. For the Larger Rondonia control volume, the atmospheric water balance can be deemed a success, at least from the reanalysis standpoint. For the monthly surface balance we again turn to the best water balance as evaluated by the criteria in Appendix A. This is achieved with point runoff, applied to Reanalysis-2, and using the OPI2 precipitation dataset. A correlation coefficient of .4989 is achieved as the seasonal precipitation cycle is lowered and the runoff/change in storage cycle is heightened. The balance is still fairly poor with a regression slope of .3 and a bias of 34.8 kg/m<sup>2</sup>-month, similar to that of the yearly case. A better method of obtaining runoff flux out of this control volume is necessary in order to compute the surface water balance, and such inconsistencies would make it impossible to conduct a total water balance for this control volume.

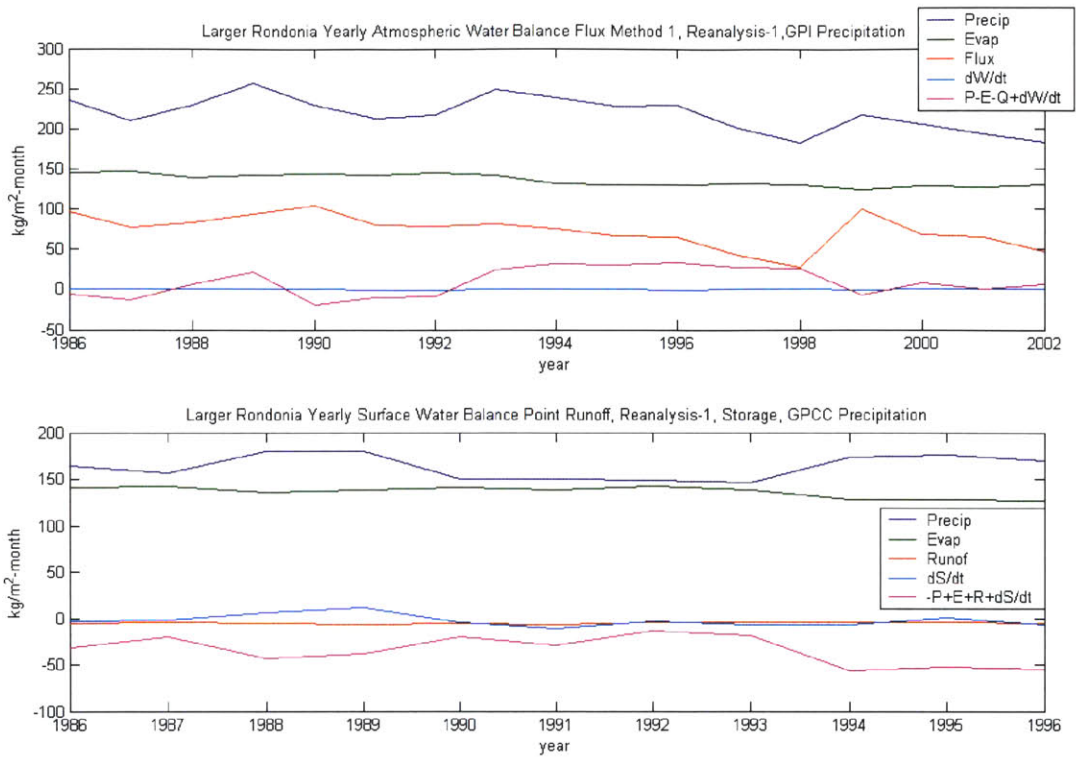


Figure 5-13: Larger Rondonia Selected Yearly Water Balances

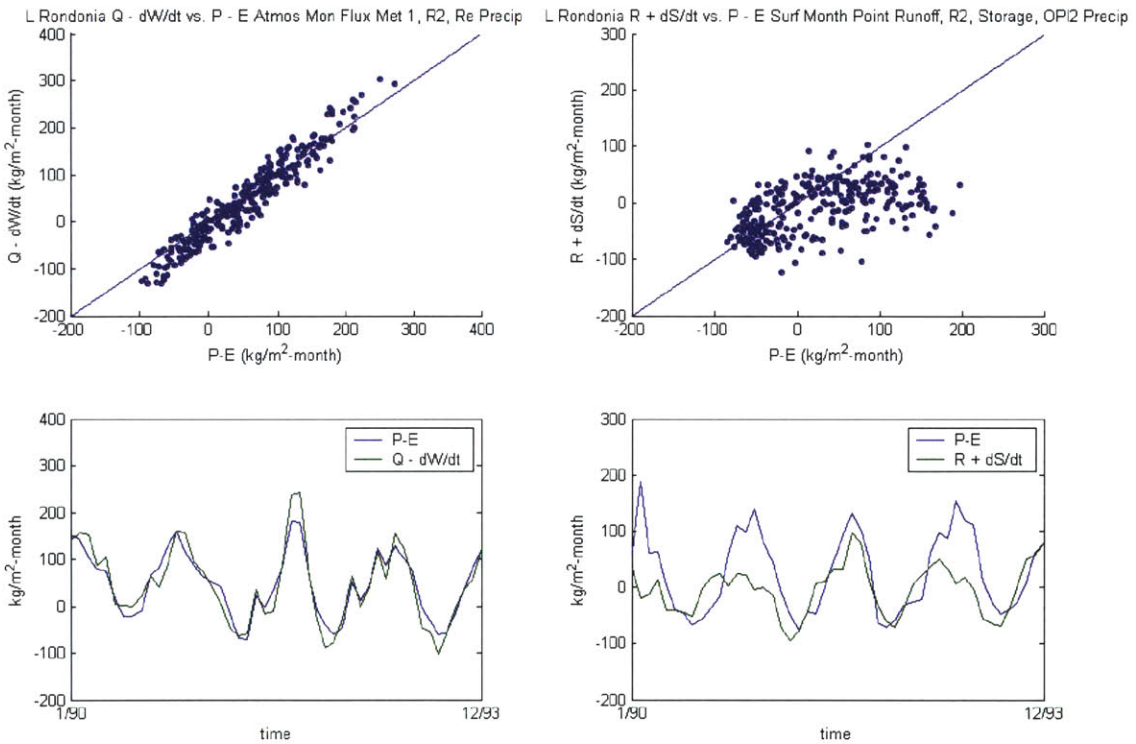


Figure 5-14: Larger Rondonia Selected Monthly Water Balances



### 5.3. Larger United States

In this section we continue to look at a large control volume, however we switch geographic regions to the United States. This region was selected as an alternative large control volume over a supposedly better observed area than the Larger Rondonia control volume. It is also a change from the tropical Larger Rondonia area to the varying climate features of the drier Western United States.

Figure 5-15 shows the yearly water balances obtained from the Reanalysis-1. The problems are almost too numerous to describe. The evaporation is higher than the precipitation over land, which is a highly unlikely phenomenon for a land mass of this size. Such a phenomenon corresponds to a slightly negative moisture flux, however the moisture flux is found to be of about the same magnitude as the precipitation and evaporation quantities. The runoff provided by the Reanalysis-1 is again near zero without much of a signal and the change in surface storage fluctuates wildly dictating the behavior of the balance term. The correlation coefficients for both balances are near zero. In Figure 5-16, the Reanalysis-2 yearly balances are presented. The properties of the water balance are just about the same, and by the evaluation criteria the balance actually gets worse. For the surface balance the correlation coefficient improves, but the absolute error increases. Bias is relatively low for the surface balance, because storage change over the long run is zero, and runoff is non-existent, therefore the bias is just the precipitation – evaporation term.

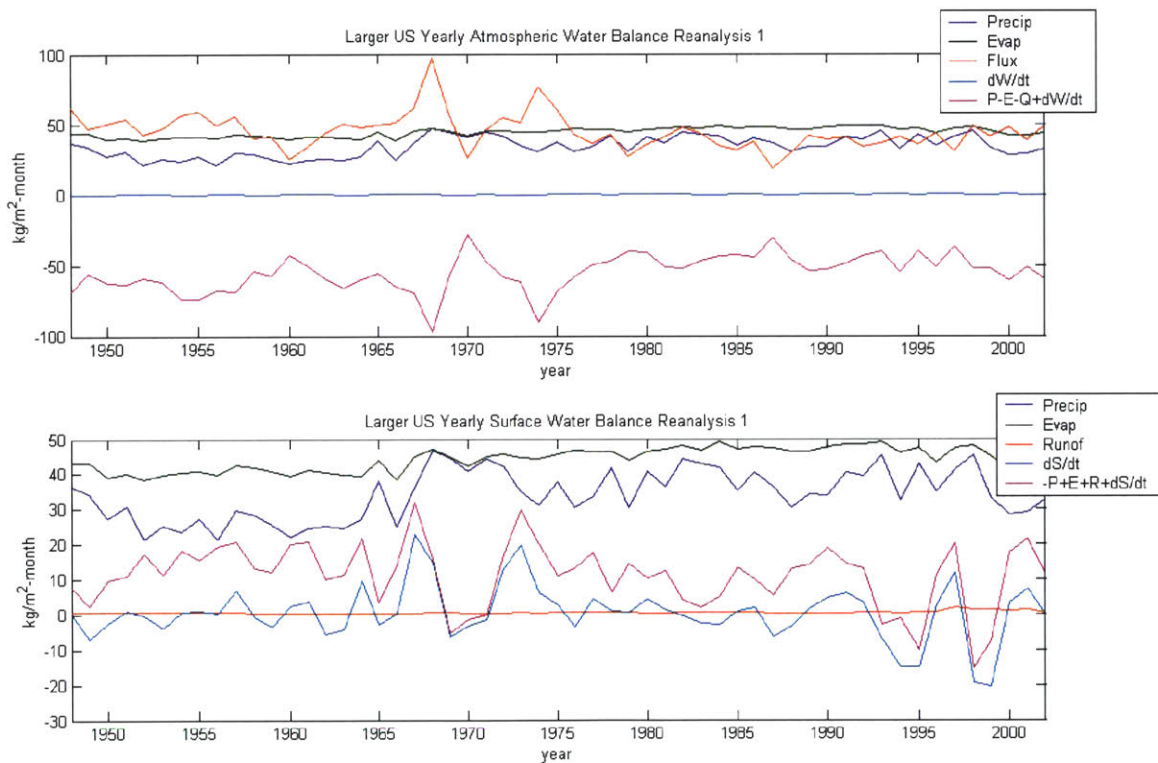


Figure 5-15: Larger US Reanalysis-1 Yearly Atmospheric and Surface Balances

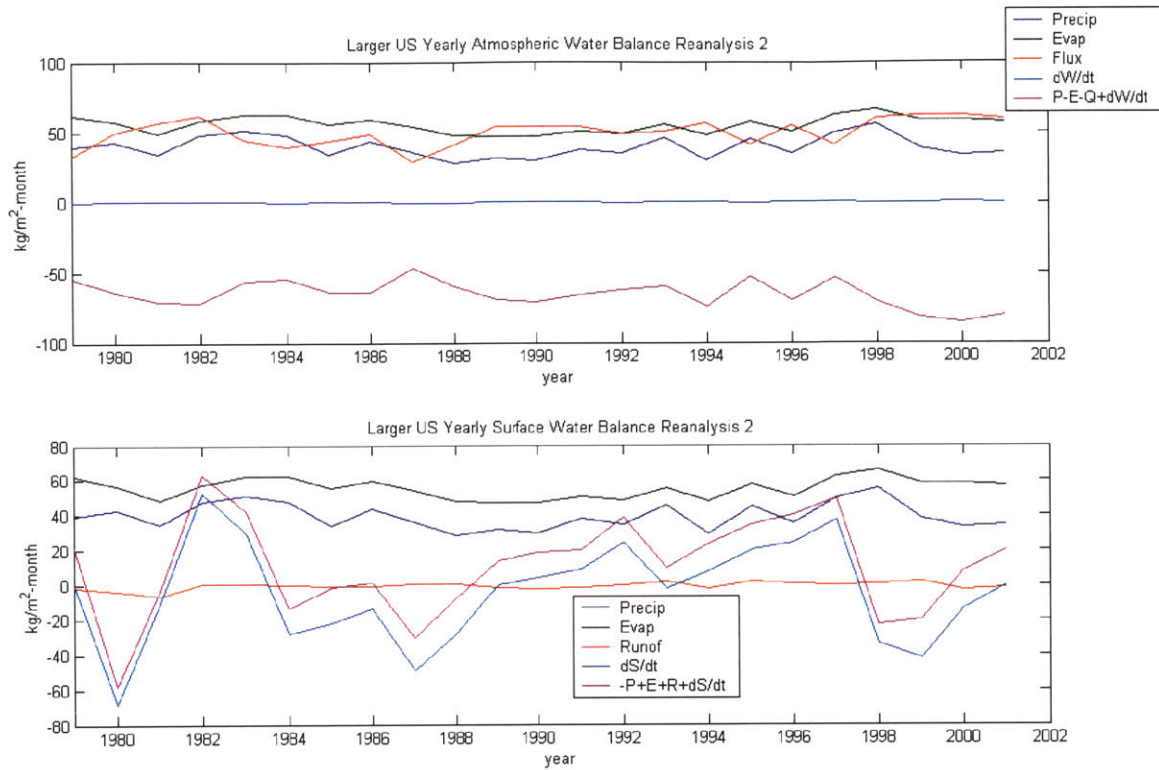


Figure 5-16: Larger US Reanalysis-2 Yearly Atmospheric and Surface Balances

In Figure 5-17 the annual precipitation time series are shown for the Larger United States control volume. The precipitation is largely consistent and appears to be a little higher in the second half of the entire time interval. This is encouraging because it appears by the water balances that the precipitation is too low, as it should be at least as large as the evaporation.

In Figure 5-18 the monthly water balances are considered for the Reanalysis-1. The seasonal cycle is not picked up at all, and instead there is a negative correlation between the side flux and the bottom flux of the atmospheric balance. The slope of the regression is  $-1.2973$  and the correlation coefficient is  $-.4539$ . For the surface balance the seasonal cycle is picked up, but likely as a result of the change in storage cycle. We observed in the yearly case that the runoff term is nearly non-existent for the Reanalysis-1. There is a strong correlation of this cycle with a correlation coefficient of  $.7673$ ; however the slope of the linear regression of this relationship is about 3. Both the moisture flux and runoff seasonal cycle are much stronger than the  $P - E$  seasonal cycle. In Figure 5-19 we see much of the same for the atmospheric balance in the Reanalysis-2. The balances are equally bad as can be discerned by the complete lack of correlation of the seasonal cycles. The surface balance however is greatly improved. Although there is a large bias, the signals are much better aligned. These observations are consistent with the results of the evaluation for this control volume in Appendix A which classifies the default monthly surface Reanalysis-2 balance as one of the best, with a correlation coefficient of  $.7616$  and an absolute error of  $21.90 \text{ kg/m}^2\text{-month}$  compared to  $51.27 \text{ kg/m}^2\text{-month}$  for the Reanalysis-1.

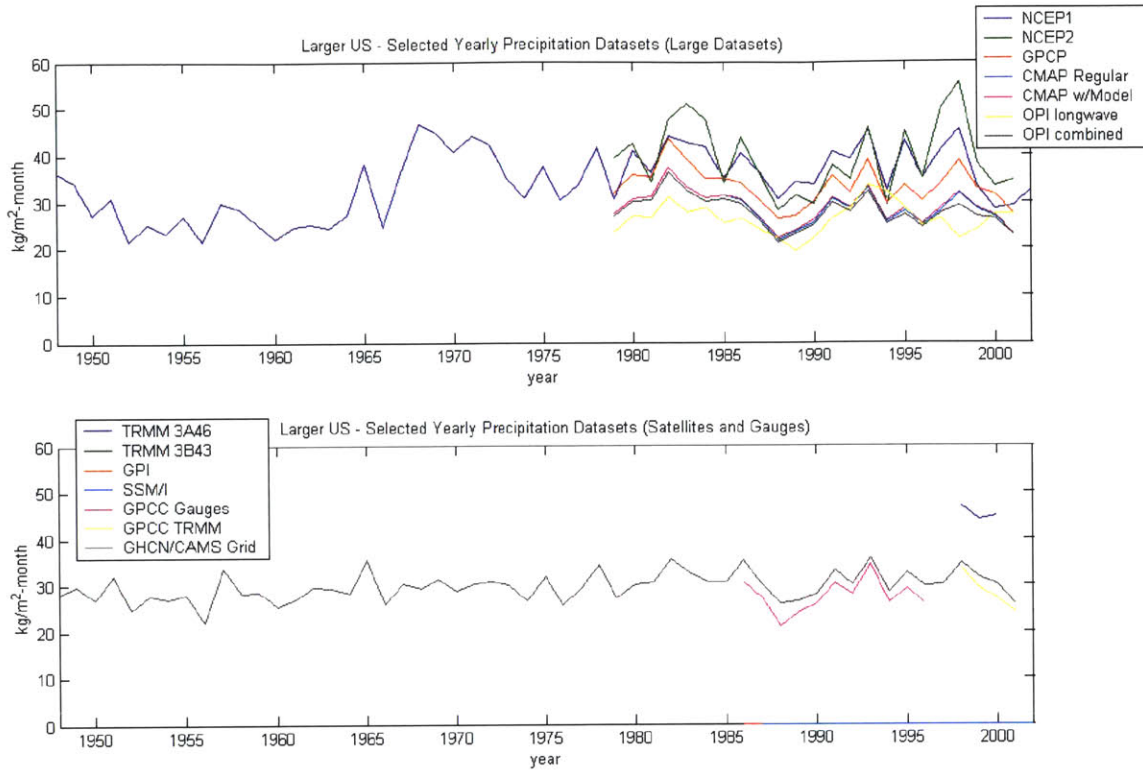


Figure 5-17: Larger US Different Sources of Yearly Precipitation

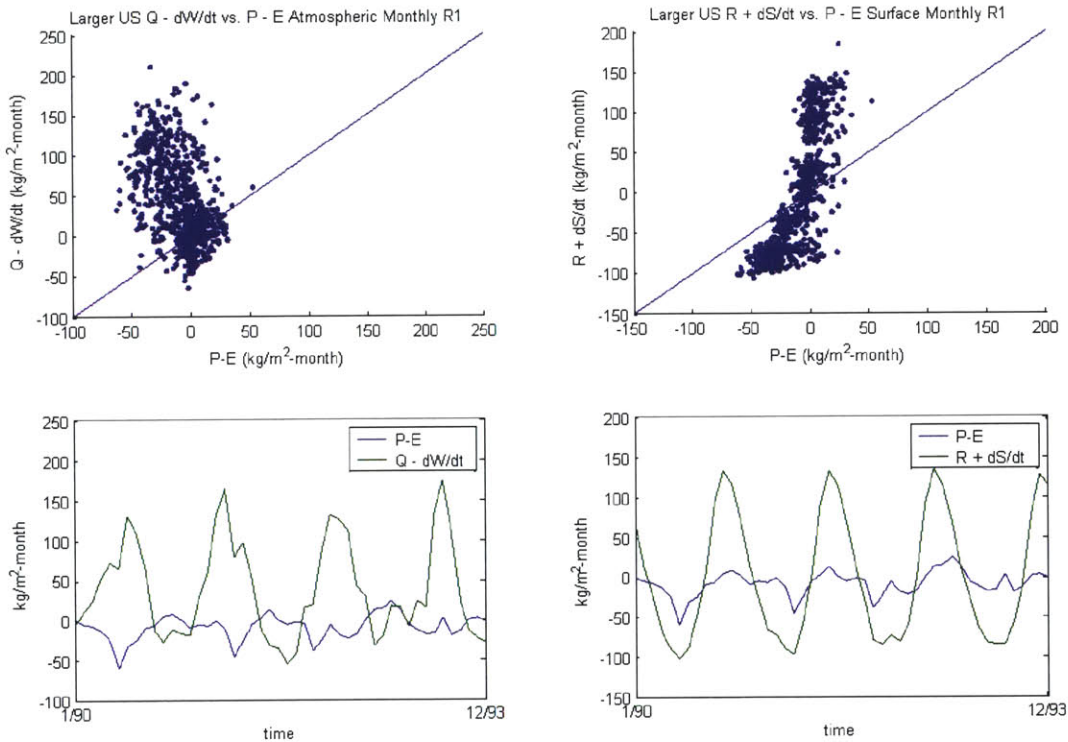


Figure 5-18: Larger US Reanalysis-1 Monthly Atmospheric and Surface Balances

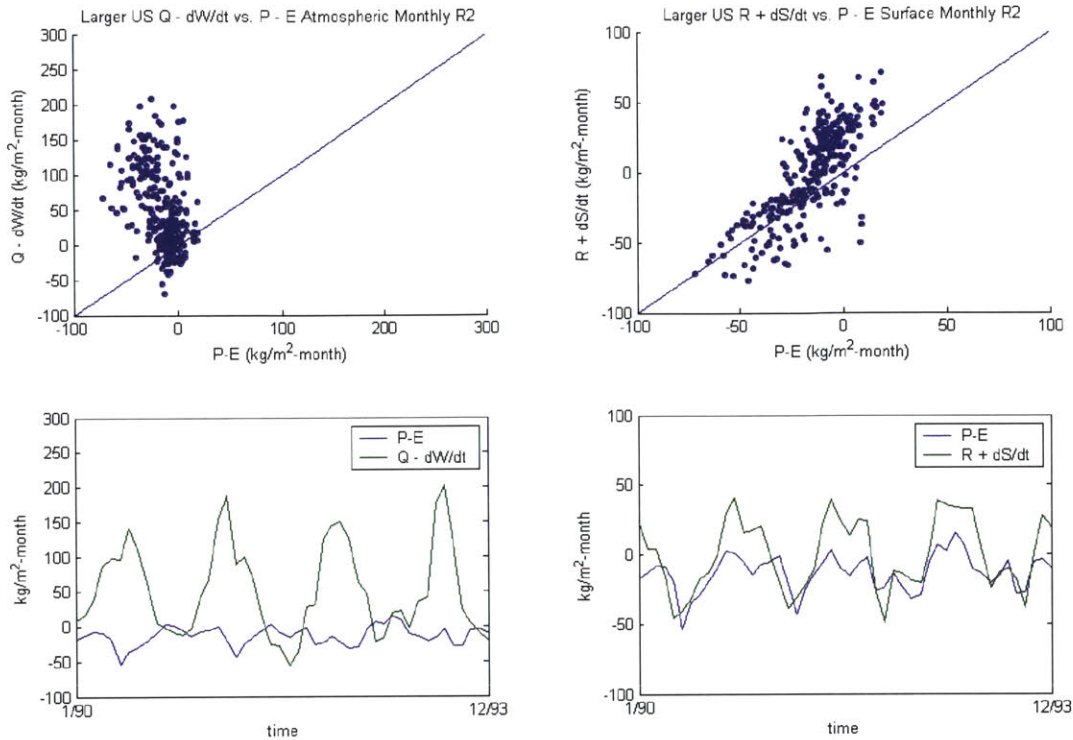


Figure 5-19: Larger US Reanalysis-2 Monthly Atmospheric and Surface Balances

Throughout this study we have observed problems with the atmospheric water balance of the Larger United States control volume. We see an unrealistically high seasonal cycle and unrealistically high annual values for the moisture flux. The most likely explanation for this is the mountainous terrain contained in this control volume. The moisture flux calculations depend on pressure level data, starting from sea level pressure, however these levels don't even exist in mountainous regions and this may lead to difficulties in the moisture flux calculation.

As a result of the location of the Larger United States control volume, several precipitation datasets are not used. Table 5-4 shows the variables which are considered for each control volume, which is similar to the previous two control volumes, however missing the 3 tropical precipitation datasets (GPI, TRMM 3B43, TRMM 3A46), and also missing the SSM/I precipitation which contained an anomalous error for this control volume. Precipitation is the only characteristic affected by the geography of the control volume.

Continuing, as we try to make corrections for the atmospheric yearly balance, we find the default reanalysis balances rank among the worst based on evaluation criterion. The top ranking balances have weak correlations and regressions. Selecting a balance with a good regression slope would mean doubling the various error statistics. Therefore we select the top ranked balance which includes a change to Method 1 and Reanalysis-1, applying the GPCP precipitation. Interestingly, for the Larger US control volume changing to Method 1 is very affective, as a regression slope of .5093 is obtained with an absolute error and bias of 21 kg/m<sup>2</sup>-month, less than half of that for the default balance.

For the yearly surface balance it is difficult to improve the balance, as all scenarios have very similar errors and very poor correlations and regression results. The most highly rated balances aren't correlated well at all, so in an attempt to find balance that is more correlated to the one to one line, we take the most highly rated water balance with a good slope. A slope of .9855 is found for the default conditions and applying the CMAP2 precipitation. The regression can be improved without offsetting the residual errors, but as can be seen by the bottom of Figure 5-20 the balance isn't improved too much.

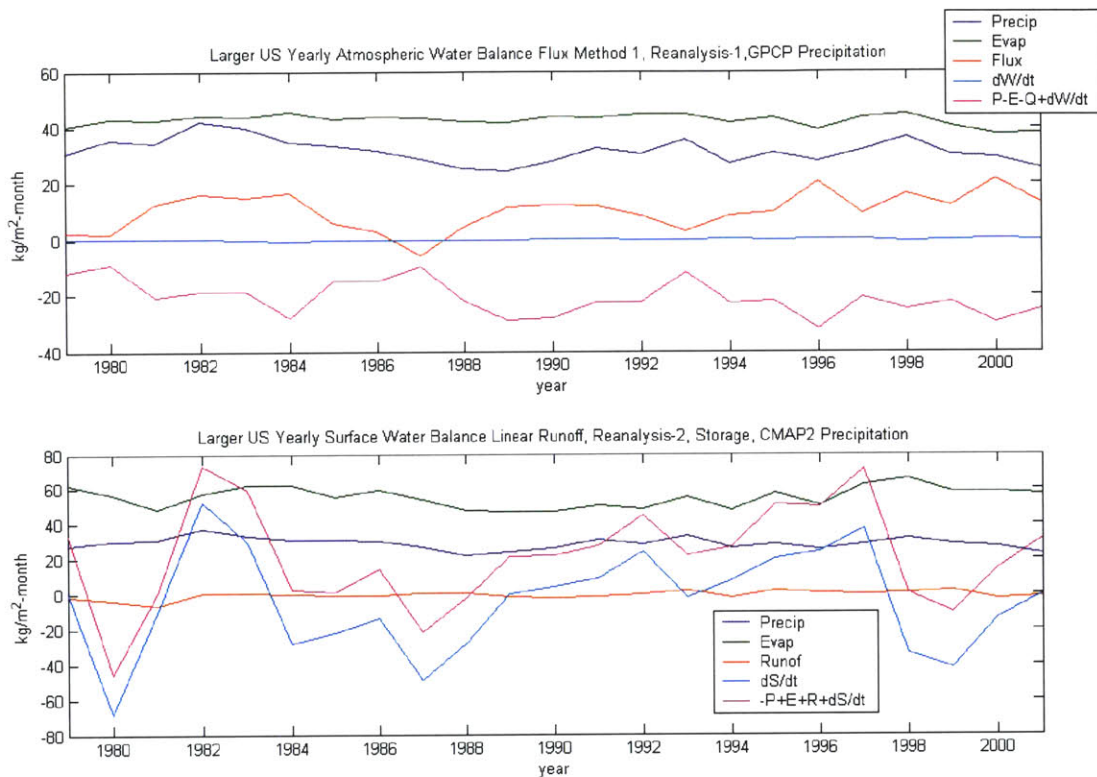


Figure 5-20: Larger US Selected Yearly Water Balances

Trying to improve the monthly atmospheric water balance is nearly impossible. For every scenario a negative correlation coefficient and slope are found. However, the less negative regression statistics do tend to correlate with the smaller residual errors and therefore it is possible to arrive at the same conditions that helped the yearly atmospheric water balance, applying Method 1 to the Reanalysis-1 with the GPCP precipitation. It is encouraging that the GPCP precipitation improves these balances, as it is a more reliable precipitation than the reanalysis. Looking at this balance in Figure 5-21, at least the opposite seasonal cycles of the two terms are aligned, and the bias is reduced to 21.03 kg/m<sup>2</sup>-month, same as in the case of the yearly balance.

In the monthly surface balance, improvement is gained with default conditions being applied to the Reanalysis-2. A comparable balance with acceptable correlation statistics is also found by changing to the point runoff method and keeping everything else as the default conditions for the

Reanalysis-2. All water balance statistics are almost exactly the same, as runoff is only slightly changed.

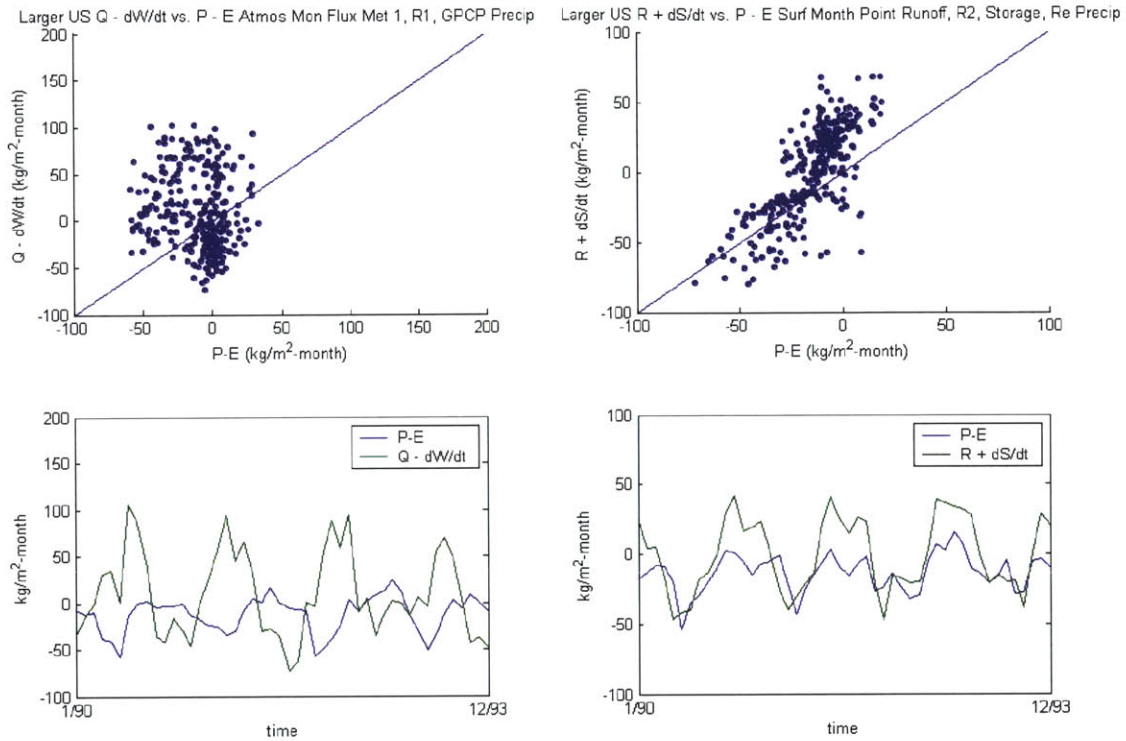


Figure 5-21: Larger US Selected Monthly Water Balances

The Larger United States water balance proves very hard to close for both the atmospheric and surface water balances. This is surprising considering the success in a remote area for the same size control volume. Problems in the closure of this balance were clearly affected by the inability of the moisture flux and runoff methodologies to compute realistic side fluxes for both water balances.

#### 5.4. Northwest Brazil

The Northwest Brazil control volume was selected for this work because it is a very remote area of the Amazon rain forest which has neither good measurement nor land change. Figure 5-22 displays the Reanalysis-1 yearly water balances for this region. Again we see a jump in the precipitation, though slight, about halfway through the time series, and a much more dramatic raise in moisture flux, which increases the negative bias of the residual over the last twenty years of the balance. The bias for the atmospheric balance is  $-45.97 \text{ kg/m}^2\text{-month}$ , however the correlation coefficient is a very high  $.8262$ , as the moisture flux signal matches up with the precipitation signal on a year by year basis. The results of the surface balance are less encouraging. While the change in storage cycle appears to correlate slightly with the

precipitation signal, the runoff does not and is negative and constant throughout almost the entire time series indicating the methodology does a poor job routing the precipitation. The result is a large negative bias and a poor correlation coefficient.

In Figure 5-23 we again see a very strong correlation between precipitation and moisture flux peaks for the Reanalysis-2 yearly atmospheric balance. The evaporation term in this balance is nearly constant for the duration of the time series, and the moisture flux is clearly overestimated in most cases. The correlation coefficient for this balance is .8424 and the slope of regression is improved to 1.37 from 1.51 in the Reanalysis-1 yearly atmospheric balance. The surface balance appears to get worse. In the Reanalysis-1 the runoff was near zero for all but the last five years of the time series. In the Reanalysis-2 runoff has magnitude, but is negative. It is likely the runoff is routed in the wrong direction here, as there is an increase in bias in the Reanalysis-2.

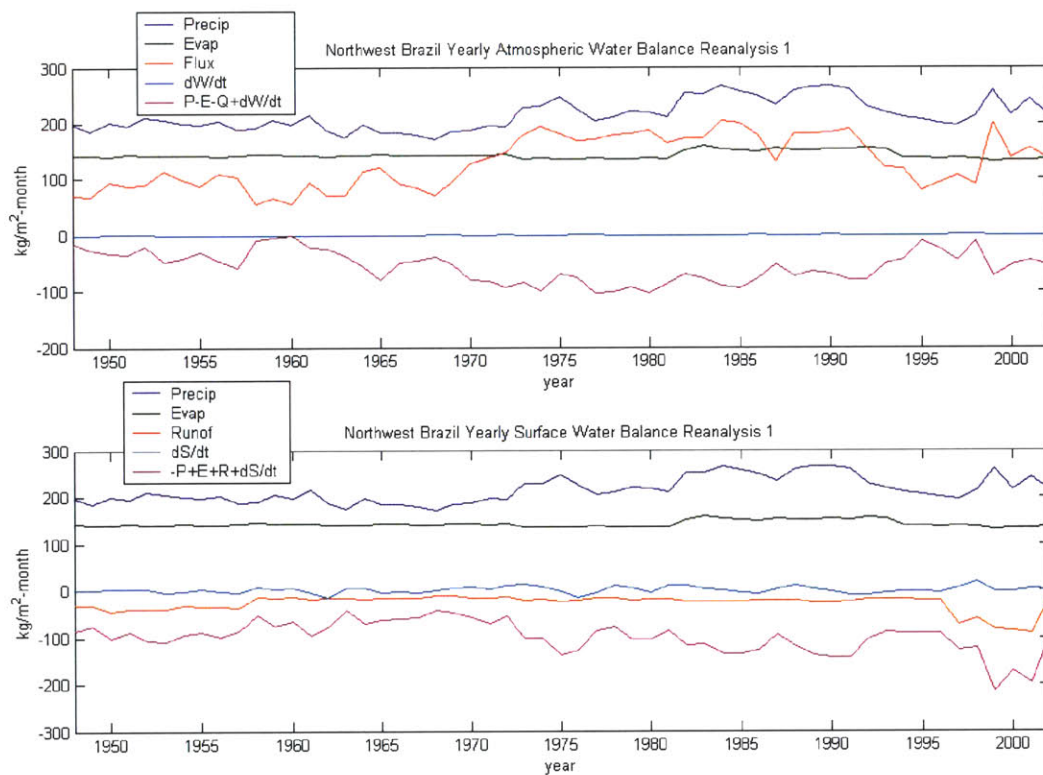


Figure 5-22: Northwest Brazil Reanalysis-1 Yearly Atmospheric and Surface Balances

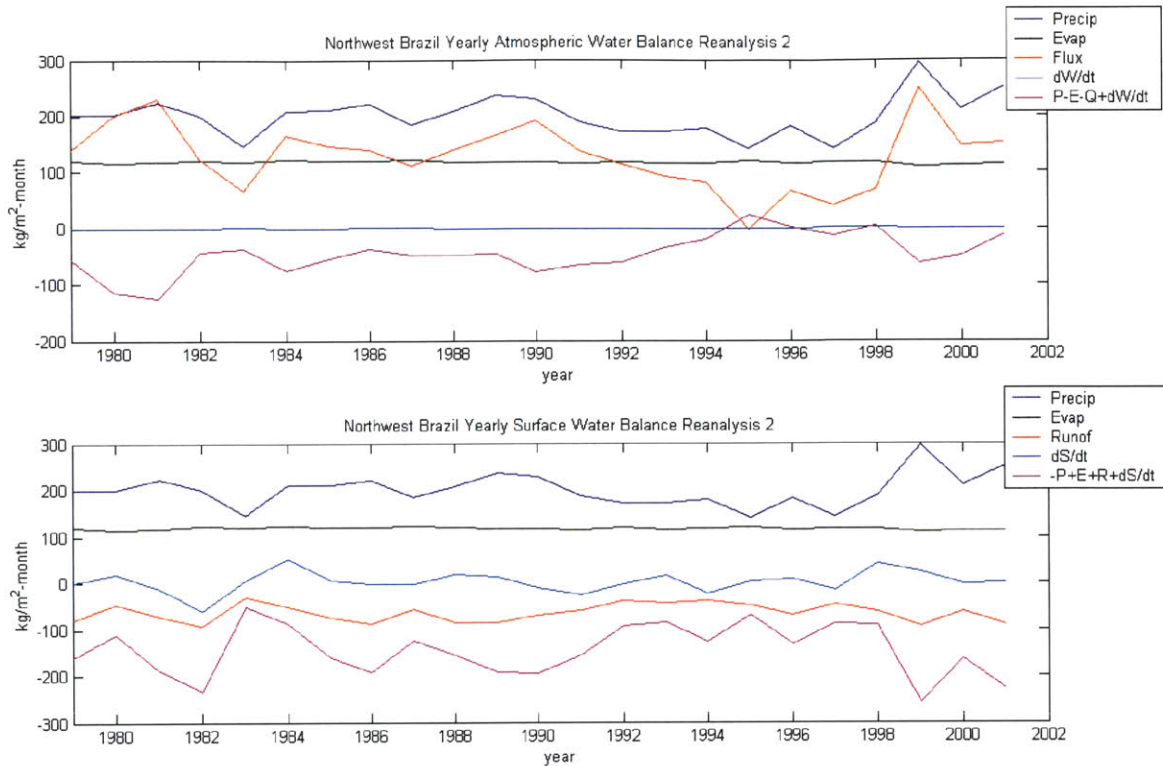


Figure 5-23: Northwest Brazil Reanalysis-2 Yearly Atmospheric and Surface Balances

The annual precipitation time series for the Northwest Brazil control volume are given in Figure 5-24. As with the other two control volumes in this region analyzed already, the Reanalysis-1 precipitation time series jumps during the second half of the time period, while more contemporary precipitation efforts are consistent with the early part of the Reanalysis-1. For this control volume however there is a sharp contrast between the Reanalysis-1 and Reanalysis-2 precipitation datasets. The Reanalysis-2 is closer to the observational precipitation data. The magnitude of precipitation for this control volume is fairly high, and there is a large amount of water flowing through this region every month. The GPI and SSM/I datasets are again higher than the other precipitation datasets; however their maximum signals do not coincide well with those from the Reanalysis-1.

Switching to the monthly time scale the Reanalysis-1 default balances are presented in Figure 5-25. In the atmospheric balance a good correlation is obtained, however the moisture flux signal is overestimated with a regression slope of 1.22 and a correlation coefficient of .8253. For all of the example points given in the bottom portion of the figure the moisture flux is higher, however signals are picked up on a point by point basis. The monthly surface balance is not very good. The runoff is routed in the wrong direction (negatively) and the seasonal cycle does not appear to be aligned.

Switching to the Reanalysis-2 in Figure 5-26, the atmospheric monthly balance is about the same, the slope of the regression is slightly increased to 1.29, but so is the correlation coefficient of .8431. The bias is reduced from  $-56.63 \text{ kg/m}^2\text{-month}$  to  $-45.93 \text{ kg/m}^2\text{-month}$ , which can be



seen by the example points. Although the evaluation criteria indicate the Reanalysis-1 is a better balance, the Reanalysis-2 is clearly comparable if not better for this control volume and the atmospheric monthly balance. The monthly surface balance for the Reanalysis-2 is interesting for this control volume. The side flux term for the surface balance appears to be greatly improved, as there is signal pick up for the example points shown in the bottom plot. However, the intensity of the water cycle in Northwest Brazil is very large and so is the scale on these graphs, thus the bias between the points is large. Examining the correlation plot, the correlation is poor, leading to a negative correlation coefficient and regression slope.

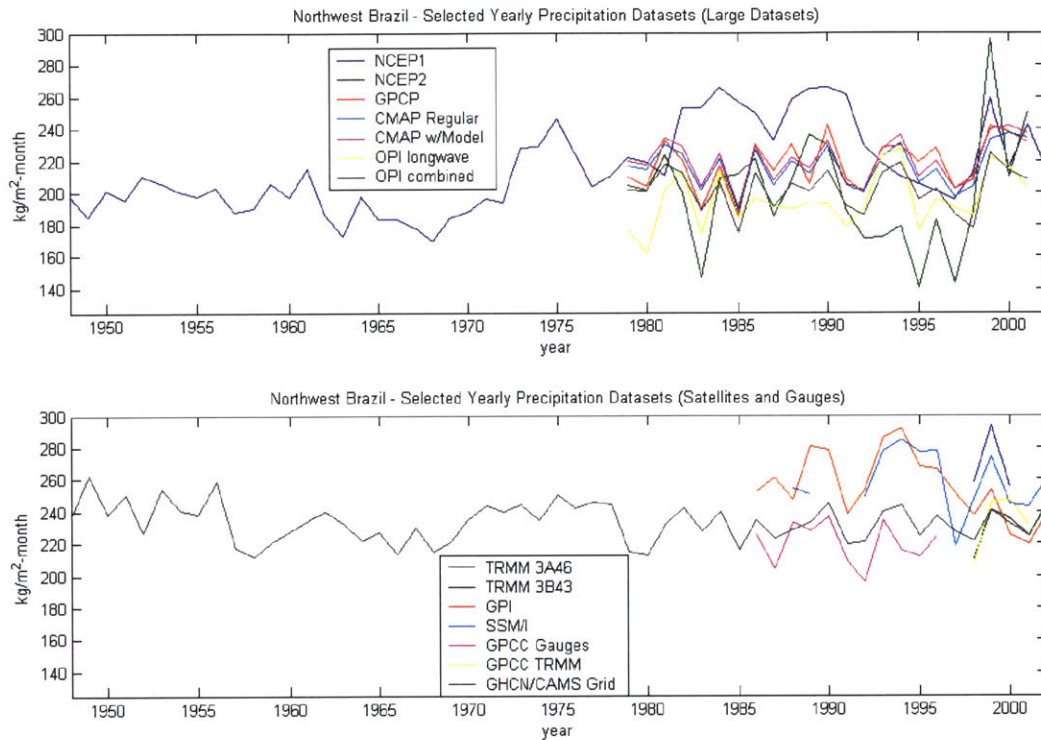


Figure 5-24: Northwest Brazil Different Sources of Yearly Precipitation

We now try to improve these water balances, especially the surface balance, by altering water balance characteristics. For the possible changes, Table 5-1 can again be used as this is one of the small tropical control volumes where all possibilities exist. The Northwest Brazil control volume is a very typical control volume in this work, as the atmospheric balance is easy to close to some to degree, while the surface balance is more difficult. In Figure 5-27 we try to improve the yearly water balances and we find by the criteria in Appendix A that for the yearly atmospheric balance, small improvements can be made. The best ranking method involves a slight shift of coordinates to the moisture flux Method 1 applied to the Reanalysis-1. This shift in coordinates lowers the moisture flux so it is almost perfectly aligned with the precipitation - evaporation and the bias is reduced to  $-4.33 \text{ kg/m}^2\text{-month}$  in a water balance with quantities very high in magnitude. This is a very good balance, and for the atmospheric case this exercise can be deemed a success.

For the yearly surface balance on the bottom half of Figure 5-27, the default reanalysis balances rank poorly, but so do all cases. In the top ranked scenario the runoff methodology is changed to the point runoff method and change in storage is eliminated, applying the balance to Reanalysis-2 using OPI2 precipitation. Through these changes positive yet small slopes and correlation coefficients are obtained, and we see small correlation between the runoff and precipitation time series in the graph, as the annual evaporation is nearly constant. The balance still has a very negative bias.

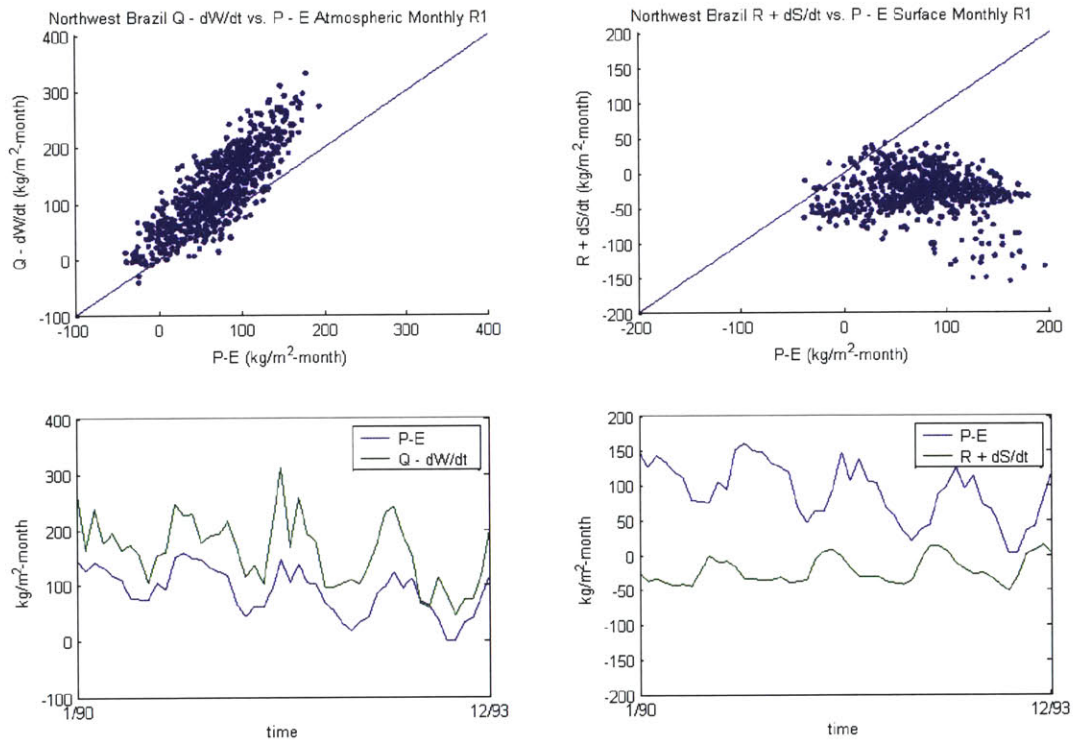


Figure 5-25: Northwest Brazil Reanalysis-1 Monthly Atmospheric and Surface Balances

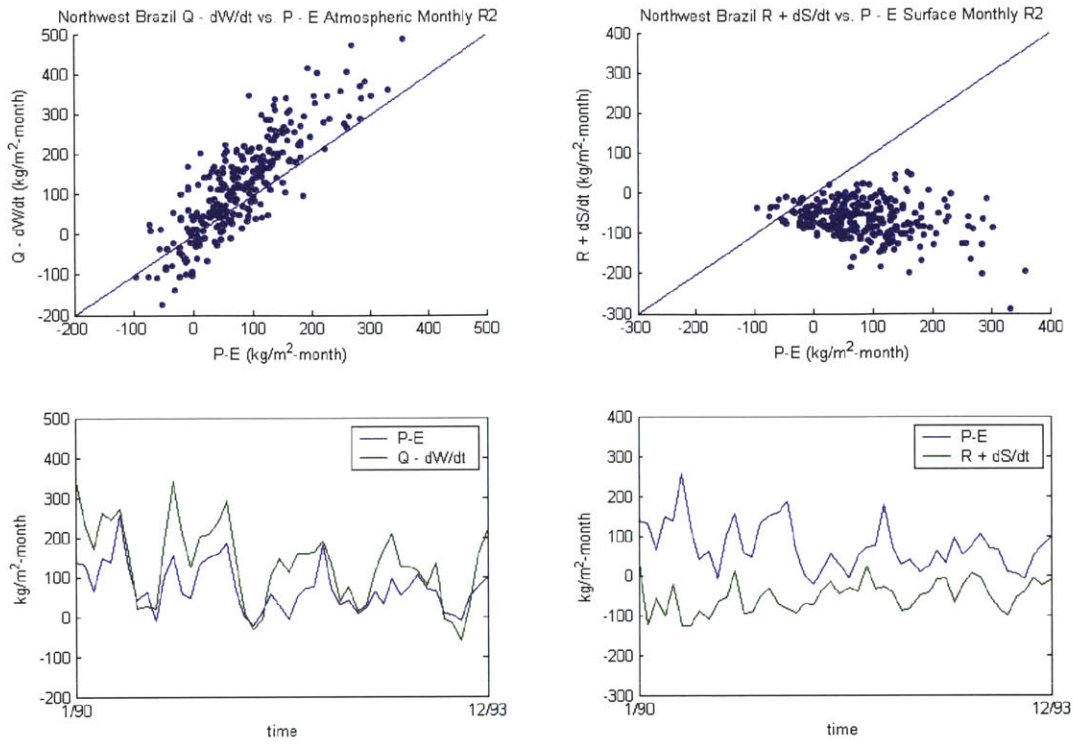


Figure 5-26: Northwest Brazil Reanalysis-2 Monthly Atmospheric and Surface Balances

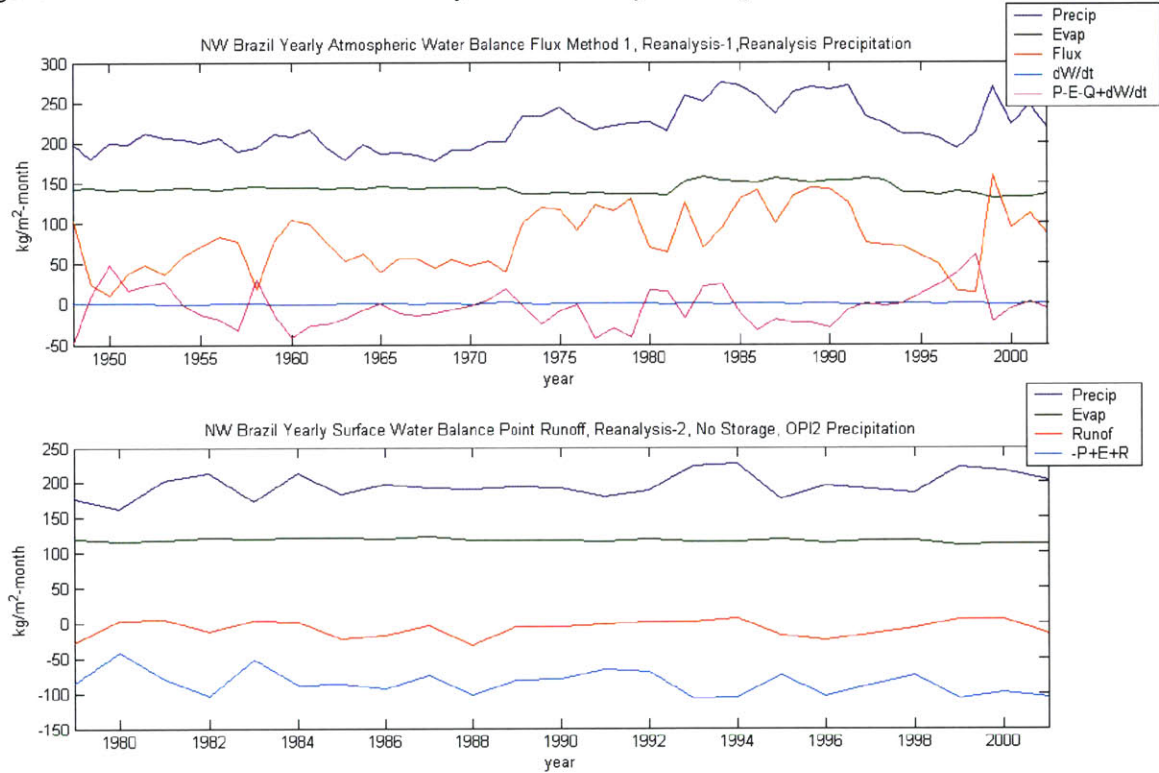


Figure 5-27: Northwest Brazil Selected Yearly Water Balances

In Figure 5-28 the monthly atmospheric water balance is improved in the same manner that the yearly water balance was improved, by applying the first moisture flux methodology to the Reanalysis-1. The bias is removed and is equivalent to that of the yearly balance. The slope of the regression is .9214. We select this set of conditions though it is ranked second by the evaluation criteria, because it is twice as long as the very similar balance achieved under the same conditions for the Reanalysis-2, which is ranked number 1. For the surface water balance there is a lot of room for improvement and correspondingly many different water balances which can be tried, however, there are still no good results. For this control volume the point runoff methodology is applied to Reanalysis-1 with change in storage, as this is the top ranked water balance based on the evaluation criteria. The point methodology allows for a much better seasonal correlation through the side flux of the surface balance is still grossly underestimated (Bias of  $-77.65 \text{ kg/m}^2\text{-month}$ , regression slope of about .21).

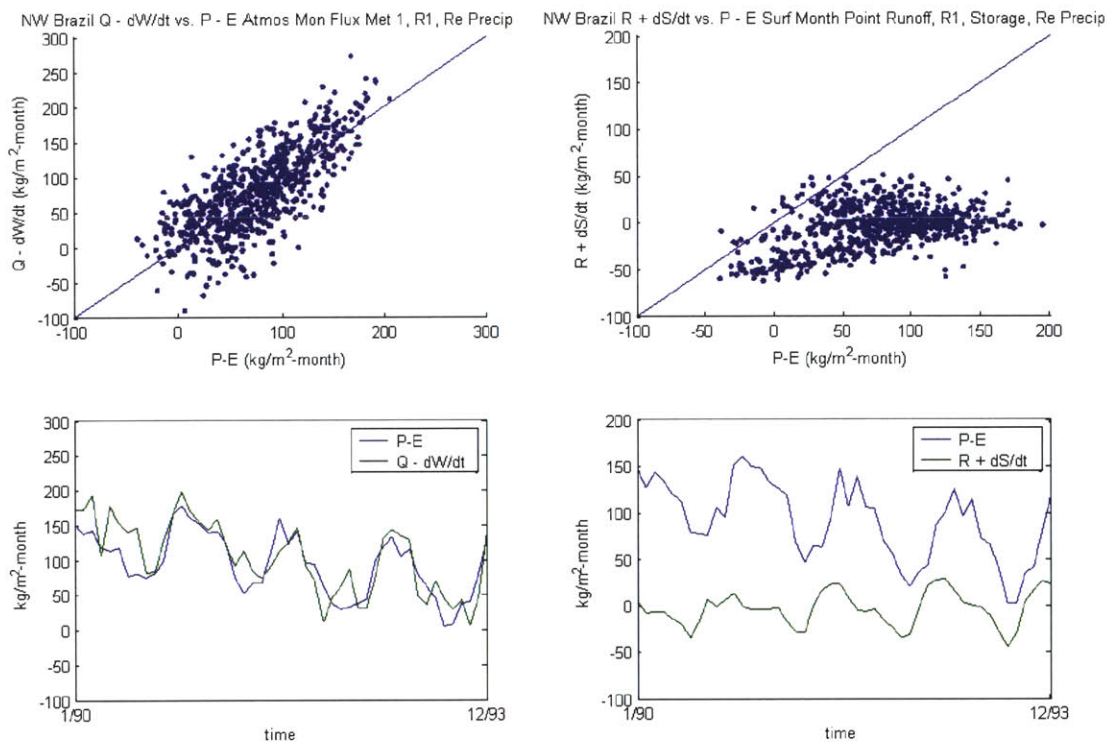


Figure 5-28: Northwest Brazil Selected Monthly Water Balances

For the Northwest Brazil control volume many of the same observations were made as for the other tropical control volumes in this region so far. A second half of the Reanalysis-1 rise in precipitation is again observed, but accounted for by the reanalysis moisture flux in the atmospheric balance. The surface balance is much harder to close than the atmospheric balance.

## 5.5. Rondonia

The Rondonia control volume was the first and original control volume selected for this work. It is an area of unique importance in the Amazon rain forest and is currently undergoing land change.

In Figure 5-29 the Reanalysis-1 yearly water balances are displayed. What is seen here is a clear jump in the precipitation between 1970 and 1975, and a corresponding jump in moisture flux. Following these shifts the balance is actually improved, thus in this case, the poor water balance and uncharacteristic negative moisture flux over the region could be explained away by unreliable data in the early half of the Reanalysis-1. However, we have consistently seen this trend over every region in the Amazon rain forest. It is unlikely that all of these observations are a coincidence. There is a distinct change in the water balance over these regions in the past fifty years. Such a discrepancy makes it even more important to continue collecting data for such studies to see if trends continue. Decent statistics are obtained for this balance with a bias of  $18.67 \text{ kg/m}^2\text{-month}$  and correlation coefficient of  $.7202$ . The question lies in the validity of the balance. For the yearly Reanalysis-1 surface balance, there is little adherence to the P-E cycle by either the runoff or storage change. Curiously in the last five years of the Reanalysis-1 the water balance is greatly improved, however this does not salvage a balance with a  $.3092$  correlation coefficient and no slope. The error is low due to the closeness of the P-E quantity to zeros.

In Figure 5-30 the Reanalysis-2 yearly water balances are displayed. The Reanalysis-2 increases the bias for the atmospheric balance. The precipitation and evaporation cycle is improved as there is less of a difference from the first half of the Reanalysis-1. However, the moisture flux is very negative, resulting in a high positive residual and a bias of  $93.50 \text{ kg/m}^2\text{-month}$ . The correlation coefficient of the balance is still a high  $.8210$ . This is indicated by the residual line being fairly flat. For the surface balance the change in storage and runoff terms do not correlate properly with the rest of the balance, and instead alternate around zero. A positive regression slope of  $.3348$  and weak  $R^2$  of  $.0759$  represent a faint linear relationship between the different sides of the water balance. Upon residual analysis, the water balance could be balanced about as well by eliminating the side flux of the surface balance entirely, as this flux provides no improvement to the balance. Closing the yearly surface balance with the reanalysis continues to be nearly impossible with the given data in remote areas, even with the Reanalysis-2, as the gridded runoff is not represented correctly by the methods used.

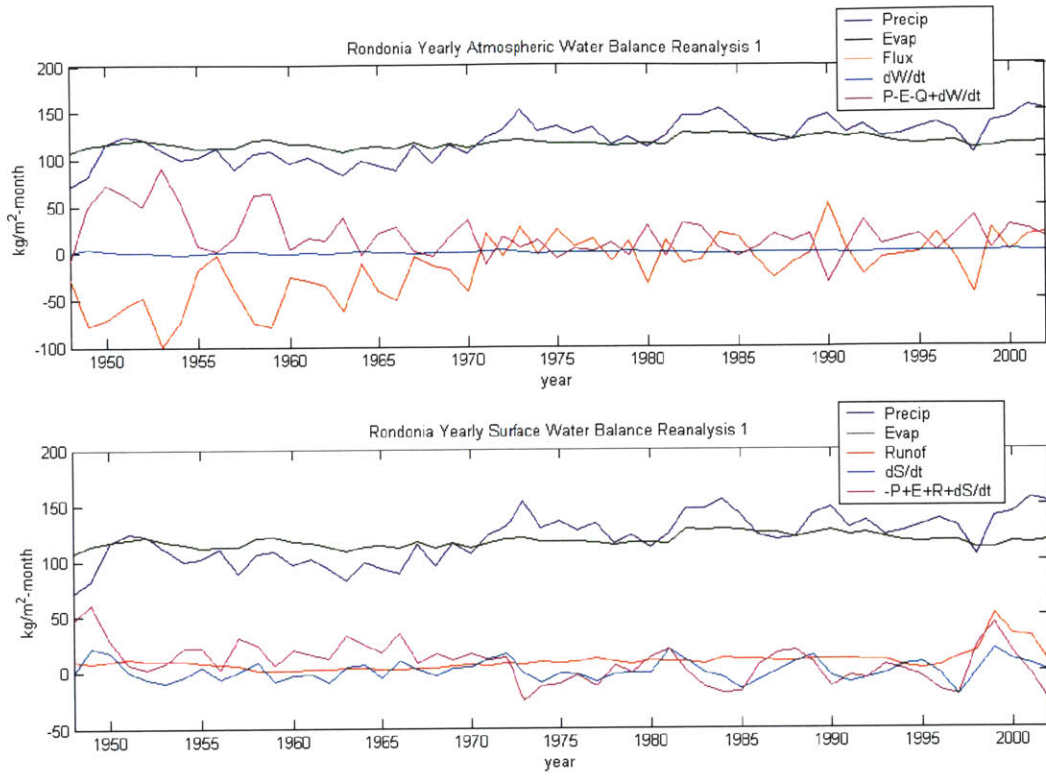


Figure 5-29: Rondonia Reanalysis-1 Yearly Atmospheric and Surface Balances

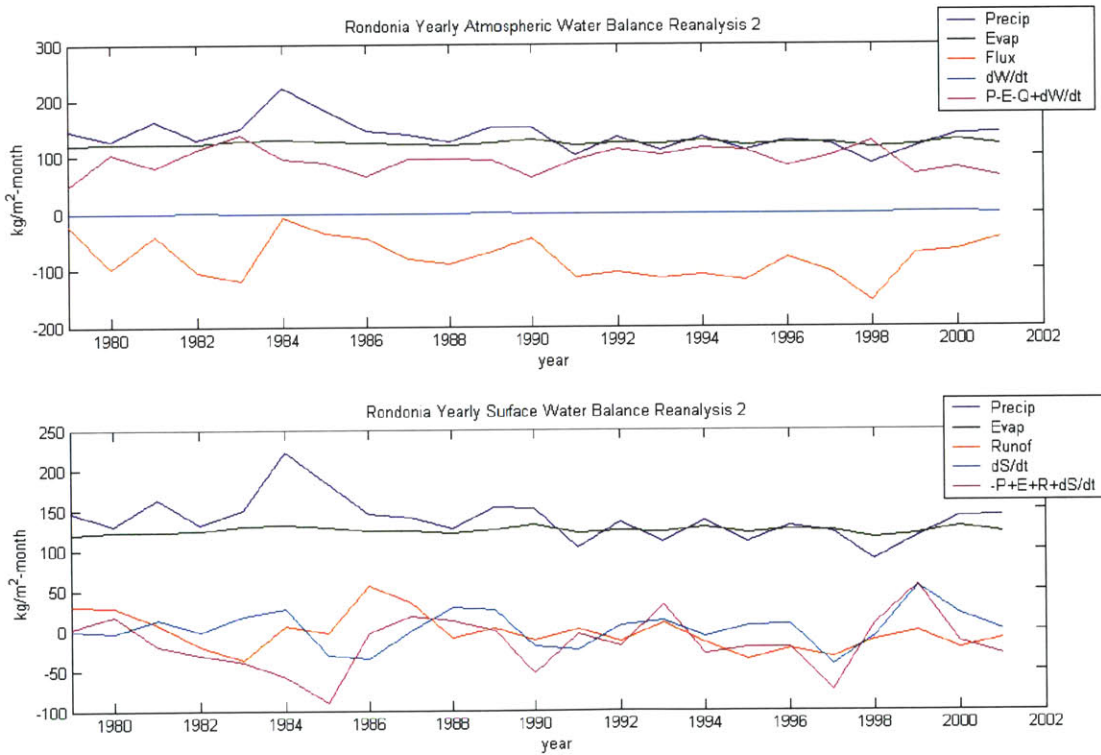


Figure 5-30: Rondonia Reanalysis-2 Yearly Atmospheric and Surface Balances

In Figure 5-31 the yearly precipitation time series are shown, and differences are seen in the Rondonia control volume compared to the other control volumes analyzed so far. The alternative precipitation datasets are not aligned with the precipitation from the first half of the analysis for this control volume, but rather with the second half of the Reanalysis-1 where the precipitation is higher. This is unusual because for other control volumes in South America alternative precipitation datasets do not agree with the second half of the Reanalysis-1.

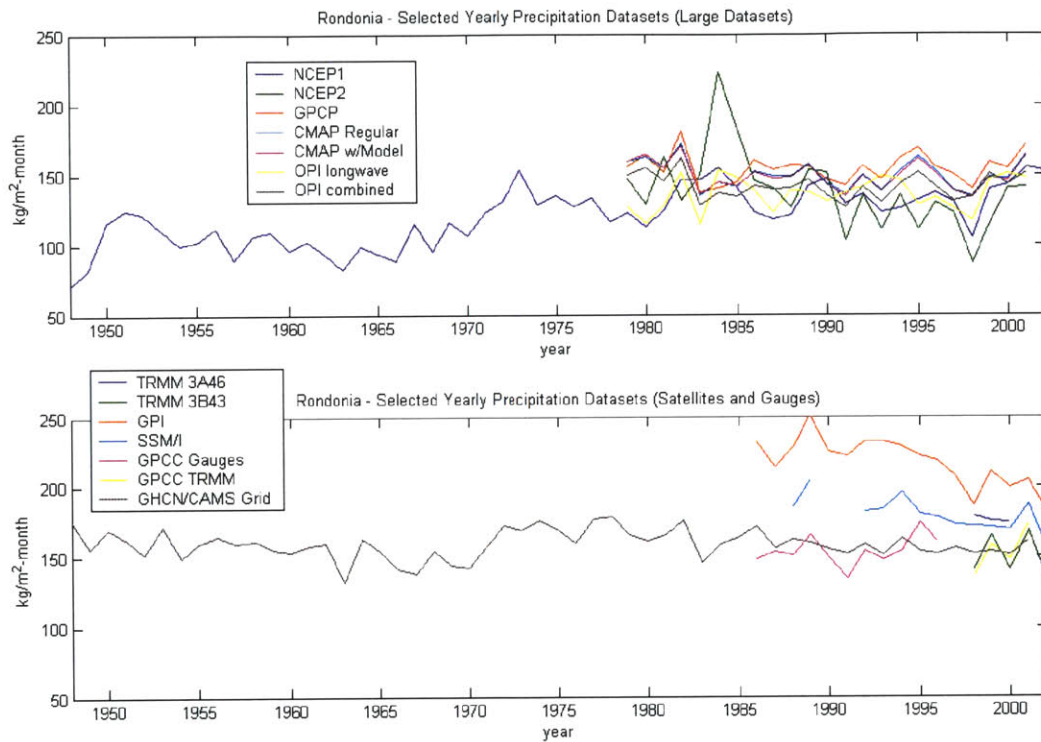


Figure 5-31: Rondonia Different Sources of Yearly Precipitation

In Figure 5-32 the monthly atmospheric and surface water balances are plotted for the Reanalysis-1. For the atmospheric balance we don't see as tight a correlation. It should be noted however, that over this control volume there is no strong seasonal cycle in the moisture flux time series as was the case with all control volumes to this point. As a result the signals presented are more like those of the yearly time scale, and so is the correlation. The correlation coefficient of this balance is only .3343. For the surface balance the seasonal cycle is uncovered with the removal of the moisture flux. There is a clear correlation of seasonal cycle shown by a correlation coefficient of .4566. This shows the effect of the seasonal cycle in improving the balance on the monthly time scale.

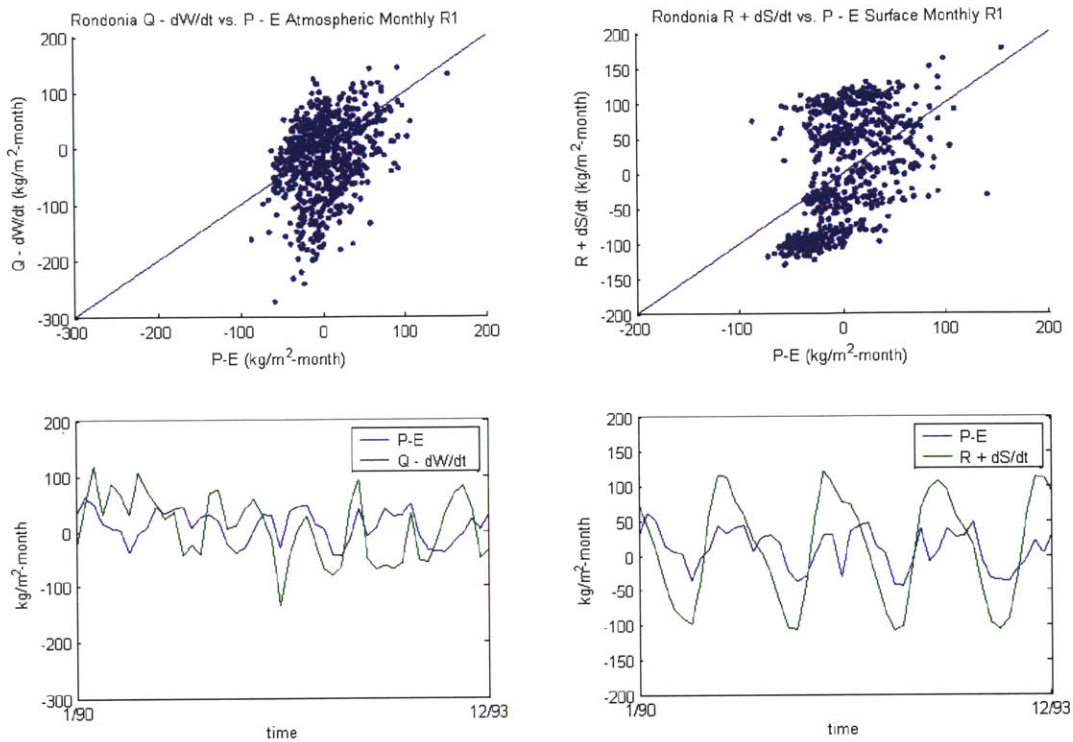


Figure 5-32: Rondonia Reanalysis-1 Monthly Atmospheric and Surface Balances

In Figure 5-33 we switch to the Reanalysis-2 on the monthly time scale. We see the negative bias in the moisture flux as observed on the yearly time scale. Although the bias is greatly increased, the correlation coefficient is also increased from the Reanalysis-1. The seasonal cycle is improved for the moisture flux, although it is way too low. The entire regression seems to be biased by an intercept as the slope of the regression is .7111 with a regression intercept of -89.44  $\text{kg/m}^2\text{-month}$  which is very close to the bias, meaning this balance could be corrected by a nudging term at each time step. The surface balance in this case looks very good, which is surprising given the condition of the yearly surface water balance in the Reanalysis-2. By the evaluation criteria this balance is slightly worse than that obtained by the Reanalysis-1 primarily due to the side flux of the surface balance not having a strong enough seasonal cycle. This is likely caused by an unsuccessful runoff calculation which may be improved by a change in methodology. A correlation coefficient of this balance is a fairly high .5428.



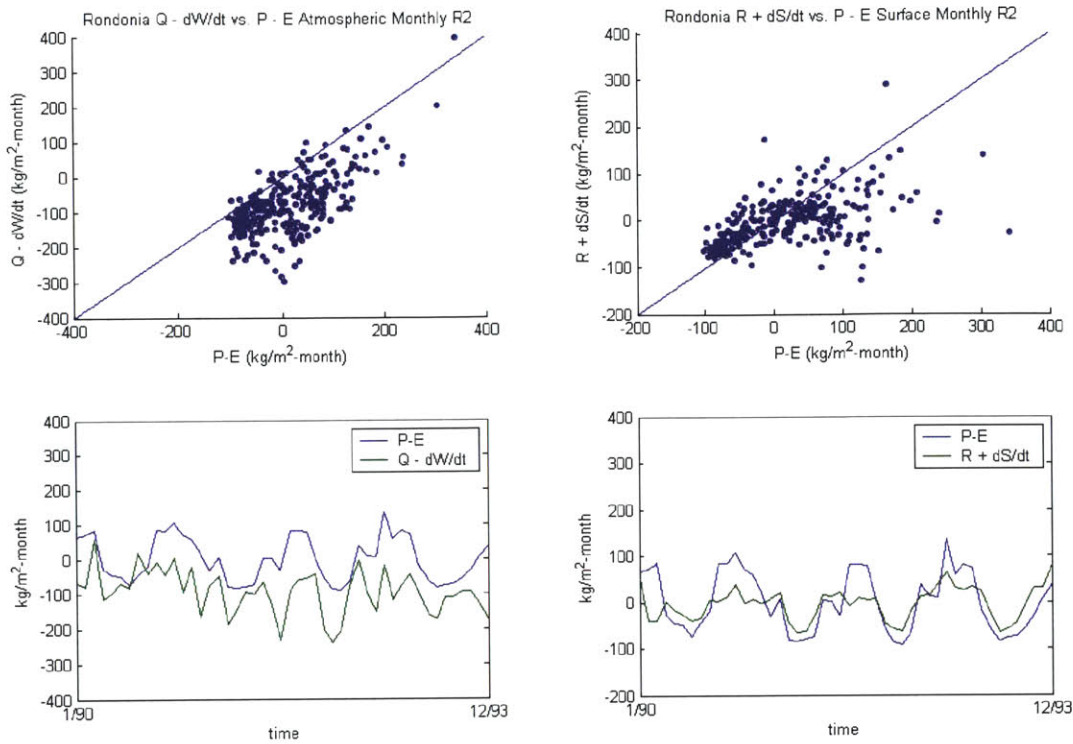


Figure 5-33: Rondonia Reanalysis-2 Monthly Atmospheric and Surface Balances

For the atmospheric yearly balance the best results by the evaluation criteria are clearly given by the default method used by the Reanalysis-1. The Reanalysis-2 balance is also one of the top balances in this scenario, but of great interest here is the success of the water balance applying the TRMM 3B43 dataset to both the atmospheric and surface balance on the yearly time scale. In the atmospheric balance this application is to the Reanalysis-1 with Method 2 moisture flux. The balances obtained are shown in the top portion of Figure 5-34. For the surface balance the TRMM 3B43 precipitation is applied to the Reanalysis-1 with the area runoff methodology. The figures indicate biases, but otherwise generally good water balances on the yearly basis with the TRMM precipitation product. The correlation coefficient for the yearly atmospheric balance is only .3551 while a strong correlation is observed in the surface balance with comparable errors and biases.

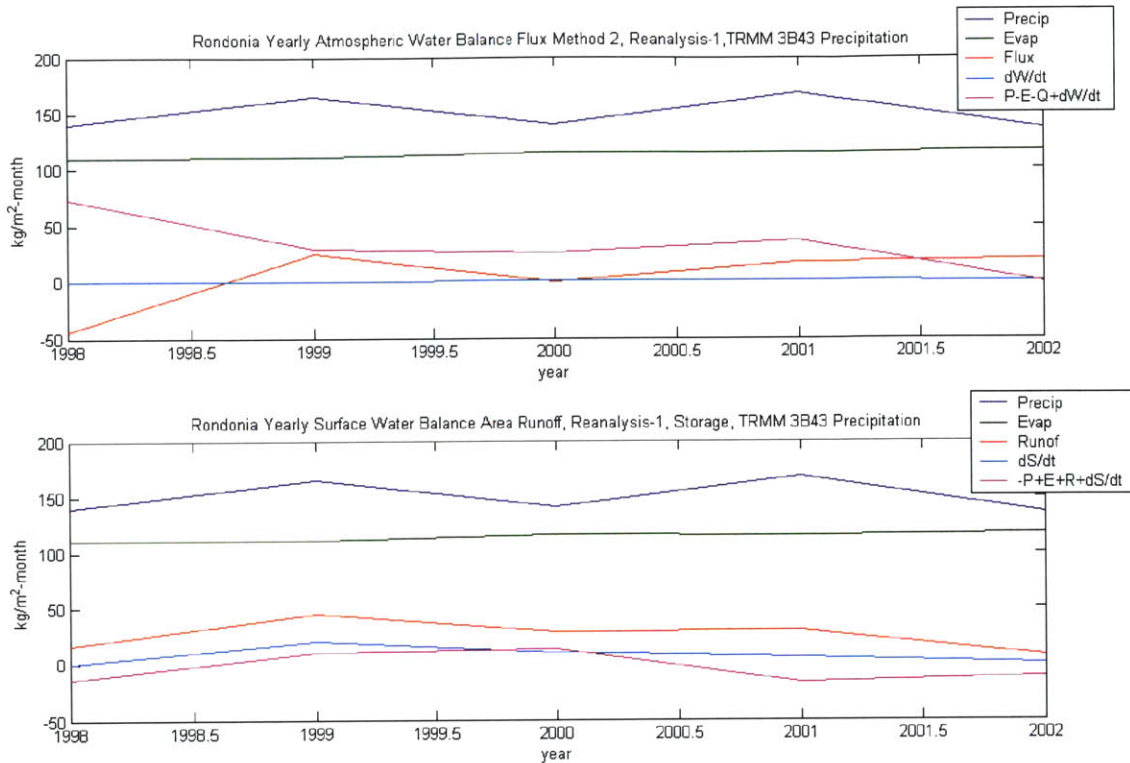


Figure 5-34: Rondonia Selected Yearly Water Balances

Next we attempt to improve the monthly water balances. For the atmospheric balance we again see the top ranked balances belong to the reanalysis under various conditions. Trying to keep with the application of TRMM products to this water balance the GPCP TRMM precipitation dataset is applied to the Reanalysis-1 for the atmospheric monthly balance shown in Figure 5-35. The balance is not improved in this case, but a comparable balance is achieved with a weak correlation of .1161. It is curious here that by the evaluation criteria the reanalysis balances for the monthly atmospheric case are far better than those obtained with alternative precipitation applied. The validity of the reanalysis precipitation has been questioned throughout this work, and it is hard to classify this precipitation as reliable despite the fact it completes the water balance.

For the monthly surface balance significant improvements can be made on the monthly time scale by using the evaluation criteria and once again a TRMM dataset completes the balance the best, and again it is the GPCP TRMM dataset. The point runoff methodology is used along with the Reanalysis-1. The balance is very visually apparent by both the scatter plot, and the plot of the compared data points in the time series. The runoff signal is captured by this configuration. It is possible to show the entire monthly time series for both balances in the bottom portion of these plots as there are only 59 points in each time series. The slope of the surface correlation is almost exactly 1 with a high regression  $R^2$  1:1 value of .6320, a quantity that is usually meaningless.

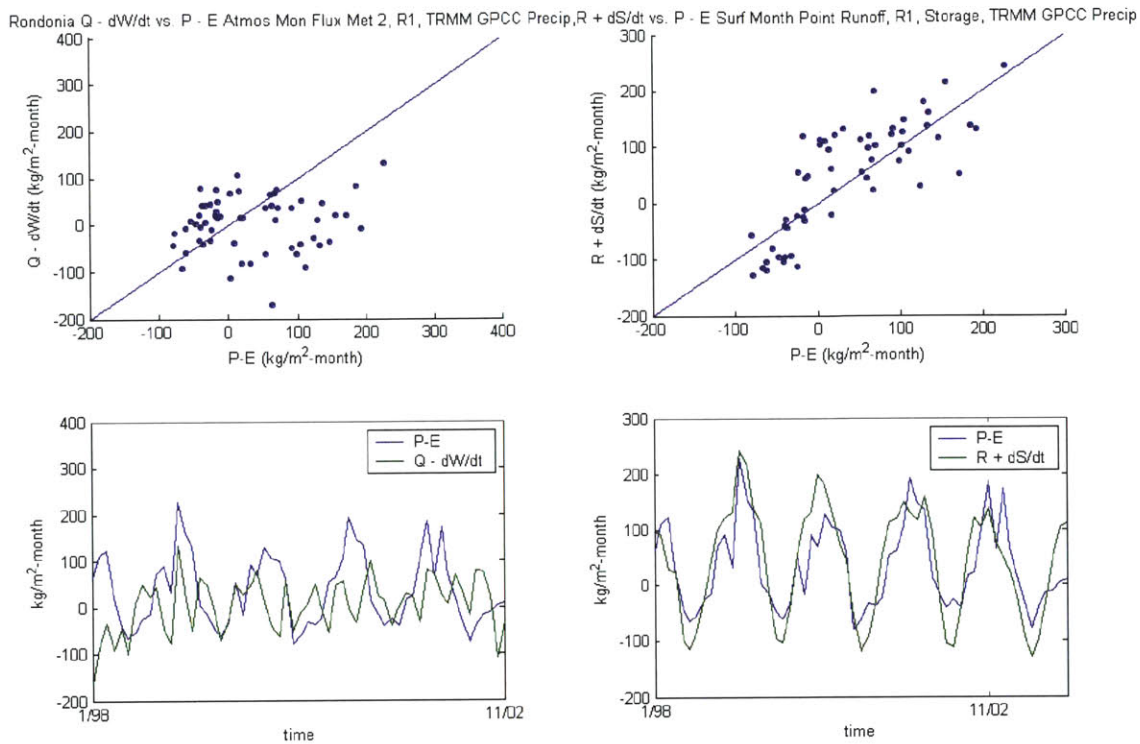


Figure 5-35: Rondonia Selected Monthly Water Balances

It is exciting to see the TRMM products be so helpful in refining the water balance for the Rondonia control volume. We do not know what to make of the water balance for this control volume based on the Reanalysis prior to 1970, as a climatic change is indicated by the best atmospheric water balance, the Reanalysis-1. It was also found that by trying many variations of the surface water balance, the balance could be greatly improved.

### 5.6. Southeast United States

The Southeast United States control volume was selected as a box control volume of the smaller 7.5 degree size, which could be used as a control in that it is in a geographic region where a lot of hydrometeorological data is known and recorded. It is also located in between 40 degree N and 40 degrees S latitude and therefore is included in the TRMM and GPI precipitation datasets. Unlike the other control volumes over the United States, this control volume contains no unusual features, and should thus prove good as a comparison to the smaller control volumes in South America.

Figure 5-36 displays the yearly water balances for the SEUS for the Reanalysis-1. For the atmospheric balance, the precipitation generally decreases in the 1960's, and a corresponding decrease is found in the moisture flux. However, the moisture flux is in general too low resulting in a positive bias of 31.90 kg/m<sup>2</sup>-month. The correlation of the slope is exactly one with an intercept of the regression equaling the bias and a correlation coefficient of .7956. This balance is very good with the exception of the bias. For the surface balance we again find a poor runoff

time series for the Reanalysis-1. A correlation coefficient of .2298 is found for the Reanalysis-1 surface balance.

Switching to the Reanalysis-2 in Figure 5-37 the bias is reduced for the yearly atmospheric control volume to 22.02 kg/m<sup>2</sup>-month, but there is still significant error in the balance compared to the magnitude of the other water balance components, and the correlation coefficient is reduced to .4736. Still the reduction in bias and general errors is encouraging. For the Reanalysis-2 surface balance the runoff is improved, but the balance gets worse. The correlation coefficient here is high at .5836, but the bias of 17.40 kg/m<sup>2</sup>-month is on par with the difference between precipitation and evaporation, meaning the surface side flux doesn't help to improve the balance. The absolute error of 29.43 kg/m<sup>2</sup>-month is significantly higher than the bias.

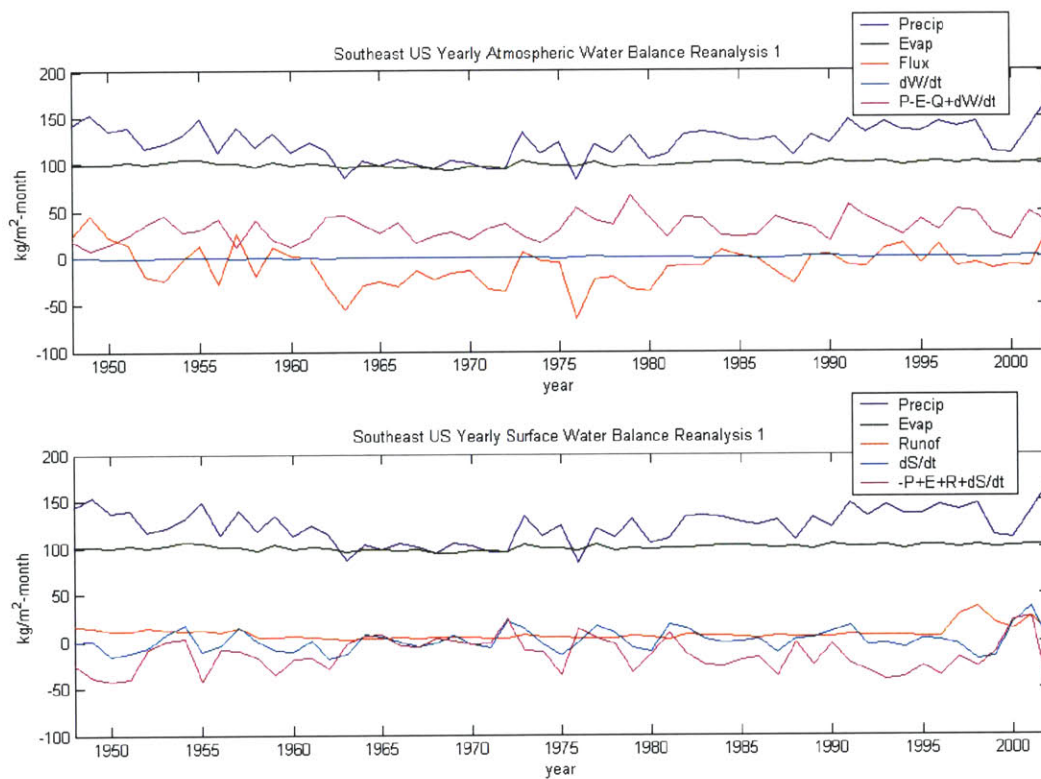


Figure 5-36: Southeast US Reanalysis-1 Yearly Atmospheric and Surface Balances

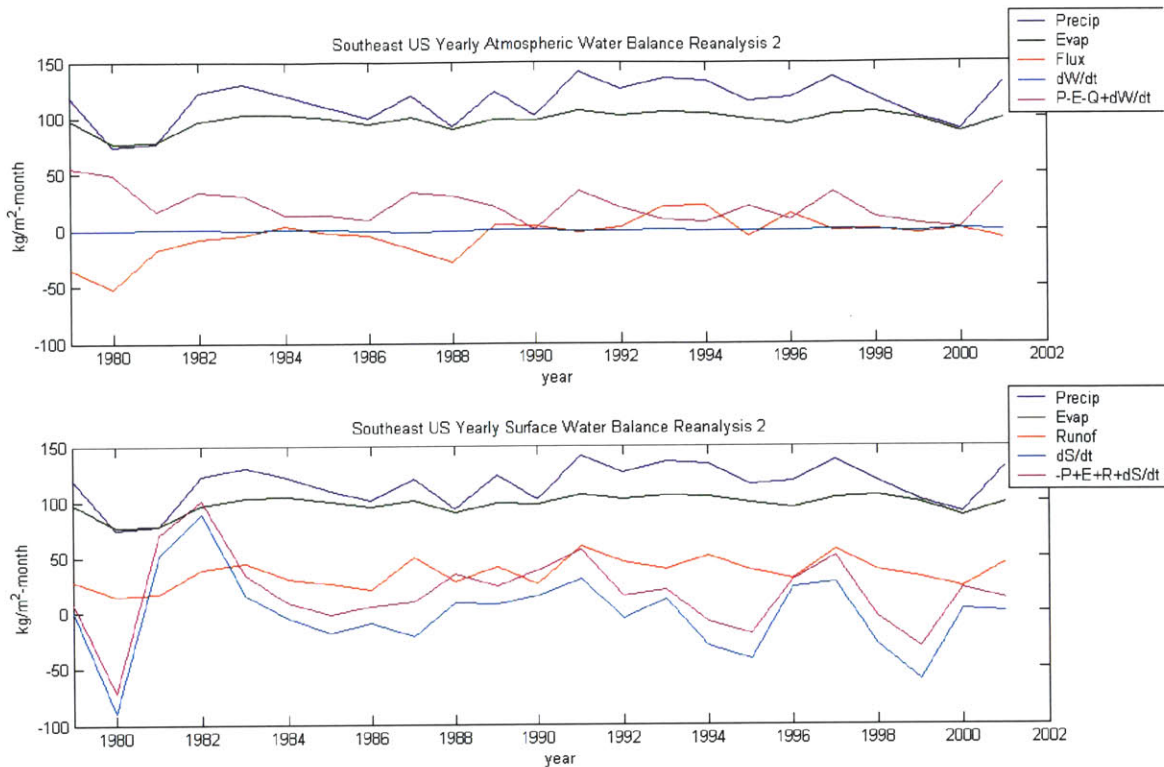


Figure 5-37: Southeast US Reanalysis-2 Yearly Atmospheric and Surface Balances

In Figure 5-38 the different annual precipitation time series for the SEUS control volume are plotted. The shift in precipitation observed in the tropical control volumes is not observed here as it wasn't observed in the Larger US control volume. Variation in the reanalysis precipitation is again observed, as there is a slight dip in the middle of the time series. Interestingly the gauge precipitation dataset does not show this dip, and this dip is consistent with the moisture flux dip in the Reanalysis-1. In this case the gauge precipitation is likely accurate because the United States has been densely gauged over the past 50 years, and most of these gauges are included in this dataset. So the precipitation trend in the reanalysis which is not backed by observational (gauges) data matches with the moisture flux of the reanalysis. The Reanalysis-2 correlates well here to the Reanalysis-1 precipitation, but all other datasets consist of a much lower precipitation values especially recently. The GPI precipitation is again observed to be high. The TRMM precipitation datasets are aligned with all the other datasets. It should be noted that the TRMM 3A46 dataset contains errors for this control volume and is not considered.

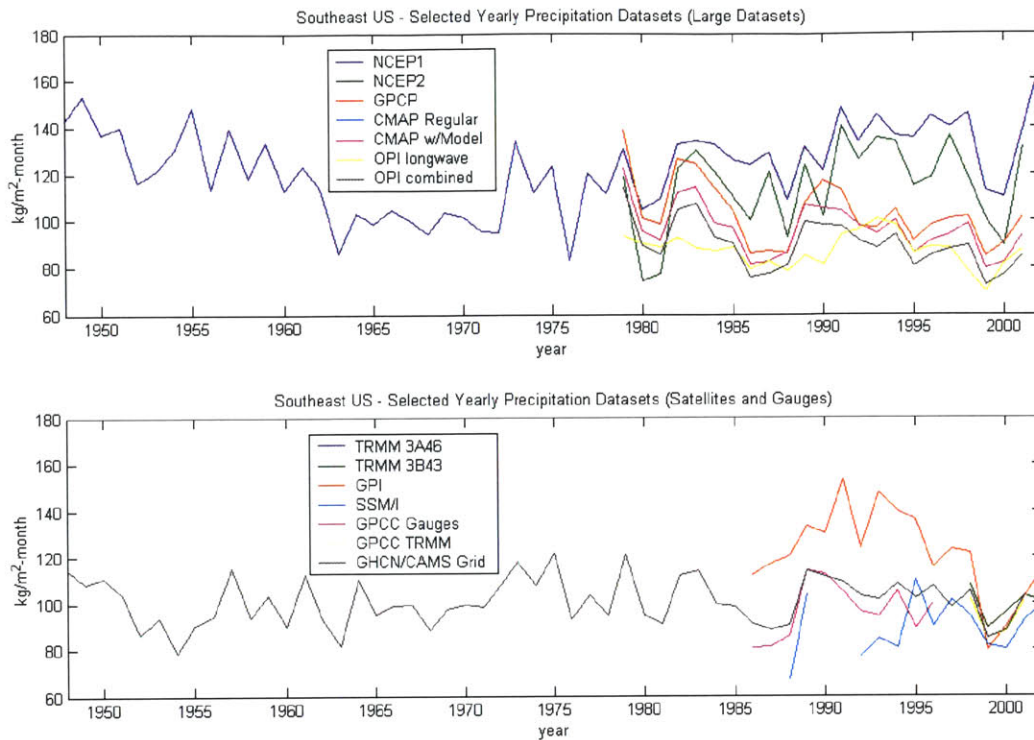


Figure 5-38: Southeast US Different Sources of Yearly Precipitation

Figure 5-39 shows the monthly water balances for the Reanalysis-1 in the Southeast United States control volume. In the monthly balance a generally good correlation is found between the two sides of both the atmospheric and surface the water balances. The seasonal cycles are not as strong for this control volume as they were in the control volumes of this size in South America. A correlation coefficient of .5553 is found for the atmospheric balance, with an equivalent bias to the yearly balance. The surface balance and the atmospheric balance seem to have similar regressions for the Reanalysis-1, as one balance is not clearly better than the other. This is encouraging because for most control volumes only one of the two balances can be computed well. A lower correlation coefficient of .2360 is found for the surface balance but signals are represented well. The correlation plot in this case does not indicate such a tight correlation.

The result for Reanalysis-2 in Figure 5-40 shows that the atmospheric balance is slightly more spread, with a smaller bias and a correlation coefficient of .4011. This water balance is inconsistent, as seen in bottom panel, where for some months there is no balance, while other months the balance is near perfect. For the surface balance, there is great improvement from Reanalysis-1 to Reanalysis-2. A good correlation is found between the two sides of the surface water balance, indicating the default linear runoff methodology works well for this control volume. This balance has a correlation coefficient here of .8228, and though there is a slight bias, the balance is well calculated and among the best by the evaluation criteria.

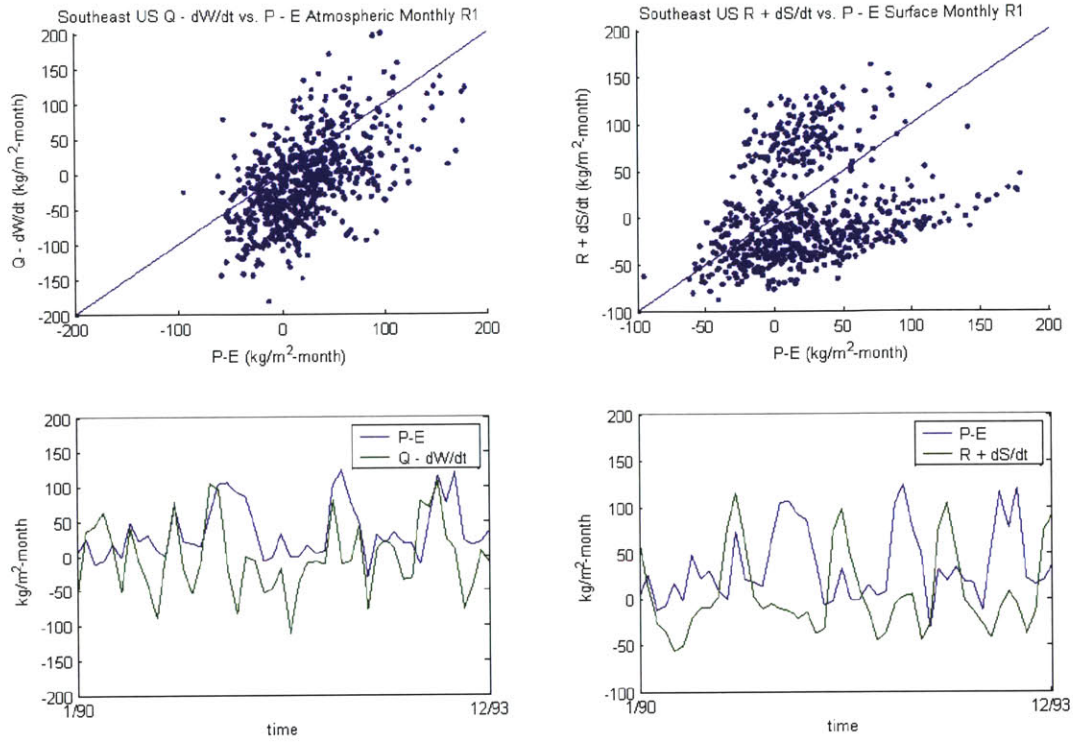


Figure 5-39: Southeast US Reanalysis-1 Monthly Atmospheric and Surface Balances

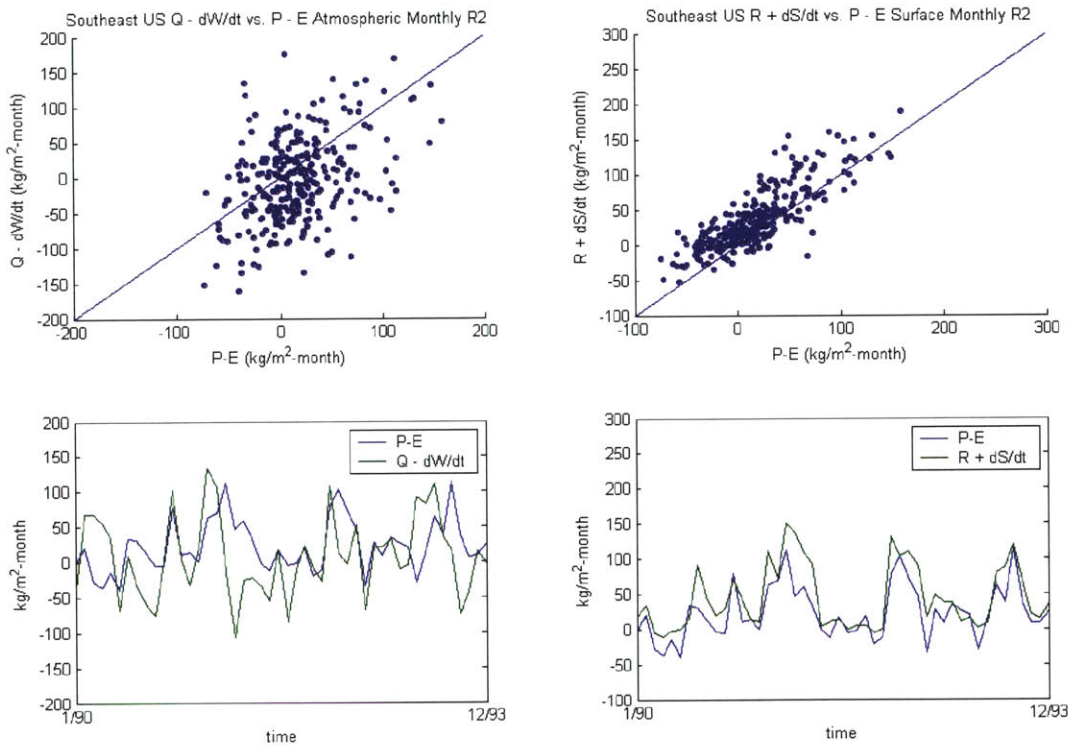


Figure 5-40: Southeast US Reanalysis-2 Monthly Atmospheric and Surface Balances

Table 5-2 shows the characteristics that can be changed in the different water balances of the SEUS control volume. The only difference between this control volume, and the control volumes represented in Table 5-1, is the elimination of the TRMM 3A46 precipitation dataset. For each of the yearly water balances there is significant room for improvement from the reanalysis. For the atmospheric balance, the bias can be completely eliminated by the use of a gauged precipitation dataset, the GPCC, to the Reanalysis-1 in Figure 5-41. This balance is taken over the GPCC TRMM dataset which is ranked highest by the criteria, because it contains more than the four years available in the GPCC TRMM product. A correlation coefficient of .4331 is found here, but the balance is unbiased ( $-1.65 \text{ kg/m}^2\text{-month}$ ) and the residual is fairly inactive. The surface balance can be improved by a change in methodology, rather than precipitation. Here storage is eliminated and the area runoff method is applied to the Reanalysis-2. This results in a very good yearly surface balance, which is rare in this study. The runoff correlates well with the P-E difference yielding a flat residual and a correlation coefficient of .8215. The bias is  $8.68 \text{ kg/m}^2\text{-month}$  with an absolute error of  $9.51 \text{ kg/m}^2\text{-month}$ , indicating a fairly constant error for all points, thus good signal pick up. The storage change for the yearly balance is bad in this control volume, which is unusual as the storage usually is the only quantity holding the surface water balance together.

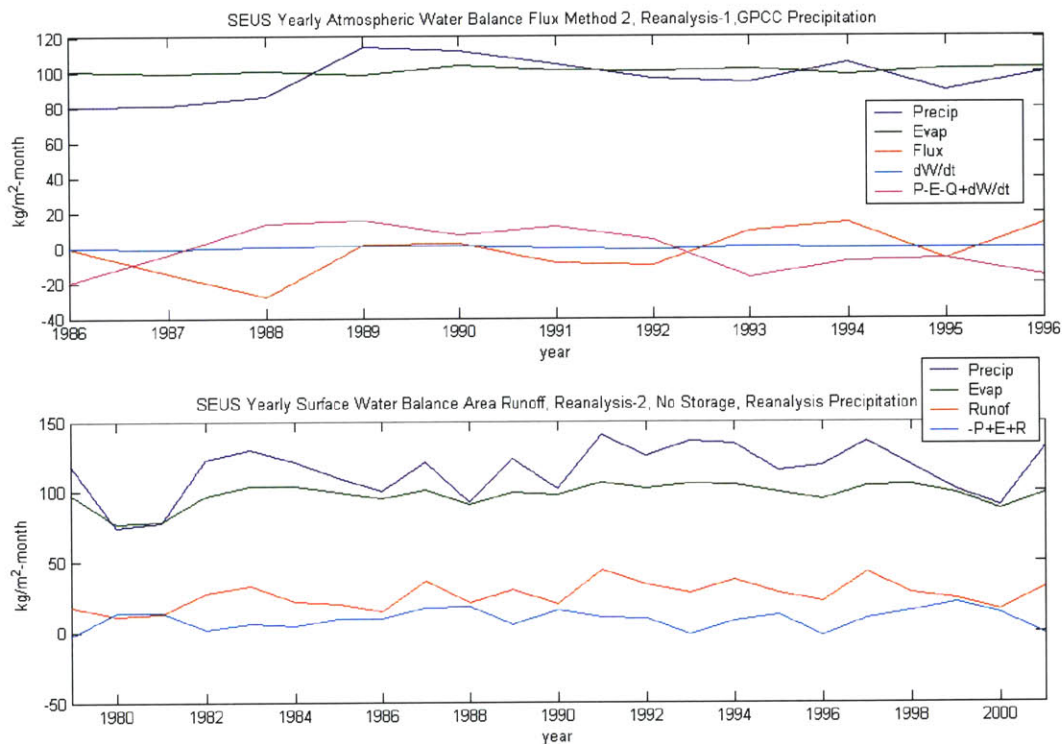


Figure 5-41: Southeast US Selected Yearly Water Balances

In Figure 5-42 improvements are made on the monthly time scale to the SEUS control volume. For the atmospheric balance the default reanalysis balances are among the best based on the evaluation criteria. Also among the best balances by the evaluation criteria are those resulting with TRMM 3B43 precipitation, among which the best of these uses the Reanalysis-2 and



moisture flux Method 2. The big difference in this balance is that the bias is eliminated (down to  $4.00 \text{ kg/m}^2\text{-month}$ ), while other water balance properties remain about the same with a correlation coefficient of .5098. The best suited water balance is passed over in this case as we are trying to use the TRMM 3B43 product as much as possible on the monthly time scale. The TRMM 3B43 precipitation dataset is consistent with the alternative precipitation datasets, and this may indicate a problem with the reanalysis.

In the case of the surface water balance we saw that the Reanalysis-2 provided a very good surface balance under default conditions. For this control volume, the balance can actually be improved by the evaluation criteria, applying the GHCN-CAMS gauge based dataset to the Reanalysis-1 and using the area runoff methodology, a 54 year and eleven month surface water balance is created which is almost as correlated and as accurate as the balance obtained previously for the Reanalysis-2. The fact that such a good monthly surface water balance can be obtained using an alternative precipitation dataset means that the side flux of the surface balance does not correlate as well with the Reanalysis-1 precipitation. This monthly surface balance is a large improvement on the Reanalysis-1 balance with a correlation of .7064 and a bias of only  $4.40 \text{ kg/m}^2\text{-month}$ .

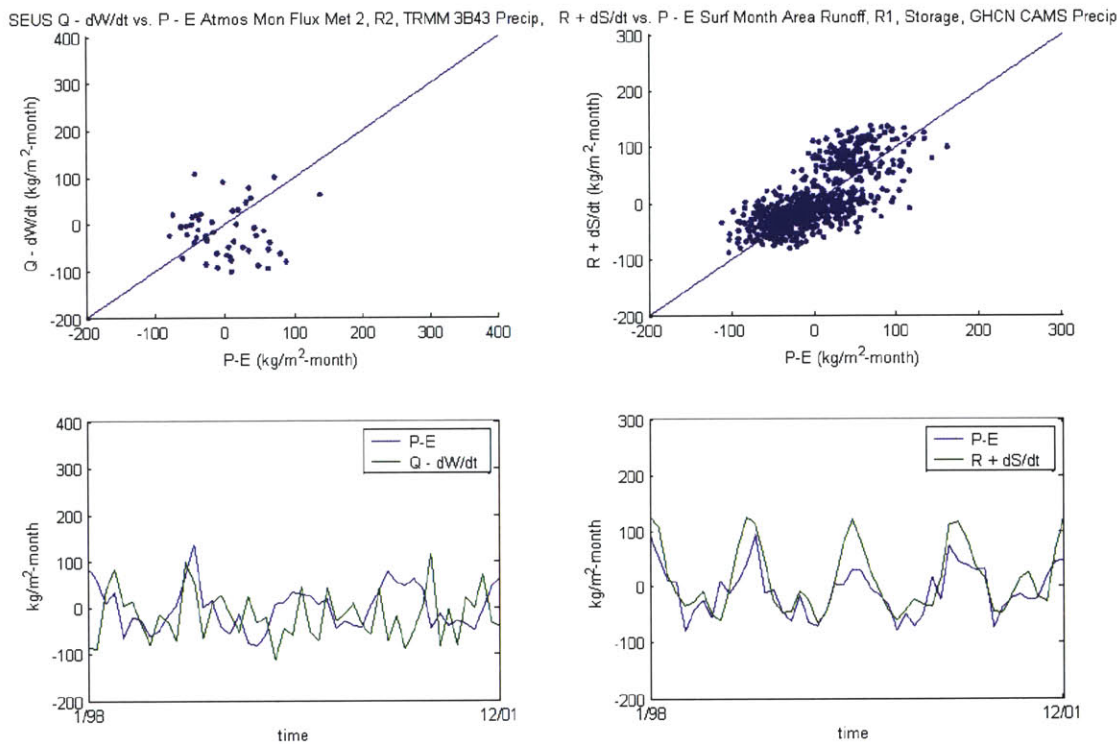


Figure 5-42: Southeast US Selected Monthly Water Balances

For the SEUS control volume the balancing of the monthly surface balance is a success. The atmospheric balance is less successful here and similar problems with the reanalysis data are found here that are found in all tropical control volumes thus far. There aren't many ways to

deal with these problems as there are a limited amount of possibilities for variations of the atmospheric water balance.

## 5.7. Southwest United States

The Southwest United States control volume is the last box sized control volume analyzed in this work. It is selected as an alternative 7.5 degree control volume with different hydrological characteristics from the other smaller box control volumes looked at so far. The Southwest United States is an arid mountainous region. However it is in the United States where accurate measurements should exist over the reanalysis time period.

Figure 5-43 shows the yearly balances for the SWUS control volume using the Reanalysis-1. In both balances we see major problems. In the atmospheric balance there are annual precipitation and evaporation of around 20 to 30 kg/m<sup>2</sup>-month. However, the moisture flux calculation is on the order of -100 kg/m<sup>2</sup>-month for each point. It is hard to tell here but there is a faint linear correlation between the P-E difference and the side flux term (mostly moisture flux on the yearly time scale), with a correlation coefficient of .1026 and a regression slope of .5483. There is clearly no balance achieved here though, as results are similar to those obtained in the Larger US control volume. The surface balance is no better. The failure of the Reanalysis-1 to pick up any runoff variation on an annual basis still exists, although this result is not wholly unrealistic given the area under consideration being largely a desert. The annual storage changes clearly govern the behavior of the residual term. A low bias is found in this balance only because the change in storage over the length of the water balance is zero.

Moving to the Reanalysis-2 in Figure 5-44 the same properties are found for the atmospheric balance. Although by the P-E data the moisture flux should be slightly negative over this region, it shouldn't be as negative as it is and the moisture flux observed is unrealistic. Any land mass would not be such a source of moisture to the atmosphere. A decent regression slope of 1.11 and correlation coefficient of .2171 are found for this control volume showing there is some correlation between the two sides of this balance, more so than there was in the Reanalysis-1, however the bias is increased by an additional 30 kg/m<sup>2</sup>-month.

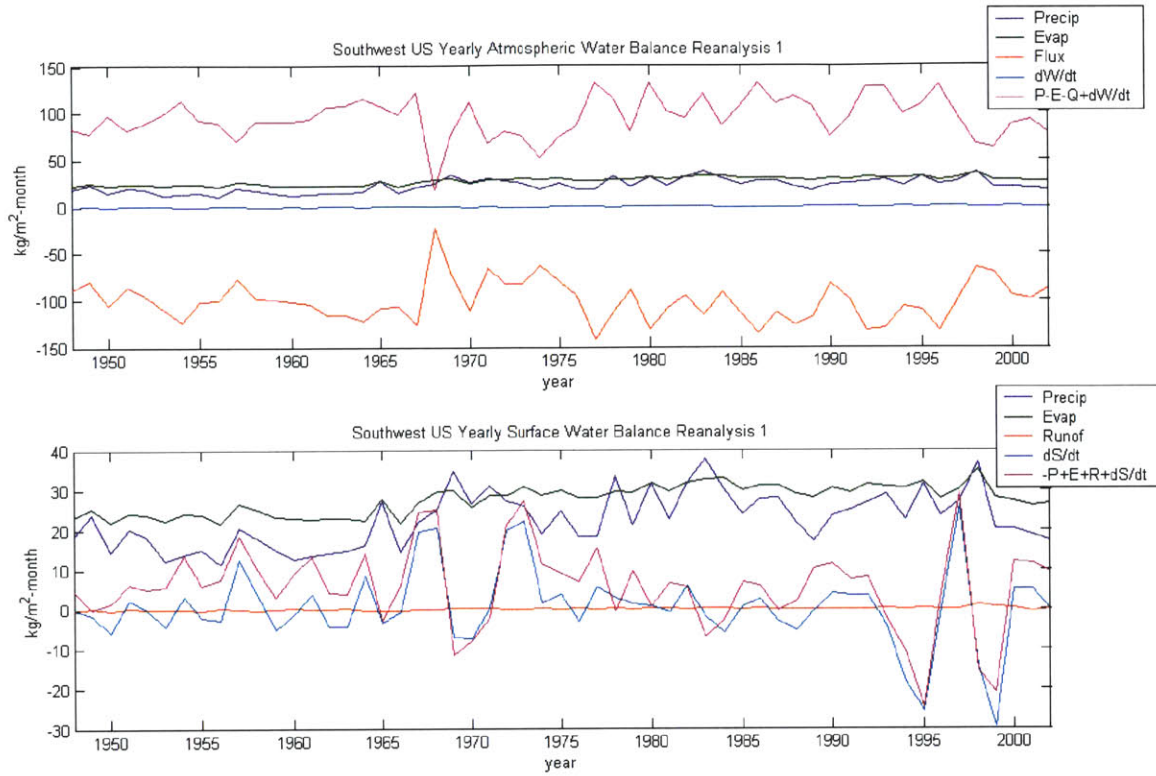


Figure 5-43: Southwest US Reanalysis-1 Yearly Atmospheric and Surface Balances

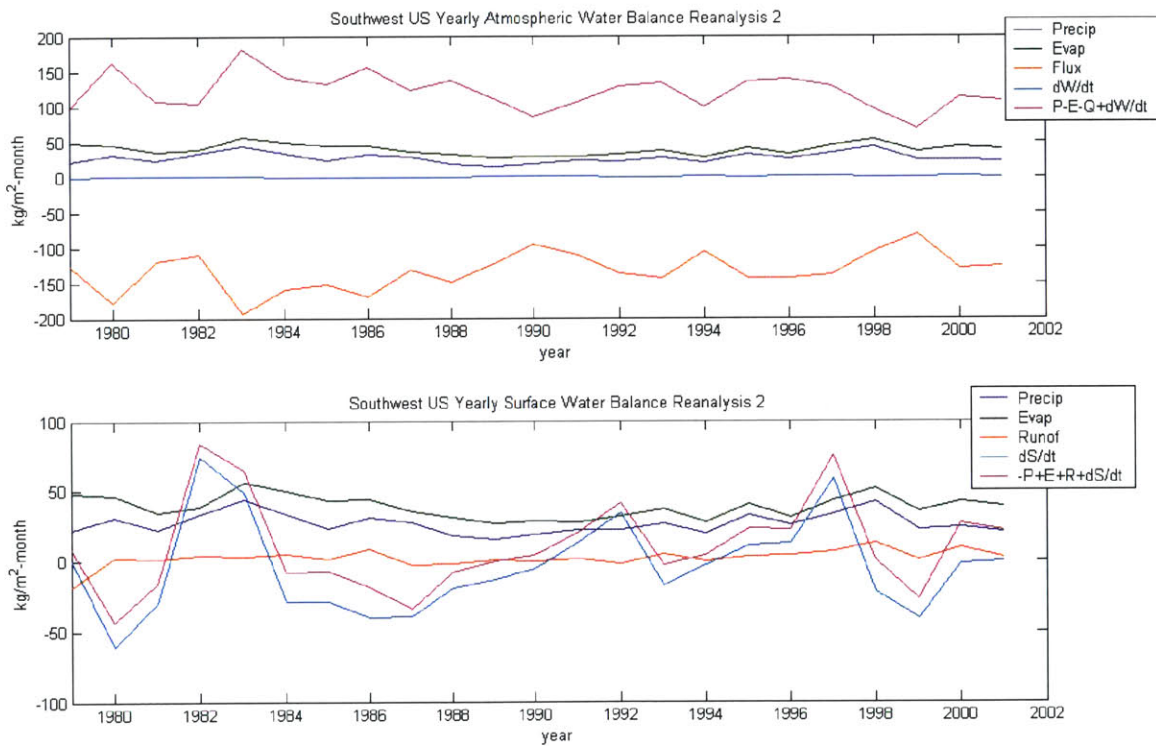


Figure 5-44: Southwest US Reanalysis-2 Yearly Atmospheric and Surface Balances

In Figure 5-45 the different sources of precipitation for the SWUS control volume are plotted as their yearly time series. The magnitude of the precipitation in the SWUS is much lower than in any other control volume looked at so far. In the top panel the larger precipitation datasets agree well on an annual basis, though variability exists which is fairly high compared to the low annual values. There again appears to be a slight shift in the Reanalysis-1 precipitation and an increase in precipitation from beginning to end. This trend can also be seen slightly in the gauge precipitation. In the annual atmospheric balance in Figure 4-129, precipitation starts out slightly below evaporation which stays constant for the time series, then gradually rises above the evaporation. Unlike in other control volumes though, there is no corresponding moisture flux trend (in other words the moisture flux matching its own reanalysis precipitation), and any trend observed would be of little meaning anyway, considering the error found in the atmospheric balance. No major differences appear in the second half of the precipitation record. It should be noted here that several precipitation datasets in the bottom portion of this graph are unavailable for this control volume. The GPI, TRMM 3B43, TRMM 3A46 do not apply here as the control volume contains points north of 40 degrees N latitude.

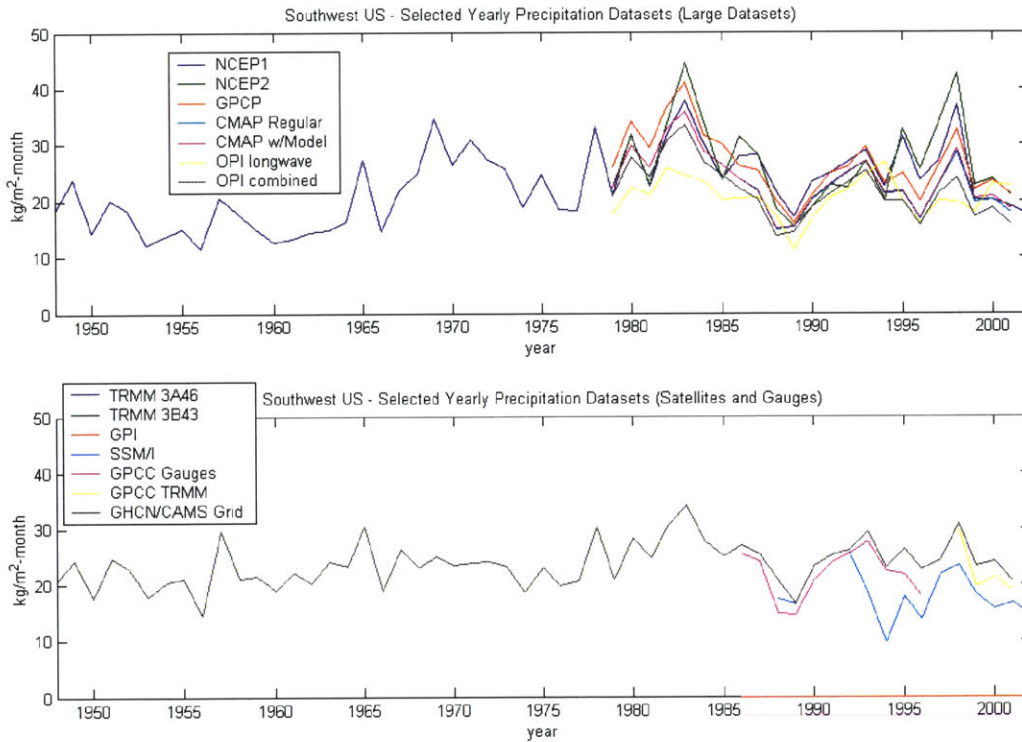


Figure 5-45: Southwest US Different Sources of Yearly Precipitation

Now considering the default reanalysis monthly water balances, the Reanalysis-1 balances are plotted in Figure 5-46. The atmospheric balance presented on the left bears no resemblance to balance, as there is no correlation between the bottom flux and the atmospheric flux. The moisture flux cycle is wildly negative. An absolute error of 122 kg/m<sup>2</sup>-month is observed, which is a significant imbalance given the magnitude of the precipitation and evaporation terms. Meaningless regression statistics and a negative correlation coefficient are also found. The

surface balance shown on the right indicates more encouraging results. The seasonal cycle is picked up resulting in a correlation between the bottom and surface fluxes. The amplitude of the surface side flux is very high in this case, likely caused by the fluctuation in surface storage which was observed in the yearly balance. Such a large cycle is unrealistic in an arid region such as this one. Encouraging balance statistics are found here, a bias of just 4.75 kg/m<sup>2</sup>-month and a correlation coefficient of .7649 with a regression slope of 2.96.

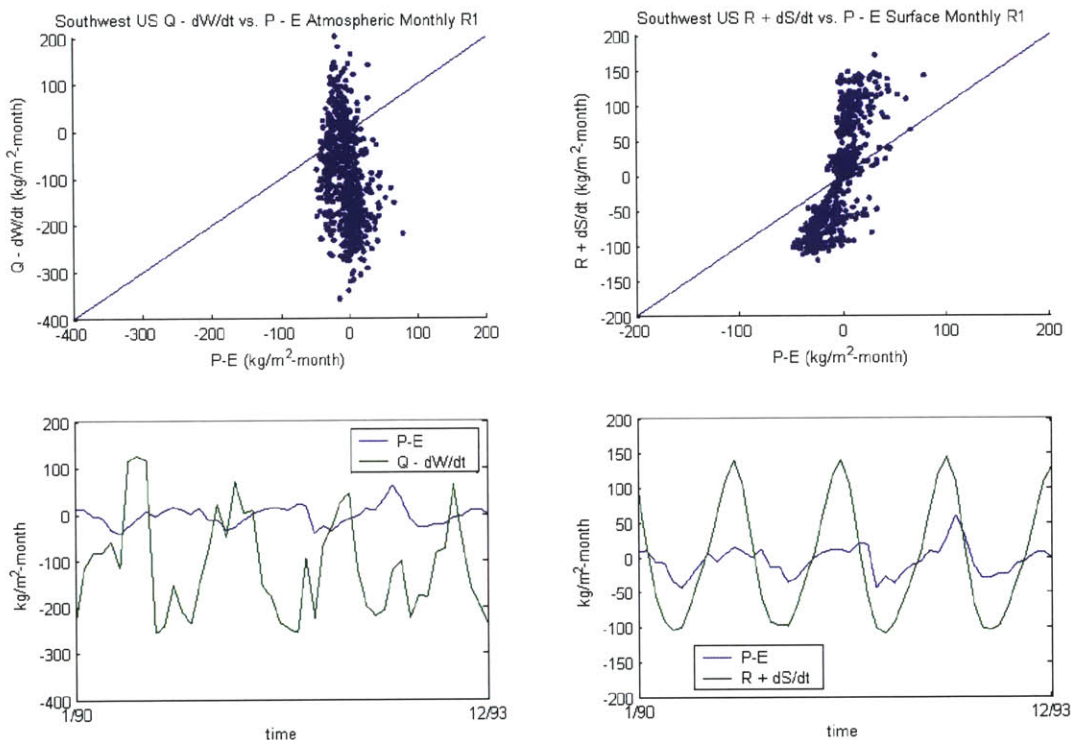


Figure 5-46: Southwest US Reanalysis-1 Monthly Atmospheric and Surface Balances

In Figure 5-47 the same balances are conducted with the Reanalysis-2 data under default conditions. Little change is observed in the atmospheric balance as the correlation presented and example plot below look nearly exactly the same as the Reanalysis-1. The absolute error in this case is increased to 141 kg/m<sup>2</sup>-month, and again unrealistic regression and correlation statistics are obtained. The interest here is the sharp improvement of the monthly surface balance between the Reanalysis-1 and the Reanalysis-2. The surface flux seasonal cycle goes from having too great amplitude to an almost correlated amplitude with the bottom flux seasonal cycle. This bias is unfortunately increased to 13.23 kg/m<sup>2</sup>-month, which is significant for the SWUS control volume, however we find a regression slope of 1.17 and a correlation coefficient of .7420.

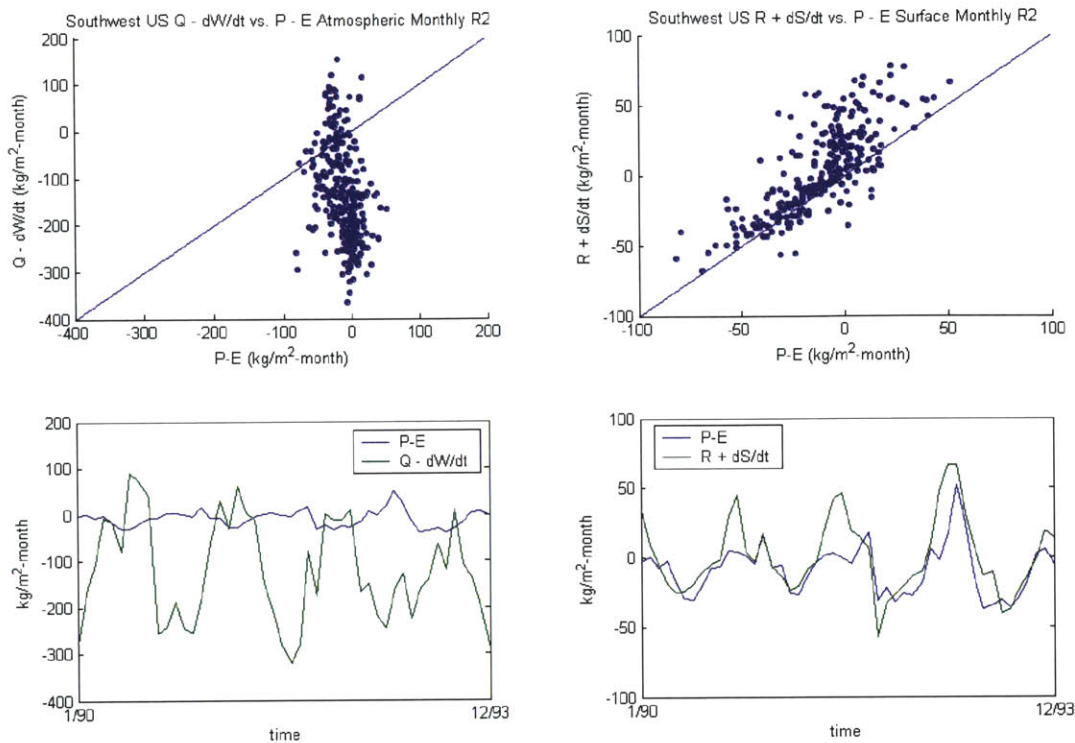


Figure 5-47: Southwest US Reanalysis-2 Monthly Atmospheric and Surface Balances

Improvements to this water balance derive from a unique amount of iterative changes in balance characteristics, as presented in Table 5-3. Three of the thirteen precipitation datasets are eliminated from consideration. No other characteristics are affected. Improving the atmospheric balance is likely an exercise in futility. Looking at the results of the evaluation, the top rated balances use the GPCC TRMM precipitation which is considered too short a time series (4 years), for a meaningful yearly balance. The next best ranked balance is unfortunately the Reanalysis-1. For comparison purposes we present the 4<sup>th</sup> highest ranked balance, applying SSM/I precipitation to the Reanalysis-1 balance, using moisture flux Method 1. A slight correlation can be visualized in the top panel (Figure 5-48) between the moisture flux and the P-E quantity as they increase slightly over the time series, but in summary the yearly atmospheric balance for this control volume can not be calculated accurately.

For the surface balance, the yearly reanalysis balances can be improved by using the evaluation criteria. Here we change to the area runoff method applied to the Reanalysis-1, with storage included, and the GPCC precipitation utilized. This scenario is selected as it is the highest ranked set of conditions which have reasonable results for each of the evaluation criteria. A bias of 3.34 kg/m<sup>2</sup>-month is obtained, an improvement over previous default conditions, and a near closure in the long term. The correlation coefficient is a weak .1913, and the inconsistencies in this water balance can be seen in the bottom portion of Figure 5-48 as the residual is still controlled by the change in storage, but to a lesser extent.

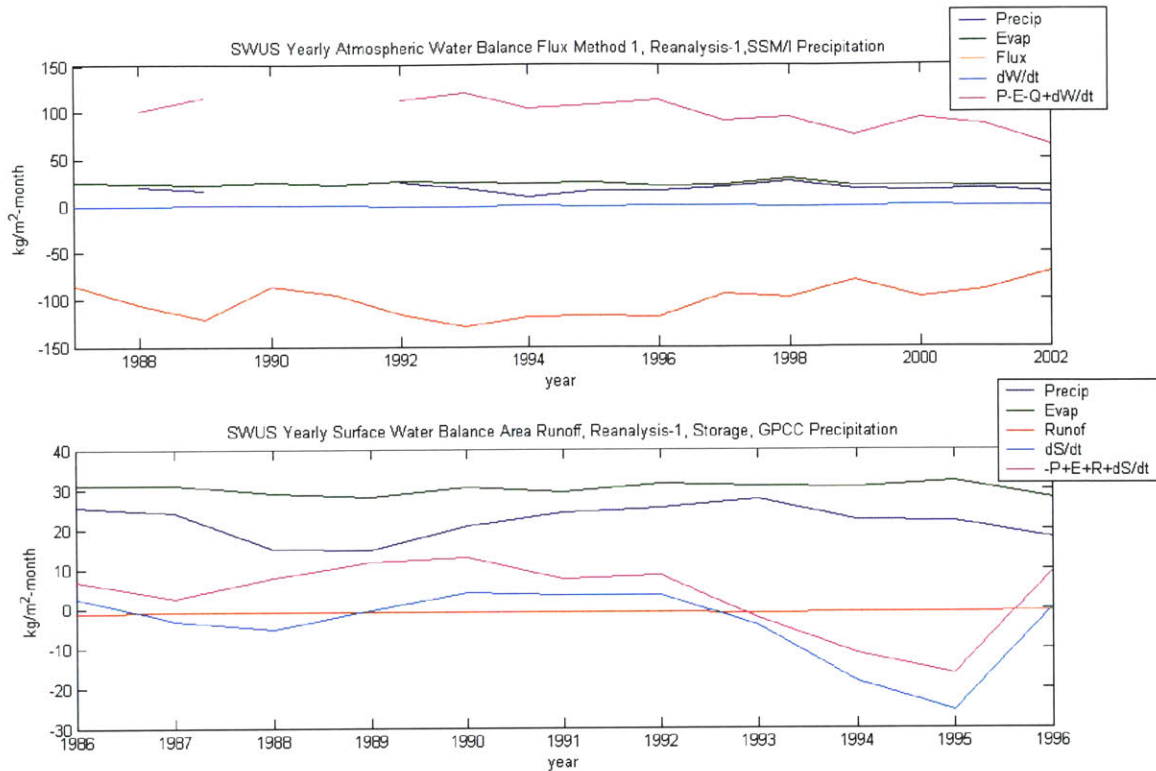


Figure 5-48: Southwest US Selected Yearly Water Balances

In Figure 5-49 improvements are presented for the monthly water balances for the SWUS control volume. For the atmospheric balance the reanalysis balances are ranked as some of the poorest by the evaluation criteria, and improvements can be made by changing to moisture flux Method 1, and using the recent gauged precipitation data set, the GPCC TRMM as applied to the Reanalysis-1. The bias of the water balance remains very high, however a regression slope of .9855 is found with a correlation of .3289. Using moisture flux methodology 2 a strong seasonal cycle was found which was out of phase and amplitude with the P-E seasonal cycle. The sample points shown on the bottom left of this plot do not have this cycle which helps make this balance better. However, with a bias of  $82 \text{ kg/m}^2\text{-month}$  being greater than either the precipitation or evaporation terms by a factor of two, we are unable to effectively close the atmospheric balance.

For the surface balance a good balance was found on the monthly time scale for the Reanalysis-2 default conditions, which ranks among the best balances by the evaluation criteria in Appendix A. This balance can be improved by changing from the linear method to area method of runoff, resulting in the top ranked monthly surface balance by the evaluation. The bias is improved to  $9.61 \text{ kg/m}^2\text{-month}$  with a lower absolute error and comparable regression statistics. The amplitude of the surface flux is reduced by the inclusion of routing data by this method, and a tight correlation results.

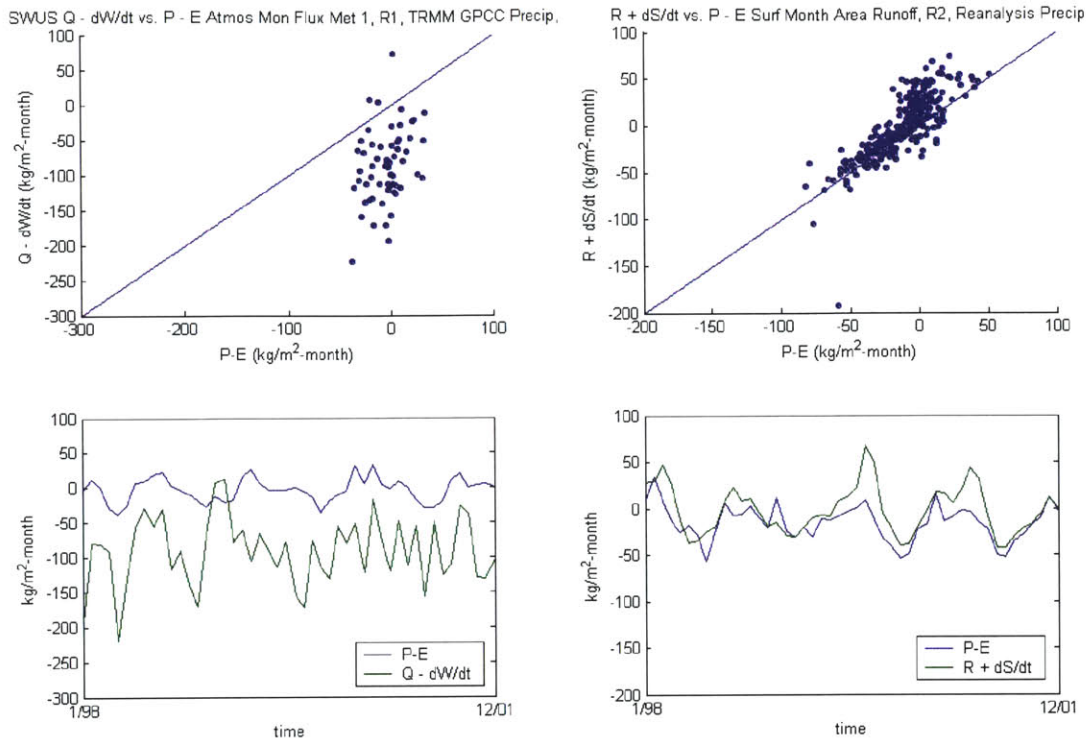


Figure 5-49: Southwest US Selected Monthly Water Balances

For the SWUS control volume, the only great success is computing the monthly surface balance. All other balances contain significant errors especially in the atmospheric balances, likely caused by the geographic features of the region. It is difficult to complete water balances in this region because of the relatively small magnitude of the side flux terms (moisture flux and runoff).

### 5.8. Amazon River Basin

The Amazon River Basin control volume is the first of four basin control volumes to be presented. This control volume is the largest considered in this study and includes several of the box control volumes which have already been discussed (Rondonia, Larger Rondonia, and Northwest Brazil). It contains very jagged edges which could present difficulties in calculating moisture flux. In the surface balance case, surface runoff data from the RIVDIS dataset is used as the default, while when improving the balance, the reanalysis runoff is used. The surface runoff limits the temporal length for the surface water balance, but not the atmospheric balance.

The default yearly balances for the Reanalysis-1 are shown in Figure 5-50. For the atmospheric balance we see signs of success. While evaporation stays constant for the most part, precipitation peaks in the early 1950's and in the second half of the balance. The peaks in the moisture flux match this behavior, but the amplitudes do not agree. The second half of the period, especially between 1975 and 1995, is relatively well balanced though year to year signals aren't picked up well. The bias of this balance is 46.43 kg/m<sup>2</sup>-month, and a good correlation coefficient of .6485 is found with a regression slope of 2.05 indicating the amplitude of moisture



flux is about double what it should be. For the surface balance we see a noticeable change from typical default balance results. Because of limitations in the surface runoff data, runoff is only available from 1970-1996, thus 27 years are represented here. The surface runoff here is much higher than was ever seen in box control volumes and the change in surface storage is no longer controlling of the balance which also was typical of the surface balance in box control volumes. As far as balance, the runoff is curiously high compared to the P-E cycle resulting in a bias of 54.96 kg/m<sup>2</sup>-month, with an identical absolute error. There is virtually no regression slope for this balance.

Considering the Reanalysis-2 in Figure 5-51 the yearly atmospheric balance looks about the same as it did in the Reanalysis-1. By the evaluation criteria, the water balance is slightly better in the Reanalysis-2 with a bias of 42.47 kg/m<sup>2</sup>-month and a greater correlation coefficient of .8590. The residual seems to be highly variable though, and the better balance is likely the result of eliminating the first half of the data. The surface balance is truncated at the beginning, by the start of the Reanalysis-2 (starts in 1979) and the end of the surface runoff data (ends in 1996). In this case the change in storage is a larger factor in the balance as the bias is reduced to 42.32 kg/m<sup>2</sup>-month and a correlation coefficient of .3186 is found, both improvements on the Reanalysis-1. The change in storage fluctuations on a year to year basis seem to be necessary in order to show a relationship between runoff and the P-E difference.

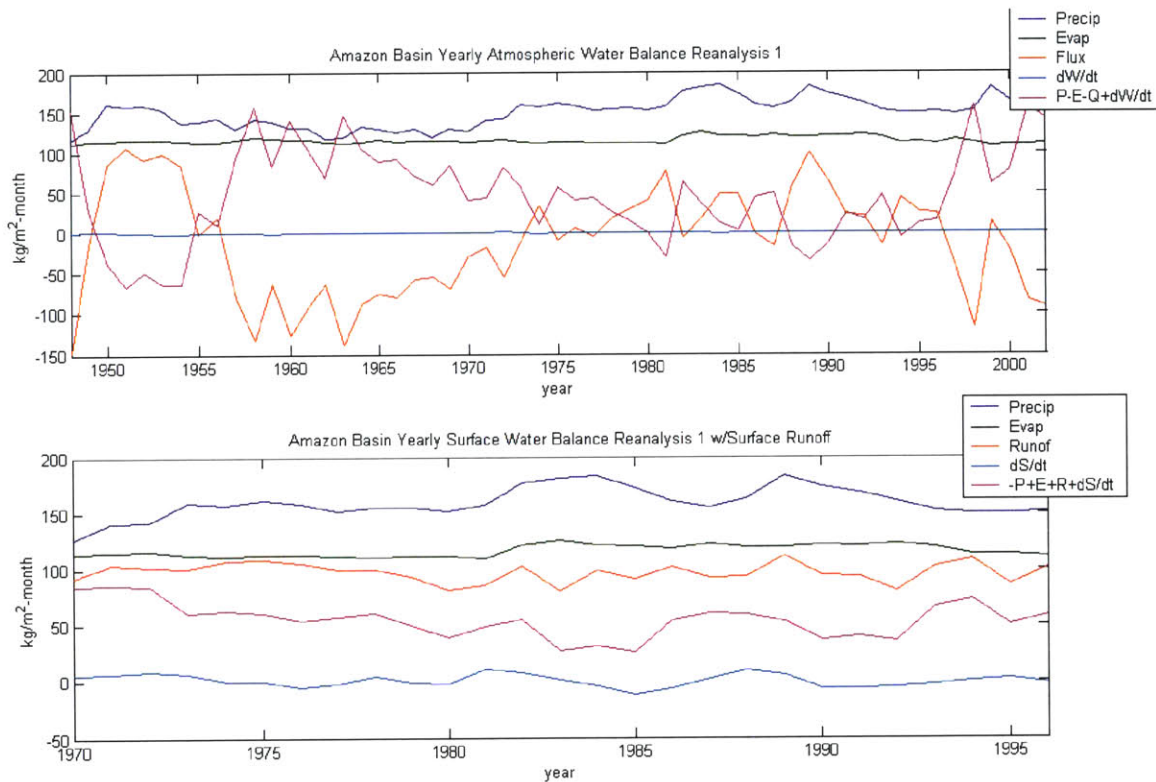


Figure 5-50: Amazon Basin Reanalysis-1 Yearly Atmospheric and Surface Balances

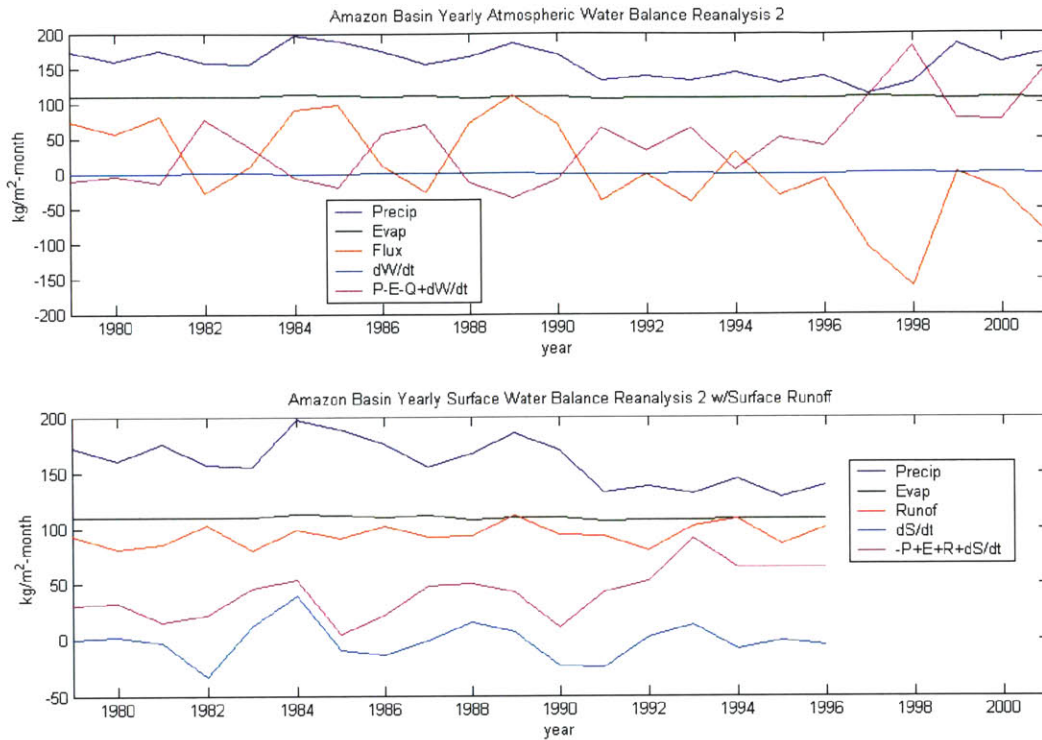


Figure 5-51: Amazon Basin Reanalysis-2 Yearly Atmospheric and Surface Balances

A closer look at the annual precipitation time series for various precipitation datasets over the Amazon Basin control volume are shown in Figure 5-52. We again see a trend repeated in all of the Amazon region control volumes. In the second half of the Reanalysis-1 time series there is an increase in annual precipitation, which is confirmed by the Reanalysis-2. Nevertheless, in the long gauge dataset and the smaller datasets no precipitation change is found. There is considerable variability among the satellite and gauge precipitation datasets which is notable. One would think such variability would be reduced in such a large control volume.

In Figure 5-53 the monthly water balances are shown for the Reanalysis-1. In both the atmospheric and surface balances, the seasonal cycle is picked up well between the P-E cycle and the side fluxes in each balance. However, other problems exist. In the atmospheric balance, a regression slope of 3.99 is found between the two sides of the balance with a correlation coefficient of .8113. This indicates the amplitude of the moisture flux seasonal cycle is way too strong. The absolute error is three times the bias indicating large errors at specific times which get averaged out over the duration of the balance. For the surface balance we see a good 1:1 correlation between the two sides of the balances, but a bias as in the yearly time scale. The absolute error and bias are nearly equal for this balance.

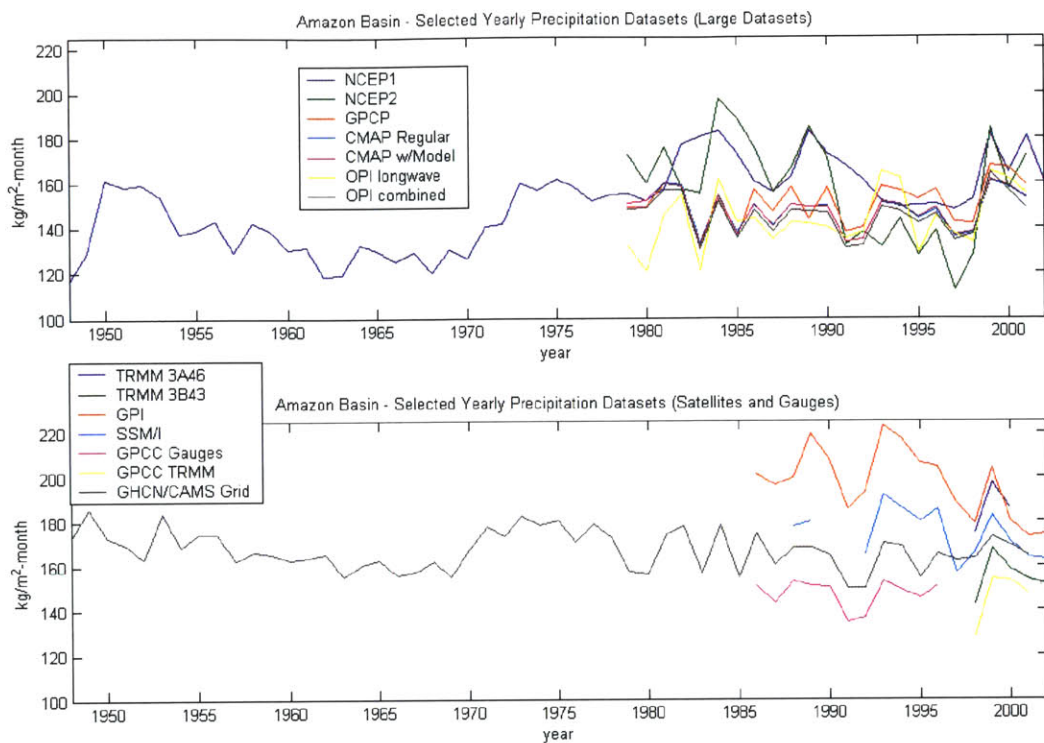


Figure 5-52: Amazon Basin Different Sources of Yearly Precipitation

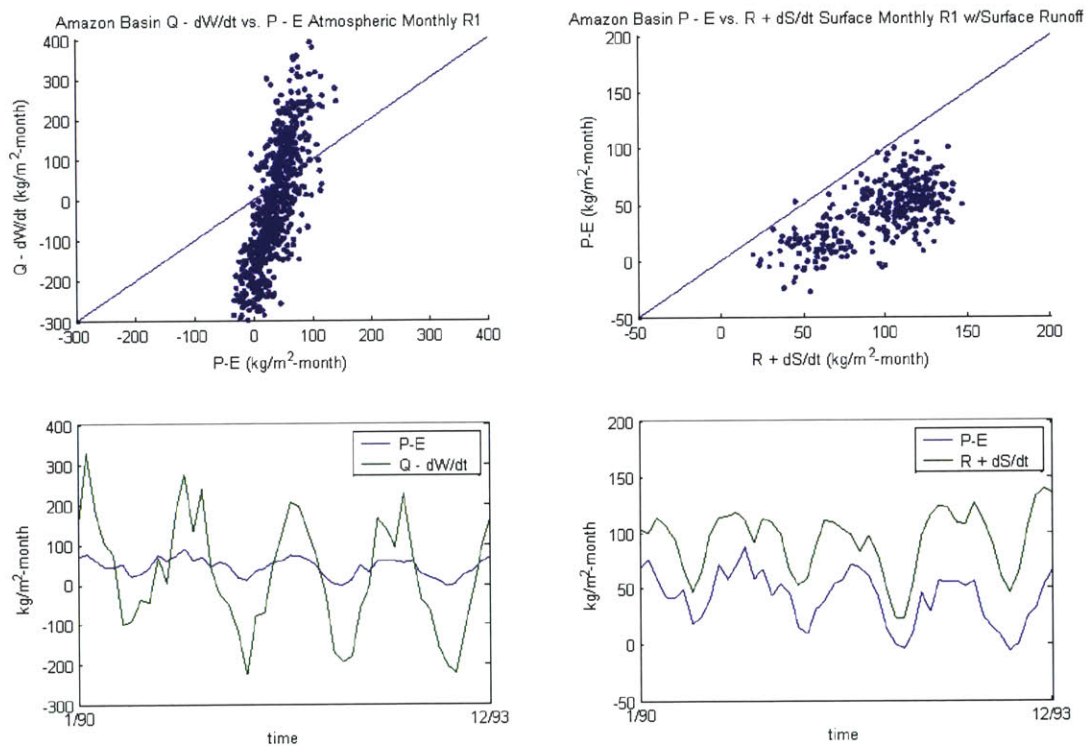


Figure 5-53: Amazon Basin Reanalysis-1 Monthly Atmospheric and Surface Balances

The Reanalysis-2 balances are shown in Figure 5-54, where many of the same observations can be made as in the Reanalysis-1 balances. In the atmospheric balance the regression slope and the correlation coefficient are improved (3.28 and .8907 respectively) as both reanalysis balances are ranked the same by the evaluation criteria. The surface balance is slightly improved in the Reanalysis-2 as the storage cycle is improved to match the surface runoff obtained with the bias being reduced by 10 kg/m<sup>2</sup>-month, but still being a significant 44.0 kg/m<sup>2</sup>-month, with an absolute error of 50.8 kg/m<sup>2</sup>-month. The seasonal cycle is captured well by all of the reanalysis balances. However this cycle is particularly strong over the Amazon River Basin and therefore should be apparent in any monthly dataset of such variables.

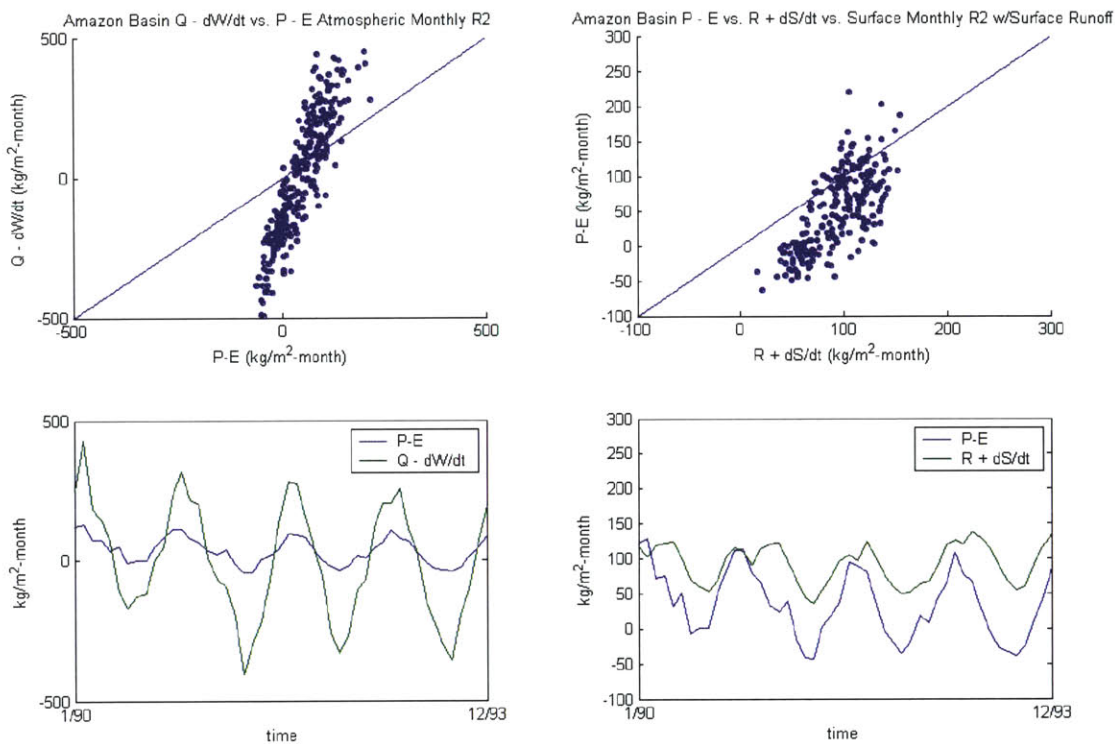


Figure 5-54: Amazon Basin Reanalysis-2 Monthly Atmospheric and Surface Balances

The Amazon and Porto Velho control volumes contain the maximum amount of scenarios for basins, with possibilities outlined in Table 5-5. For both the yearly and monthly atmospheric water balances the number of scenarios is the same as only two characteristics can be changed, the reanalysis (2) and the precipitation dataset (13). This results in 26 possibilities. For the surface balances a variety of possibilities still exist. For the monthly balance instead of three types of runoff methodologies, we in effect have two runoff methodologies, the surface runoff, or the reanalysis runoff which was derived using the linear runoff method. Because of the geometry involved with these control volumes it was deemed too computationally intensive to try every runoff methodology for each basin. There are still changes in reanalysis (2), and storage method (6), as in the case with the box control volumes. For the precipitation, we have different possibilities for each of the runoff methodologies. Using the reanalysis runoff all precipitation datasets are valid because the balance can be calculated through 2002. However, in

the case of surface runoff the TRMM precipitation datasets are useless since runoff data is only available through 1996, thus only 10 precipitation possibilities exist in this case. The result then is 276 different scenarios for the monthly balance. For the yearly balance 92 possibilities are available once lags are not considered.

Improvements to the yearly water balances are presented in Figure 5-55. For the atmospheric balance the criteria indicate the most improvement to the balance can be made by applying the CMAP precipitation dataset to the Reanalysis-1. By this application the bias is reduced to 17.0 kg/m<sup>2</sup>-month though the correlation coefficient stays low at .2676. Looking at the various scenarios for this balance good correlations tend to correspond to scenarios with high biases. The residual signal here has less amplitude than the moisture flux indicating some correlation which is impressive considering the difficulty hypothesized in calculating moisture flux through such a control volume. The surface balance is improved by eliminating change in storage and applying GPI precipitation to the Reanalysis-2 with surface runoff. The results of this water balance are very appealing. The higher GPI precipitation matches the surface runoff, which was deemed too large in the reanalysis. Change in storage over such a large basin should be zero over a number of years, and with a bias here of less the 1 kg/m<sup>2</sup>-month we have successfully completed a water balance over the Amazon River Basin using this set of conditions. A good regression slope and correlation coefficient are also found.

For the monthly cases two new water balances are presented in Figure 5-56. For the atmospheric balance the CMAP precipitation dataset is applied to the Reanalysis-1. This balance was selected as the evaluation criteria fail to capture some of the differences in bias and slope, which were the weaknesses of the original balance. The regression slope is reduced in this balance to 1.95 and the bias reduced to 18.6 kg/m<sup>2</sup>-month. All possibilities yield a slope greater than one, as the moisture flux seasonal cycle is set, and no P-E seasonal cycle can be fit to match it. It is very likely the moisture flux seasonal cycle is wrong in this balance and can't be determined by these means available in this work. The surface balance is again improved, as in the yearly balance, by the application of a satellite based precipitation dataset. In this case the SSM/I yields higher monthly precipitation numbers that match the surface runoff. A bias of 24.36 kg/m<sup>2</sup>-month and a correlation coefficient of .7516 are obtained which improve the balance. Several points in the scatter plot of this balance are outliers. These points correspond to the holes in the SSM/I data and these points are removed when the data analysis is conducted.

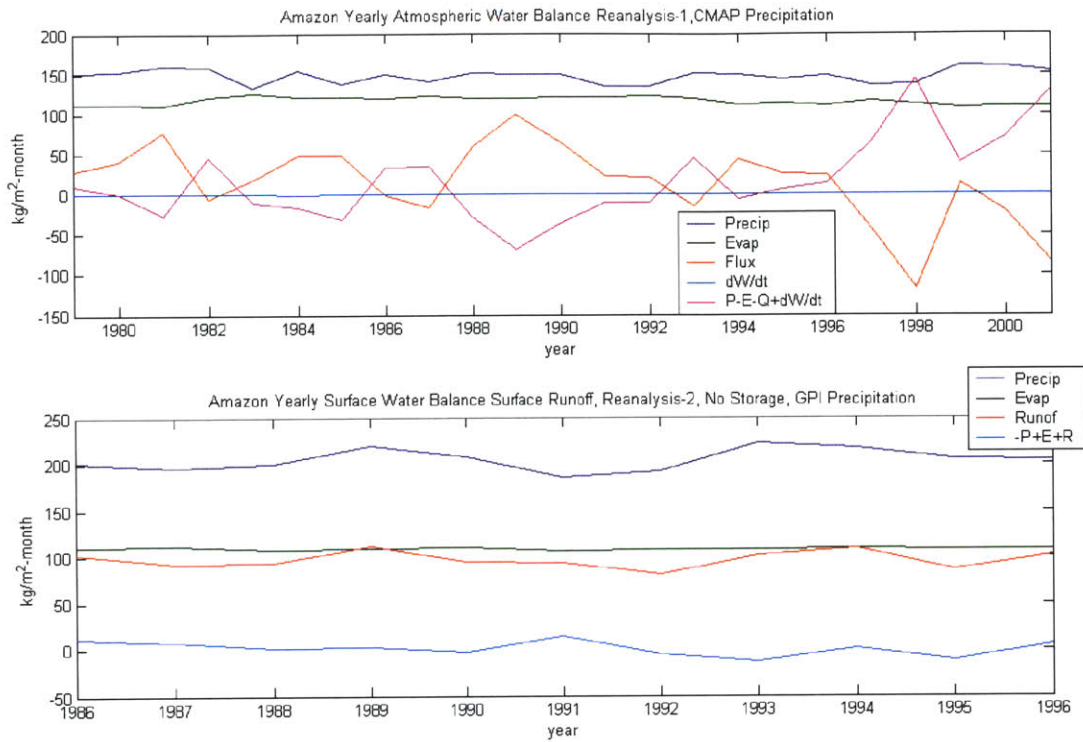


Figure 5-55: Amazon Basin Selected Yearly Water Balances

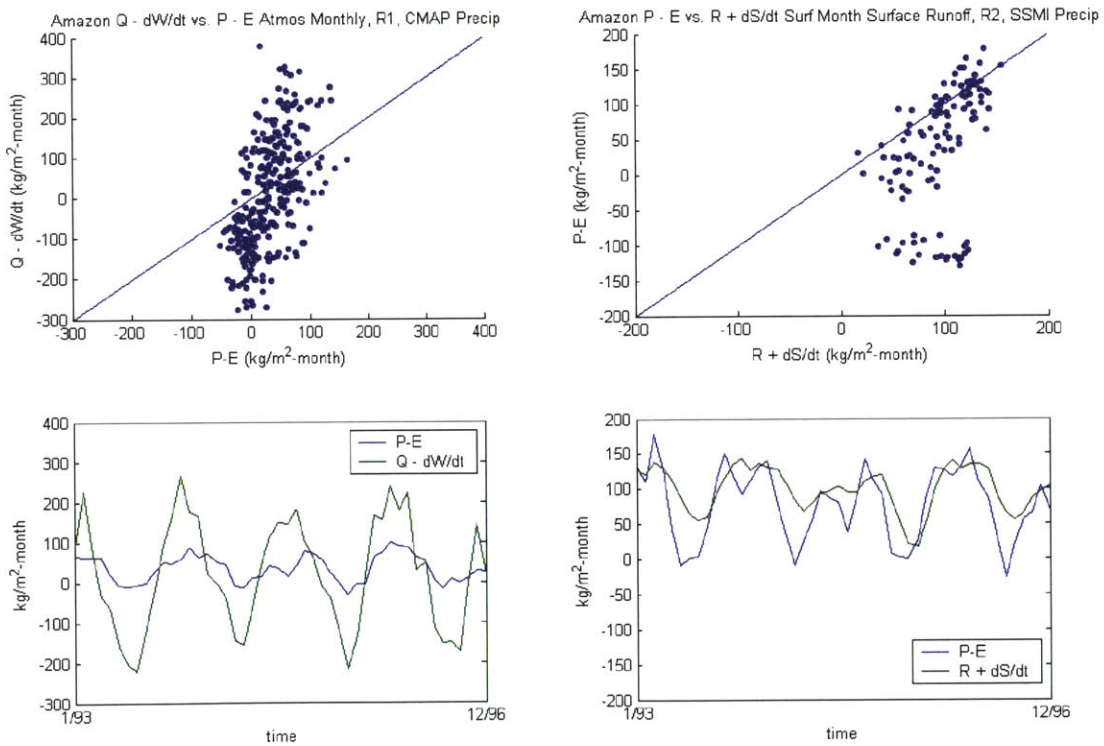


Figure 5-56: Amazon Basin Selected Monthly Water Balances

For the Amazon Basin it is difficult to close the atmospheric water balance on a consistent basis so the residuals do not change depending on the year or especially the month. It is unclear whether the relationships that do exist are based on the strong seasonal cycle of the data, or the actual correlation of the data. Biased surface balances are found in the reanalysis; however these balances can be improved especially in the yearly time scale with alternative precipitation sources. Such a surface balance utilizes only evaporation and change in storage from the reanalysis.

## 5.9. Porto Velho River Basin

The Porto Velho basin control volume is a sub basin of the Amazon Basin control volume comprised of the upper reaches of the Madeira River. A gauging station of this river is located in Porto Velho, Brazil in the northern portion of Rondonia. Therefore much of the Porto Velho control volume covers the same area as the Rondonia control volume. The Porto Velho control volume is only about a 4<sup>th</sup> of the size of the Amazon basin and extends into the Andes Mountains. The length of record constraints apply to the runoff data as in the Amazon River Basin, as the RIVDIS is again used.

The default Reanalysis-1 yearly balances are shown in Figure 5-57. In the atmospheric balance some of the same properties are found as in the Amazon basin. The moisture flux signal correlates well with the precipitation signal; however the amplitude of the moisture flux is much too strong on a year to year basis. This relationship breaks down in the second half of the balance. As in the rest of South America, precipitation seems to increase with time for this control volume. The signal in the two sides of this balance is picked up well as the correlation coefficient of this balance is .7538, however the regression slope is a high 3.73, with a RMSE of 118.2 kg/m<sup>2</sup>-month. For the surface balance, the runoff is again high compared to differences in precipitation and evaporation, resulting in a significant residual. The surface balance seems a more accurate way to assess the water balance as errors are less significant. The surface runoff data should match the P-E cycle well in the long run; however this is not seen in either the Amazon or Porto Velho basins. A bias of 32.18 kg/m<sup>2</sup>-month is observed in the Reanalysis-1 surface balance with a correlation coefficient of .3174.

Moving to the Reanalysis-2 in Figure 5-58 the moisture flux time series improves, but is still wildly amplified compared to the P-E difference. The regression slope is reduced to 2.20 with a weaker correlation coefficient of .5314 and an increased RMSE of 139.4 kg/m<sup>2</sup>-month. Again, it is difficult to calculate the moisture flux for a basin sized control volume as the resolution of the boundary is finer than the resolution of the data used for the calculation. The Reanalysis-2 surface balance is a clear improvement from the Reanalysis-1. The bias is reduced to -5.71 kg/m<sup>2</sup>-month which is low compared to magnitude of values of different quantities in the water balance of this control volume. The correlation is weak here at .2941, and the residual appears to be controlled by the change in storage. Perhaps this balance can be improved by removing this storage.

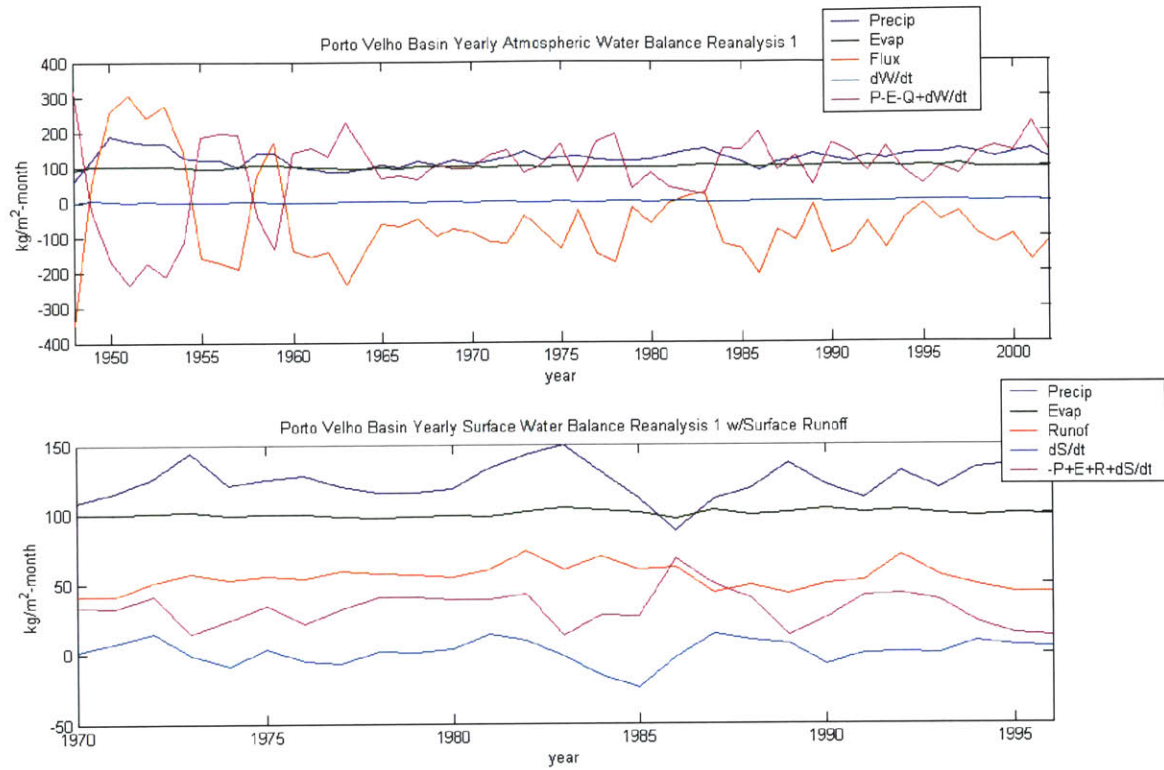


Figure 5-57: P. Velho Basin Reanalysis-1 Yearly Atmospheric and Surface Balances

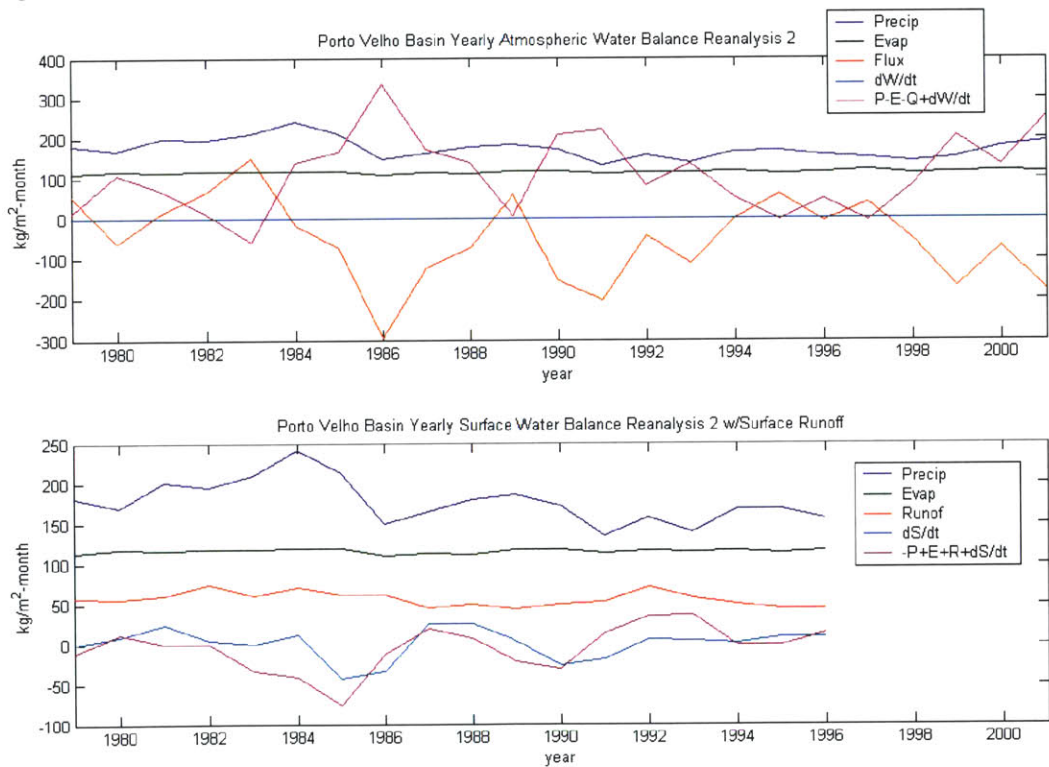


Figure 5-58: P. Velho Basin Reanalysis-2 Yearly Atmospheric and Surface Balances



The annual precipitation time series for the Porto Velho Basin are shown in Figure 5-59. As stated before we see the slight increase in the Reanalysis-1 precipitation. The trend is not observed in the GHCN-CAMS dataset, however in this case the gauge dataset is likely of not great quality given the remoteness of this control volume. Alternative large precipitation datasets continue to follow in line with the early Reanalysis. Large discrepancies are found with increased precipitation in the satellite precipitation datasets and in the Reanalysis-2. As in the case of the Amazon basin, these higher precipitation readings correspond well to the surface runoff data provided, which should be fairly accurate. There is discrepancy among satellite estimates. The GPI seems to fit the surface balance well, but the more modern TRMM 3B43 does not.

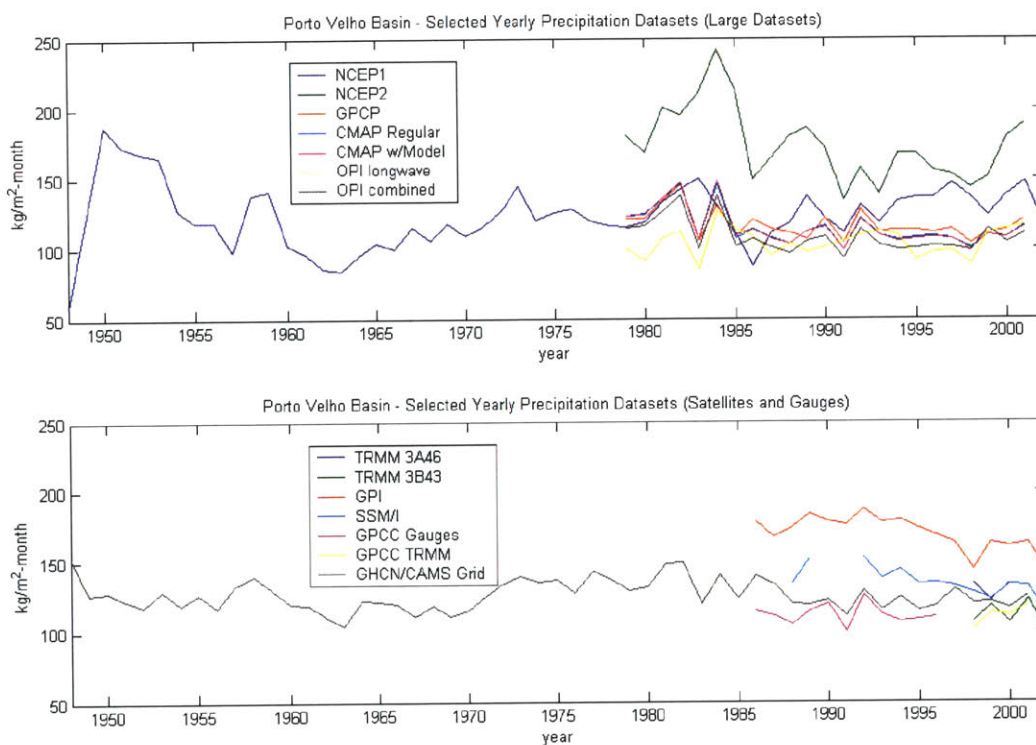


Figure 5-59: P. Velho Basin Different Sources of Yearly Precipitation

In Figure 5-60 the default monthly balances for the Reanalysis-1 are presented. In the example plot of the atmospheric balance a strong negative bias and exaggerated amplitude are seen. The regression slope of this balance is found to be 3.37 though a correlation coefficient of .6099 is found as there is good point to point correlation in this water balance. A very large RMSE is found, 205.5 kg/m<sup>2</sup>-month, for this balance. For the surface balance a similar problem arises with the side flux having a higher amplitude than the P-E cycle. Because of the strength of the seasonal cycle the correlation coefficient is found to be .7634, but with a regression slope of only .38 and a bias of 31.21 kg/m<sup>2</sup>-month.

In the Reanalysis-2 balances, shown in Figure 5-61 changes are observed in both balances. In the atmospheric balance point to point correlation is again seen with a regression slope of 2.41

(improved), but an RMSE of  $270.9 \text{ kg/m}^2\text{-month}$  with similar residual statistics. The moisture flux in the two reanalysis looks very similar, as the P-E cycle is changed in Reanalysis-2. Improvement is found in the monthly surface balance of the Reanalysis-2, it ranked highly out of the many subjected to the evaluation criteria. The bias and amplitude problems of the Reanalysis-1 are eliminated due to the amplification of the precipitation signal in Reanalysis-2, which indicates a large increase in rain during the peak rainy season in the basin. In addition the change in storage appears to help the balance on a point to point basis. Because there is improvement on both sides of the balance in this case it is hard to believe that the Reanalysis-2 precipitation is wrong, though it is very inconsistent with the larger precipitation datasets tested. A bias of  $-5.91 \text{ kg/m}^2\text{-month}$  and a correlation coefficient of  $.8736$  are observed. The absolute error found is  $40.94 \text{ kg/m}^2\text{-month}$ , which is comparable to that found in the Reanalysis-1.

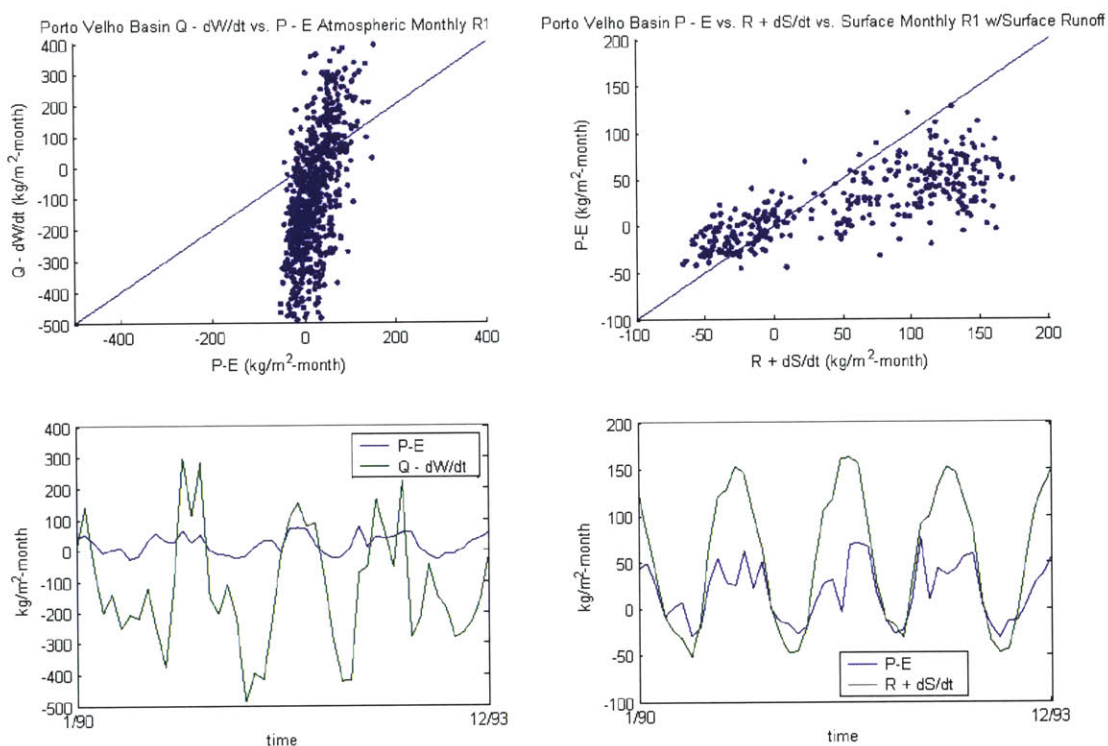


Figure 5-60: P. Velho Basin Reanalysis-1 Monthly Atmospheric and Surface Balances

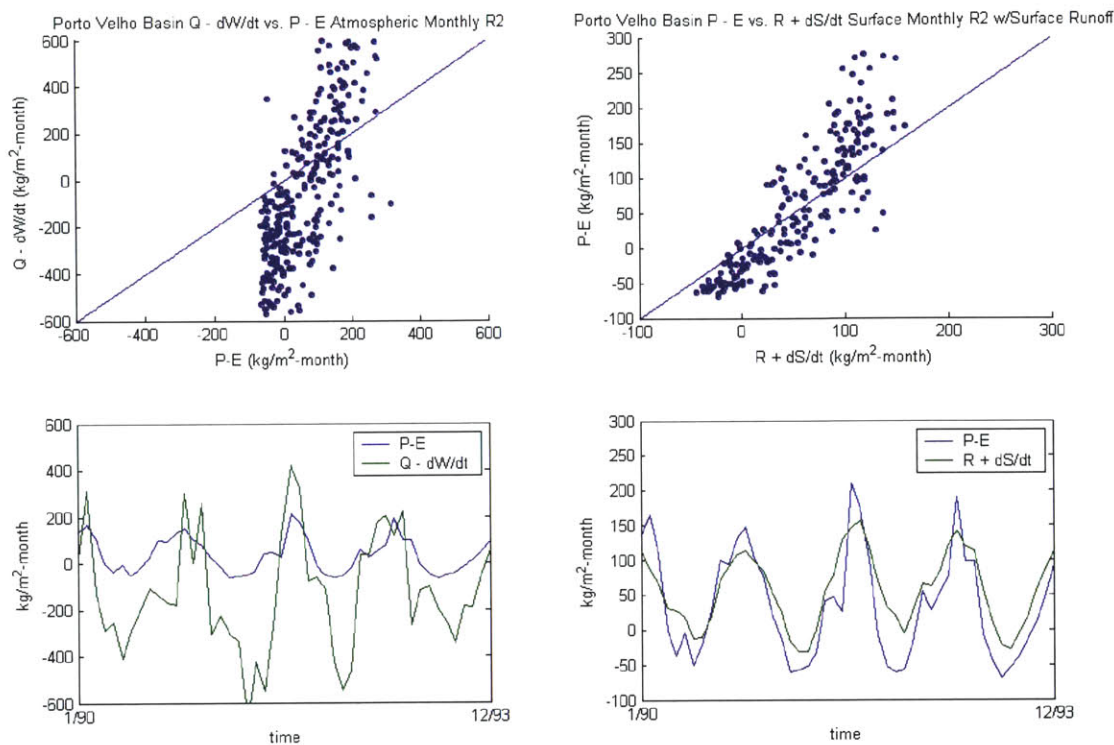


Figure 5-61: P. Velho Basin Reanalysis-2 Monthly Atmospheric and Surface Balances

The characteristics of the water balance that can be changed to improve the Porto Velho Basin control volume balance can be found in Table 5-5 and are the same as those used in the Amazon Basin control volume. In Figure 5-62 two new yearly balances are shown. In the yearly atmospheric balance, the default reanalysis balances ranked in the middle of the pack of the possible balances and the best balance was found by applying the SSM/I precipitation to the Reanalysis-2. This is unusual and a look at the plot can explain why. The SSM/I precipitation is low over this control volume and the moisture flux is biased negatively, therefore a better balance is found with the application of this precipitation, as the bias is shrunk to  $66.18 \text{ kg/m}^2\text{-month}$ . The correlation coefficient is found to be a low  $.1095$  though the balance correlates to a 1:1 slope. As anticipated the surface balance can be improved by changing the precipitation dataset to the high GPI precipitation and removing the storage term, using the Reanalysis-2 for just evaporation, as P and R are now both being found by alternative means. A bias of  $-11.43 \text{ kg/m}^2\text{-month}$  is found with a correlation coefficient of  $.7386$  and an RMSE of just  $12.73 \text{ kg/m}^2\text{-month}$ . The error is low given the magnitude of the terms involved and there is good correlation in signal observed.

Turning to the monthly balance in Figure 5-63 the default reanalysis balances again rank in the middle of the pack. The top ranked balance is found by applying the GPCP precipitation to the Reanalysis-1. The GPCP has a much stronger seasonal cycle than the reanalysis precipitation datasets resulting in a regression slope of only  $1.86$  with a correlation coefficient still fairly high at  $.5107$ . Unfortunately the RMSE and absolute error are still over  $200 \text{ kg/m}^2\text{-month}$  with no significant improvement in bias. For the surface balance small improvements are seen especially

in terms of absolute error (now 31.44 kg/m<sup>2</sup>-month) and regression slope (.81) when the SSM/I precipitation is applied to the Reanalysis-1. The seasonal cycles match very well and more importantly the results look realistic. Again outliers exist on this correlation plot due to errors which were not considered in the statistical analysis.

The Porto Velho control volume balanced in a manner similar to the of the Amazon control volume. A much too strong amplitude was found on both time scales in the moisture flux calculations which generally offset the results. It will be tested in the next two sections whether the strong seasonal cycle is the only reason for water balance correlation in a basin as seasonal cycles are not as prevalent in the other two basin control volumes. The surface balances found over this control volume were impressive, however, the precipitation datasets used to obtain these balances are under scrutiny as the GPI, Reanalysis-2, and SSM/I precipitation datasets all indicate above average precipitation.

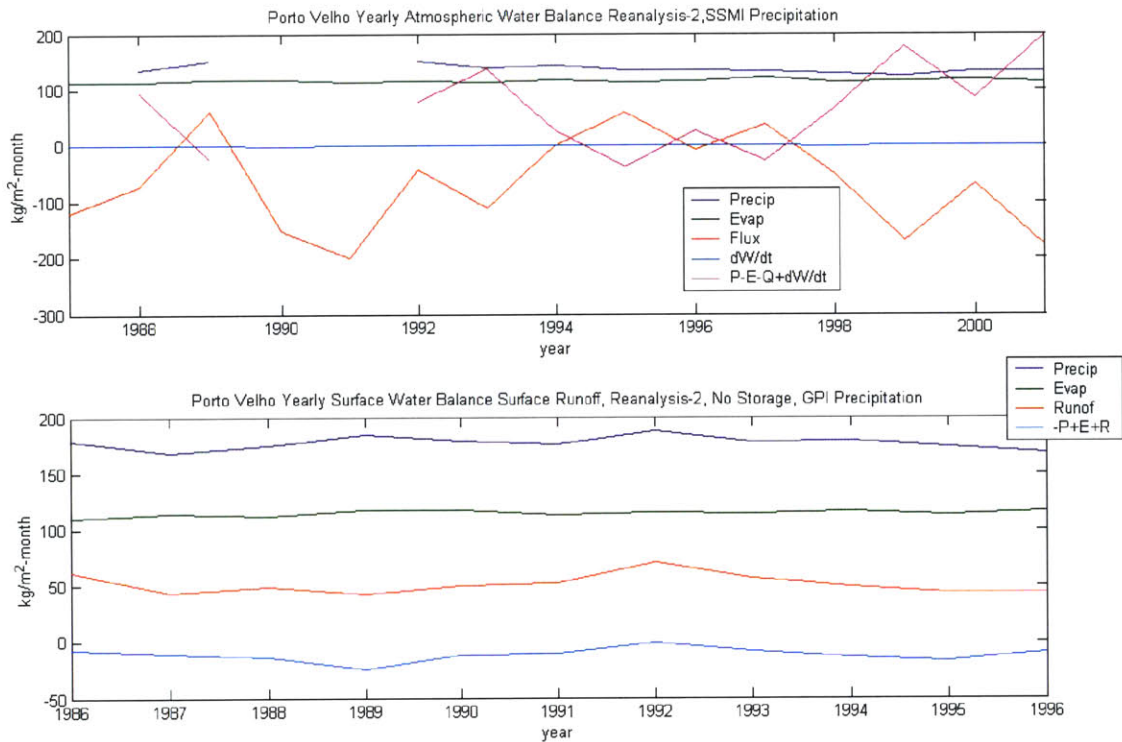


Figure 5-62: P. Velho Basin Selected Yearly Water Balances

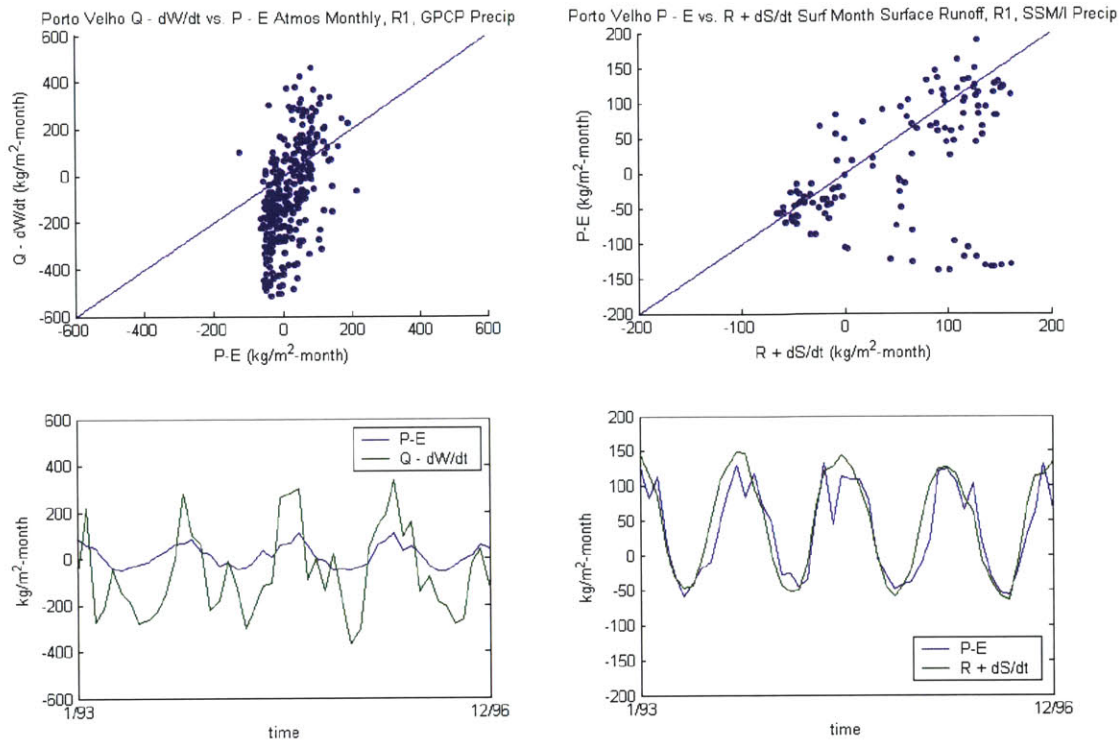


Figure 5-63: P. Velho Basin Selected Monthly Water Balances

## 5.10. Mississippi River Basin

The Mississippi River basin control volume was selected for this work as it is a large basin control volume in the general area and of comparable size to the Larger US control volume. It is located in the United States and Canada therefore more accurate data should be available for the balance in this control volume. It does not share the tropical properties of the balances in South America which have balanced better than control volumes in this region thus far. As with the previous two basins and in general throughout this work for the basin control volumes, surface runoff is used as the default although reanalysis data is added as an alternative to try and improve the balance. In this case and for the Mississippi River basin USGS runoff is used from 1948 to 1996.

In Figure 5-64 the yearly water balances are presented for the Reanalysis-1. The results of neither balance are encouraging. In both balances the annual evaporation typically exceeds the precipitation. This is unusual over a land mass over a long period of time. Furthermore, in the atmospheric balance, while the moisture flux for the scenario described should be near zero, the moisture flux is consistently larger than either the precipitation or evaporation. The result is a bias of  $-143 \text{ kg/m}^2\text{-month}$ . Ironically, a regression slope of 1.03 and correlation coefficient of .3605 are found with a regression intercept approximately equal to the bias. This indicates there is a signal relation in this balance between the moisture flux and the P-E difference. The surface balance yields realistic numbers for all components of the water balance, but the anomaly of precipitation being less than evaporation on an annual basis throws the balance off. The bias of

this balance is  $22.91 \text{ kg/m}^2\text{-month}$  with a correlation coefficient of .3159. The residual term in the balance appears to have as much variability as the terms, meaning a good balance is not achieved.

For the Reanalysis-2 similar results are found as shown in Figure 5-65. The precipitation and evaporation are nearly equal throughout the record. The moisture flux again doesn't correspond to the precipitation and evaporation series, and the balance would be much better off if this quantity was zero and just the change in precipitable water was considered. The bias is increased and the correlation coefficient decreases. For the surface balance the runoff alone seems to correlate well with the P-E cycle. The addition of the change in storage term from the Reanalysis-2 dictates the behavior of the entire balance, as the residual is almost completely dependent on the change in storage.

A look at the annual precipitation time series is interesting for the Mississippi Basin. Although we did not see a precipitation shift as has been the case for numerous reanalysis precipitation datasets, the precipitation for this control volume seemed lower than would be expected. Looking at Figure 5-66 most of the precipitation datasets available are consistent with the Reanalysis-1 and Reanalysis-2 precipitation datasets. In all cases 3 precipitation datasets: GPI, TRMM 3B43, and the TRMM 3A46 are not useful since they are for tropical regions. The TRMM GPCC product can be used for the atmospheric balance as it is not spatially restricted; however the duration of the record is out of the range of the surface balance.

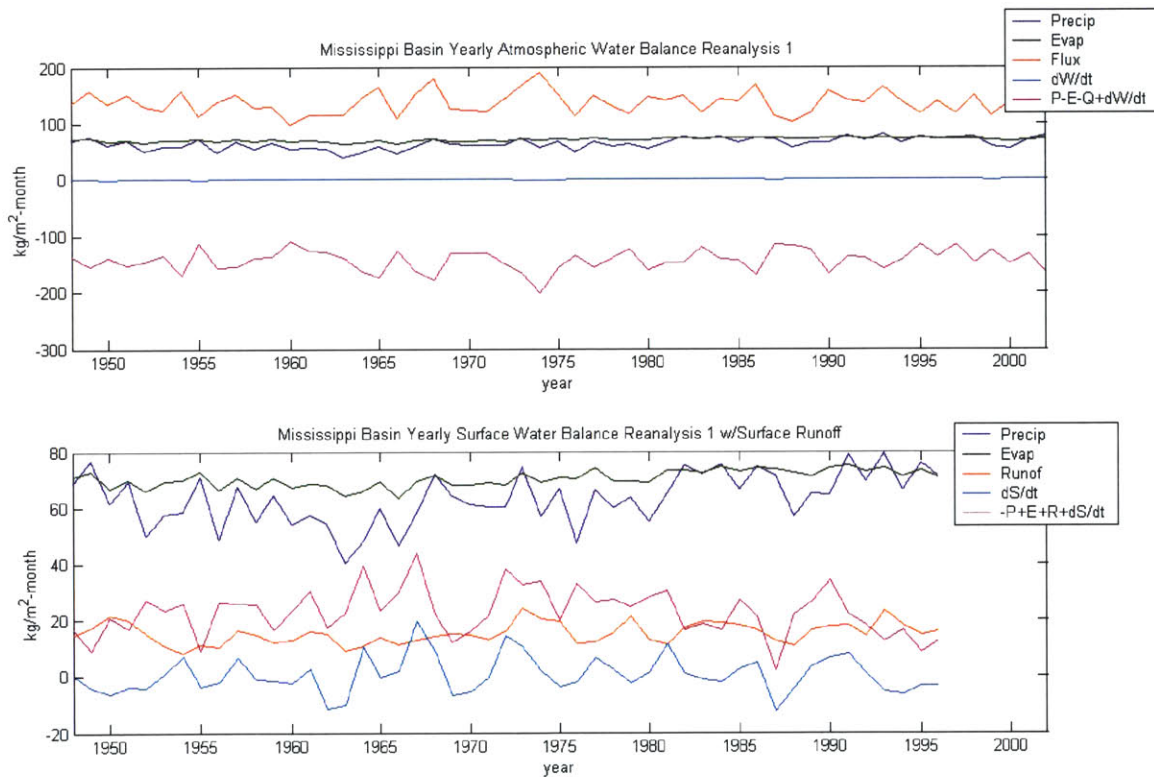


Figure 5-64: Miss. Basin Reanalysis-1 Yearly Atmospheric and Surface Balances

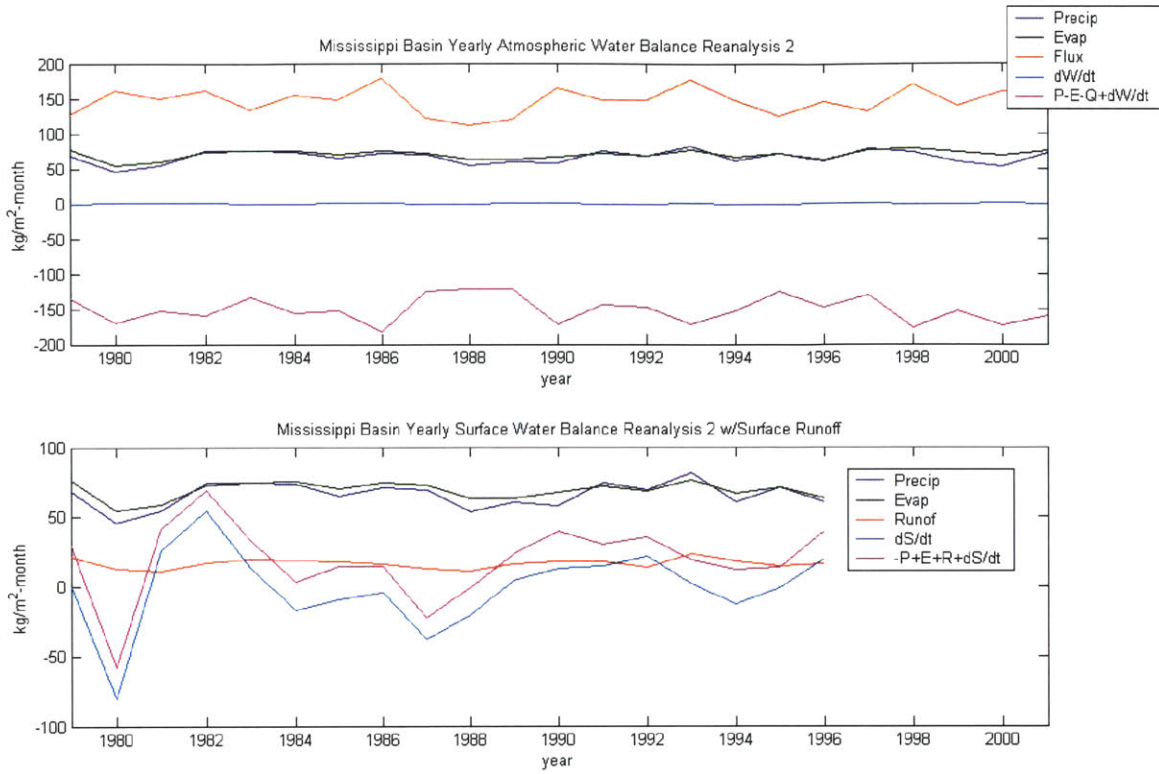


Figure 5-65: Miss. Basin Reanalysis-2 Yearly Atmospheric and Surface Balances

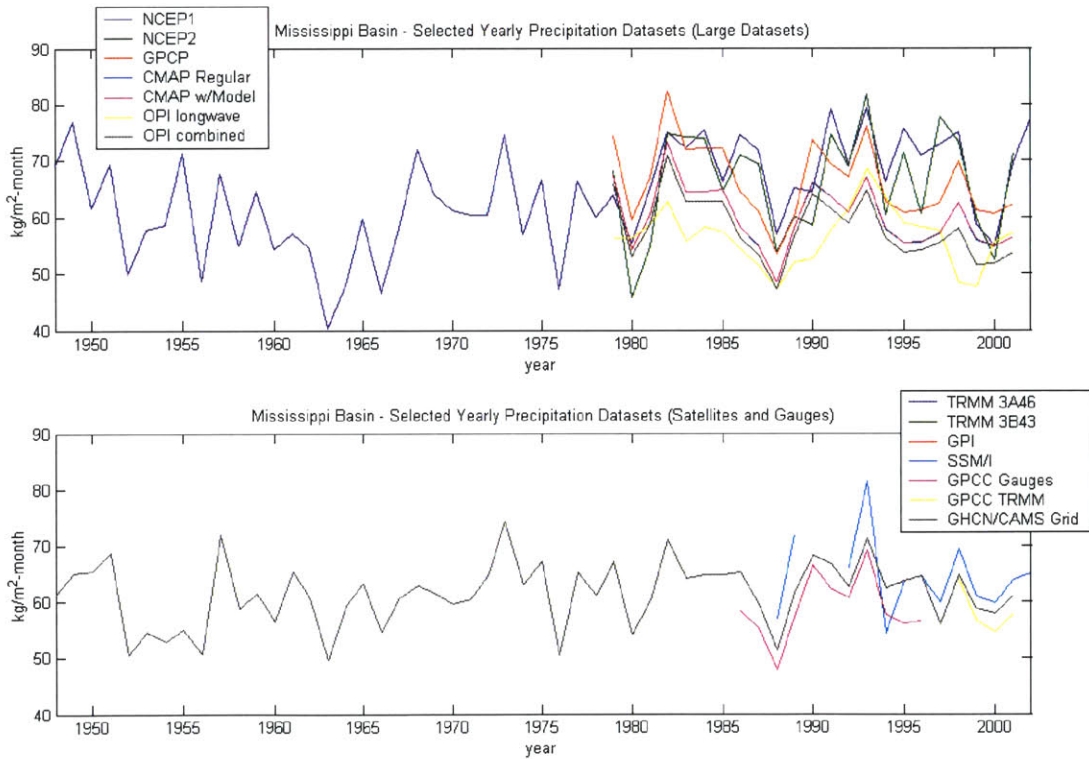


Figure 5-66: Miss. Basin Different Sources of Yearly Precipitation

Turning to the default monthly balance, the Reanalysis-1 balances are shown in Figure 5-67. As was predicted the Mississippi Basin control volume is different from the basin control volumes in the South American rain forest and thus there is no longer a strong seasonal cycle in the precipitation and evaporation cycle. As shown in the example plot, a weaker seasonal cycle still exists. There is a slight correlation between signals of the two terms of the atmospheric and surface balances which can not be seen in the upper correlation plots, but can be seen in the lower example plots. The peaks in each time series line up and a correlation coefficient of .1004 is found, but with an RMSE of 163 kg/m<sup>2</sup>-month for the atmospheric balance. In the surface balance a very large and regular divergence flux exists, which is odd. It appears that the surface storage is not flattening the seasonal cycle of the runoff as it probably should, as the runoff data is known to be fairly accurate and have a strong seasonal cycle, stronger than that of the P-E seasonal cycle. It will be interesting to see if storage considerations or other datasets can help this balance. A correlation coefficient of .5621 is found, however with an RMSE of 53.63 kg/m<sup>2</sup>-month which is on the order of magnitude of the actual quantities being balanced.

Switching to the Reanalysis-2 in Figure 5-68, the same general characteristics are observed as in the Reanalysis-1. There is slight correlation of peaks of the atmospheric balance which can only be seen on the bottom left and not in scatter plot. The RMSE is enlarged to 168 kg/m<sup>2</sup>-month and the correlation coefficient rises to .2349. In the surface balance there is improvement. The correlation coefficient remains about the same at .4948, but the RMSE is reduced to 27.46 kg/m<sup>2</sup>-month, still significant, but half of that found in the Reanalysis-1. What is odd is in the Reanalysis-2 the reanalysis change in surface storage makes the balance worse on the yearly time scale but on the monthly time scale it makes the balance better. It may be important to note that two datasets are being combined to arrive at the surface divergence flux in these balances, the surface runoff and the reanalysis change in storage term.



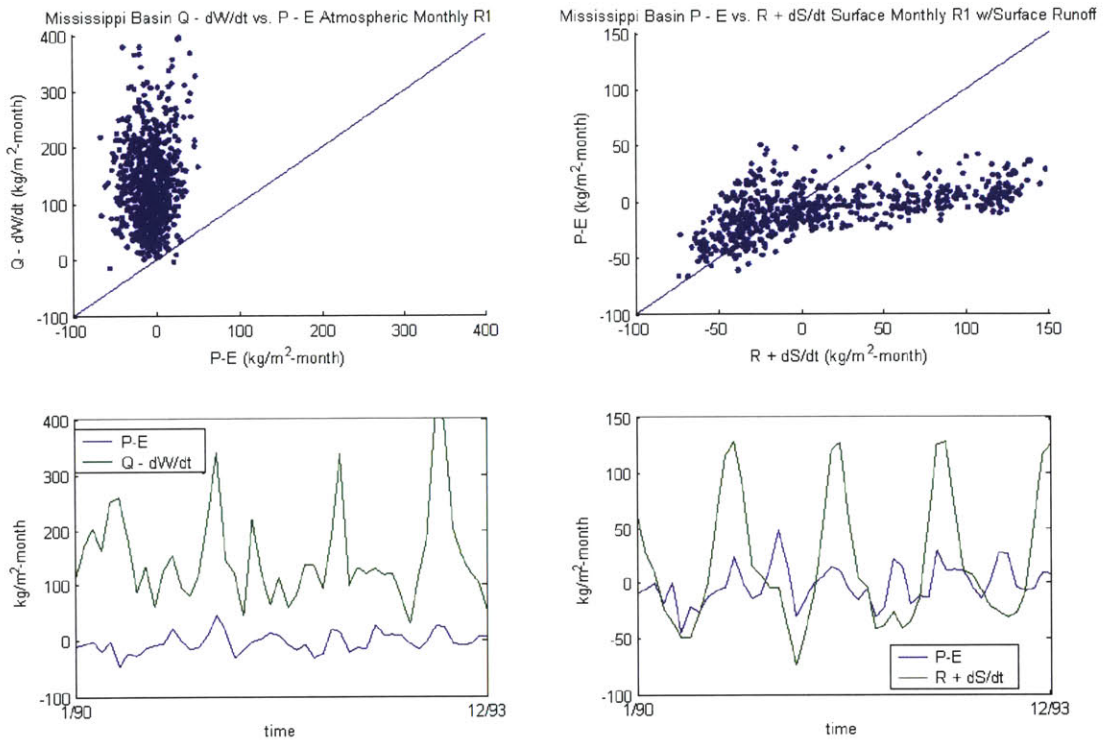


Figure 5-67: Miss. Basin Reanalysis-1 Monthly Atmospheric and Surface Balances

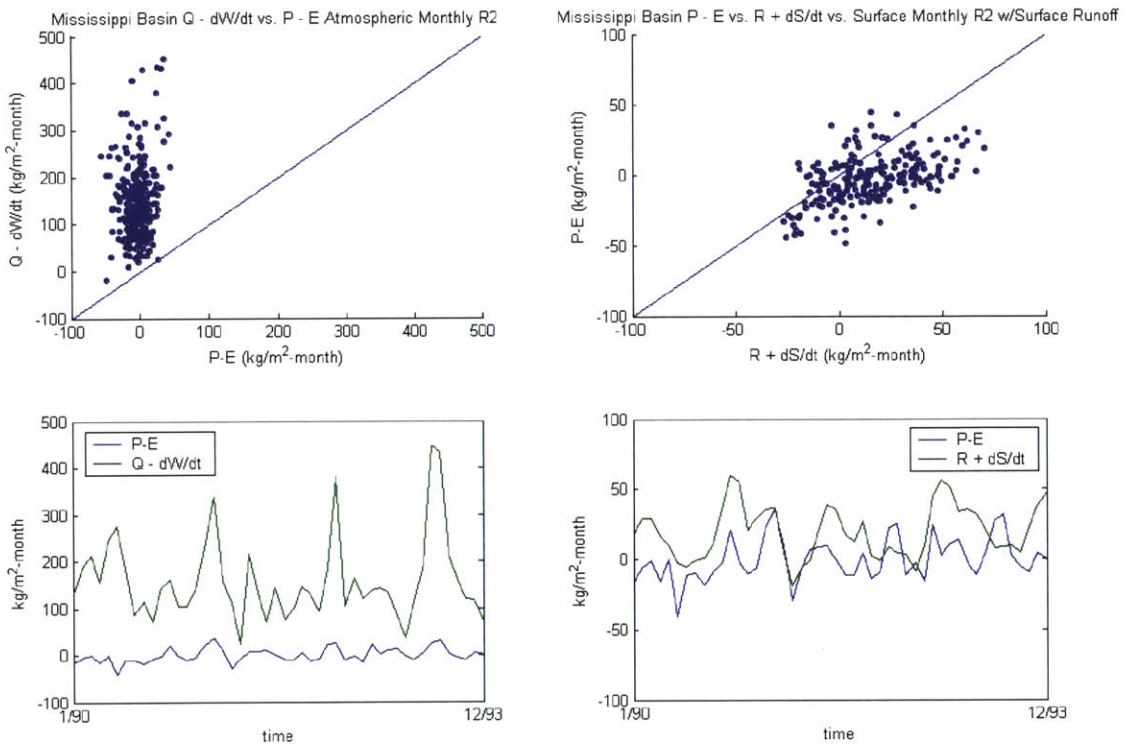


Figure 5-68: Miss. Basin Reanalysis-2 Monthly Atmospheric and Surface Balances

Possible improvement scenarios are outlined in Table 5-6. For the atmospheric balance, three precipitation datasets are eliminated, as part of this control volume extends past their bounds. The possibilities for each time scale are reduced to 20. The surface balance possibilities are again broken into two sections. Using surface runoff nine precipitation datasets are combined with six storage methods and two reanalysis. With the reanalysis runoff ten precipitation datasets can be used. The same conditions apply for the yearly time scale reducing the storage consideration from six to two.

Improvements in the yearly balances are shown in Figure 5-69. For the atmospheric balance, the Reanalysis-1 balance was highly ranked by the evaluation criteria, and the balance can be slightly improved by GPCP or SSM/I precipitation data. The GPCP precipitation application is presented here. The improvements are so minor they are almost not noteworthy as the residual statistics are reduced each by a mere 3 kg/m<sup>2</sup>-month and the correlation is only slightly improved. The balance is still not very good. For the surface balance significant improvements can be made by utilizing the Reanalysis-2 runoff instead of the USGS surface runoff for this control volume. This is very interesting because it shows the reanalysis change in surface storage balances with the reanalysis runoff, and not the more accurate surface runoff. The conditions of including storage and using the SSM/I precipitation are chosen because all five of the evaluation criteria need to be accounted for. Higher ranked balances contain regression slopes that differ substantially from one. The bias is reduced to near zero in this balance proving the reanalysis surface data balances over the long term for this control volume, with no outside data being introduced.

Attempts to improve the monthly balance are shown in Figure 5-70. In the atmospheric balance the reanalysis contains some of the top ranked balances based on the evaluation criteria and there are no great improvements that can be made. The one balance ranked better than both of the reanalyses involves applying the SSM/I precipitation to the Reanalysis-1. The bias and absolute error are again reduced just slightly though the correlation coefficient is greatly improved. The monthly surface balance is greatly improved by applying the GHCN-CAMS precipitation to the Reanalysis-2, and changing to reanalysis runoff. As shown on the right of Figure 4-156 excellent agreement is obtained between the two sides of the water balance with a bias of just 3.22 kg/m<sup>2</sup>-month and a correlation coefficient of .7305.

For the Mississippi River Basin control volume, seasonal cycle correlation is observed between terms of the atmospheric water balance. However the balance is far too biased to be considered successful. For the surface balances a much better balance is obtained when the reanalysis derived runoff is applied to the water balance, as no bias is detected. It is disappointing though that for a control volume such as the Mississippi River basin, the use of actual observed data (gauges and surface runoff), can not be effectively combined to produce a water balance. Instead the best balance is derived by the reanalysis model.

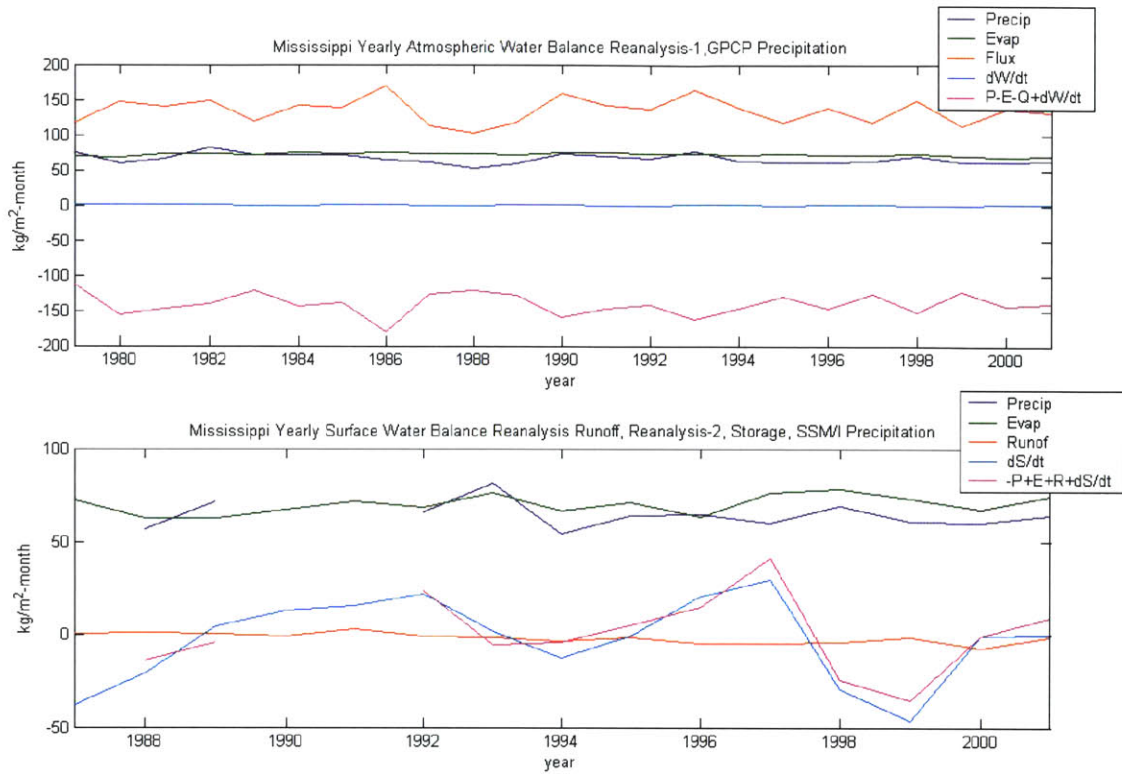


Figure 5-69: Miss. Basin Selected Yearly Water Balances

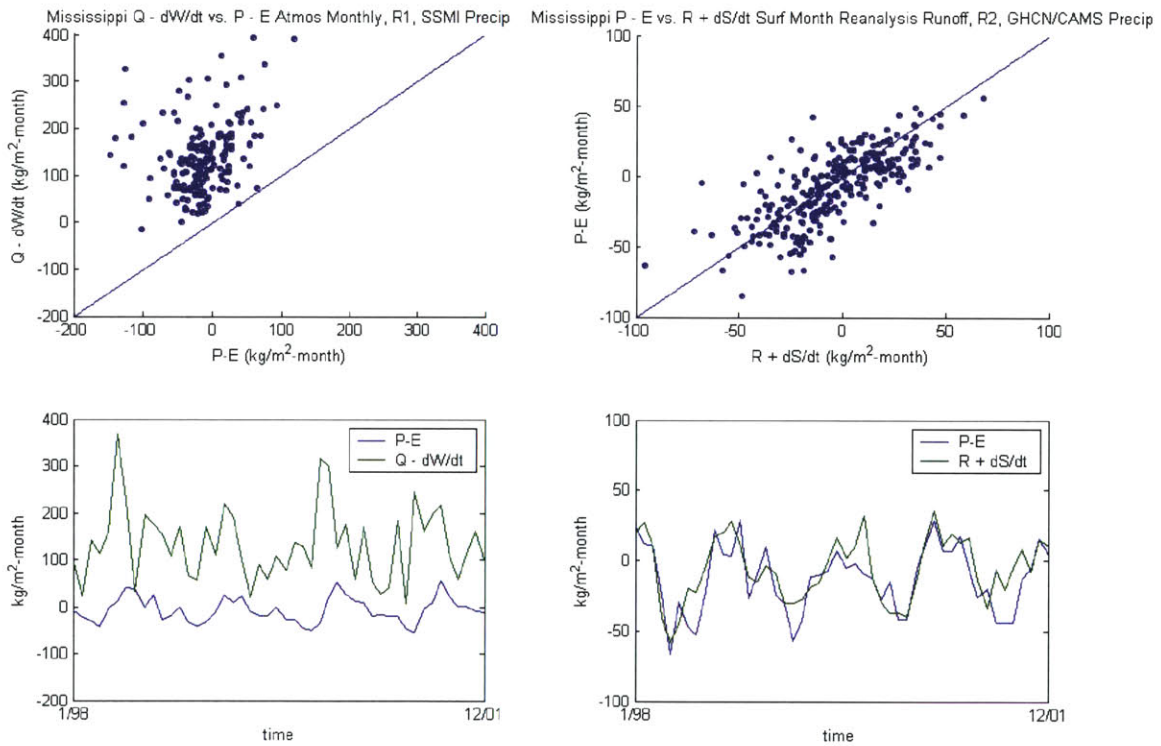


Figure 5-70: Miss. Basin Selected Monthly Water Balances

## 5.11. Arkansas River Basin

The Arkansas River Basin control volume was chosen as the smaller basin considered in North America. It is a sub basin of the Mississippi and has the advantage that it is entirely enclosed beneath 40 degree N latitude, thus it is entirely visible by the TRMM precipitation radar and the TRMM 3B43 and GPI datasets can be used. The size of this balance compared to the grid size of the datasets being applied to it makes it unlikely a balance will be obtained, however the exercise is still carried out to see the results, as for all control volumes considered so far encouraging results have been obtained for some portion of the water balance over each region.

The yearly water balances for the Arkansas Basin control volume found by the default conditions for the Reanalysis-1 are presented in Figure 5-71. The results of the atmospheric balance are very similar to those obtained in the Mississippi River balance. Annual evaporation is observed to be higher than annual precipitation for every year of the fifty five year time series. The moisture flux is large and positive not corresponding to the P-E. A bias of  $-144.6 \text{ kg/m}^2\text{-month}$  is found, with no correlation between the two sides of the balance as the correlation coefficient is .0238. The results of the surface balance are also very similar to the Mississippi River basin. The indication of the P-E cycle is that the runoff of the basin should be negative, which on a yearly basis is unrealistic because the control volume in theory is a basin, and a river such as the Arkansas isn't going to go dry over a period of a year. Switching focus to the balance the discrepancy in sign of the runoff makes it impossible to obtain a balance, and the residual term is not flat indicating correlation. The RMSE of this balance is  $22.86 \text{ kg/m}^2\text{-month}$ , larger than the magnitude of the precipitation term, with a weak correlation coefficient of .2161.

In the Reanalysis-2 (Figure 5-72) we see similar adjustments to the Mississippi River Basin in the Arkansas River Basin. The difference between P-E is reduced to the point that it is almost eliminated. However, for the atmospheric balance the moisture flux remains about the same indicating continuity between the Reanalysis-1 and the Reanalysis-2. The moisture flux here is even higher, resulting in a bias of  $-193.42 \text{ kg/m}^2\text{-month}$  and even less correlation. There is little hope to improve this balance as the moisture flux term is simply unrealistic. In the surface balance case the P-E quantity seems to match better in terms of magnitude to the runoff, but the surface storage term appears to throw the balance off on a point to point basis and the RMSE is increased to  $32.92 \text{ kg/m}^2\text{-month}$ . There is no correlation between the side fluxes of this control volume with the P-E quantity at each point as the reanalysis pretty much fails to compute a yearly water balance.

The annual precipitation time series for the Arkansas Basin control volume are presented in Figure 4-159. Again a slight increase is observed in the precipitation of the Reanalysis-1, which is matched by the larger precipitation datasets. This trend is not observed in the gauge precipitation dataset. The only precipitation datasets which can not be used at some point in this balance is the TRMM 3A46 because it contains errors.

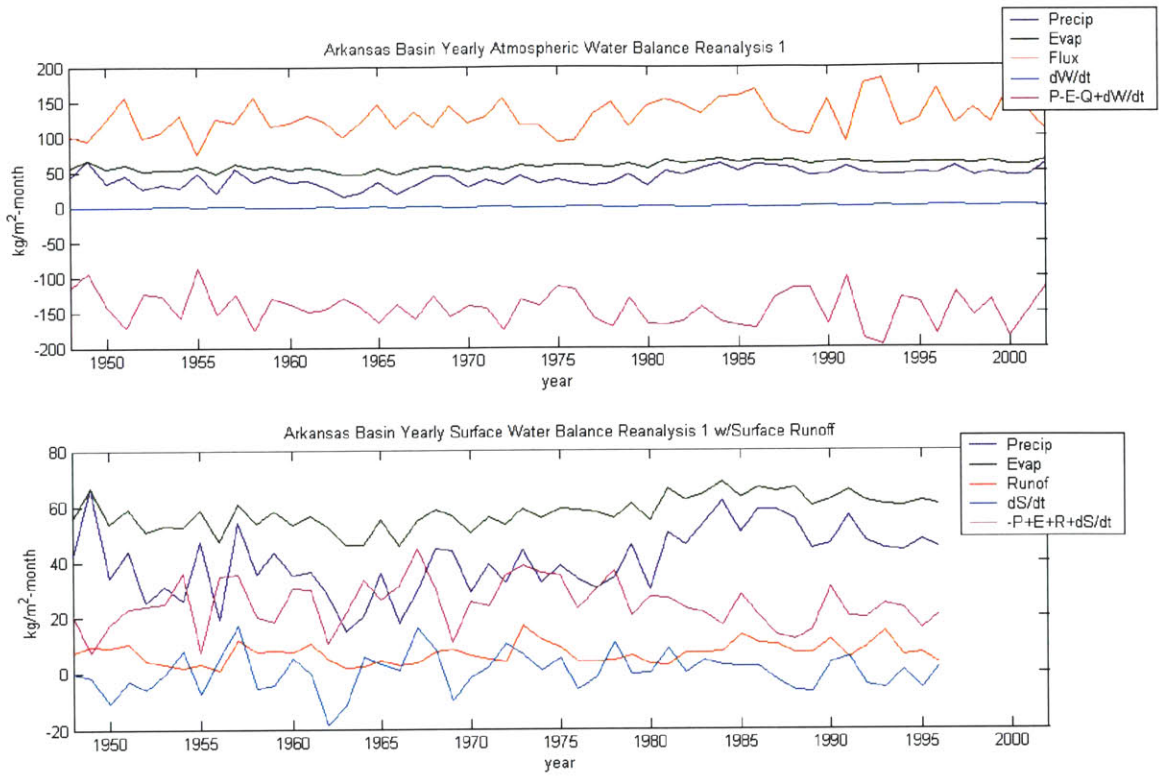


Figure 5-71: Arkansas Basin Reanalysis-1 Yearly Atmospheric and Surface Balances

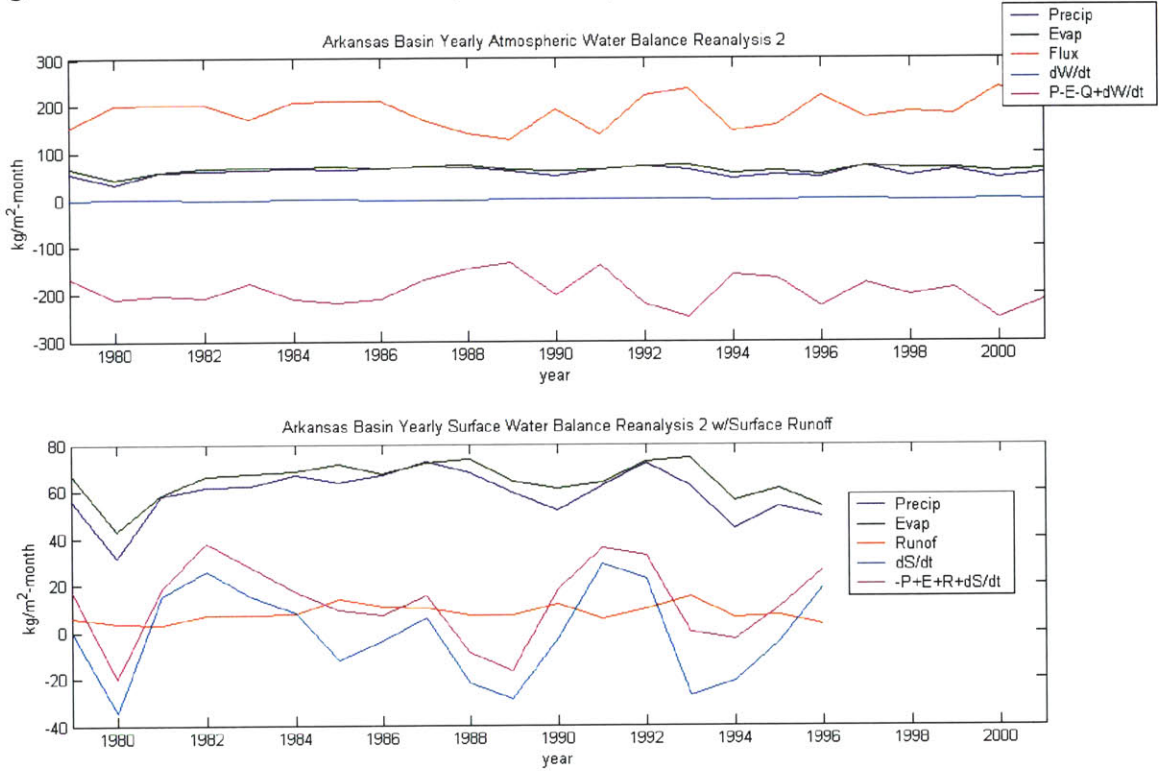


Figure 5-72: Arkansas Basin Reanalysis-2 Yearly Atmospheric and Surface Balances

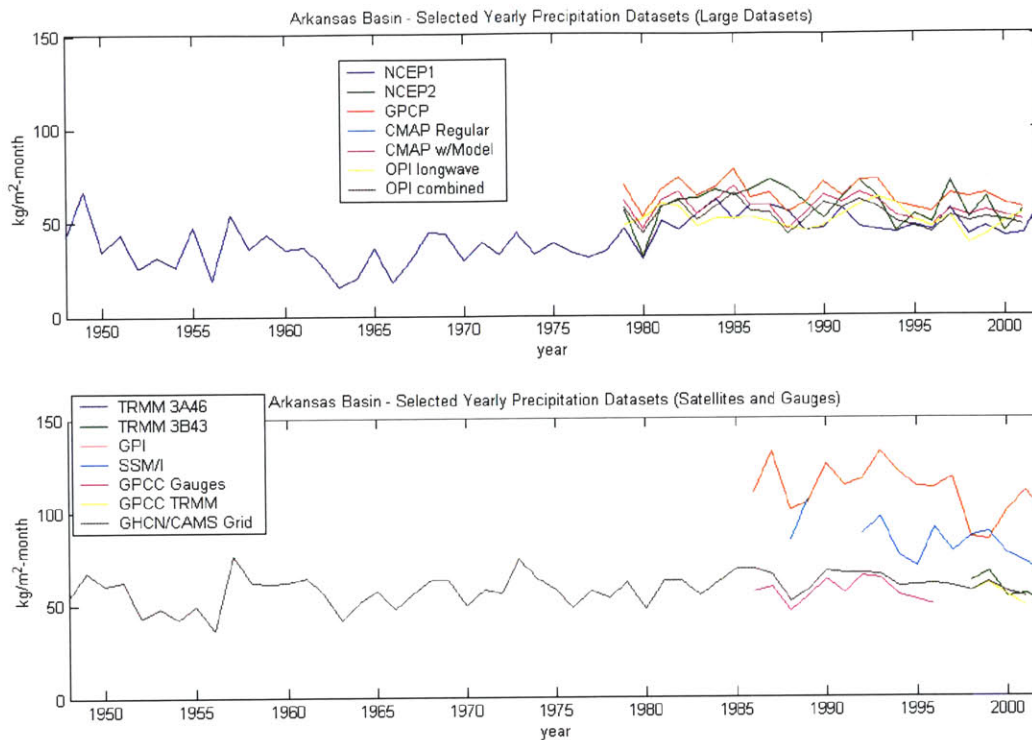


Figure 5-73: Arkansas Basin Different Sources of Yearly Precipitation

Turning to the monthly case the default Reanalysis-1 balances are presented in Figure 5-74. On the monthly time scale the atmospheric balance appears to be a little different than on the yearly time scale. The bias remains the same and there is virtually no correlation between the two quantities plotted on the top left. The surface balance is a different, as on the monthly time scale the balance looks much better as the seasonal cycle is picked up in both sides of the balance and correlation is observed, though also a bias in favor of the surface side flux is observed. A very strong correlation of .6663 is found, showing on a monthly basis the data consistently finds peaks and troughs in the data for both sides of the balance.

The Reanalysis-2 atmospheric balance shown in Figure 5-75 is even worse then the Reanalysis-1 balance. The atmospheric water balance over the Arkansas River basin is a failure. For the surface balance the correlation does decrease for the Reanalysis-2 to .4043 which is unusual, but the bias is also improved as was observed in the yearly time scale. A decent water balance can be obtained for the surface balance on a monthly basis, but this balance is shown to be less complete when aggregated to the yearly time scale, and the only property which can be extracted from these balance are strong monthly signal correlations between the two sides of the balance.

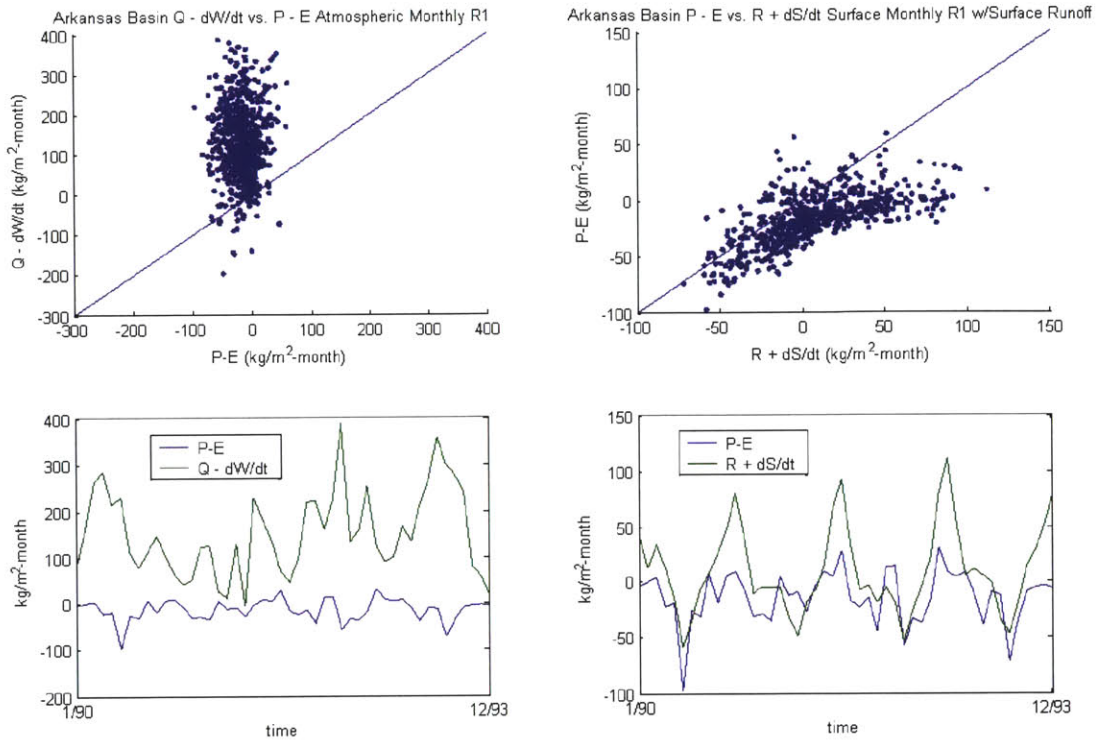


Figure 5-74: Arkansas Basin Reanalysis-1 Monthly Atmospheric and Surface Balances

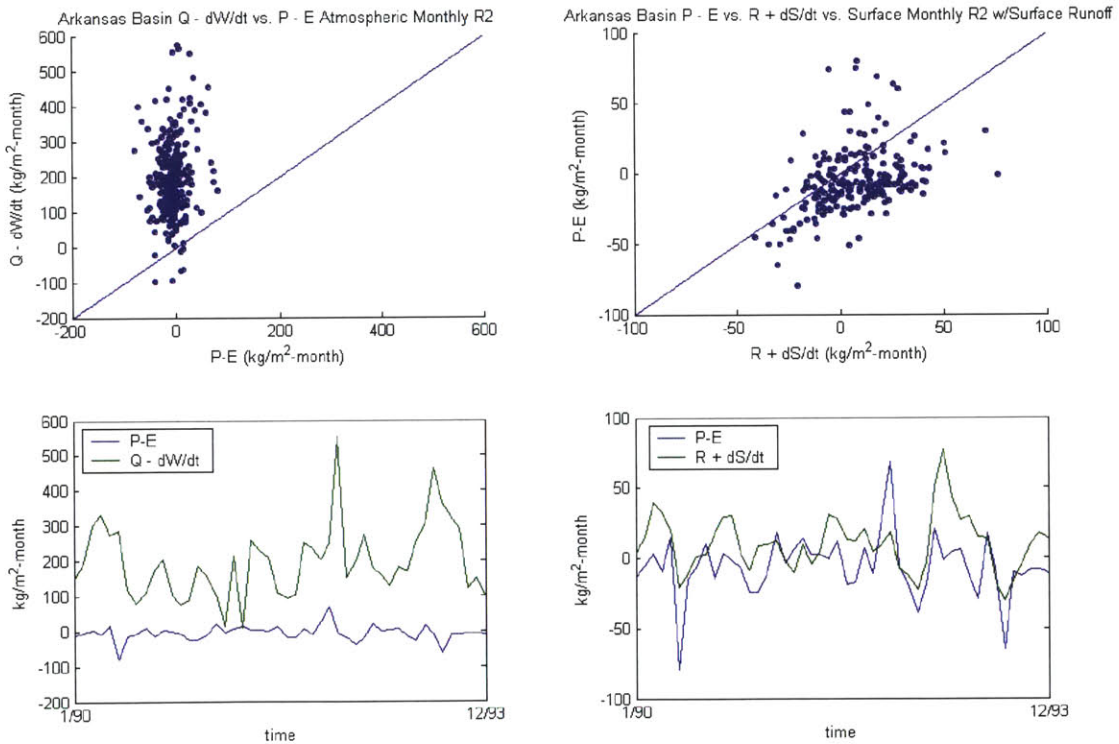


Figure 5-75: Arkansas Basin Reanalysis-2 Monthly Atmospheric and Surface Balances

The Arkansas Basin control volume is unique in the amount of water balance characteristics which can be altered to try and improve the balance. Although it is not a South American water balance, it can be classified as a tropical water balance in terms of the precipitation datasets. There are therefore 24 possible balances for the atmospheric balances outlined in Table 5-7. For the surface balance we can again change reanalysis (2), storage considerations (6, monthly/2, yearly), and runoff data source (2), in addition to precipitation dataset. None of the TRMM datasets can be used for the surface runoff case, and all but TRMM 3A46 can be used for the reanalysis runoff case.

The best yearly water balances for the Arkansas River Basin based on the evaluation criteria are presented in Figure 5-76. In the atmospheric balance the focus of the improvement is only on reducing the bias, which therefore results in the inclusion of the GPI precipitation. This precipitation is much larger in magnitude than any of the other precipitation datasets. The accuracy of this dataset is unknown, but it reduces the atmospheric bias  $-86.28 \text{ kg/m}^2\text{-month}$  with a correlation coefficient of .4143 and a regression slope of 1.20. Thus, the balance is improved in terms of correlation greatly, but not balanced. For the surface balance the application of CMAP2 precipitation and the removal of storage improved the bias and the correlation of the balance. The correlation coefficient is improved to .5746. Such improvements are difficult to see on a yearly plot.

For the monthly time scale both balances are improved by the application of GPCP precipitation data. In the atmospheric balance this application is more important than reducing the bias as a higher correlation is obtained. In general this improvement is unsuccessful at actually obtaining a water balance. For the surface balance, surface runoff is applied to the Reanalysis-2 with the GPCP precipitation a good correlation and scatter plot is obtained in Figure 5-77 on the top right. A correlation coefficient of .5746 is obtained, with a bias of just  $5.83 \text{ kg/m}^2\text{-month}$ . There is some spread in the scatter plot, but a generally good balance from month to month. The change in precipitation dataset helps the balance.

For the Arkansas Basin control volume the limits appear to be reached in terms of water balance calculations. The atmospheric balances show traces of correlation, but terrible biases that make the balance unsuccessful. The results of the surface balance are better, but the correlations on the monthly time scale are still not as good as those observed in other balances.



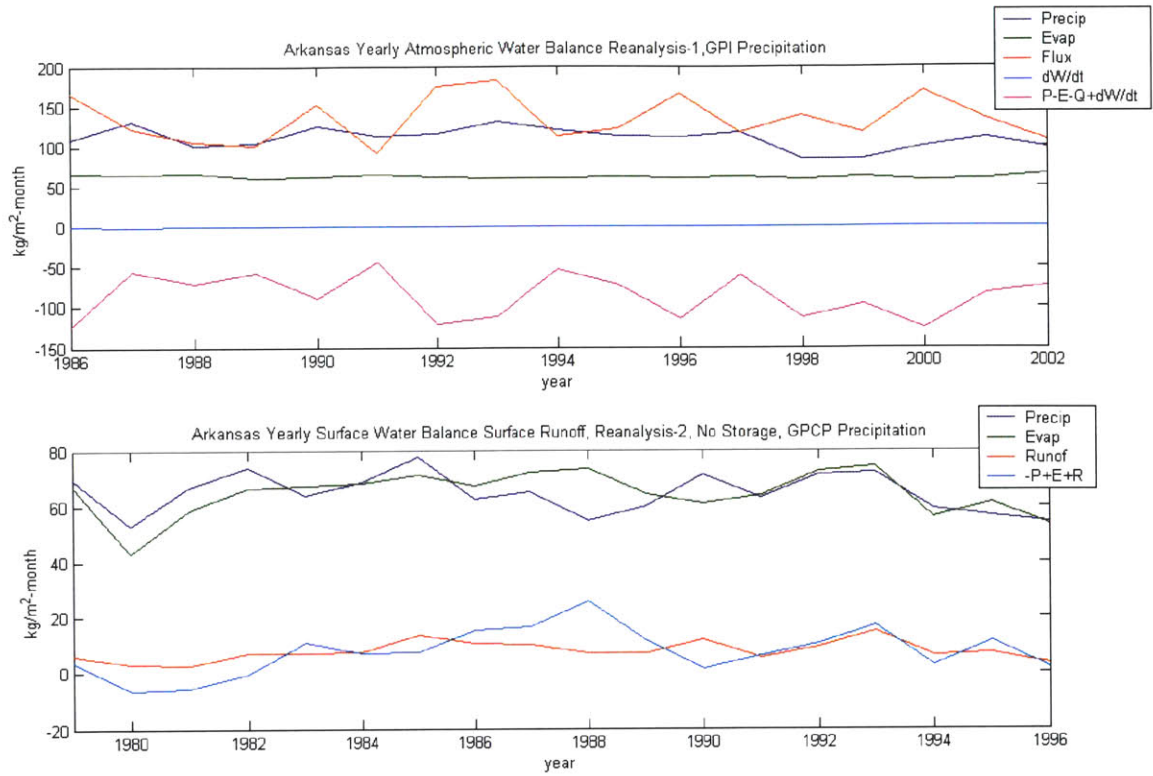


Figure 5-76: Arkansas Basin Selected Yearly Water Balances

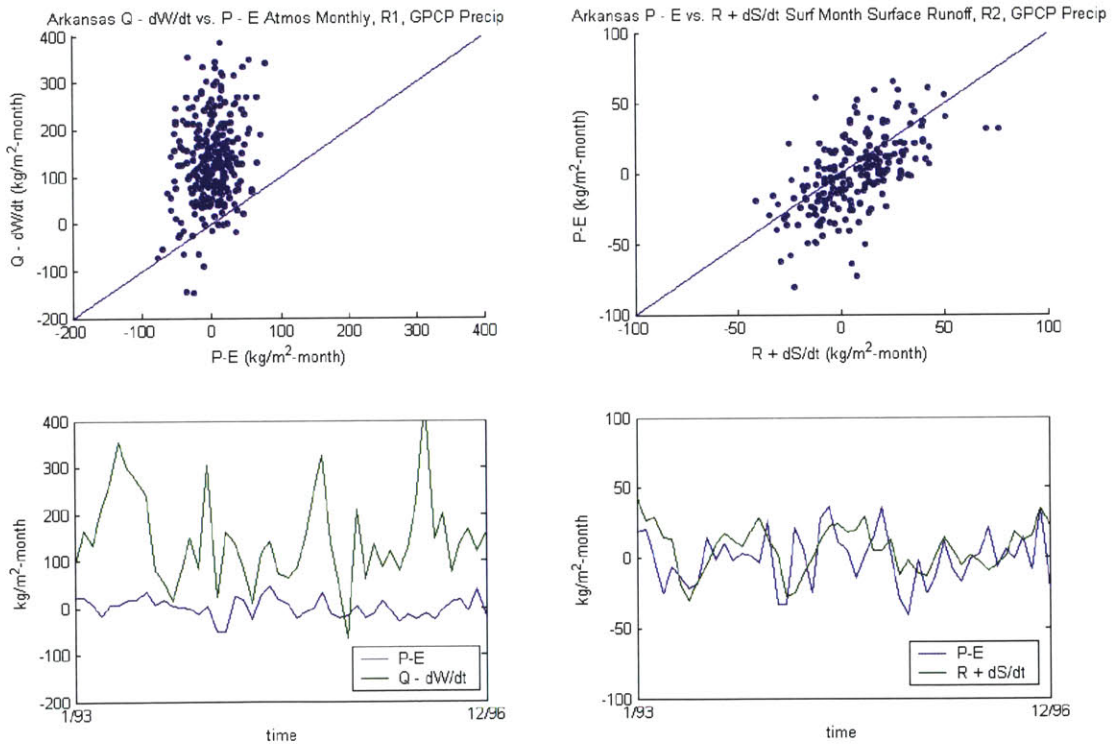


Figure 5-77: Arkansas Basin Selected Monthly Water Balances

## 6. Global Data Analysis

In this section the entire globe between the latitudes of 60 N and 60 S will be broken into equally sized regional boxes of 30, 15, and 7.5 degrees on a side and for each box the atmospheric water balance will be evaluated. Areas north and south of these latitudes are excluded due to a number of reasons. These reasons include the difficulty in computing water balances due to longitudinal geometries at high latitudes and the incompleteness of most precipitation datasets at these high latitudes. Such a comprehensive analysis is possible only for the atmospheric water balance and not the surface water balance. In order to compute a surface water balance change in storage and runoff must be calculated, however over oceans these calculations are not possible in the context of this work. Thus, in the global case we are simply looking at one atmospheric mass balance equation:

$$\frac{dW}{dt} = -P + E + Q \quad \text{Eq. (6-1)}$$

Since only the atmospheric water balance is considered, only a couple of comparisons are possible.

$$P - E - Q + \frac{dW}{dt} = \text{residual} \quad \text{Eq. (6-2)}$$

$$P - E = Q - \frac{dW}{dt} \quad \text{Eq. (6-3)}$$

When doing the global analysis the statistics that have been created before can be plotted on a contour map and can be spatially analyzed. On a coarser resolution it is difficult to recognize this “map” as the world, however at finer resolutions continental features become much more prevalent, and the data is more useful in a geographical sense. It is likely though that finer resolution water balances will be less accurate, as errors are smoothed with increased spatial averaging. The hope of such an analysis is that there would be no large differences between the closure of the water balance between land and sea or between any set of regions. Unfortunately, this is rarely the case.

The global data analysis will be composed of 4 sections. In Section 6.1 the general water balance characteristics for the reanalysis, considering all water balance analysis tools which have been utilized and applied in this work, will be shown. In Section 6.2 alternative precipitation datasets which are available on a global scale will be used in the global atmospheric water balances. In Section 6.3 the global water balances will be analyzed on a seasonal and decadal basis, and finally, in Section 6.4 land and sea points will be contrasted against each other on a global basis.

## 6.1. Water Balance Property Data Analysis

The first look at global data in this work focuses completely on the reanalysis. Maps are made of all water balance properties for the Reanalysis-1 and Reanalysis-2 for the monthly and yearly time scales. The purpose of this section is to show errors in computation of the global atmospheric water balance spatially.

### 6.1.1. Global Atmospheric Absolute Error

Although water balances are being compared from different geographic regions, which before were analyzed by normalized residual statistics, it is useful to use non-normalized residual statistics because the actual magnitude of the imbalances are of interest. The first such analysis involves mapping the absolute error of the atmospheric water balance at three resolutions. Looking at the 30 degree case in Figure 6-1, improvements are found in absolute error when the time scale is changed from monthly to yearly. There is little improvement however in the Reanalysis-2 from the Reanalysis-1. Problem areas include Eastern Asia and Southern Africa for all cases. Increasing the resolution to 15 degrees in Figure 6-2 the general observations made for the 30 degree case are maintained. Note that the change in the colorbar in this figure has doubled, thus error increases in all regions by a factor of two. At the finer resolution there errors are more tightly constrained to land points. Reducing the resolution again to 7.5 degrees in Figure 6-3 we again double the colorbar to get approximately the same picture, though better resolved. Higher absolute errors are seen in points mostly over various land bodies specifically over Central Asia.

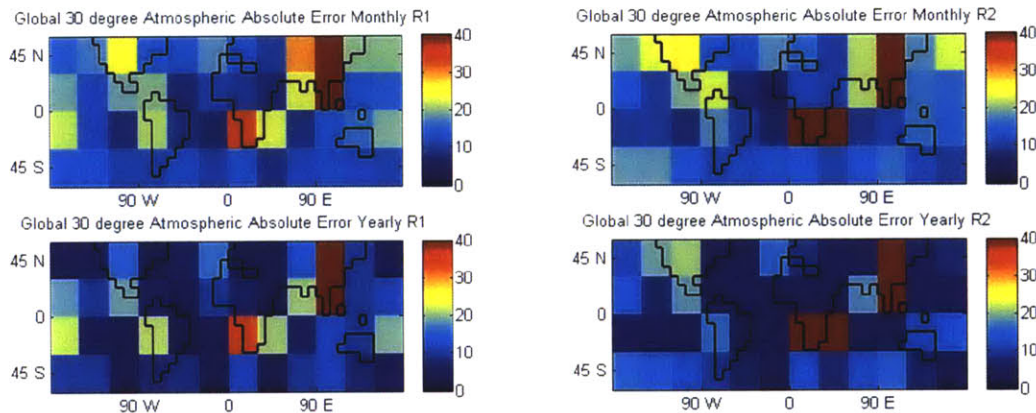


Figure 6-1: 30 degree Absolute Error Reanalysis Balances

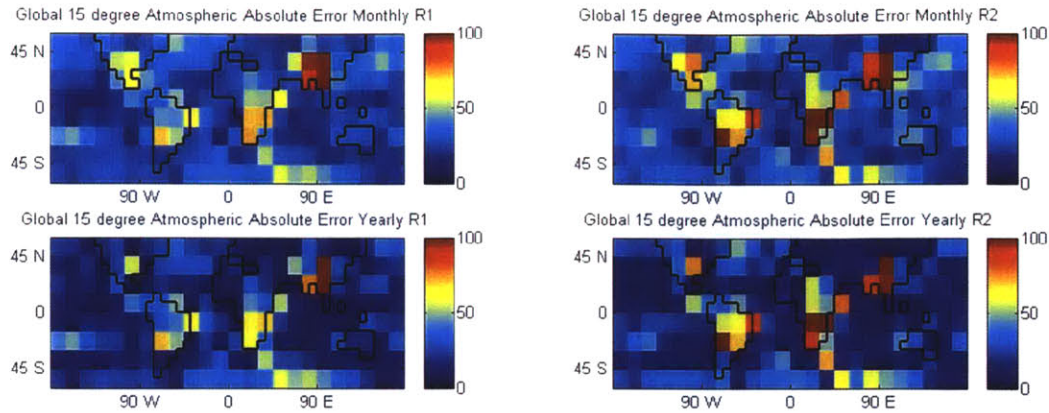


Figure 6-2: 15 degree Absolute Error Reanalysis Balances

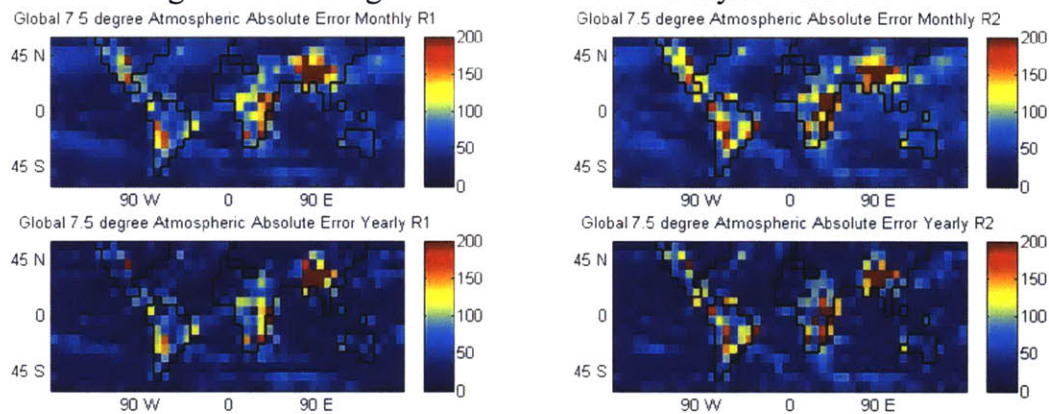


Figure 6-3: 7.5 degree Absolute Error Reanalysis Balances

### 6.1.2. Global Normalized Atmospheric Absolute Error

Normalizing the absolute error allows a better comparison of regions. The global atmospheric water balance is composed of many unique regions, some with a very active water cycle (i.e. rain forests) and some with very inactive water cycles (i.e. deserts). Errors in more inactive water cycles such as those over the desert are highlighted by this analysis. Figure 6-4 shows the 30 degree normalized absolute error. Again general improvements are seen with change in time scale, but there is little difference between reanalysis. Other problem areas are seen now at the 30 degree level over North and South America in more mountainous regions which in the case of North America are likely drier. The resolution is increased to 15 degree in Figure 6-5. Clear bands of absolute error greater than one, can be seen through Africa into Asia, and in the mountainous regions of North and South America. These errors are likely small in magnitude, but when normalized by the precipitation grow large. These errors are intensified and better resolved in Figure 6-6. Clear imbalances are seen over a large portion of Asia and the Western United States. Over the ocean, the balances seem to agree more in the tropical regions, which makes sense in that the normalization quantity is likely larger over the tropics.

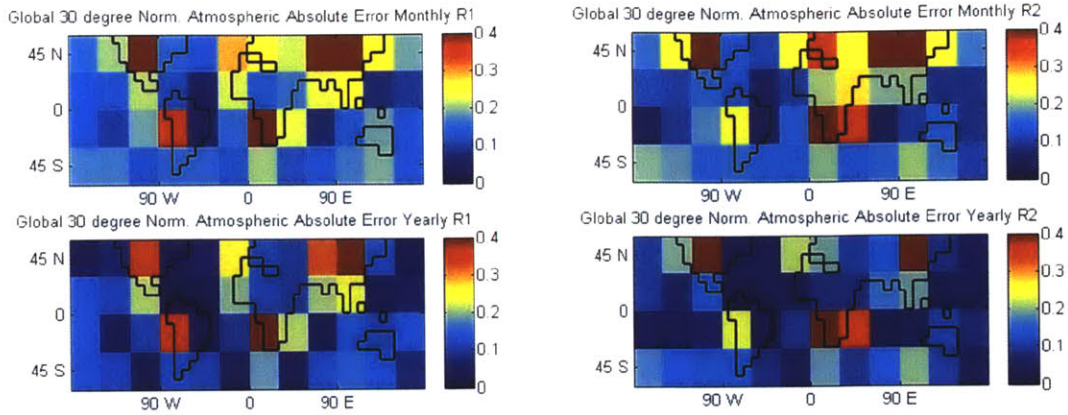


Figure 6-4: 30 degree Normalized Absolute Error Reanalysis Balances

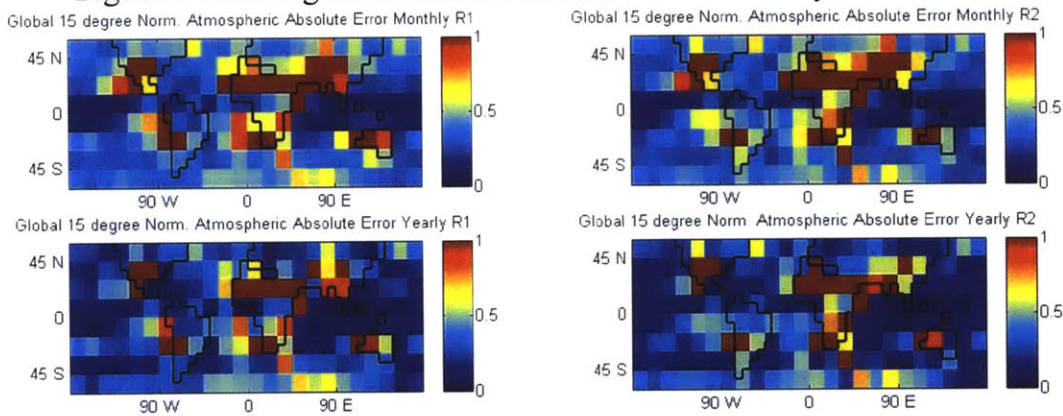


Figure 6-5: 15 degree Normalized Absolute Error Reanalysis Balances

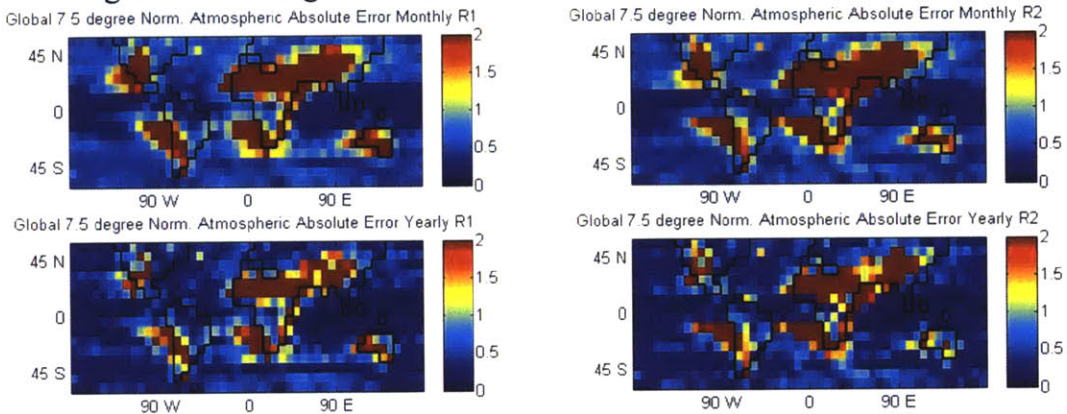


Figure 6-6: 7.5 degree Normalized Absolute Error Reanalysis Balances

### 6.1.3. Global Atmospheric Bias Error

Of particular importance in the global study is the bias. This is in essence the departure from balance for the entire time series of each control volume, and as before the monthly and yearly biases are the same. From this analysis we can find the areas of the world where there is a surplus (red) and where there is a deficit (blue) in the atmospheric water balance. In Figure 6-7 the 30 degree resolution is presented and relatively small biases are observed for all portions of

the globe. Problem areas in both reanalyses include Eastern Asia and Southern Africa, and in general equatorial regions look less biased. There are few noticeable differences between reanalyses.

Increasing the resolution to 15 degrees in Figure 6-8, problem areas are again observed over Asia, and Africa, with additional significant errors over South America. Interestingly, there is no pattern to these errors, they are both positive and negative and scattered about the land masses randomly. These errors tend to be intensified in the Reanalysis-2 and both reanalyses show breakdowns in the north most and south-most control volumes. Errors in general are much higher on the fifteen degree resolution as the bounds of the colorbar are more than doubled. Moving to the 7.5 degree resolution in Figure 6-9, the colorbar bounds remain the same. In general larger positive and negative biases are observed over land more than over water. There is a general surplus in the southern most control volumes but other geographic patterns are not observed. Only certain portions of the land masses are unbalanced. Northern land masses appear better balanced.

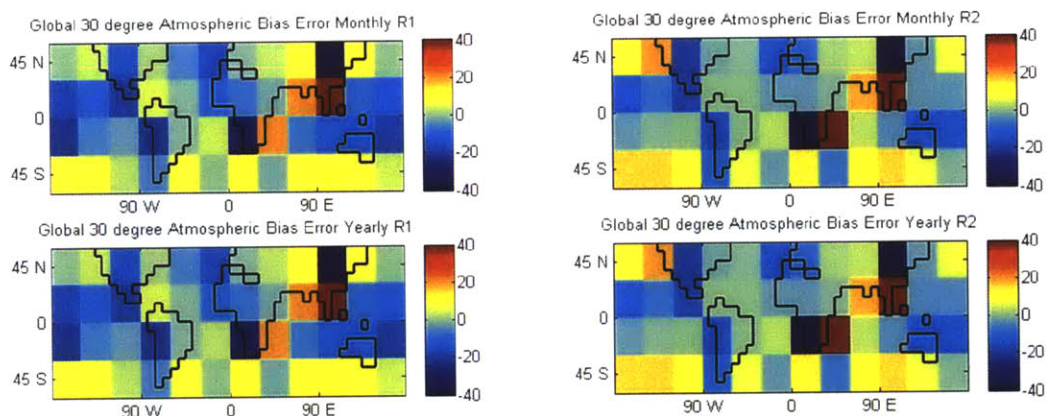


Figure 6-7: 30 degree Bias Error Reanalysis Balances

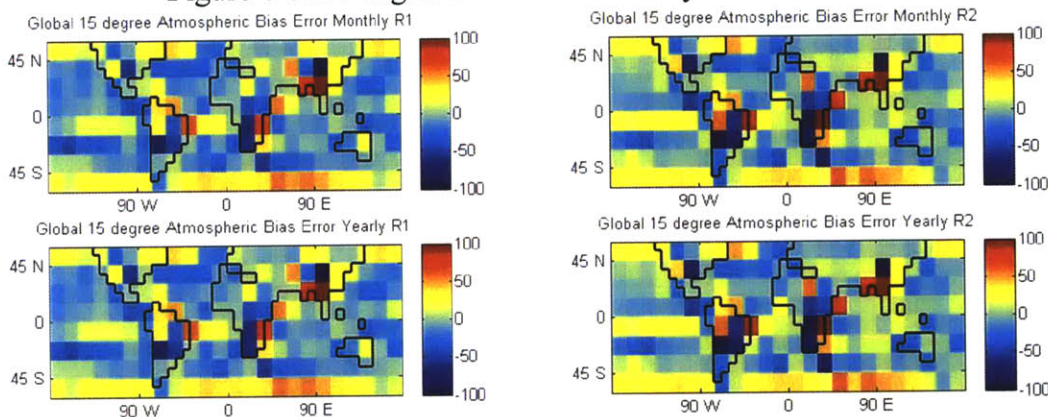


Figure 6-8: 15 degree Bias Error Reanalysis Balances

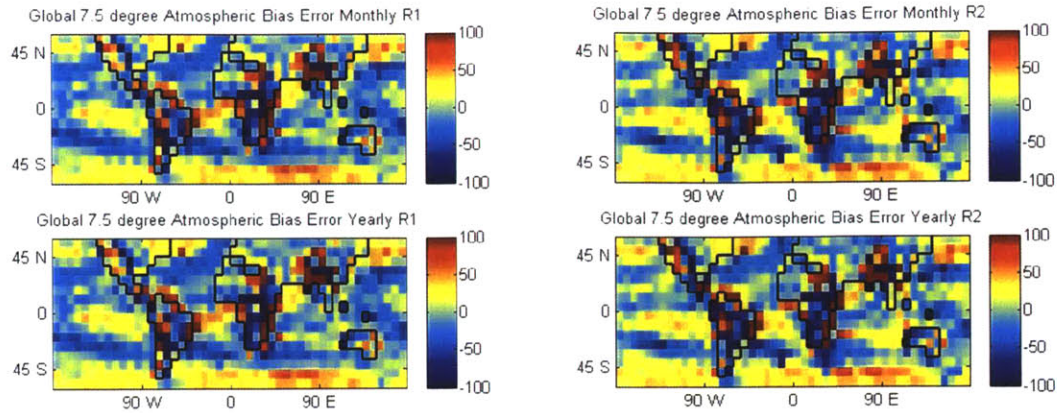


Figure 6-9: 7.5 degree Bias Error Reanalysis Balances

#### 6.1.4. Global Normalized Atmospheric Bias Error

In this section the biases are normalized. In Figure 6-10 the normalized bias is presented at the 30 degree resolution. Here more dramatic imbalances are highlighted. Large normalized deficits occur in the same places as were observed in the last section; however the surpluses which were observed are no longer as bad, as the precipitation magnitude is high over these areas. Additional deficits are found over the mountainous regions of North America and South America.

In Figure 6-11 the resolution is reduced to 15 degrees and additional problem areas are discovered. The western portions of South America and North America, as well as parts of all other land masses contain normalized biases on the order of magnitude of the precipitation of that region. Again, there is no pattern to the sign of these biases; they are both positive and negative. The only real observation which can be made is that these biases all occur over land. There is also a larger difference in the normalized case between good and bad areas when looking at the normalized error and this observation is intensified in Figure 6-12 for the 7.5 degree resolution. Again, significant normalized biases greater than 1 occur mostly over land which show no pattern as surplus and deficit control volumes are scattered randomly over land. In North and South America, these biases tend to occur over the western parts of the continent on and off the land masses. The land stretching from Northern Africa into Central Asia also shows significant highs and lows for this property, but no pattern. These biases over the Eastern Hemisphere correspond to very arid regions, and are intensified by normalization.

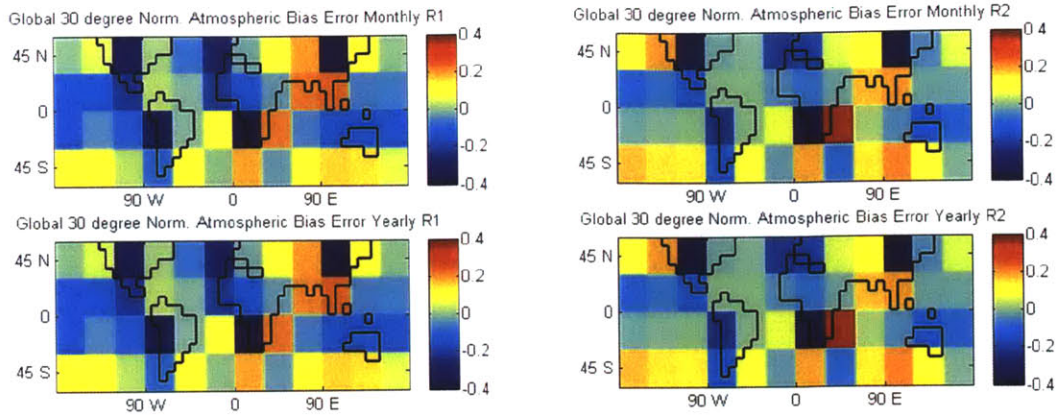


Figure 6-10: 30 degree Normalized Bias Error Reanalysis Balances

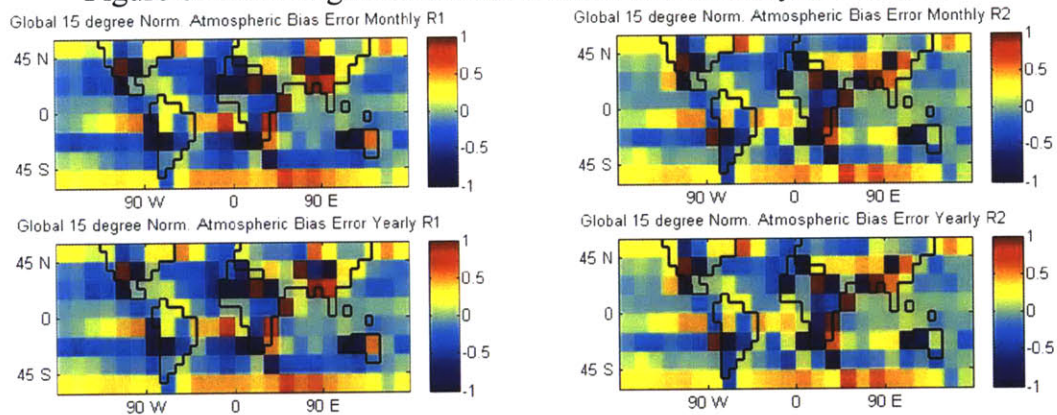


Figure 6-11: 15 degree Normalized Bias Error Reanalysis Balances

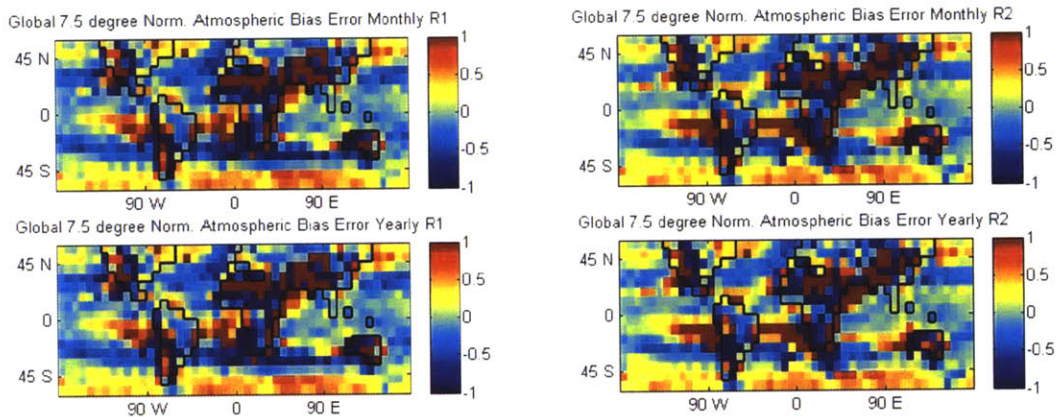


Figure 6-12: 7.5 degree Normalized Bias Error Reanalysis Balances

### 6.1.5. Global Atmospheric Regression Slope

In the next three sections results of the regression calculation for the atmospheric two sided water balance are presented. Of the most importance in the regression study is the evaluation of the slope. A slope above one results in a larger signal in moisture flux (given that change in precipitable water has little variation) than in the P-E quantity, and a slope below one means the



reverse. In Figure 6-13 the regression slopes for the 30 degree resolution are shown. Several control volumes control volumes yield slopes which differ significantly from one. These points are more common over land than water and don't show much of a pattern as far as being either high or low. The 15 degree resolution of the regression slope is shown in Figure 6-14. There are many light blue and yellow points in both the 30 and 15 degree resolution indicating slope slightly below and slightly above one especially over the oceans. These results indicate fairly good regression slopes.

In Figure 6-15 a much more detailed picture of the different regression slopes is shown at the 7.5 degree resolution. The same scale is used at all resolutions in order to display the decay of the slope from the coarse resolution to fine resolution. For the 30 degree resolution only a few tiles show the maximum and minimum slope of the color bar provided. With the 7.5 degree resolution many more points are dark blue and dark red, indicating slopes at or over the color bar bounds.

Subtle differences can be detected between the monthly and yearly time scales and between the Reanalysis-1 and Reanalysis-2 across all resolutions, as the yearly balances tend to have slopes with a further departure from one especially over land in the 15 and 7.5 degree resolutions. Fewer points are present for these regressions, and there is no seasonal cycle. Such differences can be specifically seen over the South American continent for both reanalysis (where strong seasonal cycles have been observed previously) across the different time scales.

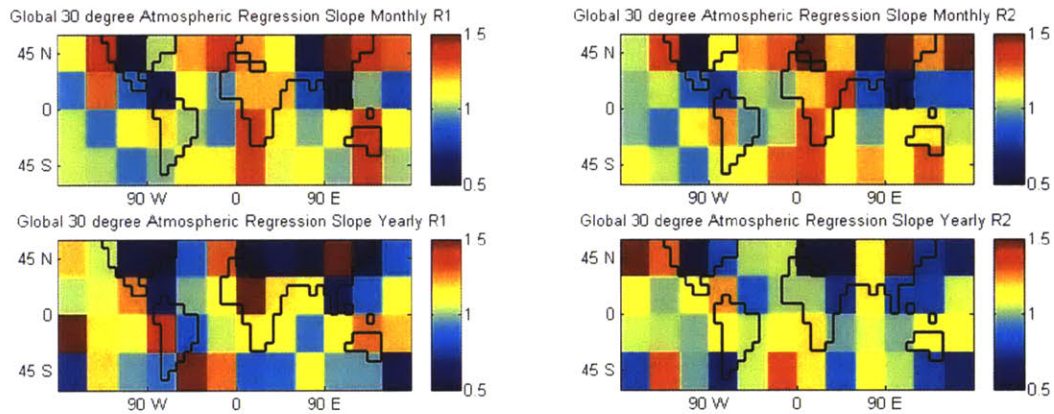


Figure 6-13: 30 degree Regression Slope Reanalysis Balances

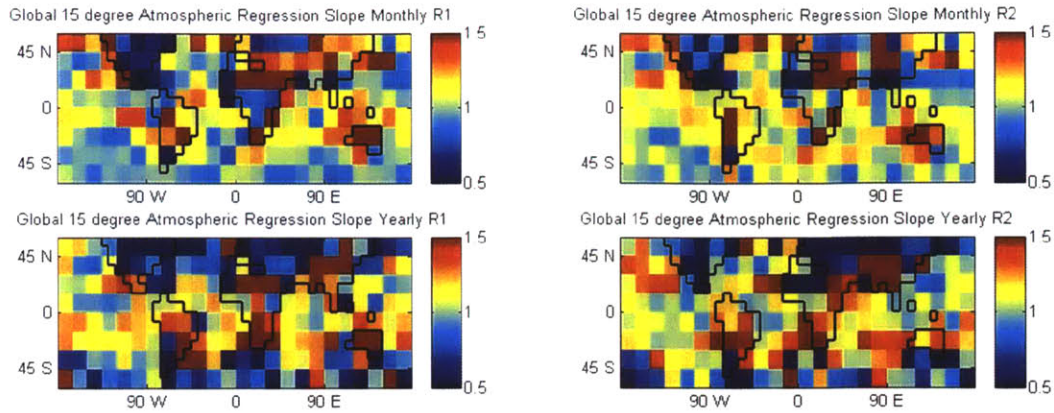


Figure 6-14: 15 degree Regression Slope Reanalysis Balances

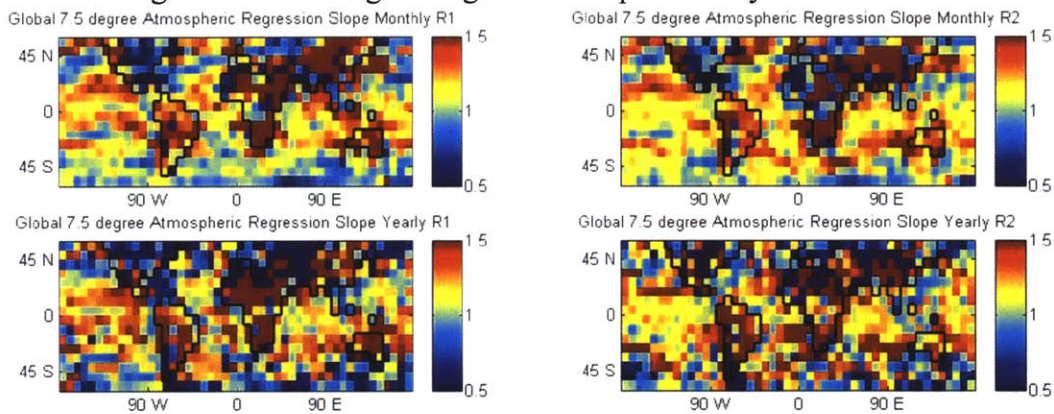


Figure 6-15: 7.5 degree Regression Slope Reanalysis Balances

### 6.1.6. Global Atmospheric Regression Intercept

The regression intercept is essentially a measure of bias in the regression, and since there is a better measure of bias in the residual statistics it is often bypassed. It is interesting though to look at this statistic for initial reanalysis conditions on a global map to observe trends and compare to with the other regression statistics. In Figure 6-16 the regression intercept is mapped for the 30 degree resolution. The desirable intercept of zero is represented by green blocks, while blue and red indicate strong negative and positive intercepts respectively. At this resolution there aren't many points which show regression intercepts over  $10 \text{ kg/m}^2\text{-month}$ . There is little difference between regression intercept across time scale or reanalysis. The distribution of errors appears totally random.

In Figure 6-17 the 15 degree regression intercepts are shown. The scale of the colorbar is doubled in order to obtain a similar picture as before, and again there are few obvious patterns in terms of regression intercepts. This is alarming over areas like the tropical ocean where water balances of this nature are suppose to be very successful. The same can be said for Figure 6-18 where the bounds of the colorbar are again doubled. Errors at the 7.5 degree resolution are more concentrated over land masses and to the north and south, but clear conclusions can not be drawn. Other water balance evaluation criteria are able to detect errors in the water balance more clearly, a further reason to disregard regression intercept in the global analysis.

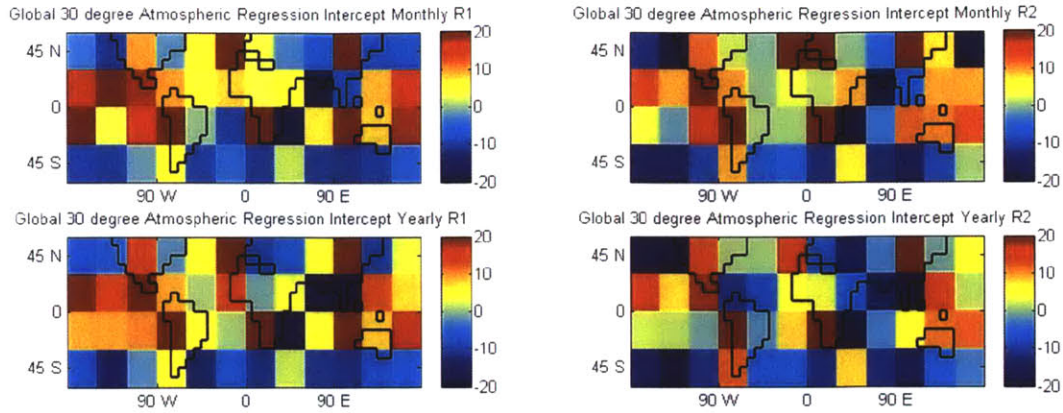


Figure 6-16: 30 degree Regression Intercept Reanalysis Balances

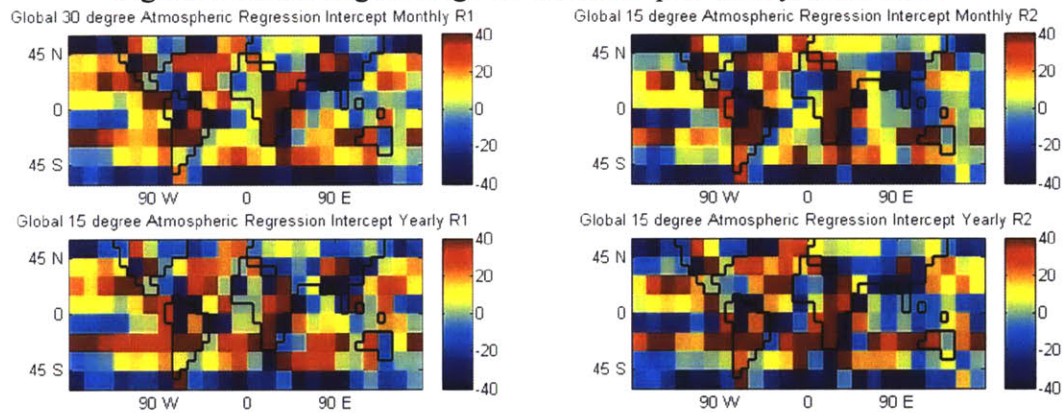


Figure 6-17: 15 degree Regression Intercept Reanalysis Balances

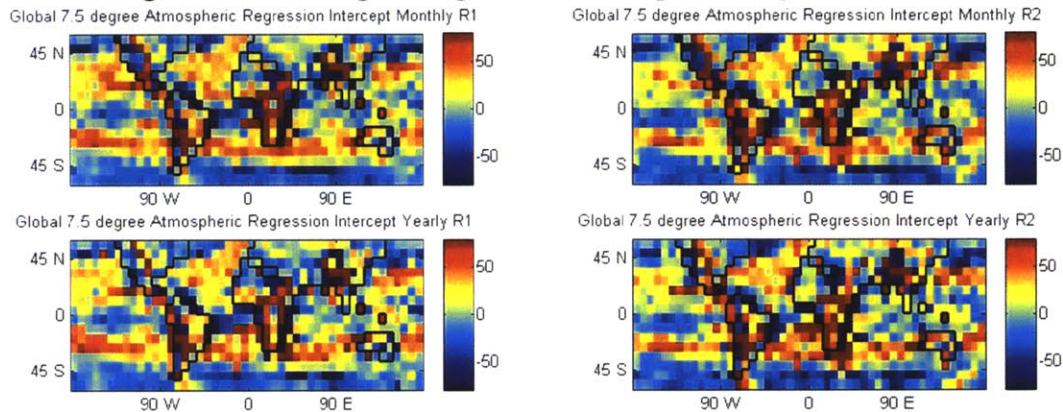


Figure 6-18: 7.5 degree Regression Intercept Reanalysis Balances

### 6.1.7. Global Atmospheric Regression $R^2$

In this section, the  $R^2$  values corresponding to the atmospheric regression calculated for each point in the global water balance map are presented. Therefore, results in these plots are dependent on the regression slope and intercept which were obtained for each control volume in the previous two sections. Very high  $R^2$  values are found almost exclusively in the tropical regions for the 30 degree resolution. This relationship is strongest in the monthly Reanalysis-2

case and weakest in the yearly Reanalysis-1 case. There is a major contrast between the middle two rows in each plot and the top and bottom row of Figure 6-19. The properties of these tropical regressions are varied, but they all correspond to good correlation even over the yearly time scale, an interesting observation which is unique to the 30 degree resolution.

When the resolution is increased to 15 and 7.5 degrees the strongest relationships are again seen clearly in the monthly Reanalysis-2, and the weakest in the yearly Reanalysis-1, shown in Figures 6-20 and 6-21. However, as resolution is increased, the strong  $R^2$  values are more limited to the oceans than the land in the tropical regions. In the coarse 30 degree resolution all of the control volumes in the tropical regions have at least some ocean in them. At finer resolutions when points can be better classified as land or ocean, the land points don't perform as well.

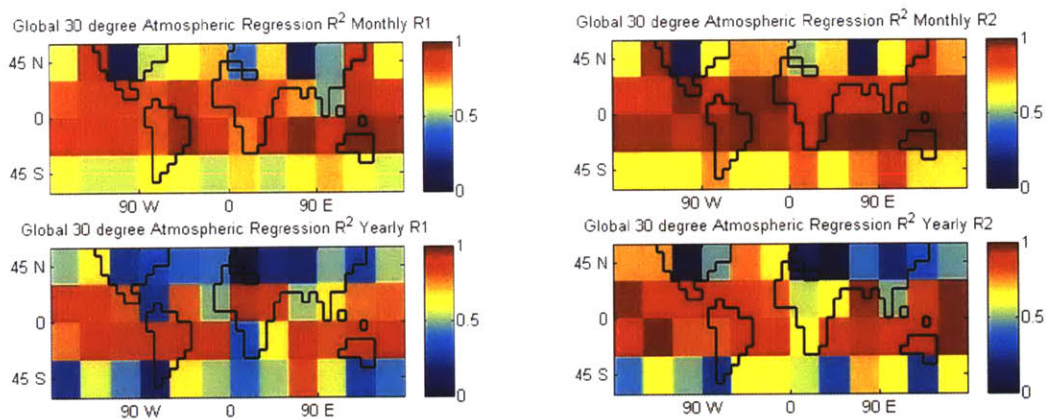


Figure 6-19: 30 degree Regression  $R^2$  Reanalysis Balances

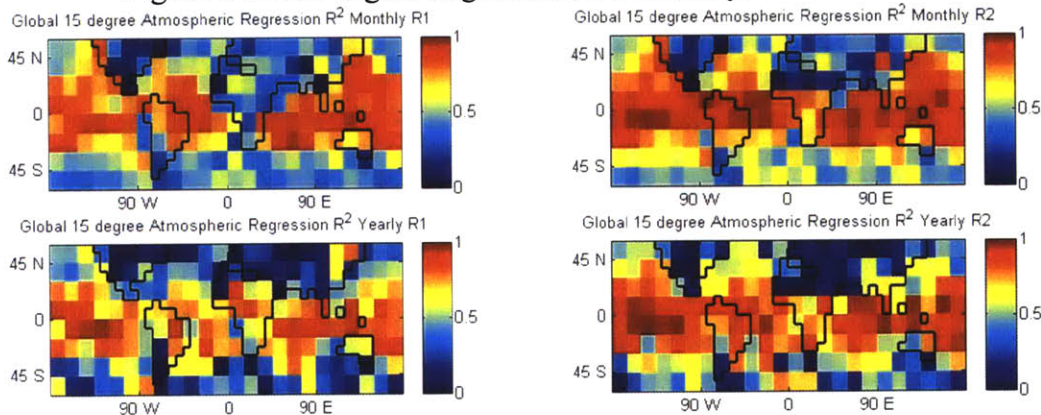


Figure 6-20: 15 degree Regression  $R^2$  Reanalysis Balances

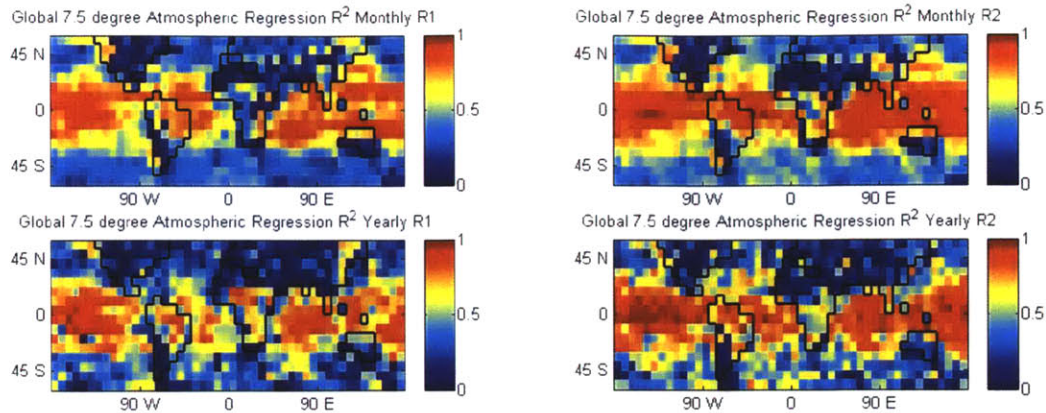


Figure 6-21: 7.5 degree Regression R<sup>2</sup> Reanalysis Balances

### 6.1.8. Global Atmospheric Regression R<sup>2</sup> 1:1

In Figure 6-22 the R<sup>2</sup> 1:1 values are shown for the 30 degree resolution control volumes. For the monthly time scale many tropical control volumes adhere well to the 1:1, with the Reanalysis-2 having more of these solid red areas. Few such points are found for the yearly time scale even at the 30 degree resolution. The strong seasonal cycle governs the R<sup>2</sup> 1:1 term in tropical regions as it did for the regression line R<sup>2</sup> term. In Figure 6-23, showing the 15 degree resolution, fewer good areas exist on the monthly time scale and those that do exist are primarily over the tropical oceans. The locations of these good R<sup>2</sup> 1:1 results extend north and south in the oceans in the Reanalysis-2. Some ocean points yield good results for the yearly times scale and more so in the Reanalysis-2, but the global map is dominated by blue and therefore significantly negative R<sup>2</sup> 1:1 values. In Figure 6-24 the same observations can be made for the 7.5 degree resolution as for the 15 degree resolution. Over the oceans and over large tropical areas (30 degrees), the R<sup>2</sup> 1:1 measure is an effective tool in evaluating the water balance.

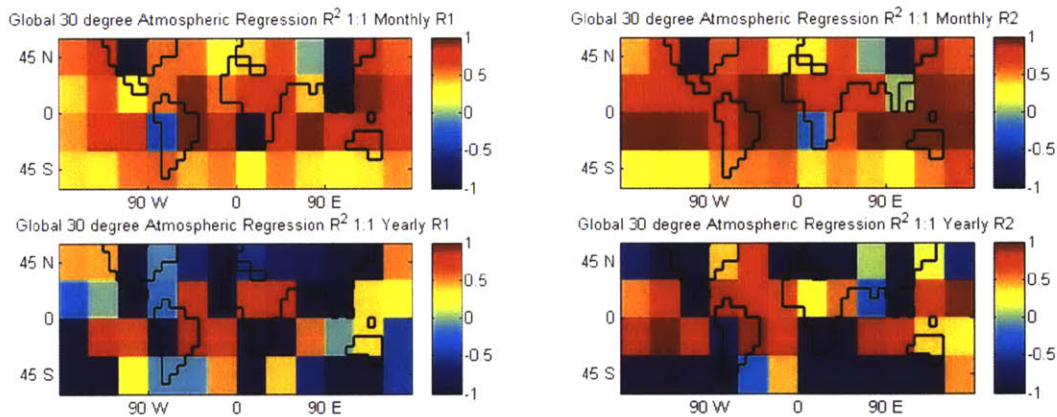


Figure 6-22: 30 degree Regression R<sup>2</sup> 1:1 Reanalysis Balances

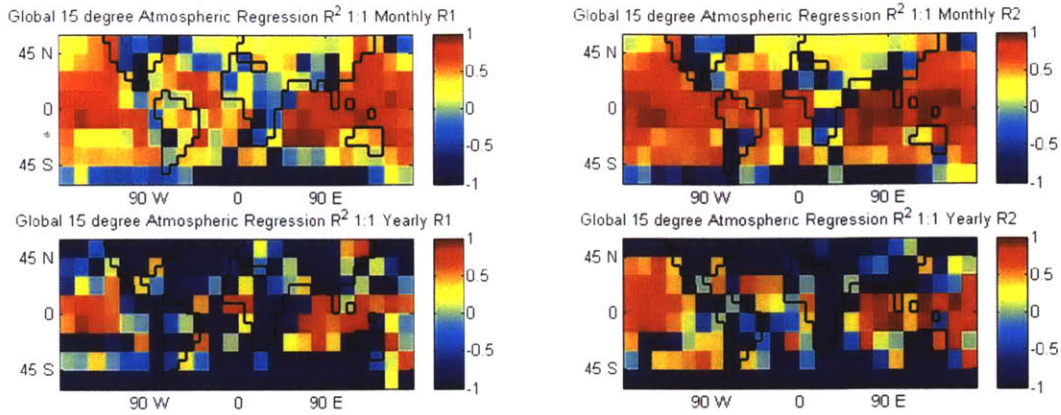


Figure 6-23: 15 degree Regression  $R^2$  1:1 Reanalysis Balances

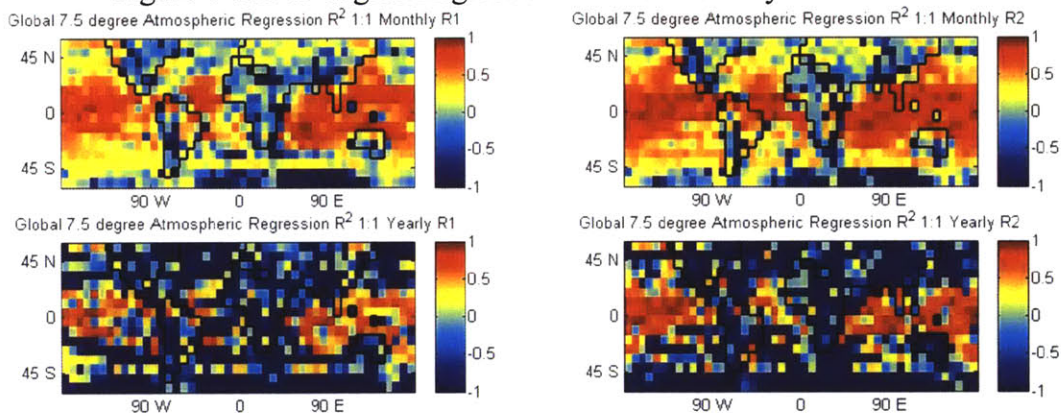


Figure 6-24: 7.5 degree Regression  $R^2$  1:1 Reanalysis Balances

### 6.1.9. Global Atmospheric Correlation Coefficient

The correlation coefficient and  $R^2$  metrics are very similar as they both measure the adherence of the two sided water balance to a linear relationship. Therefore the results of maps of the 30 degree correlation coefficient in Figure 6-25 are no surprise. Again the tropical regions show very good correlations while other points showing less correlation. For most balances the correlation coefficient is greater than .5 for all control volumes at the 30 degree resolution and monthly time scale, with the exception of two control volumes which are completely over land. Increasing the resolution to 15 degrees in Figure 6-26 all tropical regions still show a good correlation in the monthly Reanalysis-2 balance regardless of land or water. At the 7.5 degree scale in Figure 6-27 the correlations begin to break down in the tropical regions over land. Particularly weak correlations are found over northern landmasses at 15 and 7.5 degree resolution.

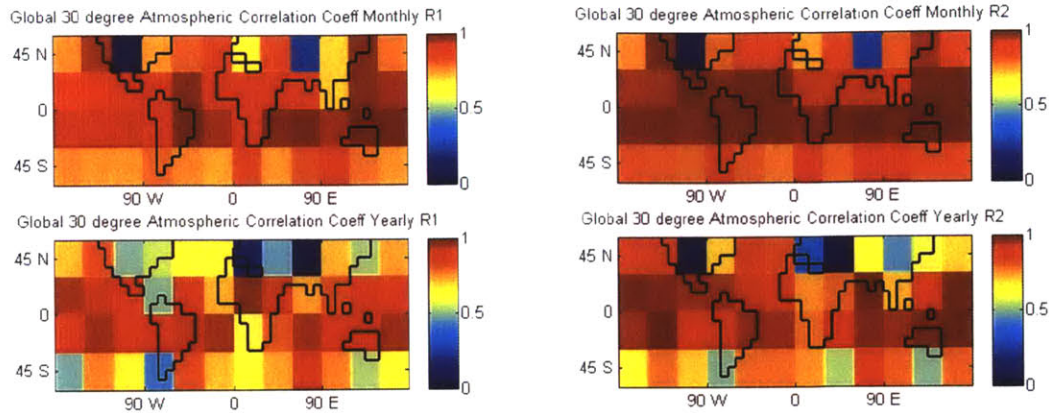


Figure 6-25: 30 degree Correlation Coefficient Reanalysis Balances

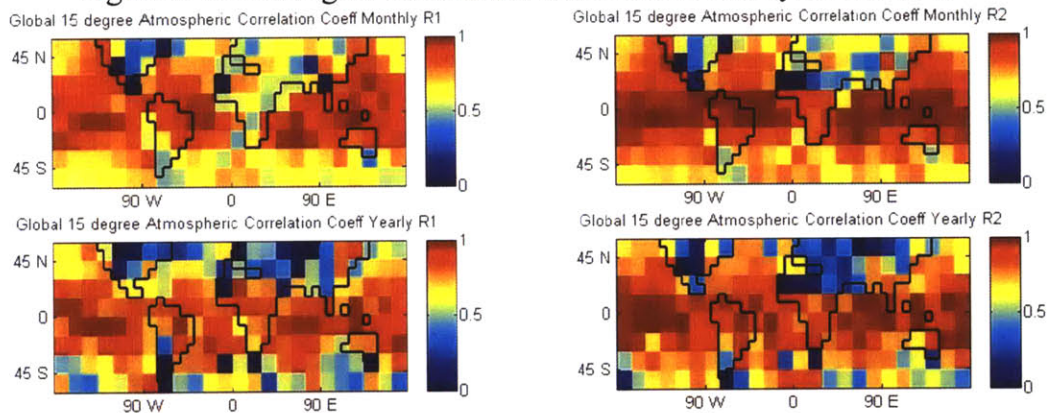


Figure 6-26: 15 degree Correlation Coefficient Reanalysis Balances

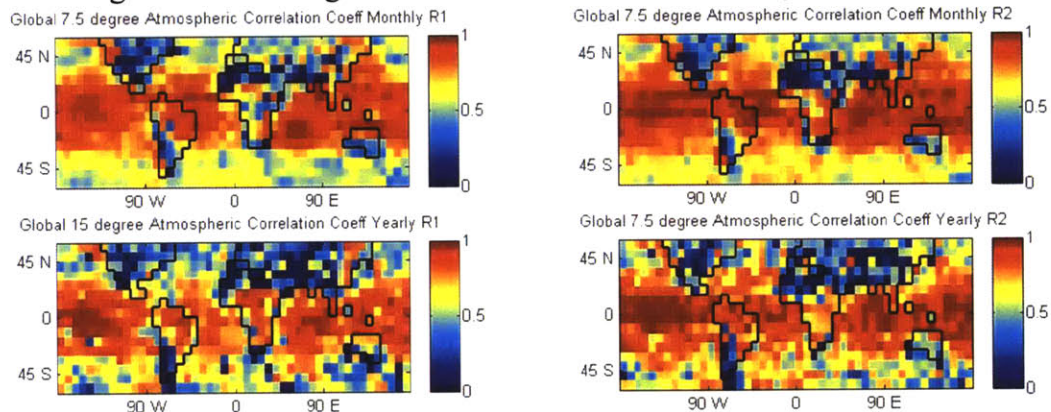


Figure 6-27: 7.5 degree Correlation Coefficient Reanalysis Balances

### 6.1.10. Global Atmospheric Root Mean Square Error

The root mean square error is an excellent tool for the analysis of the atmospheric balance. The RMSE is always positive thus the magnitudes of the error are easy to recognize on a global map. In Figures 6-28 through 6-30 the root mean square error is mapped for the global atmospheric water balance at 30, 15, and 7.5 degrees respectively. At the 30 degree resolution many regions yield RMSEs of less than 10 kg/m<sup>2</sup>-month especially on the yearly time scale. This is a very

low error. Problem areas arise where residual statistics indicated problems earlier, in Eastern Asia and Southern Africa. Errors are approximately doubled when resolution is increased from 30 degrees to 15 degrees as indicated by the change in colorbar. On the 7.5 degree resolution and monthly time scale, significant errors are found over almost all land masses. These large errors are indiscriminate to geographic features as they cover both mountainous and flat lands. Such errors are nowhere near as high or concentrated to land on the yearly time scale. On the yearly time scale many water balances over land have the same error as water balances over ocean. In the case of RMSE the best balances are found in the yearly Reanalysis-2 and the worst in the monthly Reanalysis-1 because of variability which exists in the monthly time scale.

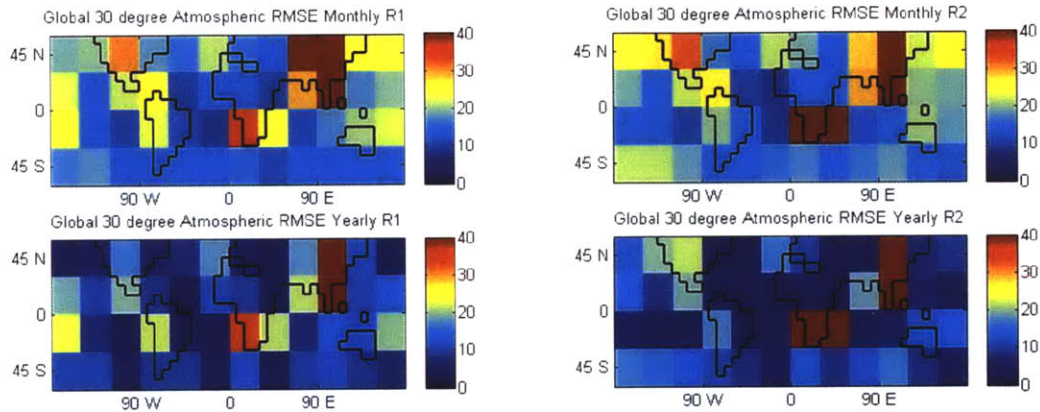


Figure 6-28: 30 degree RMSE Reanalysis Balances

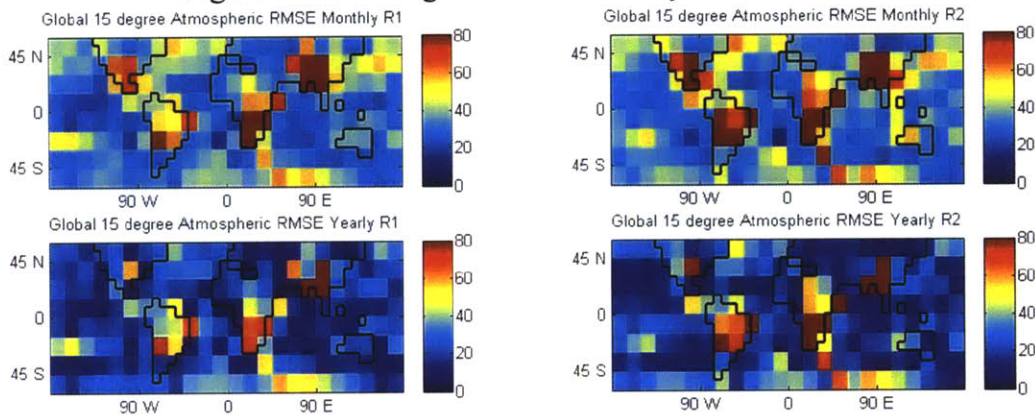


Figure 6-29: 15 degree RMSE Reanalysis Balances



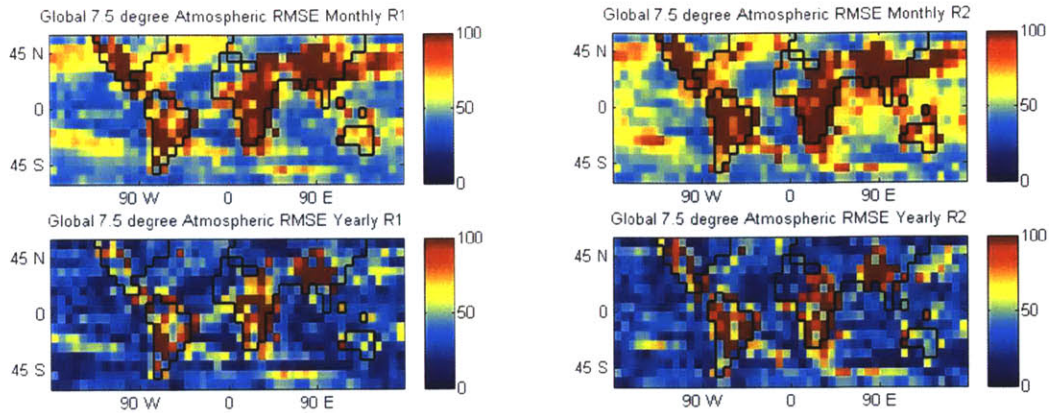


Figure 6-30: 7.5 RMSE Reanalysis Balances

### 6.1.11. Global Atmospheric Normalized Root Mean Square Error

The final analysis tool which can be applied to the global atmospheric water balance is the normalized version of the RMSE. This is displayed at various resolutions over Figures 6-31 through 6-33. For the 30 degree resolution higher normalized error exists generally over land, without much of a pattern, but at the 15 and 7.5 degree resolutions these errors become increasingly concentrated and localized. Specifically at the 7.5 degree resolution there is a large contrast between normalized errors greater than 2 (dark red) and normalized errors less than 1 (blue). These areas are not limited to land mass, although they persist over the African and Asian continents, they also exist in wake of continents specifically in the Southern Hemisphere. It is hard to ascertain the reason for the ocean errors in these locations as ocean water balances have behaved very regularly for most other properties. These areas of the ocean must be drier and clearly more error prone. The normalized absolute error and bias were a prior indicator.

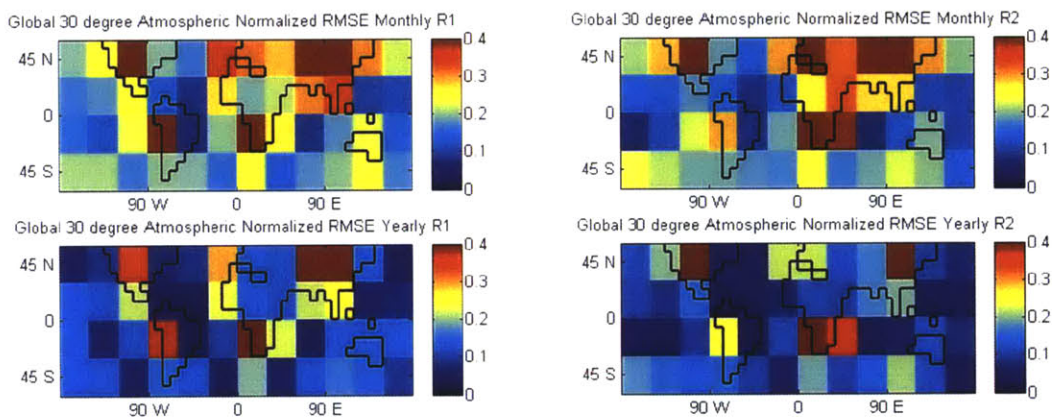


Figure 6-31: 30 degree NRMSE Reanalysis Balances

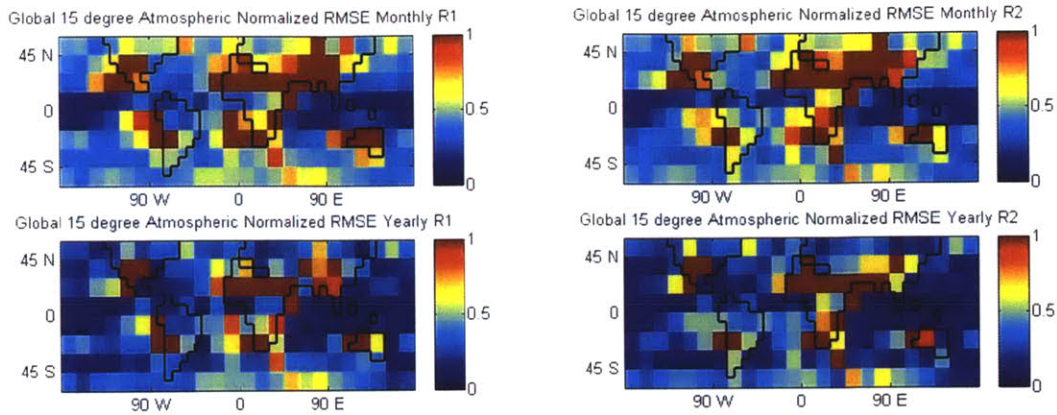


Figure 6-32: 15 degree NRMSE Reanalysis Balances

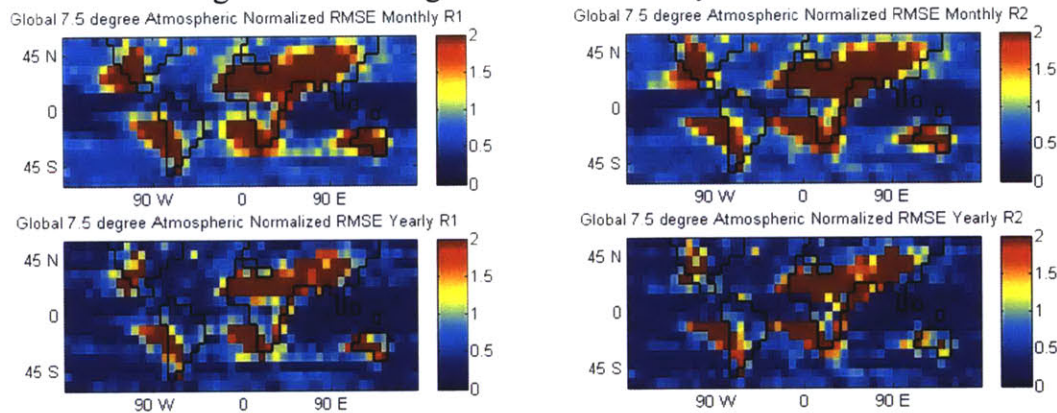


Figure 6-33: 7.5 degree NRMSE Reanalysis Balances

## 6.2. Precipitation Set Data Analysis

In the previous section the different properties of the global atmospheric water balance were looked at and evaluated by looking at spatial maps. The results on each spatial map can also be averaged over the control volumes in either the global (60 N to 60 S) or tropical (30 N to 30 S) regions to evaluate the general characteristics of all balances. This is a necessary simplification of analysis when different precipitation datasets are applied to each water balance.

An additional six precipitation datasets can be applied to each control volume in the global atmospheric water balance: the default reanalysis precipitation, the GPCP precipitation, the SSM/I precipitation, the OPI/Longwave-Radiation and OPI/Model precipitation datasets, and the CMAP precipitation dataset. The CMAP2 precipitation dataset contains errors globally and therefore is dropped from consideration. Additionally, the GPI and TRMM 3B43 precipitation datasets can be applied to the tropical region control volumes between 30 degrees N and 30 degree S.

Thus analyzing different characteristics as applied to the global atmospheric water balance eight basic balances are created varying reanalysis (2), time scale (2), and creating separate analyzes for global and tropical regions. The tropical region analysis allows for the inclusion of two additional precipitation datasets as mentioned before and is also of interest as since during the

spatial analysis it was observed that the tropical water balances had much different characteristics, especially for the 30 degree resolution.

In Figure 6-34 the averaged absolute errors are displayed averaging over all control volumes for the different water balance characteristics. Resolution is decreased from the top row to the bottom row. In general, across each balance the reanalysis precipitation provides the lowest absolute error, particularly in the tropical regions. The precipitation errors for the global cases are generally the same for all datasets as the GPCP and SSM/I applications provide generally higher errors. The same can be said for only tropical control volumes on the right with even higher absolute error in the water balance found when the tropical specific 3B43 and GPI precipitation are applied. These effects are dampened as resolution decreases, but are generally universal.

In Figure 6-35 the normalized absolute error is displayed in the same way. Trends are generally the same with an even larger increase in SSM/I and GPI precipitation error. These precipitation datasets are both satellite based and their variability seems to increase the normalized absolute error. For the 7.5 degree resolution absolute errors for all precipitation datasets are on average of the same order of magnitude as the precipitation itself. It is known that over the oceans these errors are fairly low; therefore over land the errors must be even higher. This contrast between land and ocean points will also be analyzed in a later section.

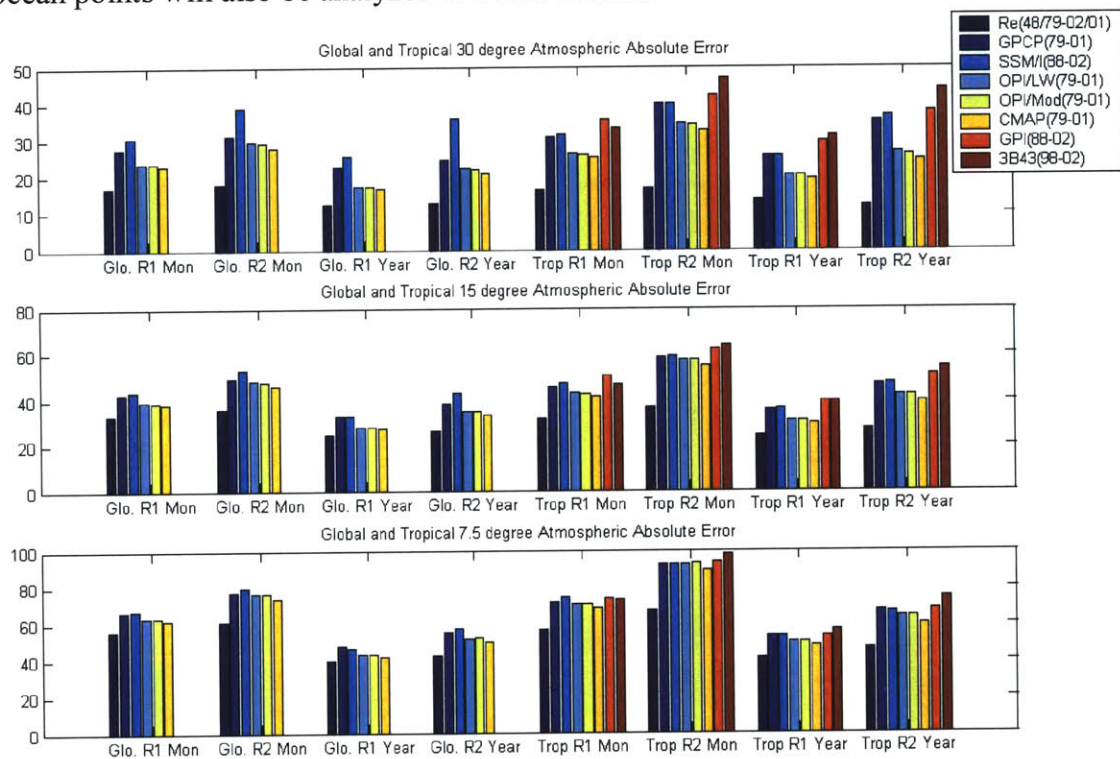


Figure 6-34: Global and Tropical Analysis of Different Precipitation Datasets – Absolute Error

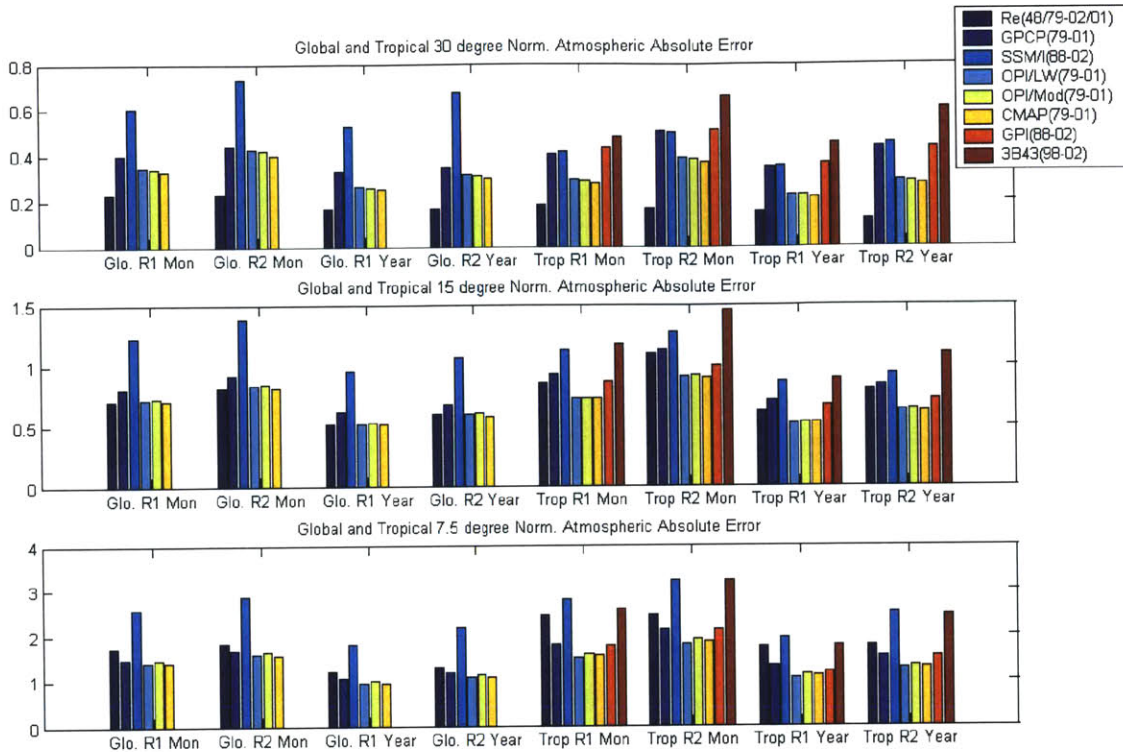


Figure 6-35: Global/Tropical Analysis of Different Precip. Datasets – Norm. Absolute Error

Of special interest in the global atmospheric balance is the averaged global bias of all water balances. If the reanalysis is truly a global product there should be little if any bias in the long term water balance, when all spatial areas are taken into account. When other precipitation datasets are introduced, biases are more likely as the water budget of the reanalysis was made to match its own precipitation data and not alternatives. The biases for different global atmospheric characteristics are given in Figure 6-36.

As suspected, the biases are lowest for the reanalysis precipitation for each situation, and across time scale bias remains constant. Bias residuals behave similarly to absolute error residuals as far as precipitation dataset, in terms of departure from zero. All biases are negative for every situation. Interestingly, bias is not only constant across time scale, but it also nearly constant across resolution. This proves that the integrity of the different datasets used in each case is maintained throughout the water balance process. The biases for the reanalysis are very low for all situations in the Reanalysis-1 and the Reanalysis-2 indicating that the water balance in the reanalysis is preserved in between both the 60 degree and 30 degree latitudes. All of the moisture flux terms which contain sizeable errors at some resolutions cancel out in the spatial averaging and the water balance nearly balance and simplifies to  $P=E$ .

This is an important observation in that the total moisture flux globally is approximately zero when complete spatial averaging is performed (as it should be), and any discrepancies are likely due to differences in moisture flux through the top and bottom boundaries (60 or 30 degrees latitude). Alternative precipitation datasets indicate much more precipitation globally within the

bands being analyzed. It would seem then that a proper balance would include increased precipitation and evaporation. The reanalysis evaporation should be higher in order to match more accurate alternative precipitation datasets.

The normalization of the bias in Figure 6-37 yields consistent results at the 30 degree and 15 degree resolutions. However, as resolution increases, biases become less negative and even approach zero. This does not mean a balance is achieved. There are many tropical regions which are dry and mountainous and these normalized errors can be very high because of very low average precipitation being used as a normalization tool. This effect will increase with increased resolution because the precipitation data can be more specific at higher resolutions and therefore have greater extremes.

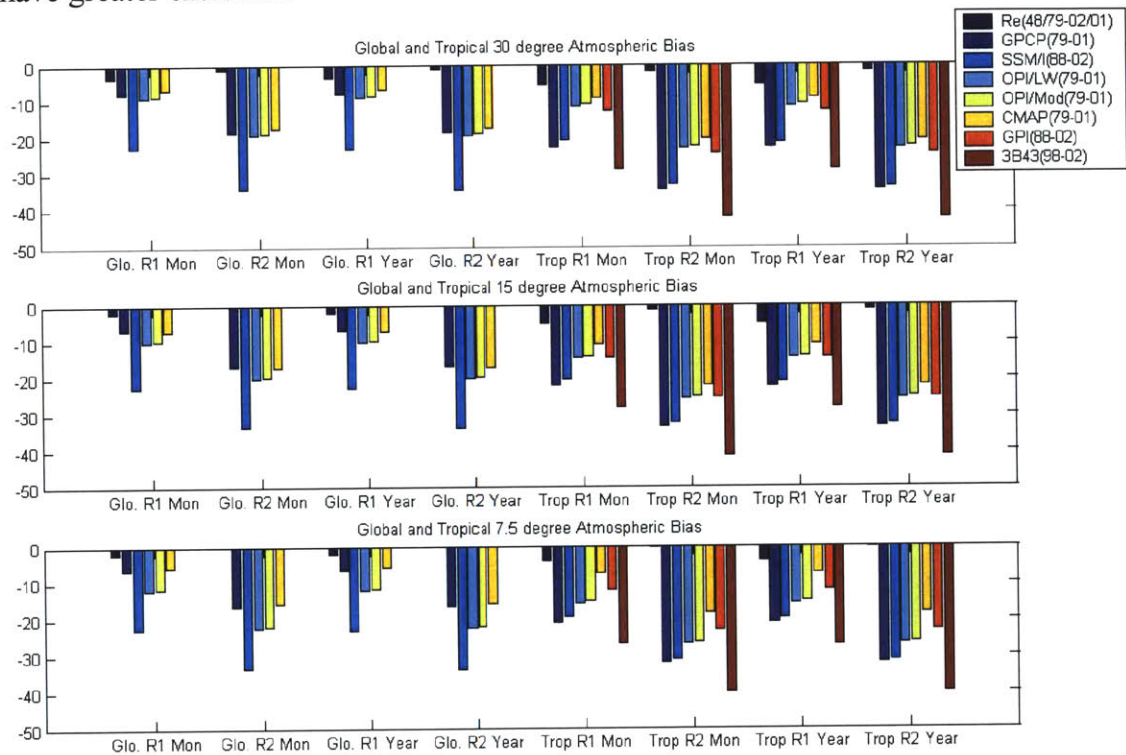


Figure 6-36: Global and Tropical Analysis of Different Precipitation Datasets – Bias Error

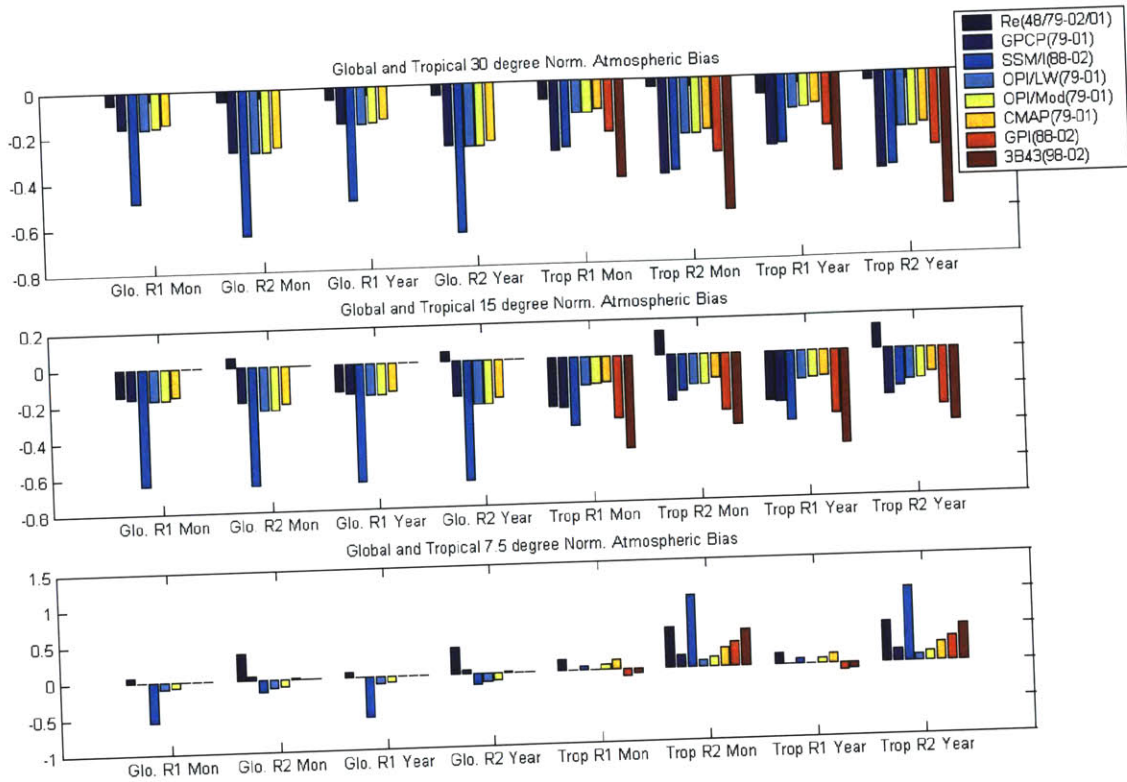


Figure 6-37: Global and Tropical Analysis of Different Precip. Datasets – Norm. Bias Error

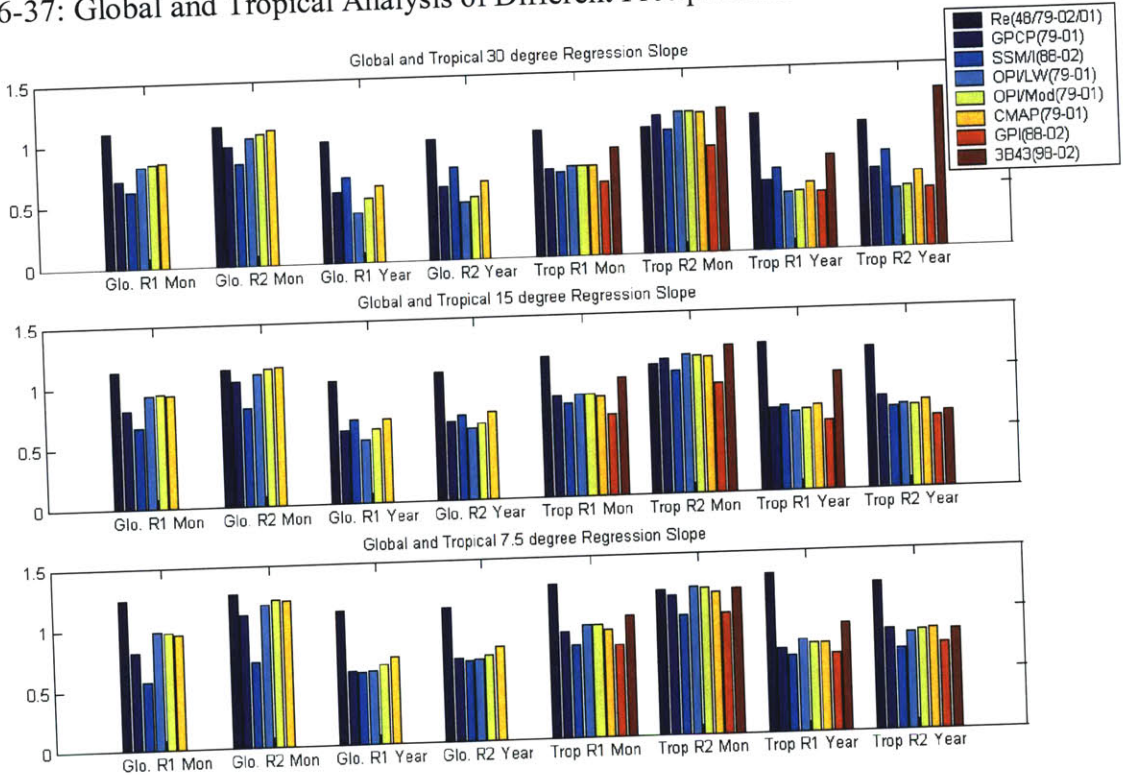


Figure 6-38: Global and Tropical Analysis of Different Precip. Datasets – Regression Slope

The regression intercept is bypassed here as it yields no new insight. The same is true for the  $R^2$  and  $R^2$  1:1 analyzes as they yield nothing that can not be evaluated by the correlation coefficient. The slope and correlation coefficient for these spatially averaged analyzes are displayed in Figure 6-38 and 6-39. Looking at the global results for slope in Figure 6-38 the reanalysis precipitation yields the slope closest to one clearly at every resolution. For all yearly cases the average slope is less than one indicating weaker signals in the moisture flux time series than the P-E time series. The distinction between precipitation datasets is clearer in the yearly time scale, and on the 7.5 and 15 degree resolutions. Slope results tend to depart more from one with increase in resolution. For the tropical cases the regression slope is generally closer to one. The tropical precipitation datasets yield comparable results to the other precipitation datasets over this area.

Turning to correlation coefficient, the reanalysis precipitation clearly provides for the best correlation of the two sides of the water balance for every situation and resolution presented. Correlation generally decreases with increased resolution which is expected. The other precipitation datasets vary in terms of effectiveness. However there is a general increase in average correlation coefficient for the tropical control volumes. A stronger seasonal cycle is more likely in tropical regions leading to higher correlation.

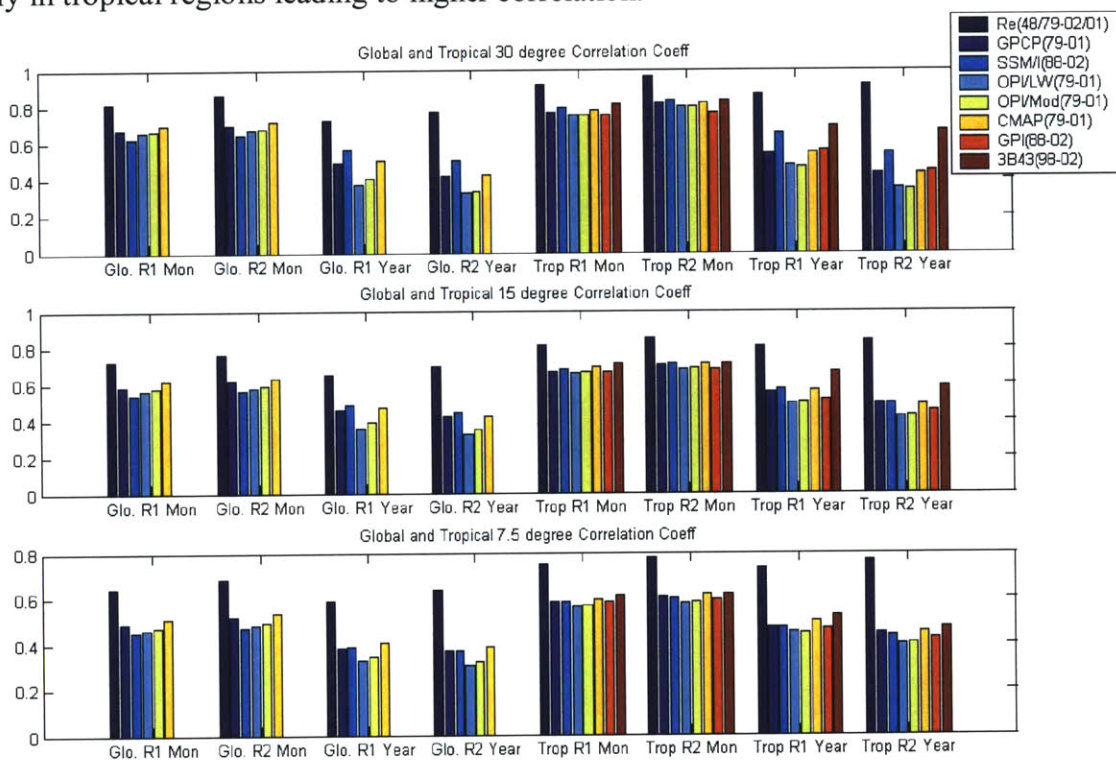


Figure 6-39: Global and Tropical Analysis of Different Precip. Datasets – Correlation Coeff

In Figures 6-40 and 6-41 the RMSE and Normalized RMSE are presented as the other water balance properties in this section. For every scenario evaluated by the RMSE the reanalysis precipitation again provides the best results. It has become clear that the reanalysis precipitation is best suited to close its own water balance, globally. In general, results of the RMSE property

are similar to the residual evaluation of the different balances especially on the monthly basis. Error nearly doubles over each resolution for both the RMSE and NRMSE. Error increases, but so did correlation when only the tropical control volumes are considered. These findings are counterintuitive and must indicate there is much more water in the tropical water balances over which error can range. NRMSEs are especially high for the SSM/I precipitation dataset. The normalizing ability of this dataset over dry regions has a large affect on this property.

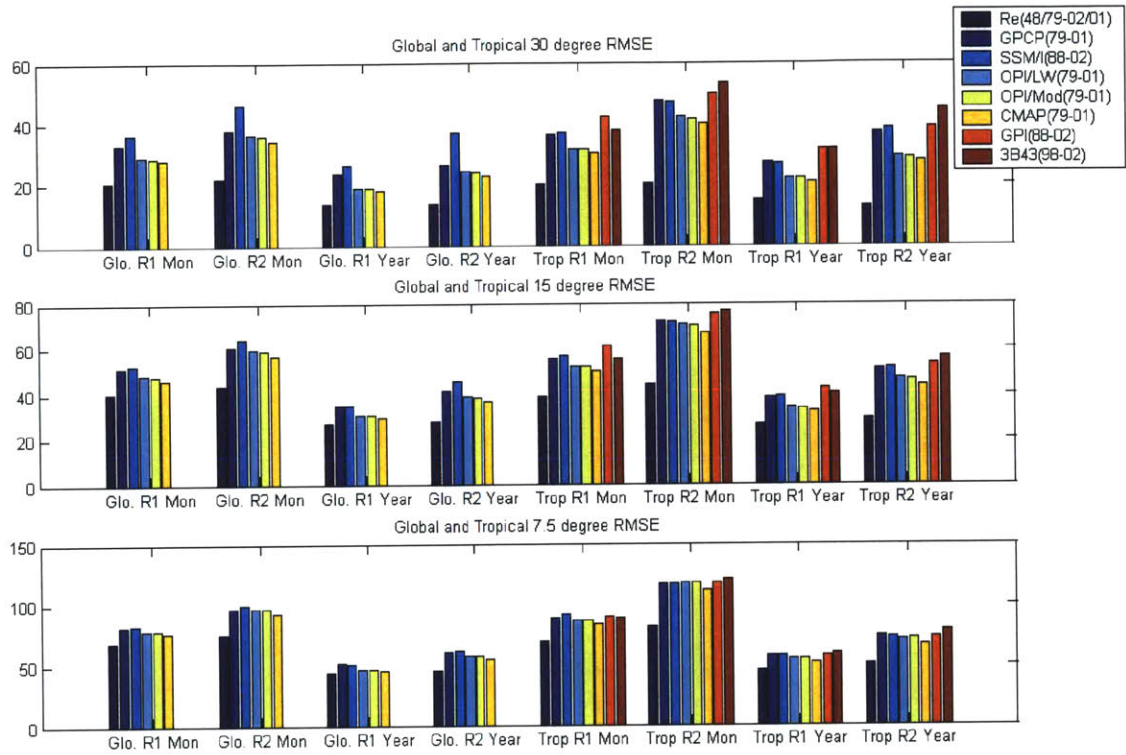


Figure 6-40: Global and Tropical Analysis of Different Precipitation Datasets - RMSE



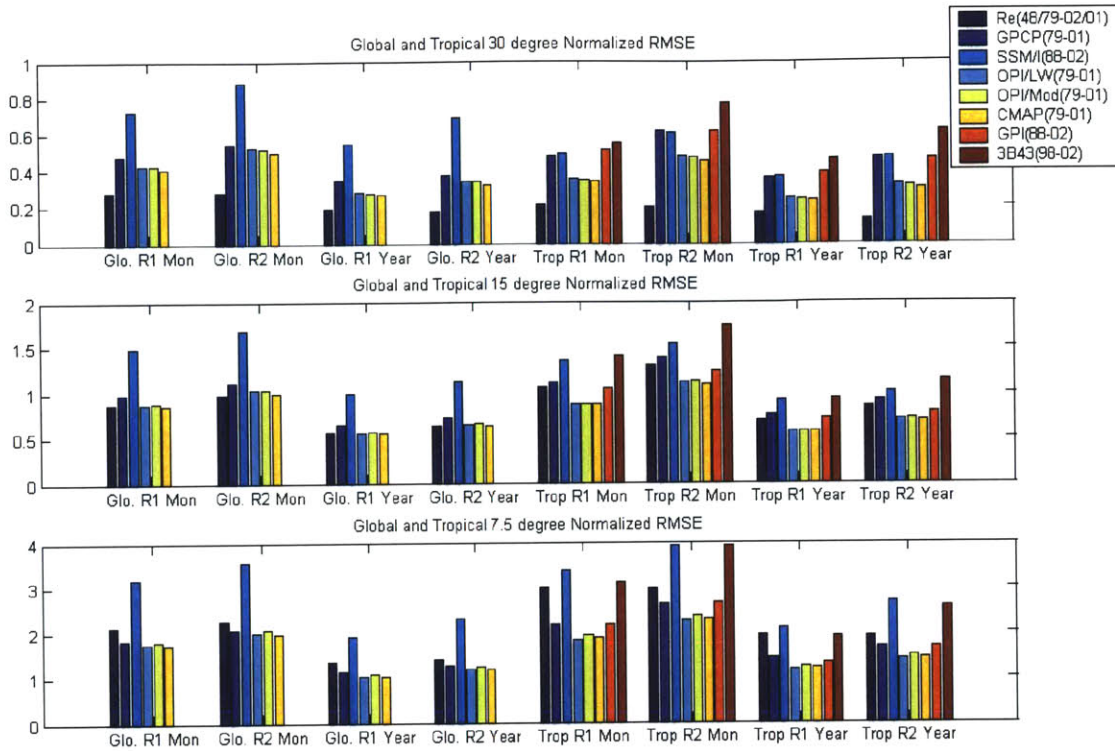


Figure 6-41: Global and Tropical Analysis of Different Precipitation Datasets – NRMSE

### 6.2.1. Spatial Precipitation Data Analysis

Each precipitation dataset can be applied to a global water balance map at each resolution to generate a unique spatial map. In other words every bar of each graph in the previous Section 6.2 represents the averaged value of the spatial mapping for that property. However, this results in an amount of maps too numerous to analyze in detail. Therefore, a number of these maps have been created varying precipitation over the Reanalysis-2 monthly and yearly time scales for the global and tropical areas and placed in Appendix B. For the Reanalysis-2 monthly and yearly time scale four precipitation datasets are applied to all of the control volumes on the global scale and plotted for each water balance property and resolution. These precipitation datasets are the Reanalysis, GPCP, SSM/I, and CMAP precipitation datasets. The GPCP and CMAP precipitation datasets are likely the best alternative precipitation datasets, while the SSM/I is a unique satellite based dataset. Additionally, the same plots are made for all precipitation datasets for the tropical region.

A few of these plots are selected for discussion. The yearly Reanalysis-2 is selected at the 7.5 degree resolution as these balances are most important to the main purpose of evaluating regional water balances at higher resolutions. Important water balance features of the global and tropical water balances are presented.

In Figures 6-42, 6-43, and 6-44 absolute error, bias, and RMSE for different precipitation datasets are presented. The figures look generally similar in terms of absolute error, with some

problem areas over land which persist regardless of precipitation dataset. The tropical ocean absolute error is affected by all three of the alternative precipitation datasets, and appears to be the major difference in the figures. The reanalysis ocean water balances were observed to be very strong in Section 6.1, but are weakened by more reliable precipitation datasets (GPCP and CMAP) here. This indicates the reanalysis model is balancing itself. These trends are also seen in Figure 6-44 in terms of RMSE. The bias maps in Figure 6-43 indicate the signs of these imbalances. For the alternative precipitation datasets biases are both high and low over land and vary without a pattern. The ocean biases which correspond to higher absolute errors are primarily negative indicating a surplus in precipitation over the ocean. This is consistent with the overall results in the previous section, and is very interesting as with the reanalysis water balances all ocean balances showed great agreement, but supplying alternative precipitation that may be better than the reanalysis provides problems over the ocean.

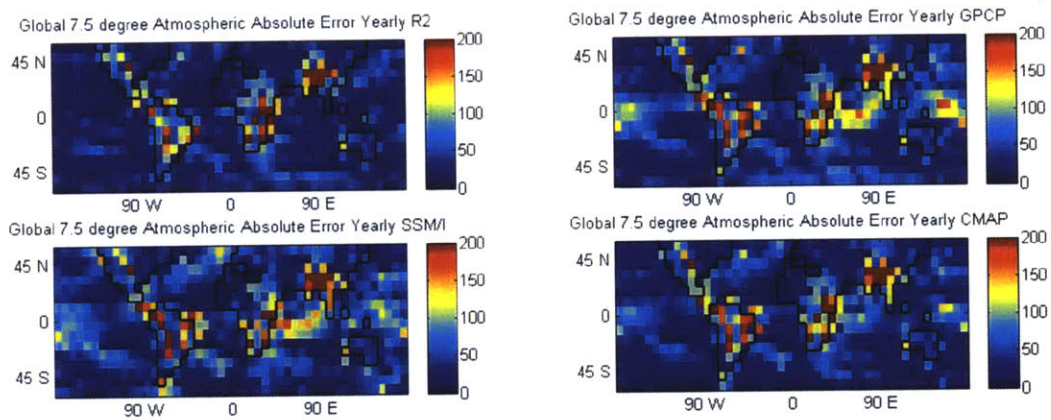


Figure 6-42: 7.5 degree Absolute Error Different Precipitation Global Yearly

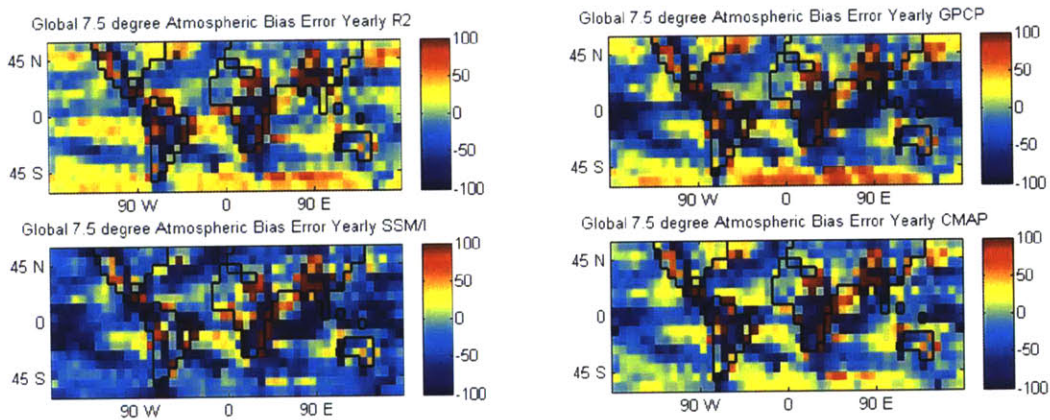


Figure 6-43: 7.5 degree Bias Error Different Precipitation Global Yearly

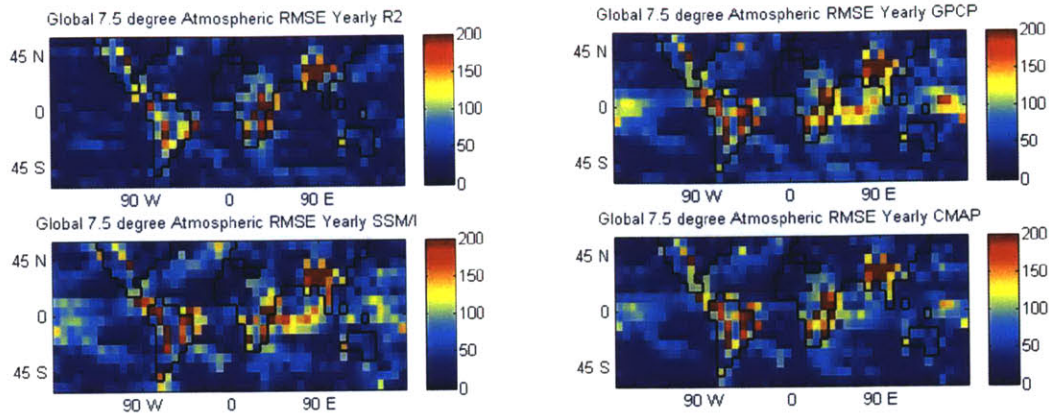


Figure 6-44: 7.5 degree RMSE Different Precipitation Global Yearly

In Figures 6-45 and 6-46 the two sided water balance is represented for different precipitation datasets. In terms of slope, there is a general departure of slope from one for all alternative precipitation datasets shown, and the effect is about the same for all three alternatives. The same can be said about the correlation coefficient. The relationship between the P-E and primarily the moisture flux term Q (as change in yearly precipitable water is nearly zero) is not represented as well when alternative precipitation datasets are applied.

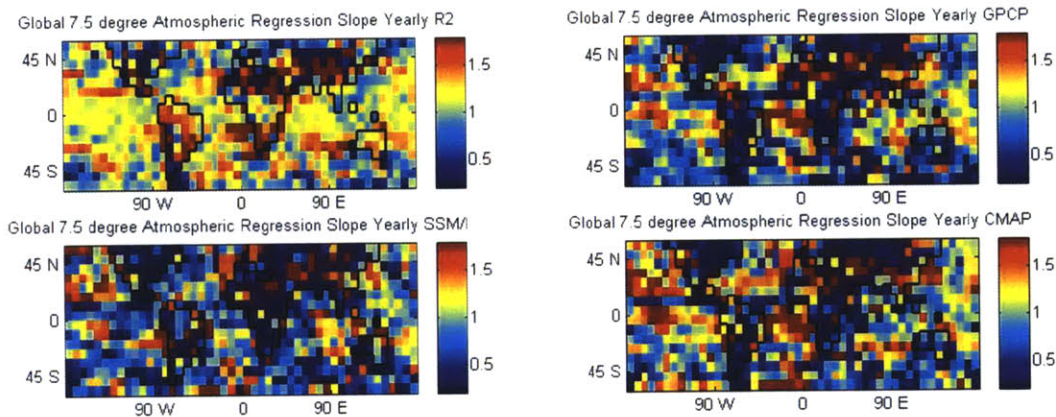


Figure 6-45: 7.5 degree Regression Slope Different Precipitation Global Yearly

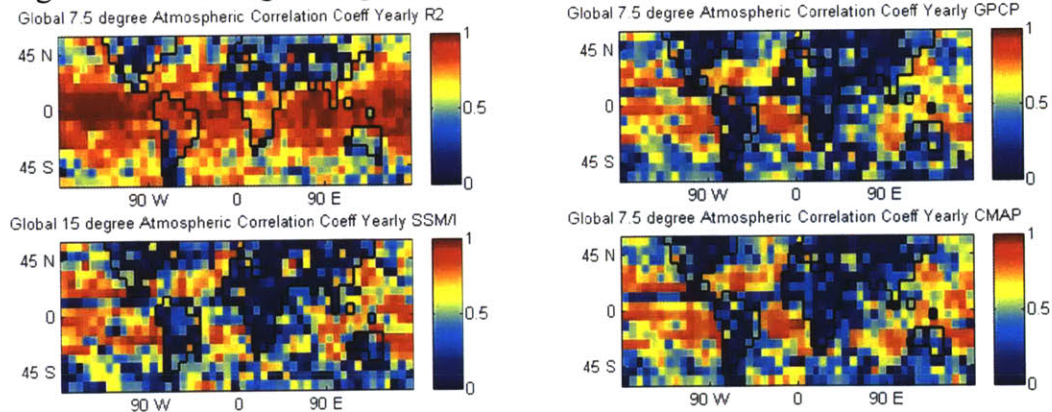


Figure 6-46: 7.5 degree Correlation Coefficient Different Precipitation Global Yearly

It would stand to reason that with so many alternatives, differences would be discernable between precipitation datasets when looking only at tropical control volumes. However, looking at the absolute error results in Figure 6-47, few differences are apparent besides the obvious deviation of all of the alternative precipitation datasets from that provided by the reanalysis. The absolute error of the water balances over tropical regions is very good for the reanalysis in most areas. The tropical precipitation datasets (3B43 and GPI) provide little additional insight to the water balance. If anything, increases in absolute error are seen over the ocean especially in the Indian and Pacific Oceans. These errors can be observed as negative biases over the Indian Ocean in Figure 6-48. The absolute errors and biases are strongest for the TRMM 3B43 dataset, perhaps due to the fact only five years of data is available for the TRMM 3B43 precipitation, and a point or two of bad data may throw the balance off. Similar spatial observations are supported by the RMSE in Figure 6-49.

Turning to the analysis of the two sided balance the regression slopes and correlation coefficients are presented for the tropical control volumes in Figures 6-50 and 6-51. Focusing on the added precipitation datasets in terms of slope, very poor regression slopes persist for specifically the TRMM 3B43 precipitation dataset. Again, these poor regressions are likely due to the fact only five points are being regressed, and a strong regression cannot be found with so few points. A similar finding is found in terms of correlation coefficient. This problem requires a look at these same figures except for the monthly time scale in Appendix B. A quick look to these figures indicates similar problems with the TRMM 3B43 on a monthly scale, where the times series consists of 60 months. The precipitation dataset is simply too short to compare with other precipitation datasets in this type of analysis.

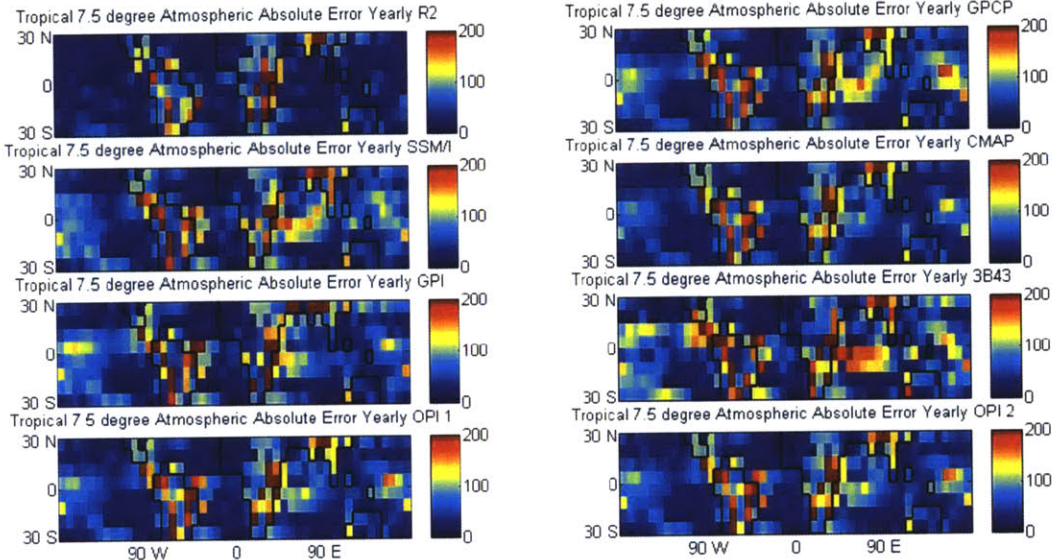


Figure 6-47: 7.5 degree Absolute Error Different Precipitation Global Yearly

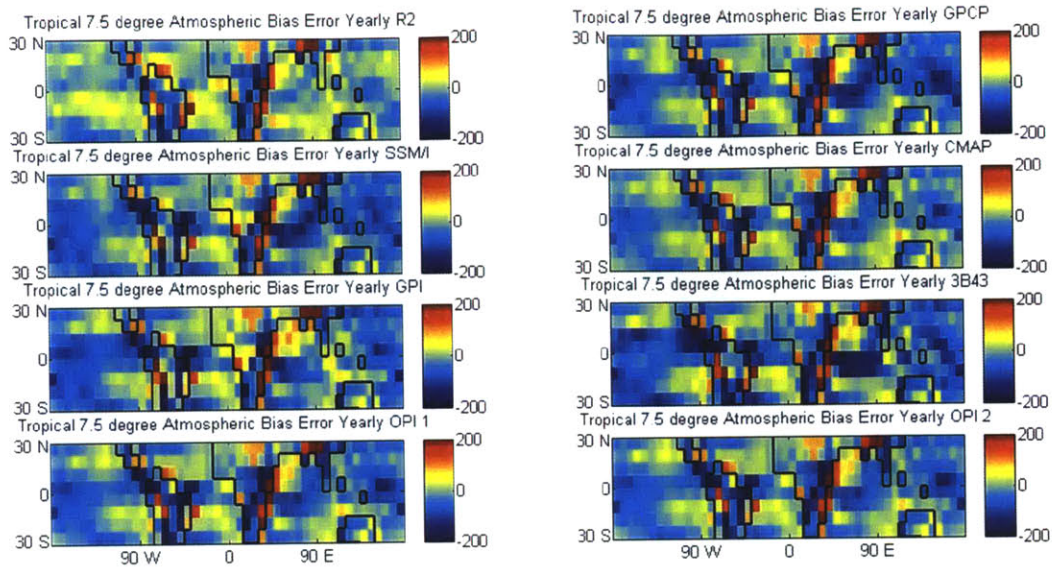


Figure 6-48: 7.5 degree Bias Error Different Precipitation Global Yearly

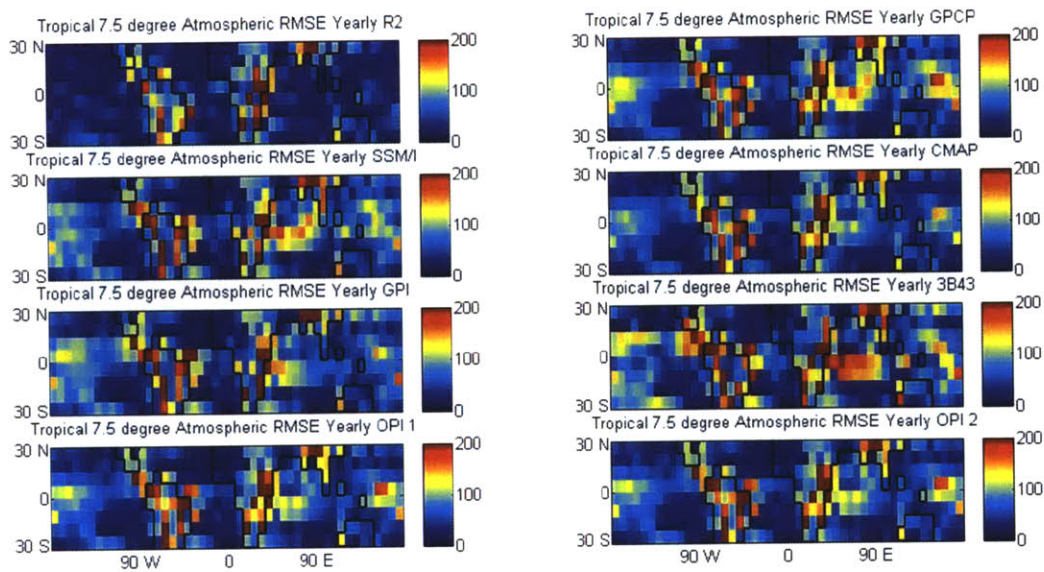


Figure 6-49: 7.5 degree RMSE Different Precipitation Global Yearly

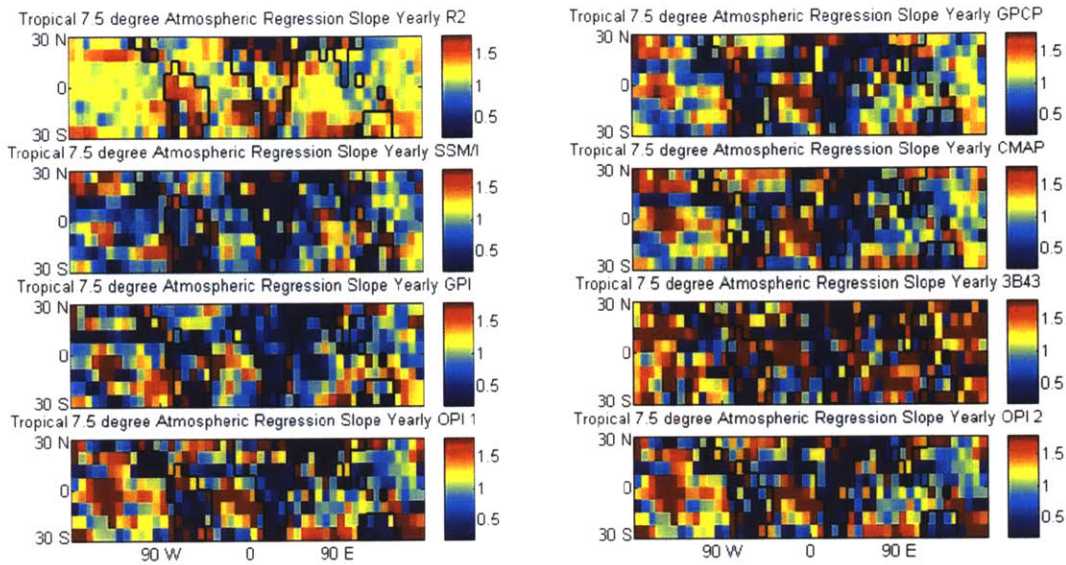


Figure 6-50: 7.5 degree Regression Slope Different Precipitation Global Yearly

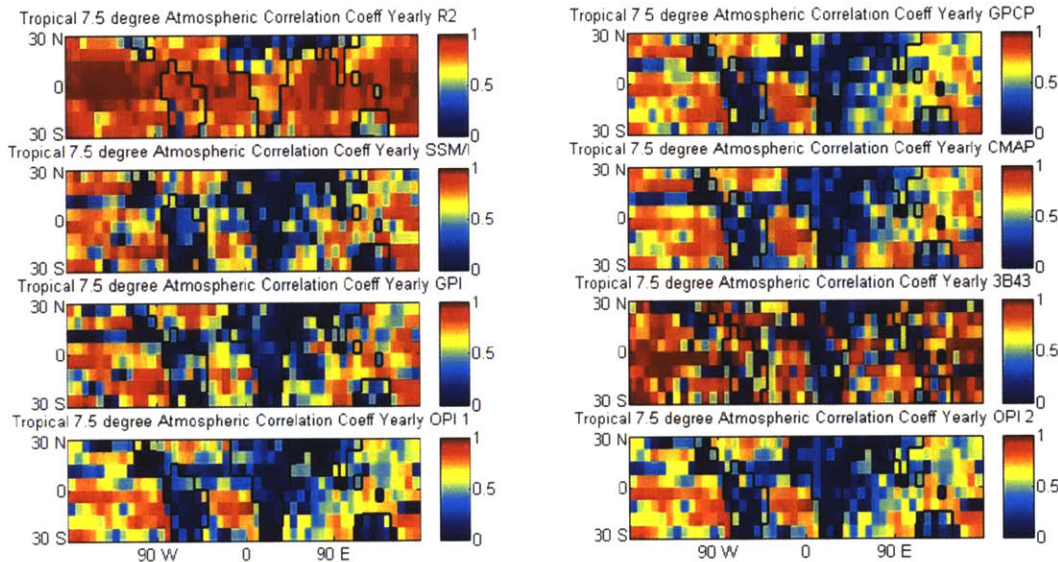


Figure 6-51: 7.5 degree Correlation Coefficient Different Precipitation Global Yearly

With the addition of alternative precipitation datasets, the reanalysis mapped water balance properties look generally better, with alternative precipitation datasets having strikingly similar properties. Key differences between the reanalysis and alternatives are found over the ocean and the errors in more reliable precipitation datasets on a global scale indicates errors in both precipitation and evaporation in the reanalysis, as on the global scale these water balance properties should equal each other. The bias analysis indicates  $P=E$  for the reanalysis, but the reanalysis evaporation is not consistent with alternative precipitation datasets.

### **6.3. Time Period Data Analysis**

The water balance can be broken into different time periods. In the case of the Reanalysis-1, this breakdown is more useful for a multi year analysis as there are fifty five years of data. The water balance here can be considered on a decadal basis and a separate evaluation can be performed every ten years for both the monthly and yearly time scales. This is particularly useful because there is a likelihood that over the last fifty years the quality of information has improved in the reanalysis and this hypothesis can be tested in terms of the atmospheric water balance.

For the Reanalysis-2, such a long time period analysis isn't as meaningful. A different analysis can be performed on the monthly time scale. The data here can be separated by month, and thus the properties of this water balance can be observed for twelve different months. It was also deemed useful to perform this same analysis by substituting the Reanalysis-2 precipitation with the TRMM 3B43 precipitation and considering only the corresponding tropical points. The Reanalysis-1 is used with the TRMM 3B43 precipitation in order to increase the length of the balance by one year.

For both of these analyses spatial differences in water balance properties are extremely limited, specifically in the evaluation of the two sided balance. Therefore these differences will not be looked into in detail, only presented for a few properties to show there are few differences. Most of the analyzes in this section deal with the overall averaging of water balance properties across all of the control volumes in the different time periods and resolutions. The seasonal analysis is performed first followed by the decadal analysis.

#### **6.3.1. Seasonal Variations in the Reanalysis-2**

In this section several water balance properties are analyzed as global averages over the different months and different resolutions of the Reanalysis-2 monthly balance. The absolute errors for these balances are displayed in Figure 6-52. Across the three different resolutions there is an increase in absolute error over the summer months peaking in July. The pattern is very regular for the seasonal cycle with minimums in the winter months and maximums in the summer months. When normalization is performed, the difference between errors in resolutions is greatly increased. Minimum errors occur here in the spring and the fall with much larger maximum errors in the summer. When only points from one month are considered especially in the 7.5 degree case, regions arise particularly in the extreme winter and summer months over which monthly precipitation is very small. Any errors in the water balance over these regions caused by evaporation and moisture flux are therefore greatly amplified by normalization with low precipitation. For instance over desert regions it may never rain in the winter over a 7.5 degree grid space.

In terms of regular bias in Figure 6-53 a general increase in bias is observed over the summer months. What is remarkable here is the magnitude of the bias is on the order of  $1 \text{ kg/m}^2\text{-month}$  over all seasons. For all months the Reanalysis-2 provides a very good water balance with hardly any overall residual. In other word  $P=E$  for every month averaged over the twenty three years of the balance. A nearly perfect balance is achieved on the global basis. When bias is normalized however, trends are seen similar to that of normalized absolute error. There is clear

interference as resolution decreases by outlier points resulting in a 12 month profile similar to the normalized absolute error at each resolution.

In Figures 6-54 and 6-55 the monthly regression and correlation statistics for this water balance analysis are given. The average regression slope for each month is very close to one and there are few differences between months. The regression slope is often most useful on the monthly time scale when all months are being analyzed. However, when only one month is analyzed, the signals on each side of the balance are unique to the month being analyzed. Interestingly this property does not degrade with resolution. What likely happens as resolution increases is that overestimates and underestimates cancel each other out. Equally impressive for the regression analysis is the regression intercept property which is on the order of magnitude and behaves like the bias, being nearly zero over all twelve months. The regression  $R^2$  and correlation coefficients for this experiment are split between figures 6-54 and 6-55 with differences between months being very small. A slight decrease is seen for the summer months which corresponds to the higher residual errors found for these months, but the differences are small. The results of the RMSE and NRMSE statistics for this balance are very similar to the absolute error results for both the normalized and non-normalized cases.

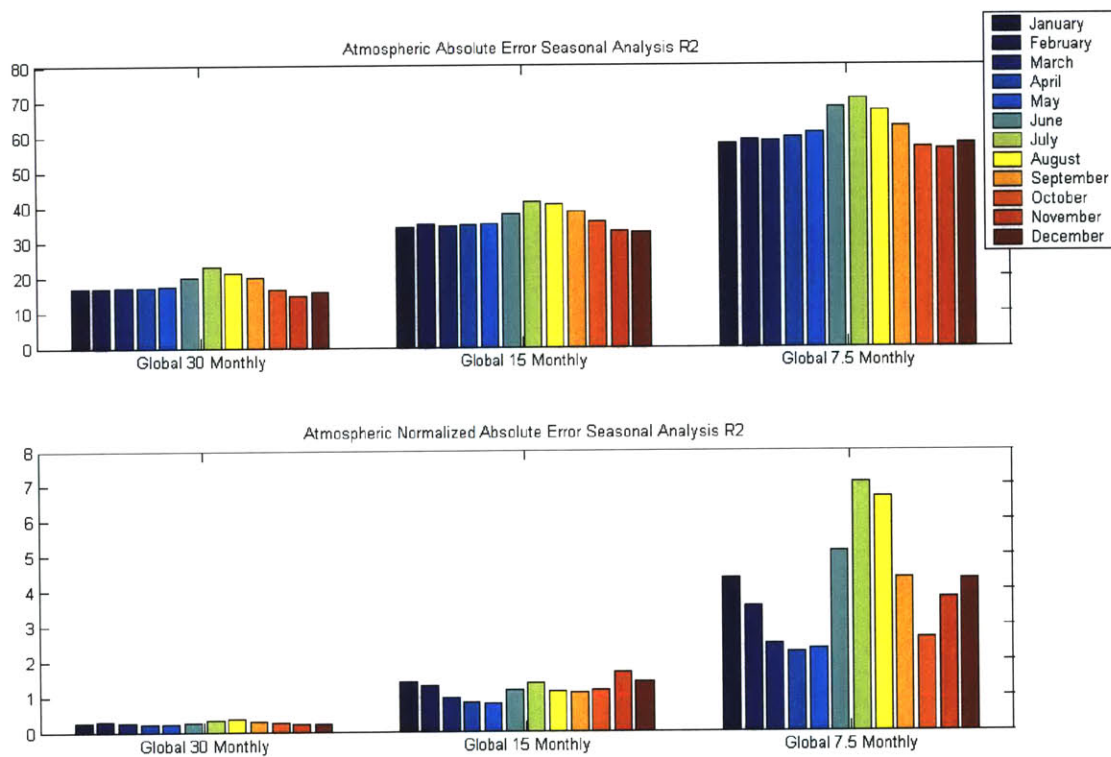


Figure 6-52: Global Analysis of Different Months – Absolute Error



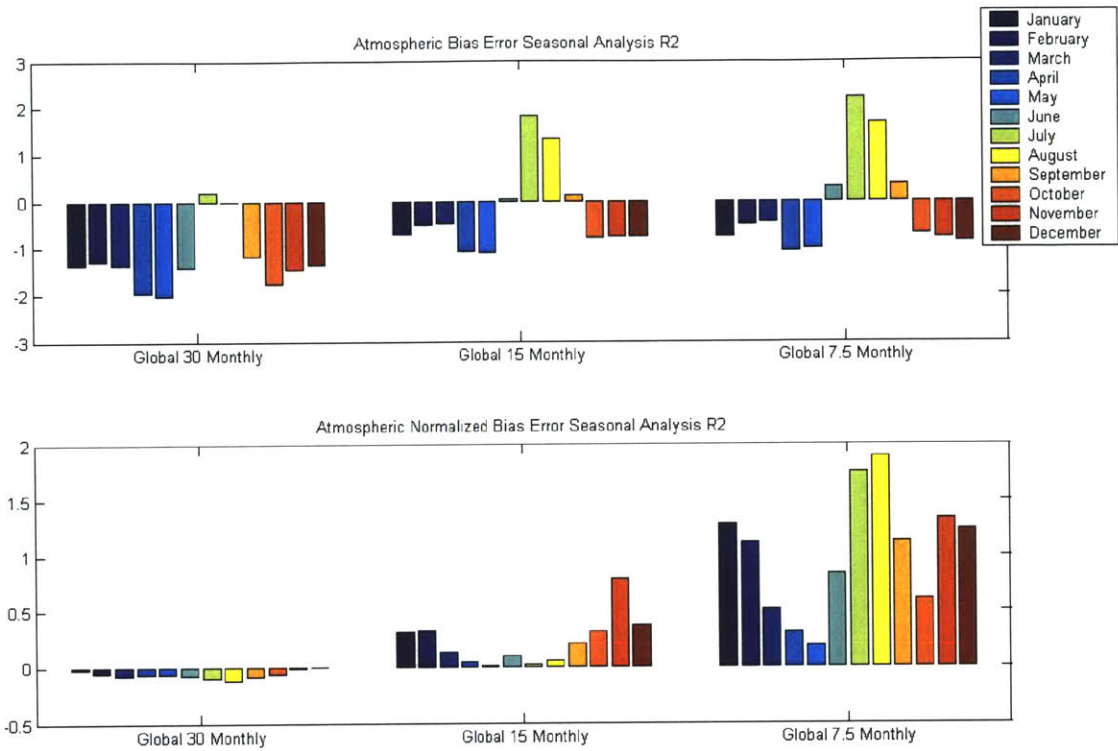


Figure 6-53: Global Analysis of Different Months – Bias Error

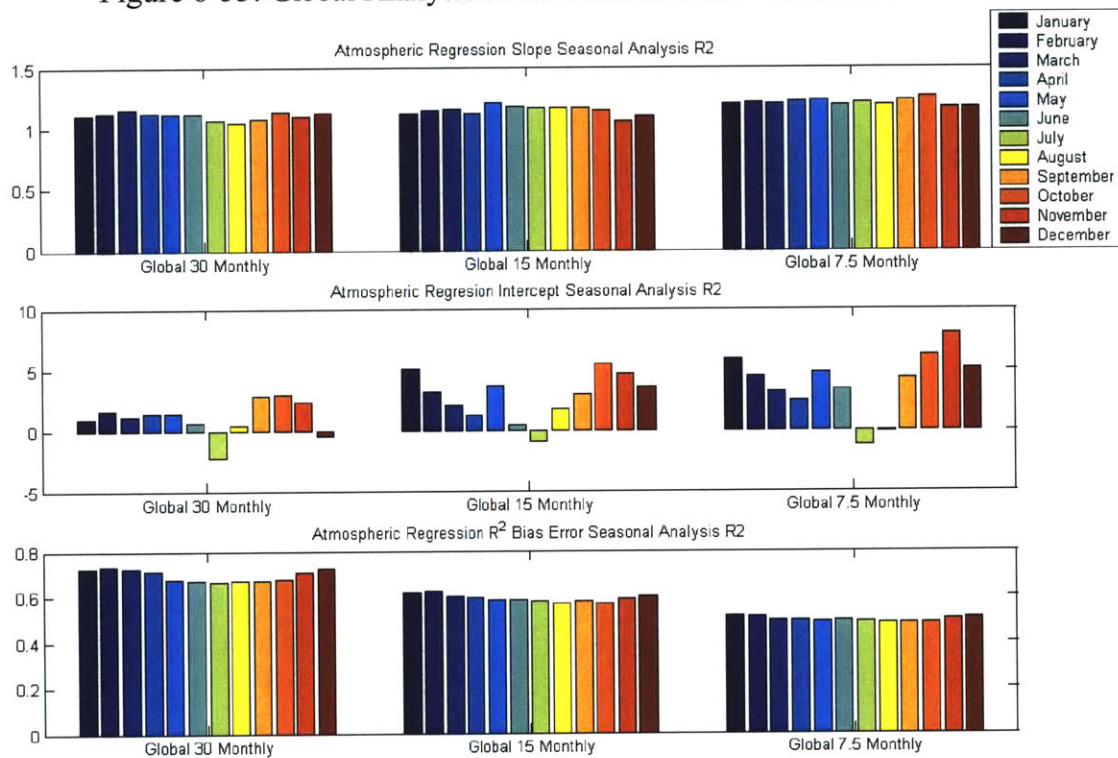


Figure 6-54: Global Analysis of Different Months – Regression Statistics

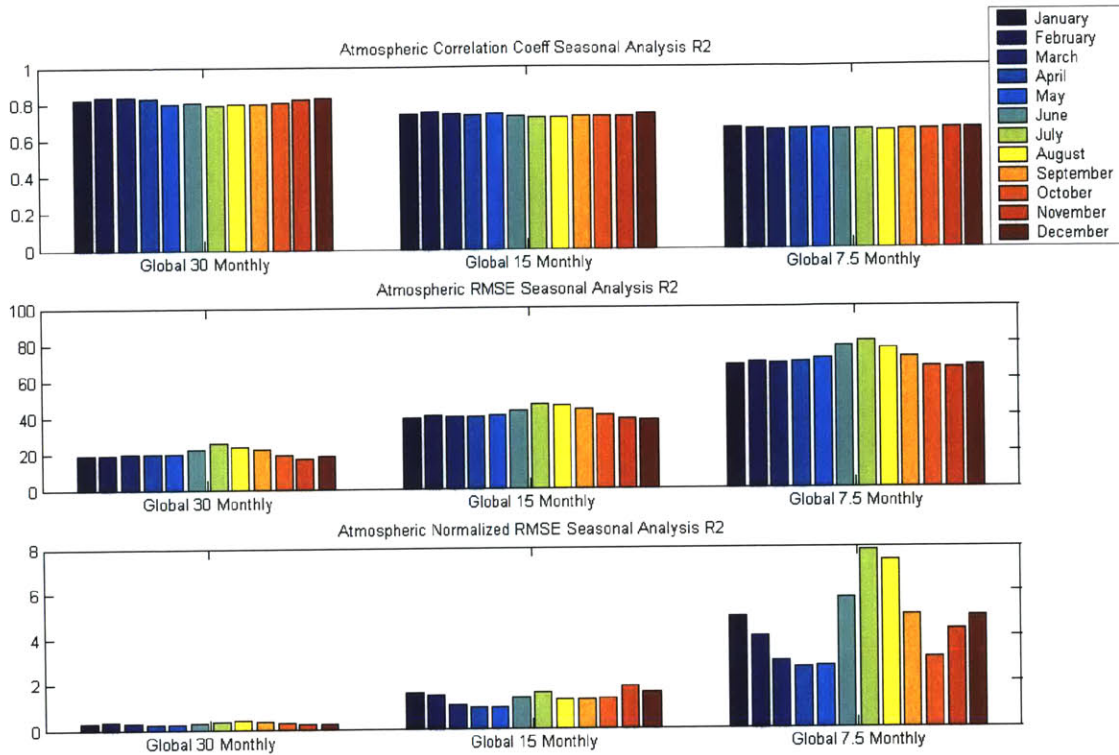


Figure 6-55: Global Analysis of Different Months – Correlation Statistics

In Figures 6-56 through 6-59 the average water balance properties of all control volumes in the tropics are presented applying the TRMM 3B43 precipitation dataset to the last five years of the Reanalysis-1 atmospheric water balance. In terms of absolute error and RMSE, which throughout the global analysis yield very similar results, errors are again very even throughout the months, increasing uniformly with resolution. Such consistency is surprising considering only five points (years) of data are available for each month analyzed. Problems again arise with normalization, but on a very random basis. Satellite precipitation can be particularly vulnerable as a normalization tool especially when only five points are being considered.

Unlike absolute error there is a clear bias trend with the application of TRMM 3B43 precipitation when points are sorted by month. Ideal conditions exist in the summer, and in general there is a surplus of precipitation being applied to each control volume. This surplus is lowered in the summertime with the TRMM 3B43 precipitation. These trends are destroyed by normalization. In terms of the regression analysis the regression slope, intercept, and  $R^2$  values are highly variable throughout the 12 months. This is caused by the low number of points being used in each regression. The clearest two sided comparison analysis is provided by the correlation coefficient which is consistently lower over all resolutions during the summer months. These results are opposite of the bias results the only other property where there were strong monthly trends are seen. It is therefore hard to make any conclusion over what months the TRMM 3B43 dataset can be applied best to atmospheric water balances. It is unlikely the total tropical atmosphere control volume is a net source or sink of water for the rest of the globe, therefore  $P=E$  is best satisfied in the summer when bias is low.

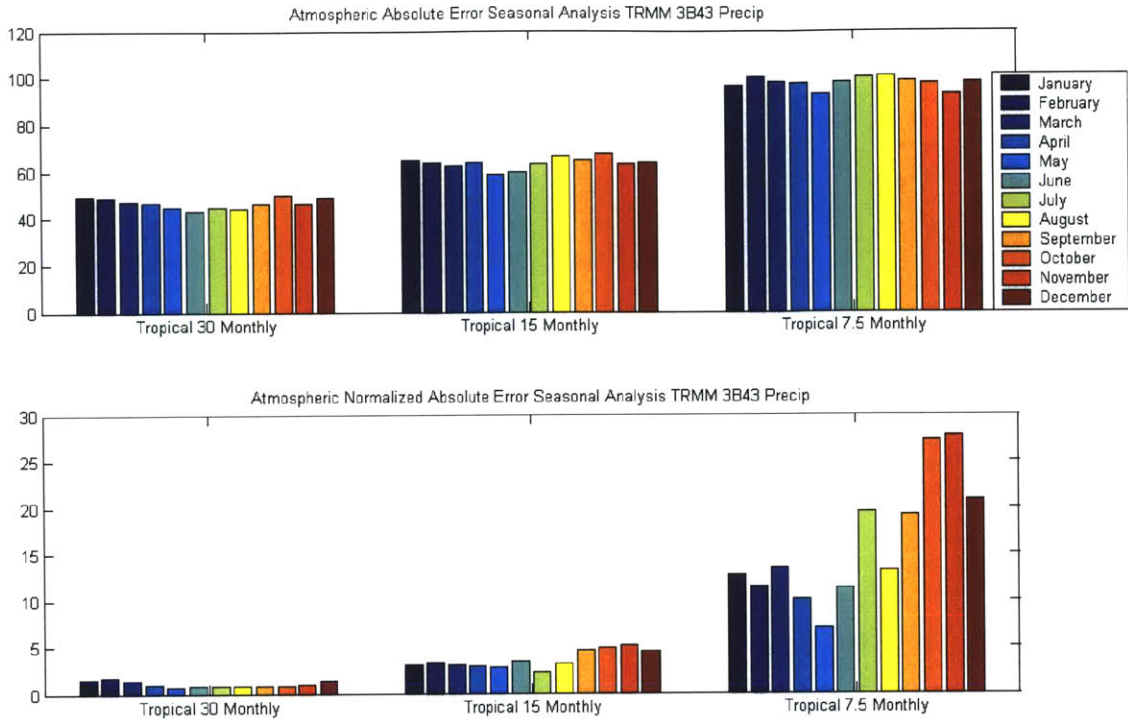


Figure 6-56: Tropical Analysis of Different Months w/ TRMM 3B43 – Absolute Error

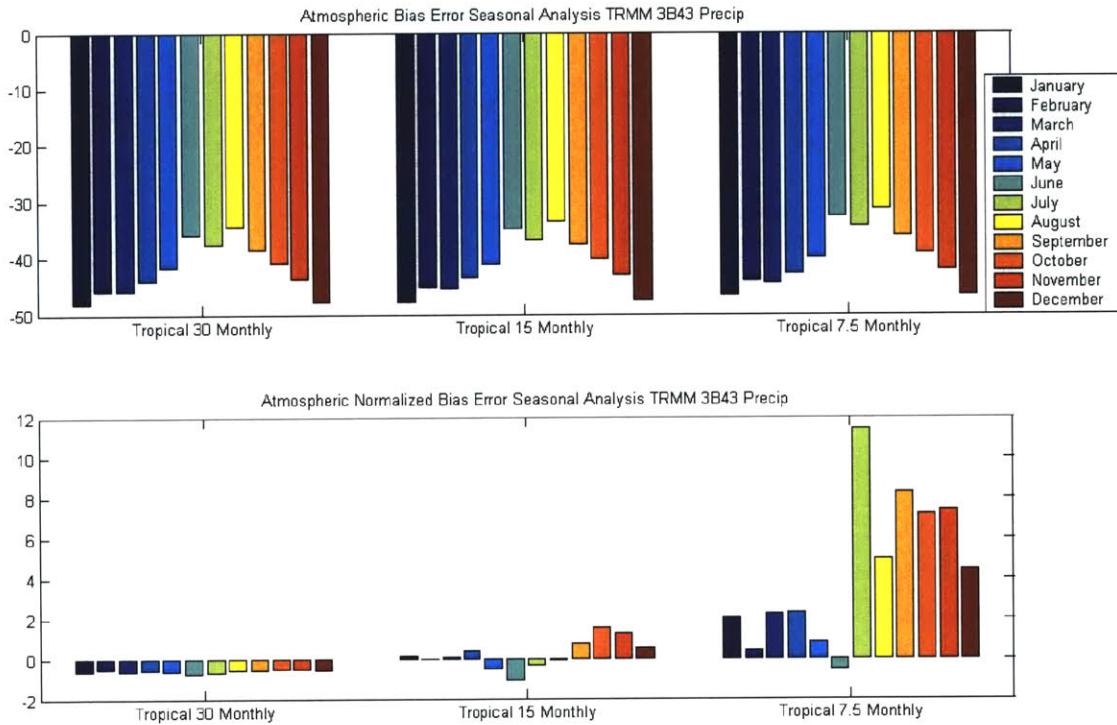


Figure 6-57: Tropical Analysis of Different Months w/ TRMM 3B43 – Bias Error

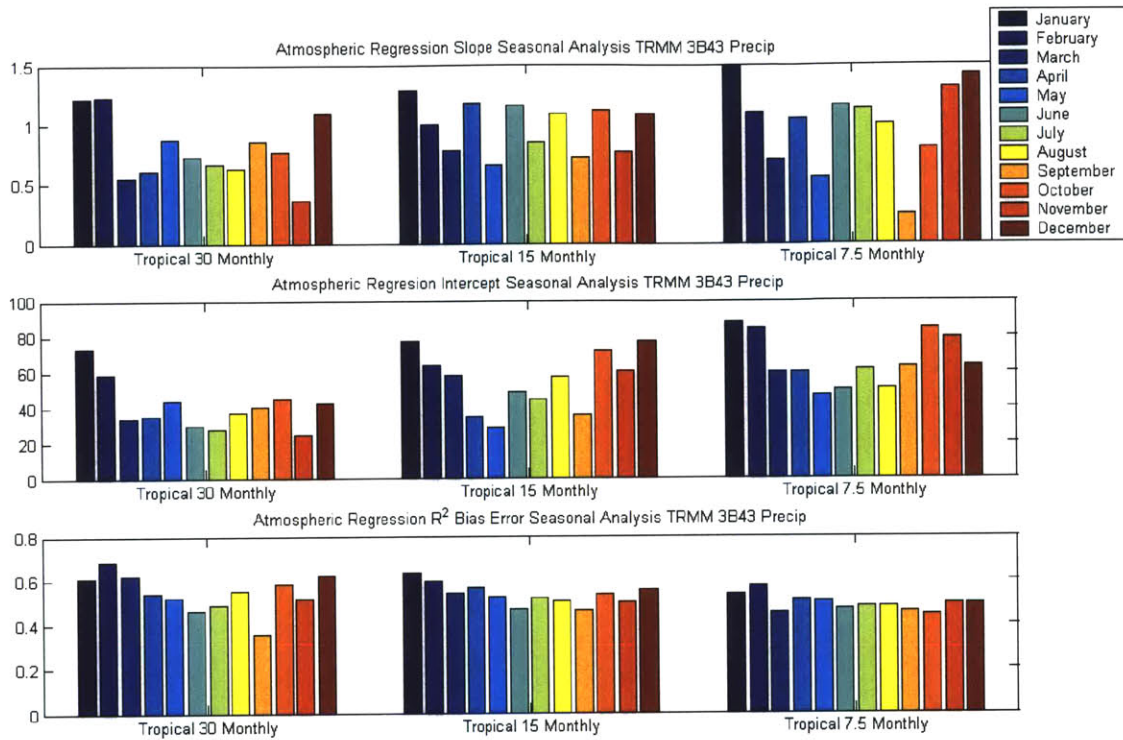


Figure 6-58: Tropical Analysis of Different Months w/ TRMM 3B43 – Regression Statistics

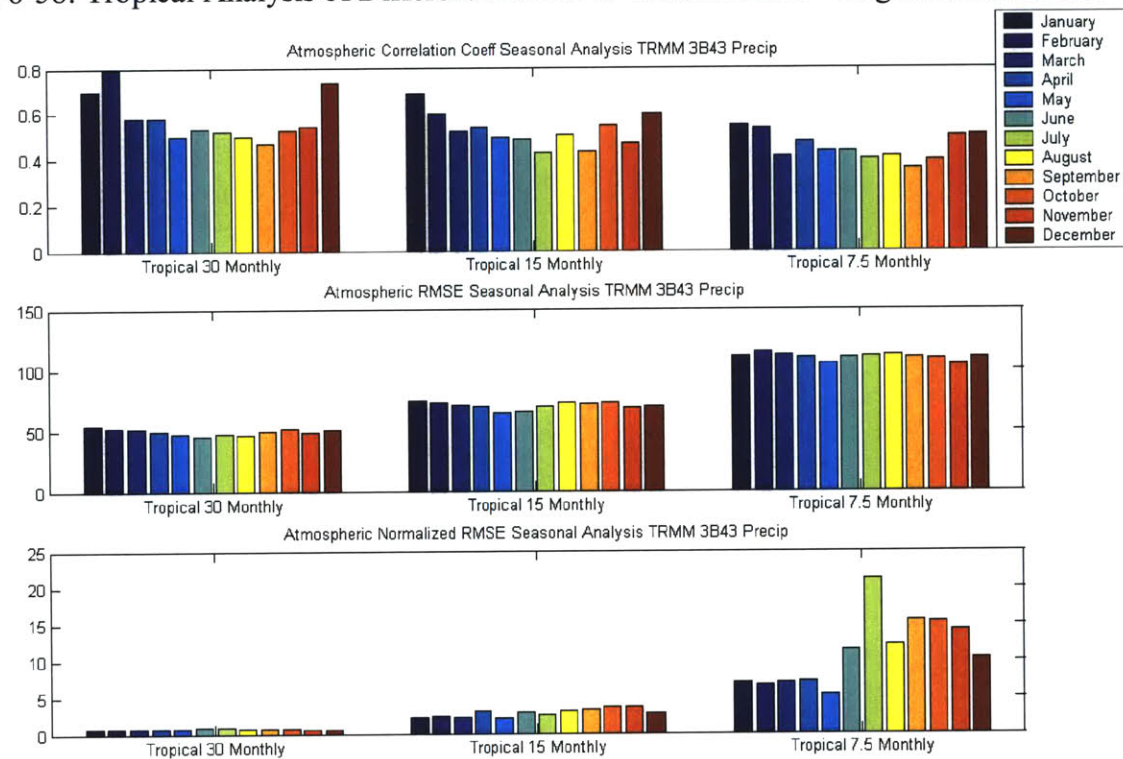


Figure 6-59: Tropical Analysis of Different Months w/ TRMM 3B43 – Regression Statistics

For the monthly analysis, getting back to the original Reanalysis-2 global conditions, spatial maps can be made of water balance properties for specific months. Differences in these maps on a monthly basis are minimal for most of the water balance properties studied. Therefore only the two non-normalized residual statistics are shown for the different resolutions. The absolute error is shown in Figures 6-60 through 6-62 and the bias is shown in Figure 6-63 through 6-65. Four months are selected to represent each season: January, April, July, and October. As displayed by these figures at each resolution there is little distinction between these properties for single months. Errors circulate throughout Asia on the 30 and 15 degree basis, and the balances are particularly bad over the southwest United States during the summertime for absolute error. Similar problems are also over Asia during the summertime on the 7.5 degree resolution. In terms of bias the same general observations apply as there is just little spatial change in the residuals generated by the water balances considering different months.

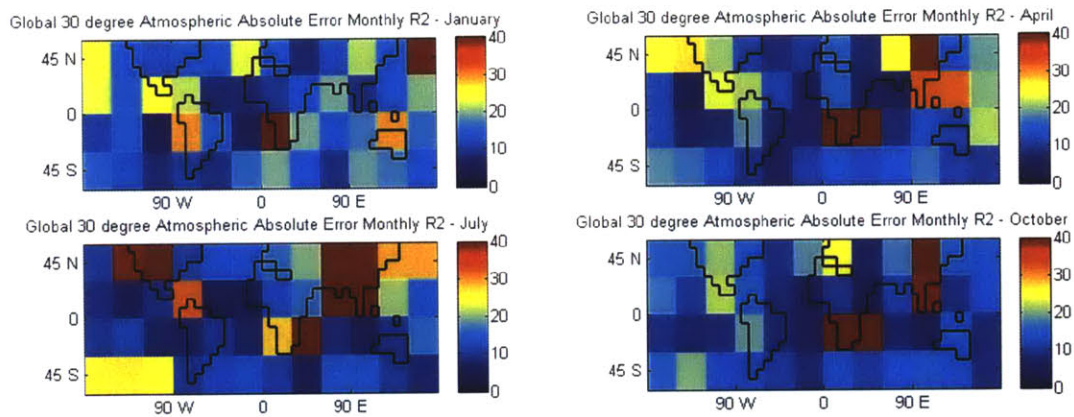


Figure 6-60: 30 Degree Absolute Errors Different Seasons Reanalysis-2, 4 Different Months

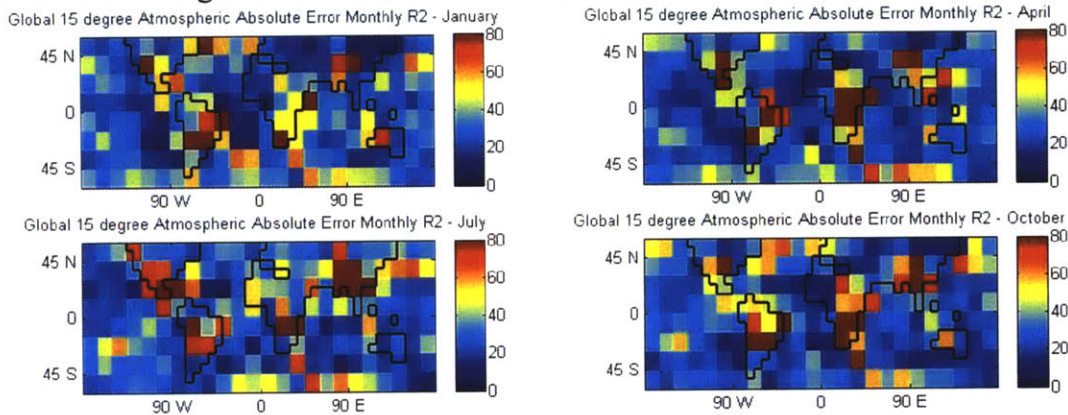


Figure 6-61: 15 Degree Absolute Errors Different Seasons Reanalysis-2, 4 Different Months

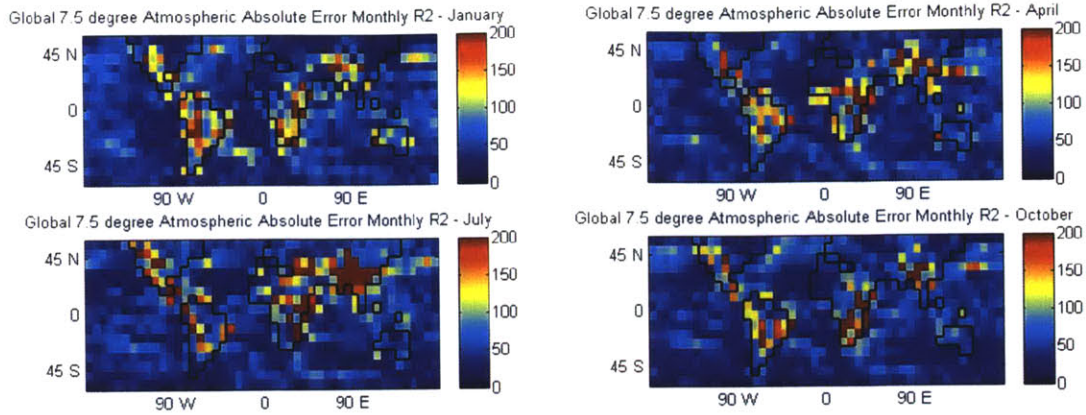


Figure 6-62: 7.5 Degree Absolute Errors Different Seasons Reanalysis-2 Monthly, 4 Different Months

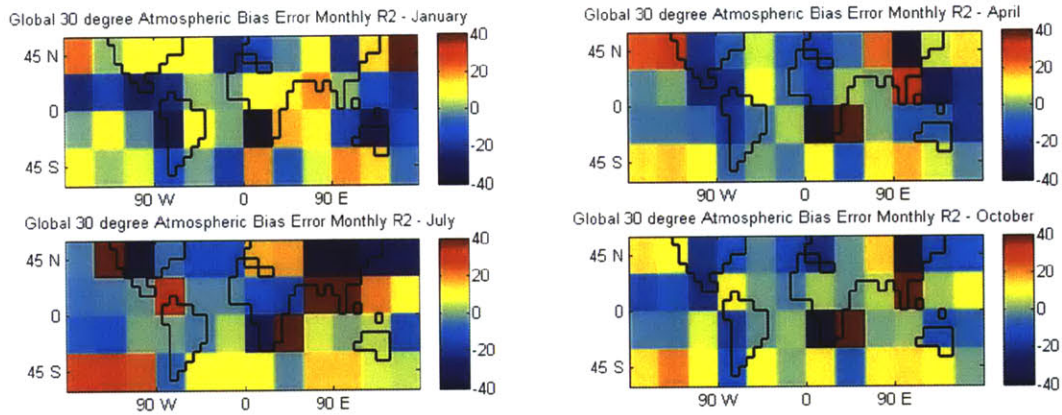


Figure 6-63: 30 Degree Bias Errors Different Seasons Reanalysis-2 Monthly, 4 Different Months

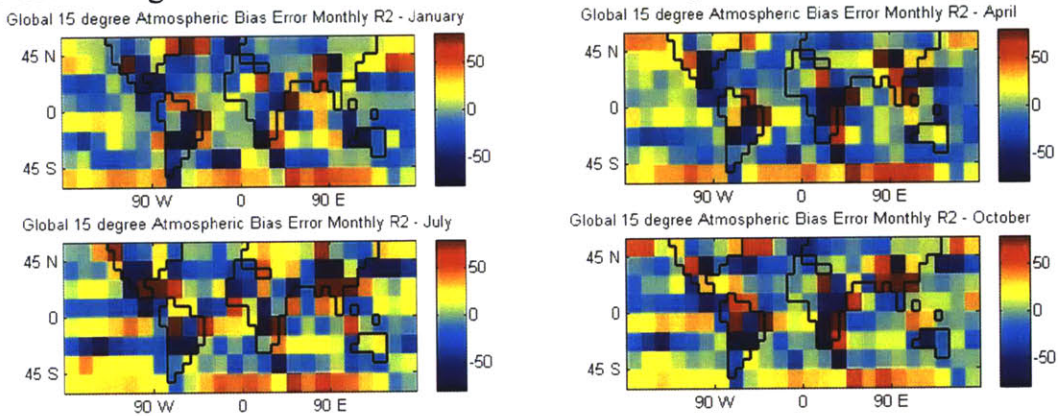


Figure 6-64: 15 Degree Bias Errors Different Seasons Reanalysis-2 Monthly, 4 Different Months

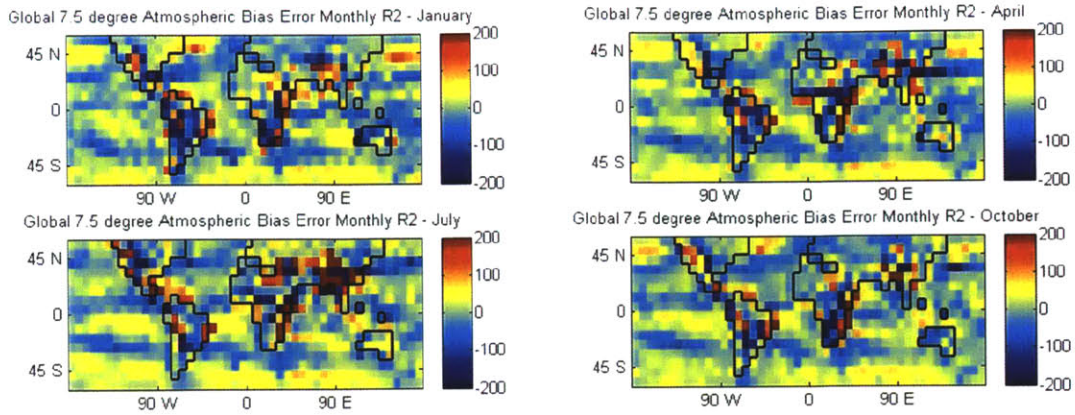


Figure 6-65: 7.5 Degree Bias Errors Different Seasons Reanalysis-2 Monthly, 4 Different Months

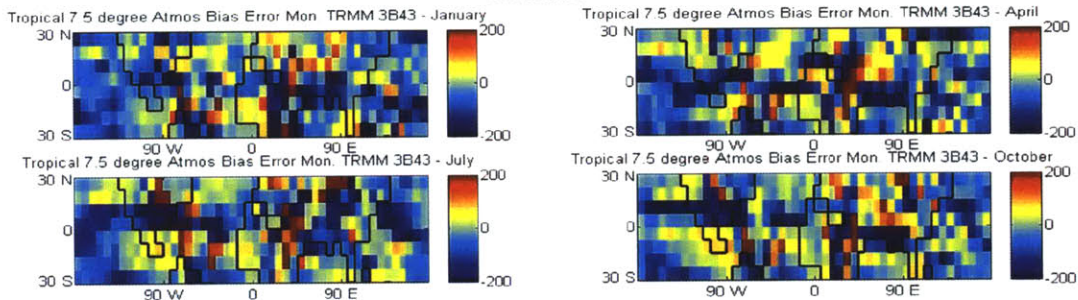


Figure 6-66: 7.5 Degree Bias Errors Different Seasons Reanalysis-2 w/ TRMM 3B43 Precipitation, Tropical

Changes in spatial properties are also not observed with the application of the TRMM 3B43 product to the tropical portions of this balance. An example spatial plot contrasting bias on the 7.5 degree resolution is presented in Figure 6-66. Bias was the property which showed the most spatial differentiation with change in month for these conditions. Subtle differences are observed between figures, but in general balances are very similar from month to month and season to season.

When the monthly Reanalysis-2 water balance is sorted by month, nearly identical water balance properties are found over each month, and spatially these properties do not change much either. Problems develop in the normalization of properties for certain months, and small conclusion can be drawn, as the balances are slightly worse on average during the summertime. Conducting the same study except considering only tropical points and using the TRMM 3B43 precipitation, bias is observed to be lowest in the summer months overall.

### 6.3.2 Decadal Variations in the Reanalysis-1

The length of the Reanalysis-1 allows for the dataset to be broken down by decade, and the water balance for each decade analyzed globally for the atmospheric case. The most useful way to view the results of this analysis is by the averaging of water balance properties over all control volumes considered. Both the yearly and monthly time scales can be analyzed in this manner at each resolution.

Four plots are made of the global spatial averages of water balances properties considering different resolutions and time scales averaging over all control volumes in Figures 6-67 through 6-70. Looking at the absolute error results in Figure 6-67, there is hardly any difference between the errors observed over different decades, while previously observed trends for time scale and resolution are preserved. Considering the normalized absolute error, on the finer resolutions the 1980's decade performs poorly. This is counterintuitive to the theory that the water balance gets better as time nears the present. In terms of bias, differences are apparent in the top of Figure 6-68. These biases are very small, as have been consistently found when averaging the global atmospheric water balance. All biases are below 5 kg/m<sup>2</sup>-month, small in comparison to typical values of water balance variables. The 1960's is consistently the most biased decade, though by a very small amount. The normalized biases behave like the normalized absolute errors.

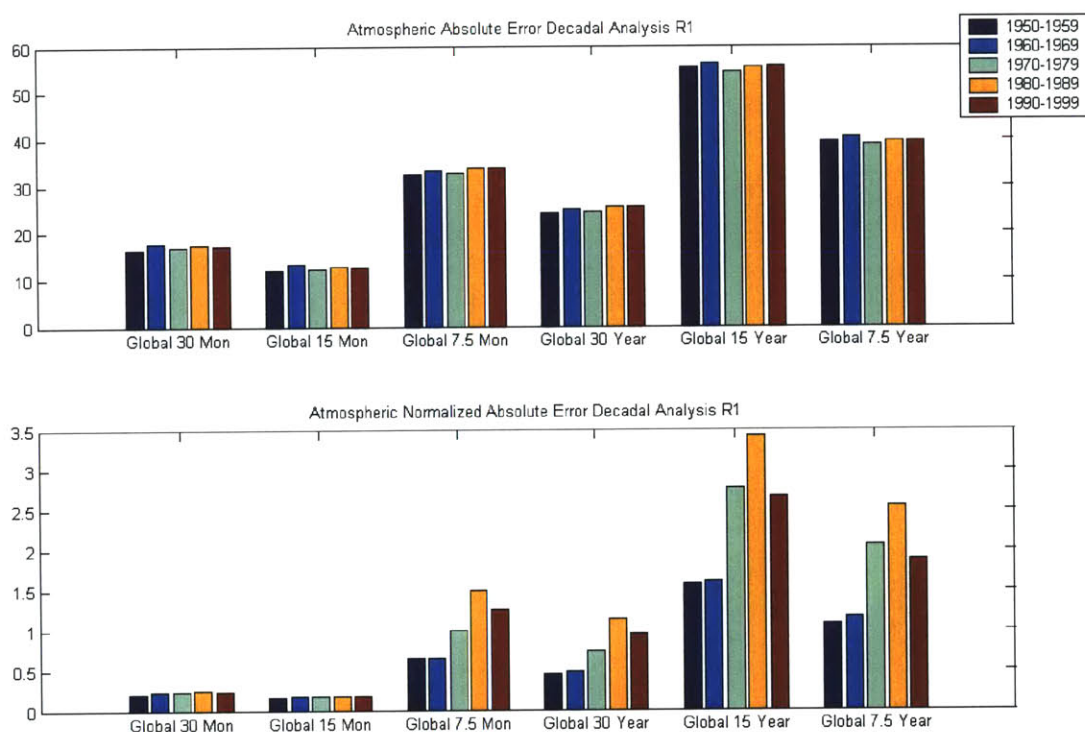


Figure 6-67: Global Analysis of Different Decades – Absolute Error



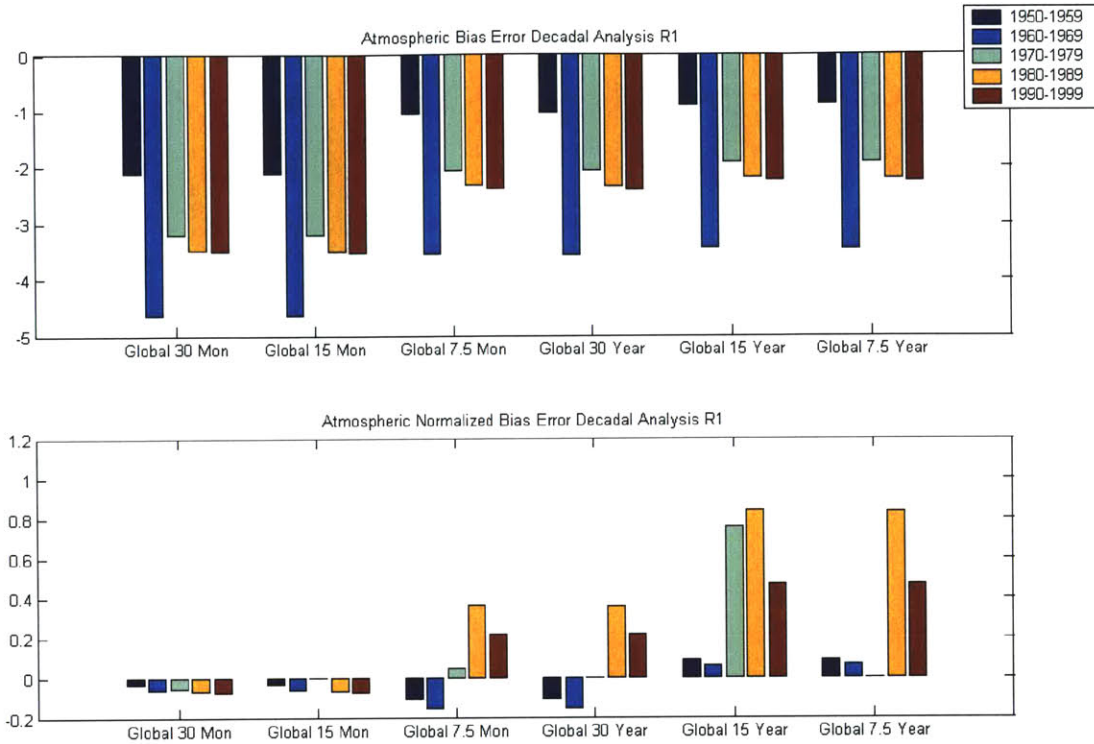


Figure 6-68: Global Analysis of Different Decades – Bias Error

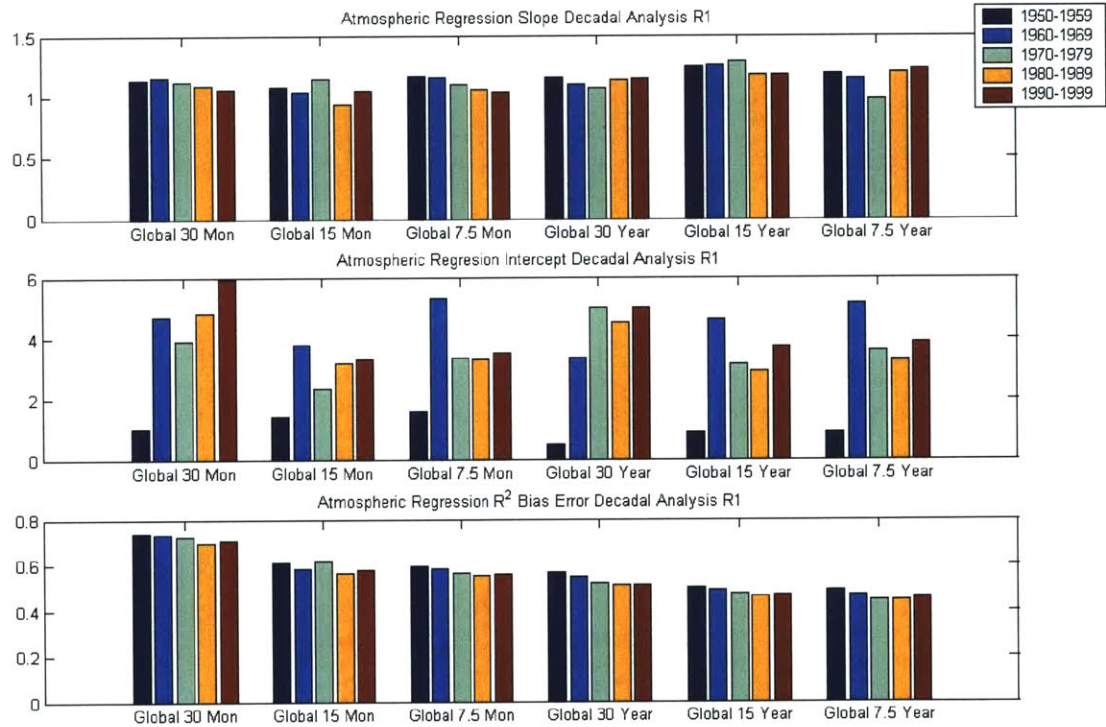


Figure 6-69: Global Analysis of Different Decades – Regression Statistics

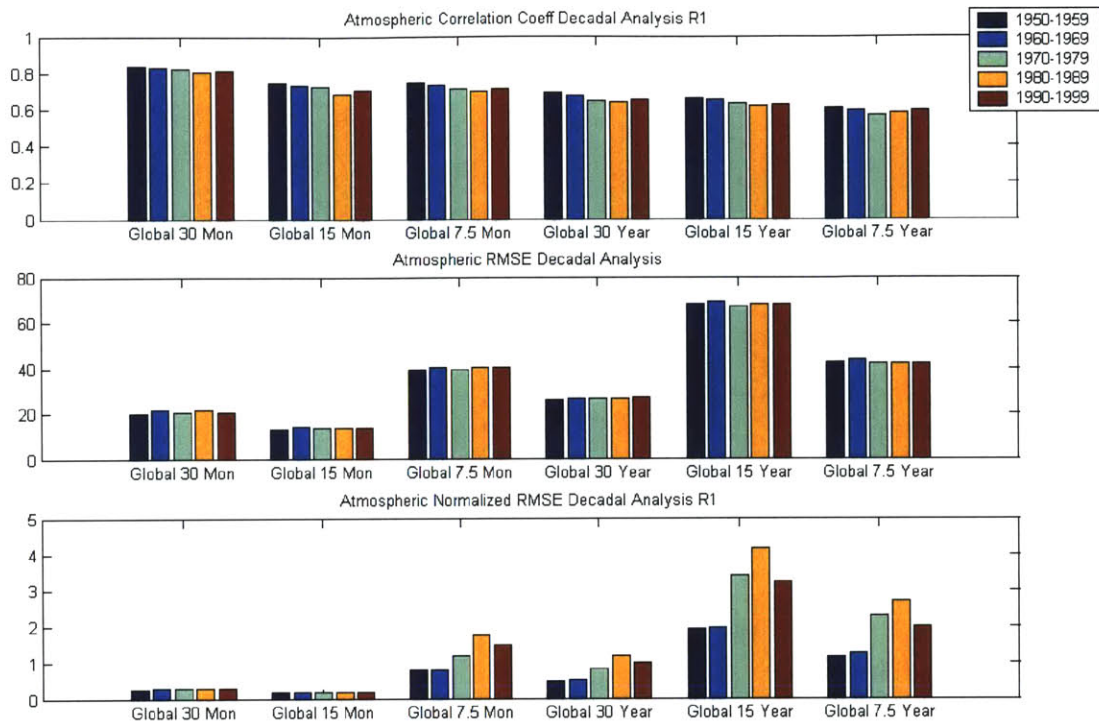


Figure 6-70: Global Analysis of Different Decades – Correlation Statistics

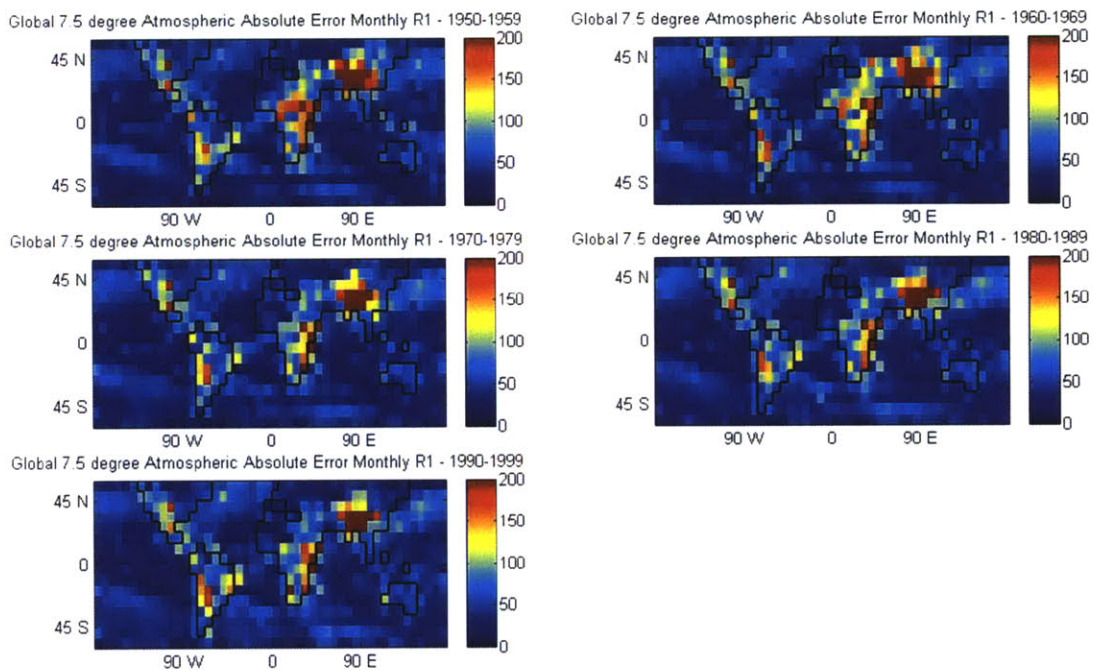


Figure 6-71: 7.5 Degree Absolute Errors Different Decades Reanalysis-1 Monthly

Similar observations apply for the regression slope and intercept. The average regression slope for each of the time scales and resolutions resides around one, and the intercept is highly governed by the bias. What is interesting in this analysis are the  $R^2$  and correlation coefficients found. Both of these properties steadily degrade over the decades for each time scale and resolution. One would expect these quantities to increase as better available data helped to complete the water balance, but the opposite is true. It is hard to speculate the reasons for this, but the trend is likely not a coincidence.

Several of these decadal comparisons were mapped spatially for different water balance properties, with no noticeable differences between decades. In Figure 6-71 an example of these plots is presented to illustrate the lack of change by decade in the reanalysis water balance. Only slight changes in absolute error over land are detected from decade to decade. It is an important observation that the reanalysis water balance is consistent across decades.

#### **6.4. Land Points vs. Sea Points**

A common theme of the global water balance data shown so far is a contrast between the errors found over land and the errors found over water. Because of the large differences between the atmospheric water balance systems over land and ocean, it stands to reason that water balance properties may be different. For instance some precipitation datasets may be better over water than over land or vice versa.

The smallest available grid over which the atmospheric water balance can be reasonably calculated is 7.5 degrees. This is a fairly large area, and it is difficult to distinguish land points from sea points on a grid of 7.5 degrees. For the pressure level fields of the reanalysis data, a 2.5 degree land grid is provided, and this grid is used to generate Figure 6-72. The percentage of land for each point on a 7.5 degree grid of the globe is found from the finer resolution 2.5 degree dataset. Many points of the 7.5 degree resolution contain both land and sea, because of the coarseness of the resolution. Expanding such an analysis to a 15 or 30 degree resolution would be useless because only a very poor and probably inaccurate land grid could be generated. Grids are classified as land if they have more than 50% of their area covered land. Figure 6-73 is created this way, for 60 N to 60 S, and Figure 6-74 is created for the tropical precipitation cases.

In this analysis, global maps are not necessary because all land and sea points are being combined and we are looking for distinction simply between these two types of points. Variables that can be considered while looking at this distinction are reanalysis, time scale, and precipitation dataset. A separate analysis of only tropical points is also conducted.

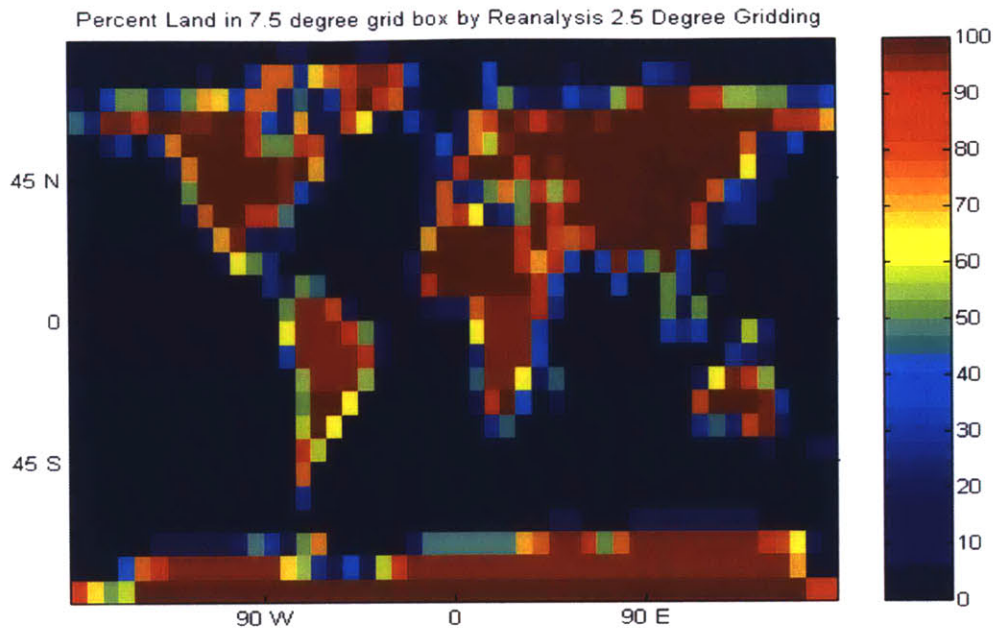


Figure 6-72 – Percentage of Land for a 7.5 Degree Resolution

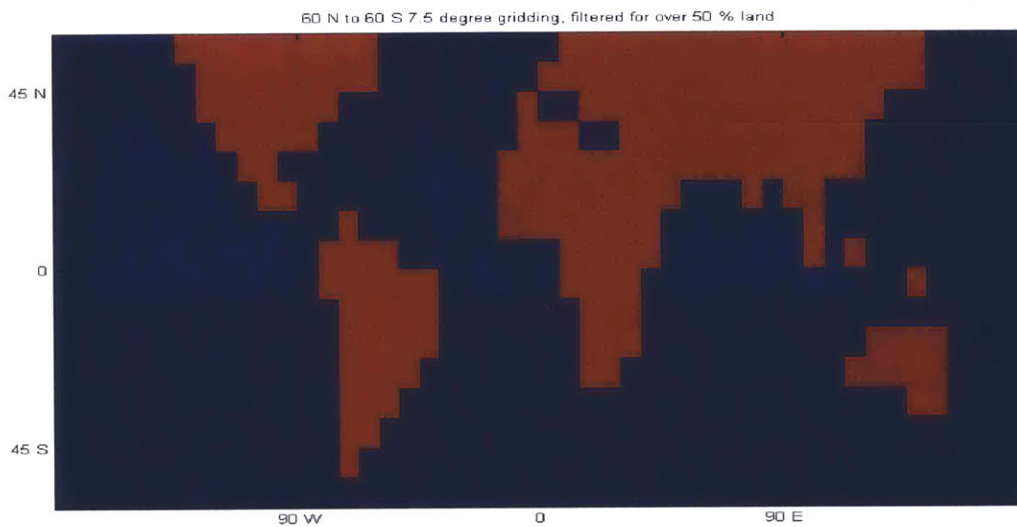


Figure 6-73: 7.5 Degree Land Grids for 60 N to 60 S

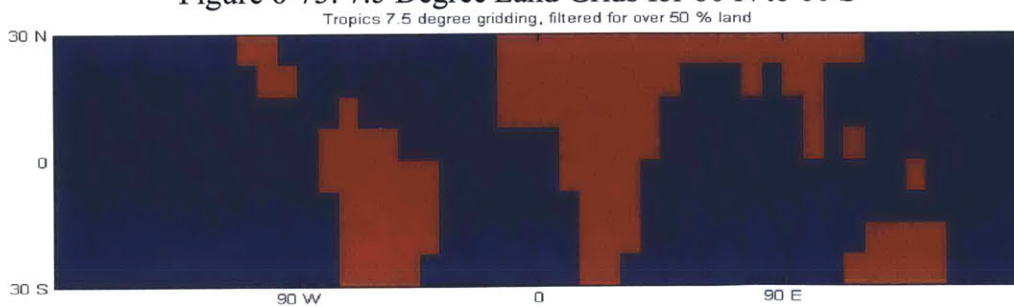


Figure 6-74: 7.5 Degree Land Grids for Tropical Datasets

In Figures 6-75 through 6-78 the water balance properties of land and sea control volumes are averaged and contrasted against each other for a variety of conditions. As expected in Figure 6-75 the absolute error for all land conditions is much higher than that for sea conditions. The difference is approximately a factor of two. At this resolution large atmospheric errors exist for both land and sea points. Based on the bias results the weakness of the SSM/I precipitation dataset is clearly an overestimation of precipitation over the ocean. A strong difference between this dataset's applicability to the land and sea balances is observed in the bias specifically. When the residual statistics are normalized in Figure 6-76, results are consistent for absolute error and most of the bias results are as well. The normalized bias for land is much higher and positive. This is probably caused by poor reanalysis normalization over sparsely precipitated land cells, likely in the desert. Such dry points are less likely over the ocean. More meaningful results when averaging over all global points are picked up by the actual residual statistics without normalization.

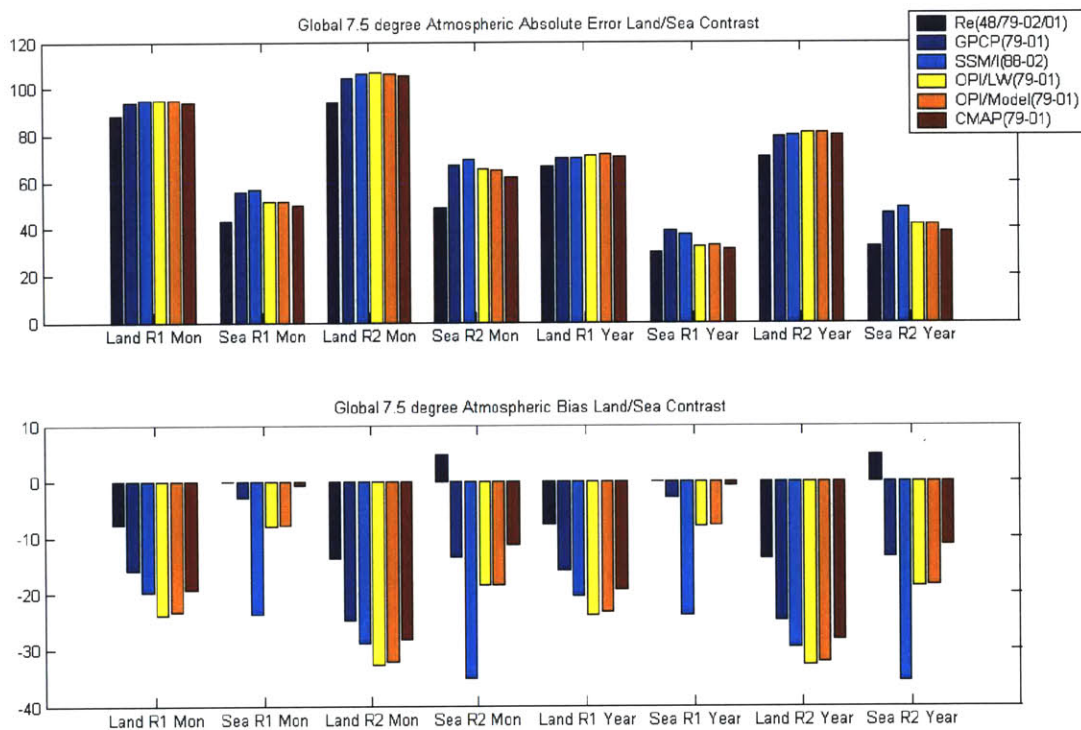


Figure 6-75: Global Analysis of the Land/Sea Contrast – Residual Statistics

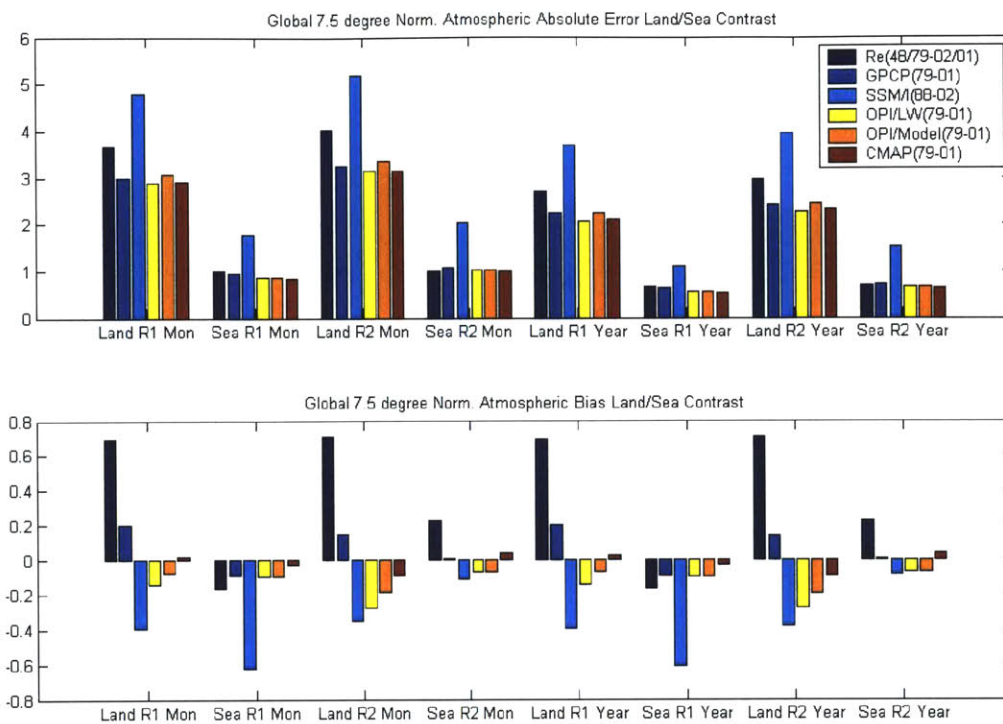


Figure 6-76: Global Analysis of the Land/Sea Contrast – Normalized Residual Statistics

In Figure 6-77 and 6-78 regression and correlation statistics are presented with the land and sea points being separated. Much better and consistent regression slopes are found for all balances with sea points, as regression slopes are very close to one on average for the ocean in all situations. Over land the reanalysis precipitation provides a good average slope, as the reanalysis precipitation matches its own water balances on average. However, there is complete disagreement in the slope property when alternative precipitation datasets are applied over land. As usual intercept results imitate balance results and RMSE results imitate absolute error results. In terms of  $R^2$  and correlation coefficient values are consistently and significantly lower over land than over sea. Particularly poor correlation is found over land points for the yearly time scale and alternative precipitation datasets.

Separating the land and sea points, differences are seen in the yearly slope and correlation using the reanalysis precipitation, however using other precipitation datasets over land with the yearly time scale provides very low correlations, and deems these precipitation datasets nearly useless to the water balance over land.

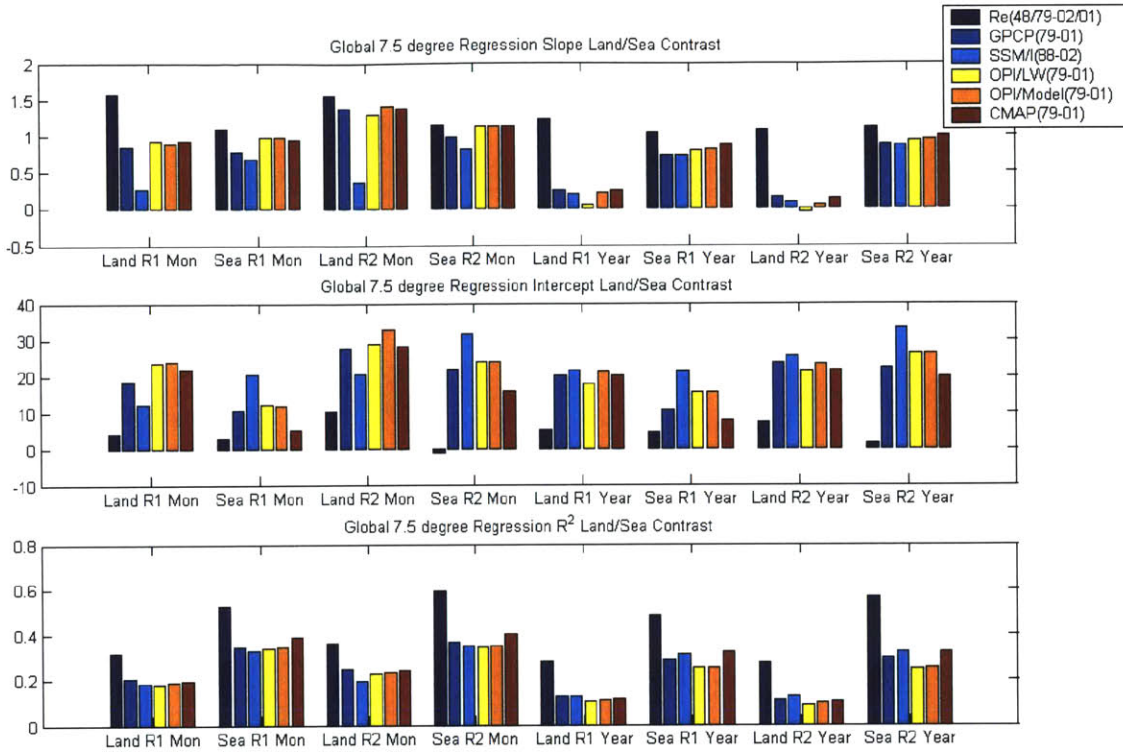


Figure 6-77: Global Analysis of the Land/Sea Contrast – Regression Statistics

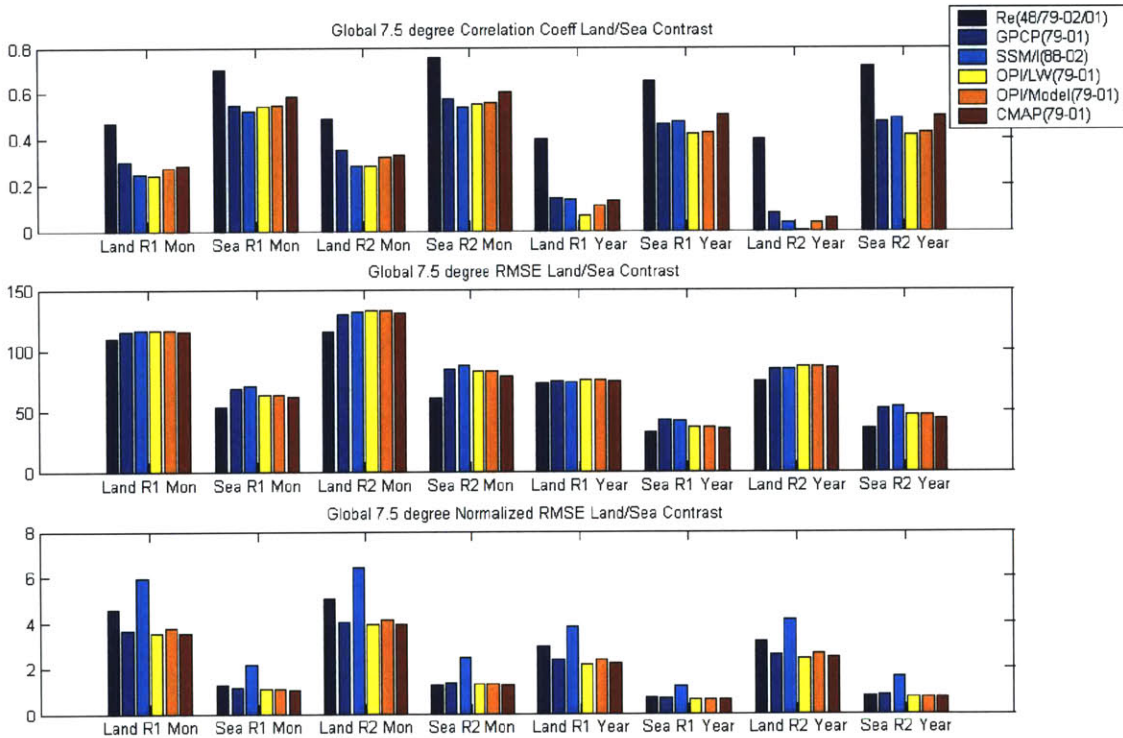


Figure 6-78: Global Analysis of the Land/Sea Contrast – Correlation Statistics

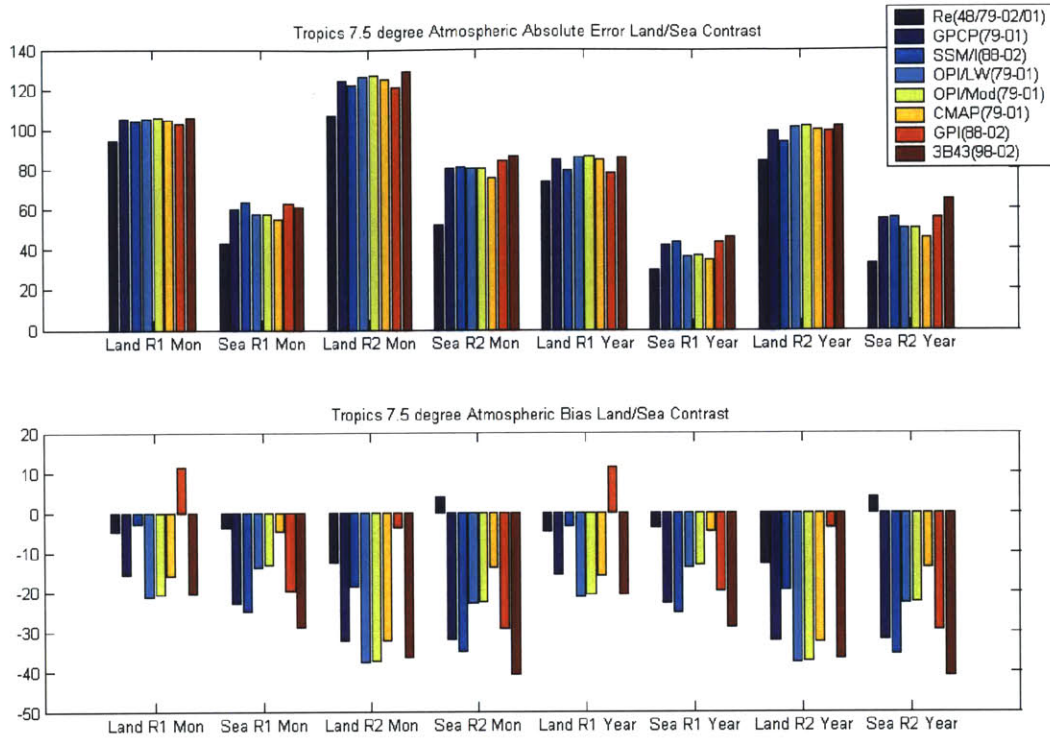


Figure 6-79: Tropical Analysis of the Land/Sea Contrast – Residual Statistics

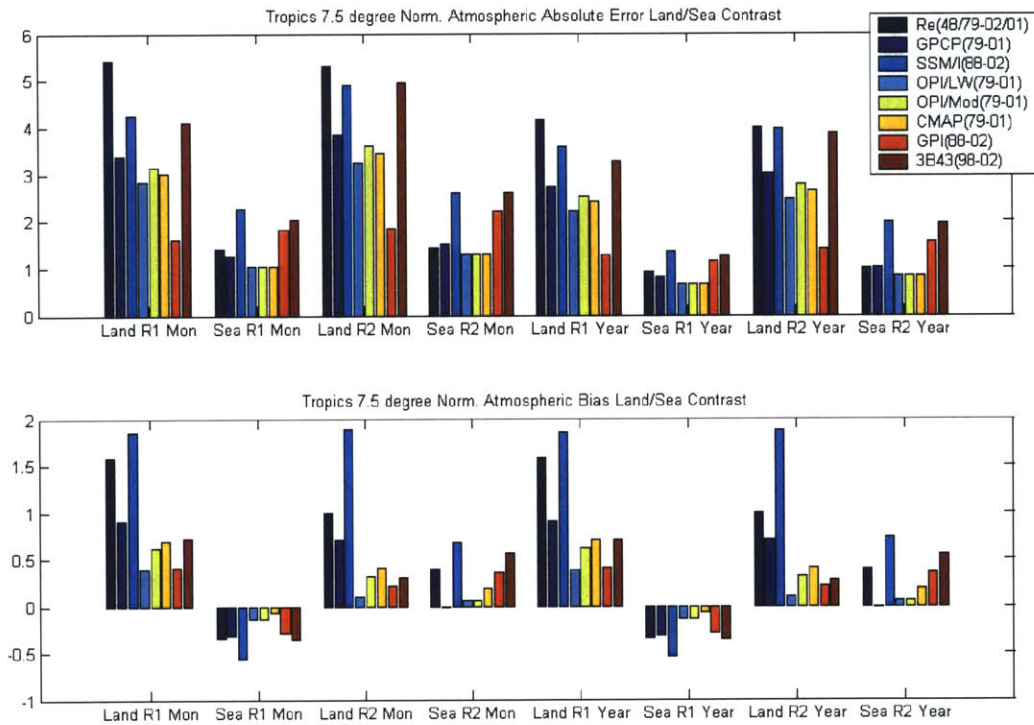


Figure 6-80: Tropical Analysis of the Land/Sea Contrast – Normalized Residual Statistics



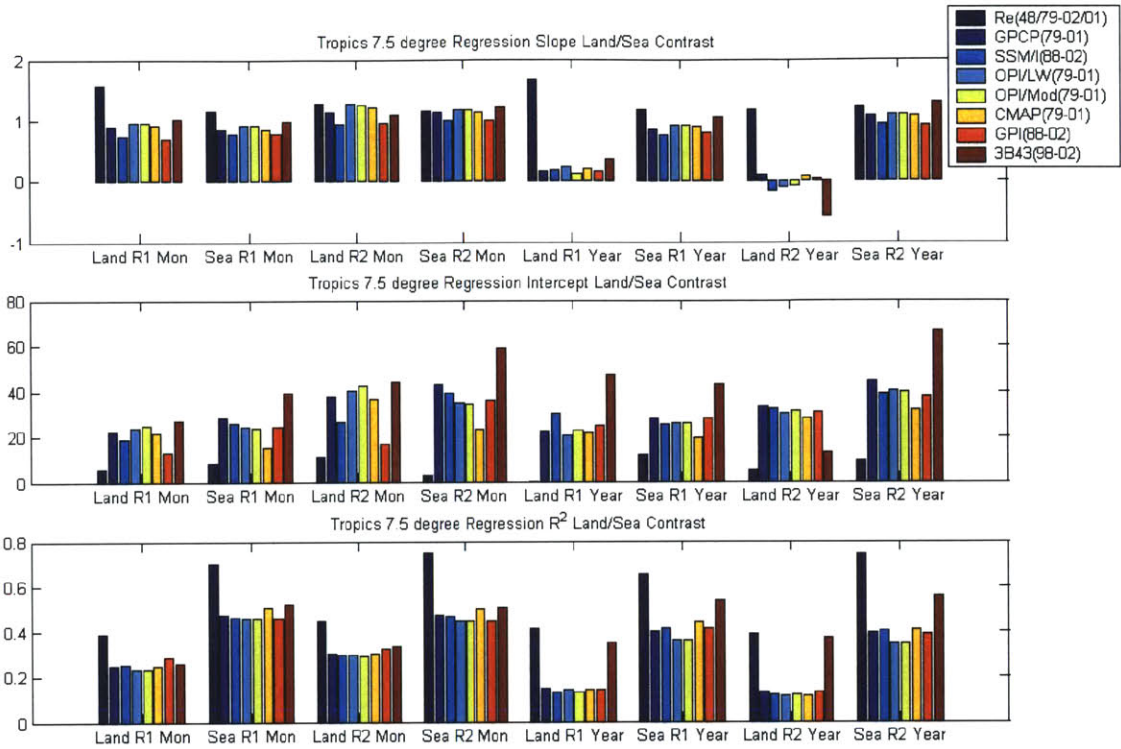


Figure 6-81: Tropical Analysis of the Land/Sea Contrast – Regression Statistics

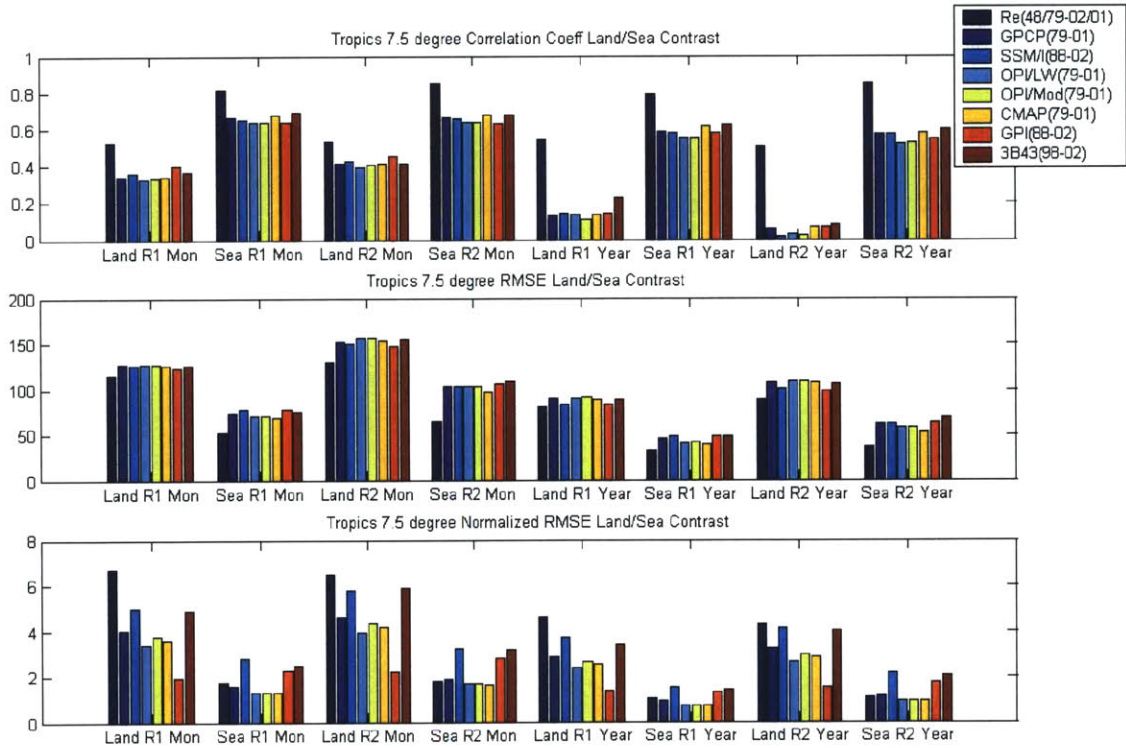


Figure 6-82: Tropical Analysis of the Land/Sea Contrast – Correlation Statistics

The same analysis applied to only the tropical points between 30 degrees north and 30 degrees south is displayed in Figure 6-79 through 6-82. The ratio of land to sea errors is about the same in terms of absolute error and RMSE, however the magnitude of errors are slightly higher in the tropical cases for all conditions. In terms of bias there is a large contrast between the GPI precipitation over land and sea, as over land biases are near zero and even positive. The normalization process again interferes with the normalized residual statistics and RMSE. Average regression slopes are again consistent with one for all conditions except for the use of alternate precipitation over land on the yearly basis. This property is not corrected in the tropical case.

The comparison of land and sea balances under different conditions in general validates what was seen in spatial plots earlier. The ocean balances are balanced much better than the land balances. About two times as well on average in terms of absolute error and regression correlation. Useful observations can be made pertaining to precipitation datasets over land and sea points especially in terms of overall bias, as certain precipitation datasets cater better to land or sea points or, in the case of the reanalysis, both.

## 7. Ground Heat Flux Research

Ground heat flux (GHF) over the land surface can be calculated using the half-order derivative method proposed by Wang and Bras (1999):

$$G(t) = \frac{I}{\sqrt{\pi}} \int_0^t \frac{dT_g(s)}{\sqrt{t-s}} \quad (\text{Eq. 7-1})$$

where  $G(t)$  is the ground heat flux at time  $t$ ,  $T_g$  is the ground temperature, and  $s$  is the integration time of the temperature time series used to find the ground heat flux.  $s$  varies between 0 and  $t$ .  $I$  is the thermal inertia of the soil material:

$$I = \sqrt{K\rho C} \quad (\text{Eq. 7-2})$$

where  $K$  is the bulk thermal conductivity ( $\text{Wm}^{-1}\text{K}^{-1}$ ),  $\rho$  is the bulk density ( $\text{kg m}^{-3}$ ), and  $C$  is the specific heat capacity ( $\text{J kg}^{-1}\text{K}^{-1}$ ).  $t=0$  represents the time when  $G=0$ , which is typically at 6 am local time. The interested reader is referred to Wang and Bras (1999) for a detailed description of the half order derivative method.

In this study the Reanalysis-1 and Ranalysis-2 skin temperature (ground temperature) data at 6 hourly, daily, and monthly time resolutions will be used to compute the global 6-hourly, daily, and monthly ground heat flux for the year 1988. A certain amount of data from the previous year 1987 is needed to calculate the ground heat flux at the beginning of 1988.

In Section 7.1, a first estimate of  $G$  will be derived with a constant  $I = 1 \times 10^3 \text{ J m}^{-2} \text{ s}^{-1/2} \text{ K}^{-1}$  using the Reanalysis-1 dataset. In Section 7.2 the spatially varying  $I$  parameter will be estimated from field observations. Section 7.3 contains an original methodology for determining the global map of  $I$ , based on the findings of Section 7.2. In Section 7.4, a global ground heat flux product will be presented.

### 7.1. Initial Findings

The only global data set on ground heat flux comes from the NCEP Reanalysis product. We'll use the skin temperature to recalculate ground heat flux using Eq. (7-1). Ground heat flux in the NCEP Reanalysis dataset, like many other fields, is derived by a global model on a 6-hourly basis, and then aggregated into daily and monthly resolutions. In this section a universal thermal inertia of  $1 \times 10^3 \text{ J m}^{-2} \text{ s}^{-1/2} \text{ K}^{-1}$  is applied to all land points to derive a ground heat flux time-series at three time scales. There are 5914 land pixels with a size of about  $1.90^\circ \times 1.88^\circ$  (lat x long).

Four measures are then utilized to compare the "estimated" ground heat flux and the "observed" ground heat flux (from the reanalysis output), namely root mean square error (RMSE), normalized root mean square error (NRMSE), regression slope and regression intercept. The root mean square error is a measure of the differences between two time series in a non biased fashion. The purpose of the initial estimates is to identify the locations where there are

substantial differences between the estimated and observed ground heat flux. Differences may exist between the two reanalysis datasets with certain geographical and temporal features. The comparison will help in producing the improved global maps of ground heat flux

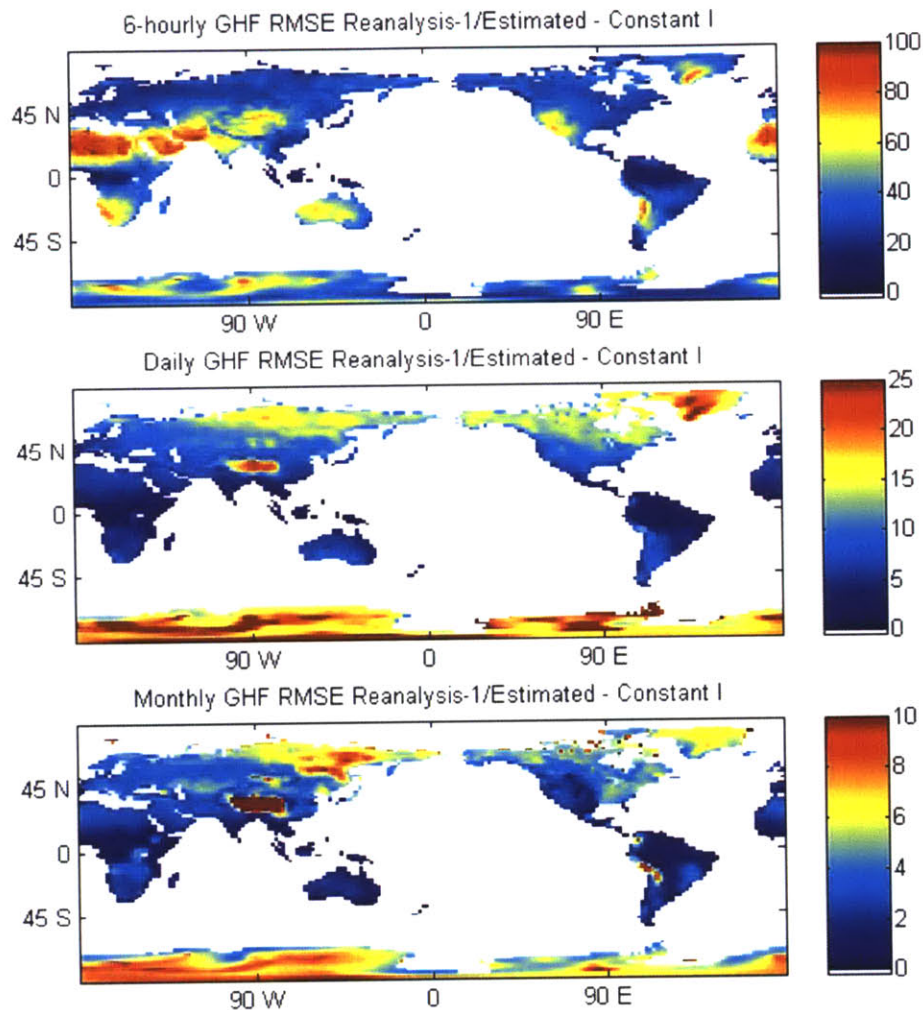


Figure 7-1: RMSE between the Estimated and Reanalysis-1 GHF – Non-Averaged Data

Figure 7-1 illustrates the RMSE between the Reanalysis-1 observed and the estimated ground heat flux at the three time resolutions using the skin temperature input. On the 6-hourly time scale, moderate differences are found throughout the globe, on the order of  $20 \text{ Wm}^{-2}$ . Large differences are found over desert regions around the world including Northern Africa, Central Asia, Australia, the Southwest United States, and the Andes mountains. On the daily and monthly time scales, these differences are reduced with significant errors over Central Asia, but predominantly over high latitude regions.

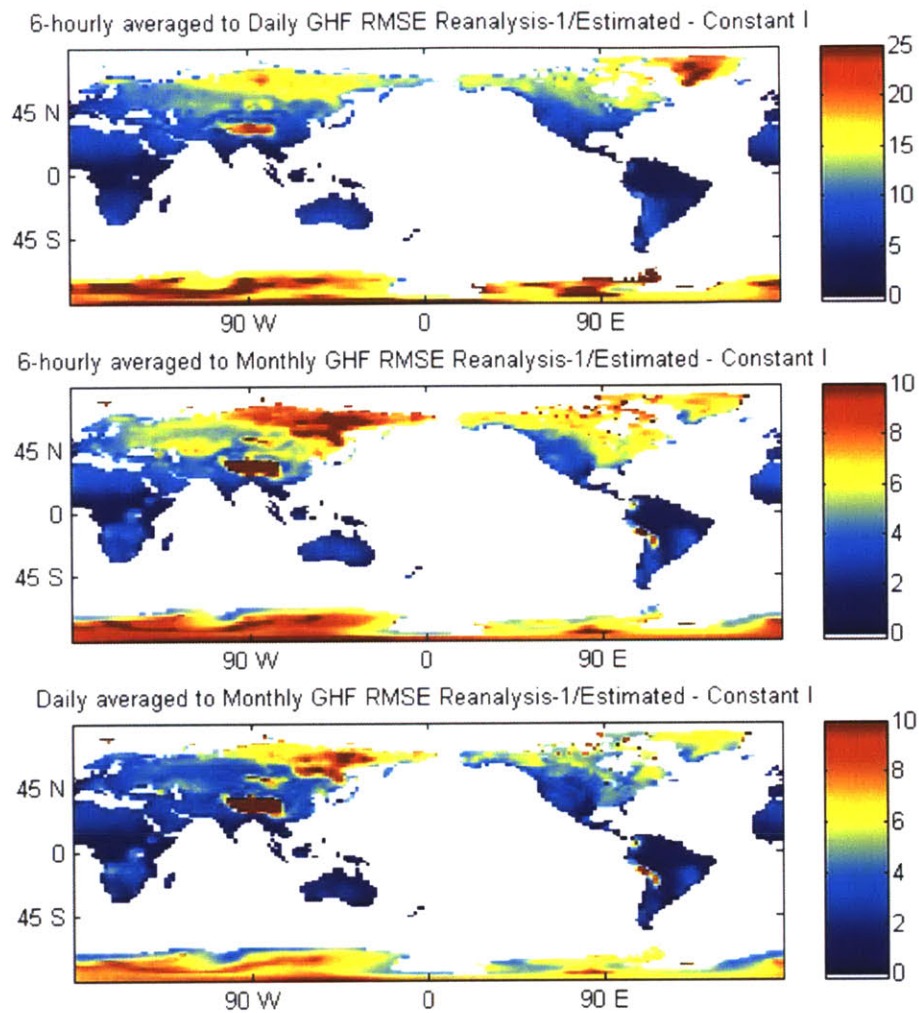


Figure 7-2: RMSE between the Estimated and Reanalysis-1 GHF – Aggregated Data

Like the reanalysis ground heat flux, 6-hourly estimated ground heat flux can be aggregated to daily and monthly time scales, resulting in three additional ground heat flux products which can be compared with the Reanalysis-1 products at the corresponding time scales. These comparisons are shown in Figure 7-2. Large differences persist for the same regions as in Figure 7-1 showing non-aggregated data. Of particular interest are the high latitude regions and Central Asia. The differences are the largest when the 6-hourly estimated ground heat flux is aggregated all the way to monthly data and then compared with the reanalysis monthly ground heat flux. In northern areas the aggregation leads to major difference between the two.

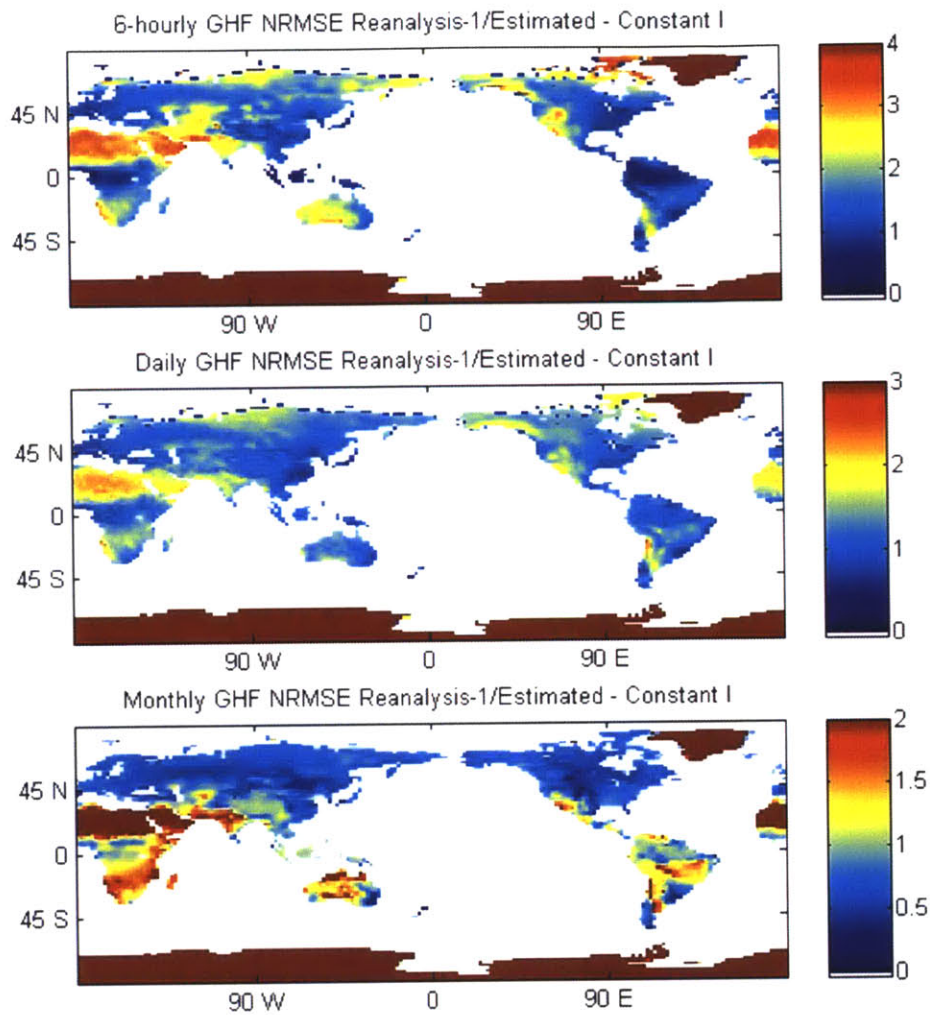


Figure 7-3: Normalized RMSE between the Estimated and Reanalysis-1 GHF – Non-Aggregated Data

Because of the variability in the magnitude of ground heat flux, normalized RMSE is used to quantify the relative error of the estimated ground heat flux to the observed ground heat flux. Normalization is performed for each dataset by dividing the RMSE by the average of the absolute value of Reanalysis-1 ground heat flux. Since the annual mean ground heat flux is close to zero it is necessary to normalize by the average of the absolute value of ground heat flux. Figure 7-3 shows that relatively large normalized differences are observed over Greenland and Antarctica. Significant errors also appear in dry desert regions as well as in the arctic regions to a lesser extent. The arctic errors do not appear to be very large on the monthly time scale, while the differences over the desert are greater. When comparing the aggregated estimated ground heat flux with the reanalysis shown in Figure 7-4, there are increased differences over Northern Africa and Australia when aggregating from the 6-hourly to the monthly time scale.

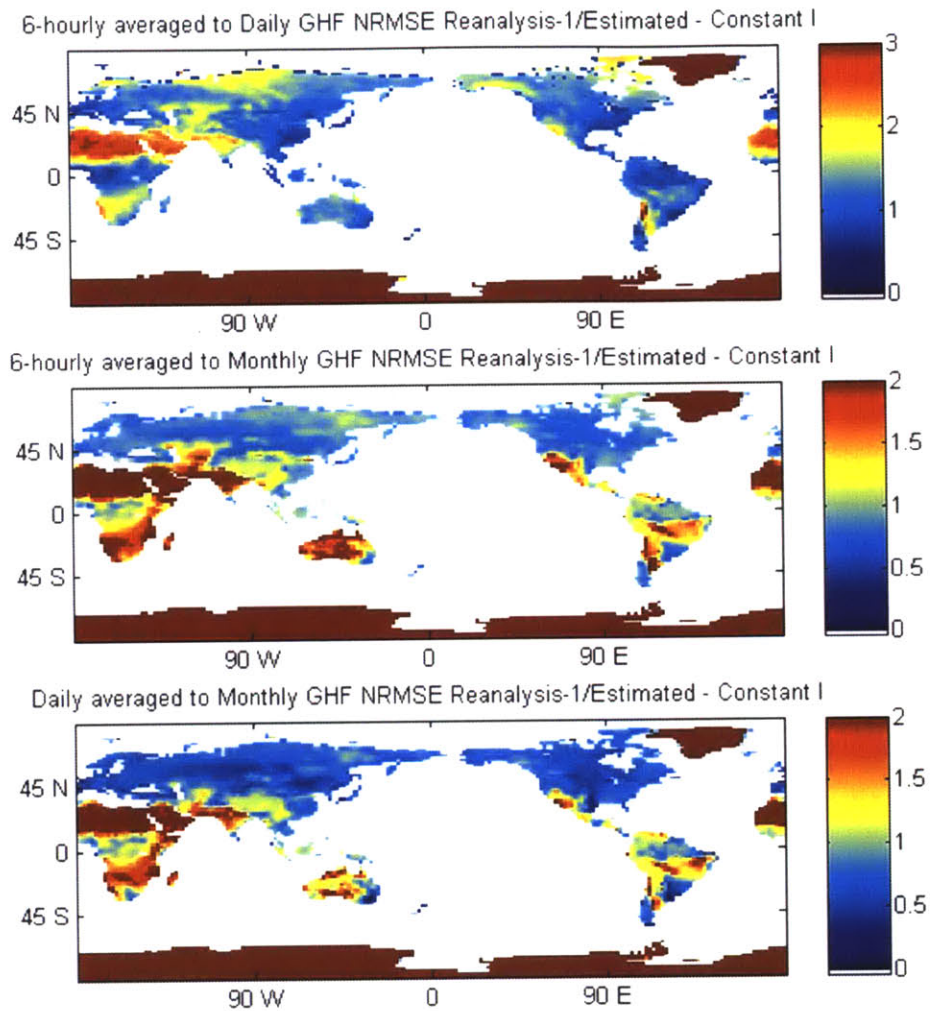


Figure 7-4: Normalized RMSE between the Estimated and Reanalysis-1 GHF – Aggregated Data

Regression slope and intercept are two useful parameters in comparing the estimated ground heat flux with the Reanalysis-1 ground heat flux. They can be obtained from the scatter plot of the reanalysis ground heat flux as the y-axis vs. the estimated ground heat flux as the x-axis. Close correlation is found in all cases, but the regression slopes vary. When the estimated ground heat flux does not follow the 1:1 line we assume the reanalysis data is more accurate than the estimated ground heat flux using a constant  $I$ , since the thermal inertia parameter is expected to vary spatially. In the following analysis, the thermal inertia in Eq. (7-1) would be adjusted to force the regression slope to be unity.

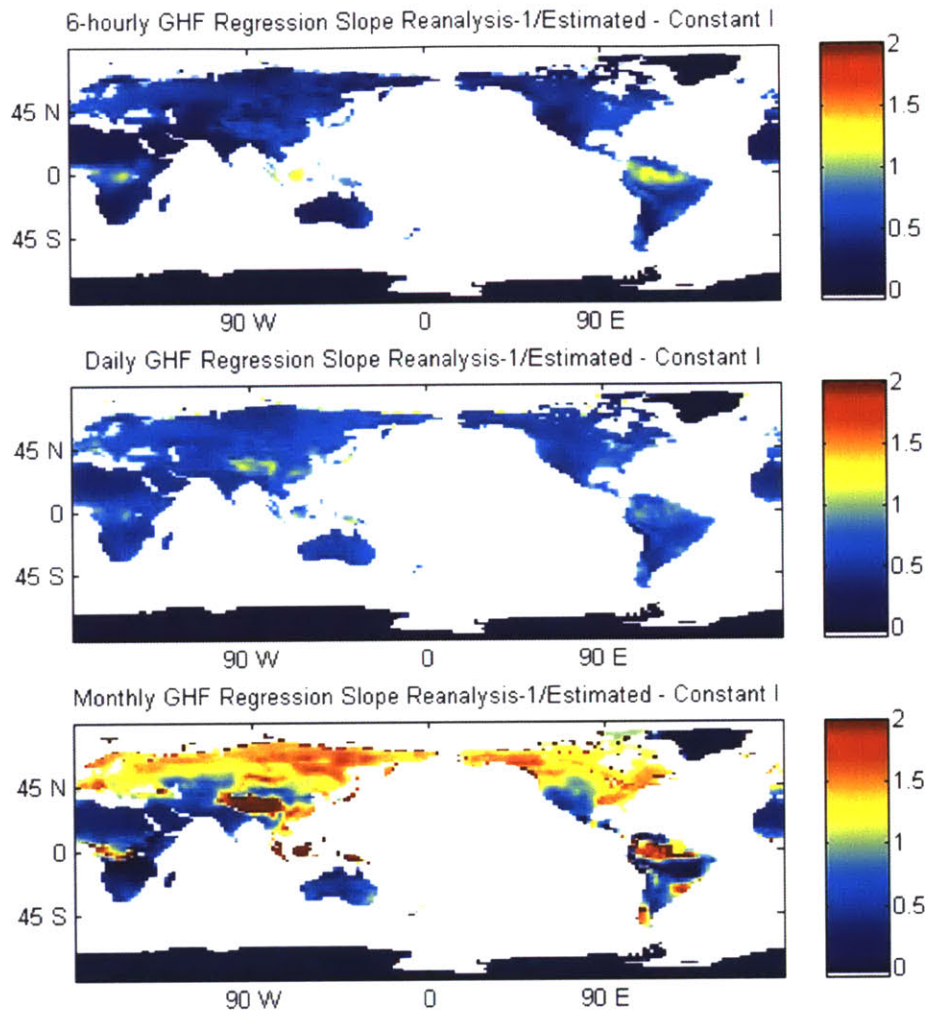


Figure 7-5: Regression Slope of Reanalysis-1 GHF vs. Estimated GHF – Non-Aggregated Data

Figure 7-5 shows the global maps of regression slopes of non-aggregated estimated ground heat flux vs. the corresponding reanalysis data at the three time resolutions. In general the regression slopes of the 6-hourly and daily data are quite low. Note that the regression slopes vary with the time resolutions of the data, while the thermal inertia should be time-independent theoretically. For example the monthly map shows greater regression slopes globally relative to the 6-hourly and daily map with significant geographical variations. The global maps of regression slopes for aggregated data are shown in Figure 7-6.



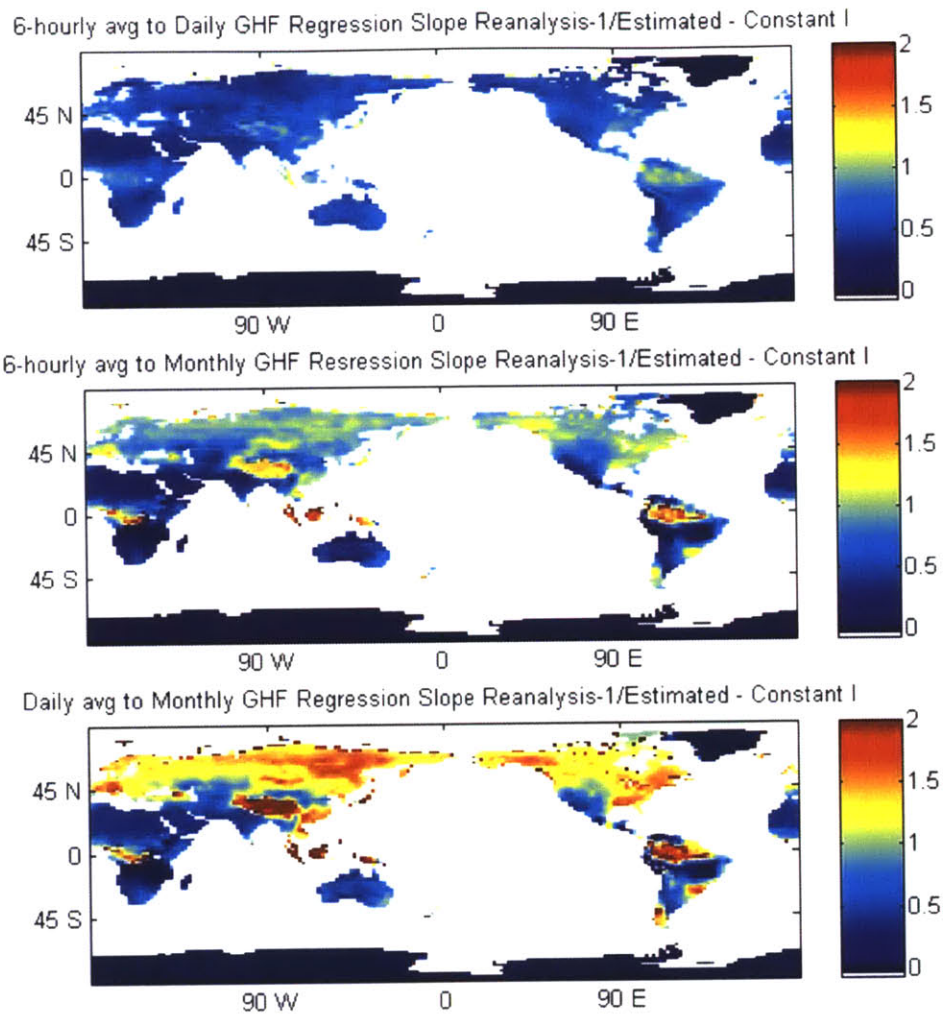


Figure 7-6: Regression Slope of Reanalysis-1 GHF vs. Estimated GHF – Aggregated Data

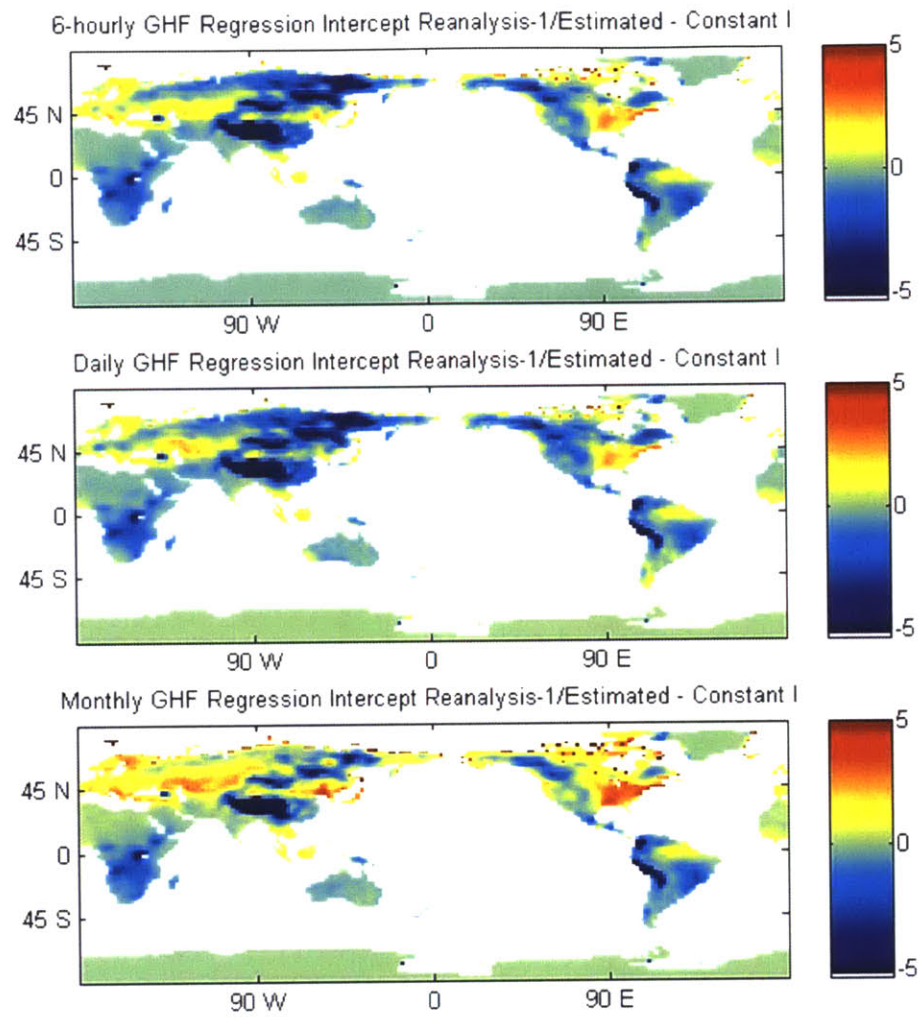


Figure 7-7: Regression Intercept of Reanalysis-1 GHF vs. Estimated GHF – Non-Aggregated Data

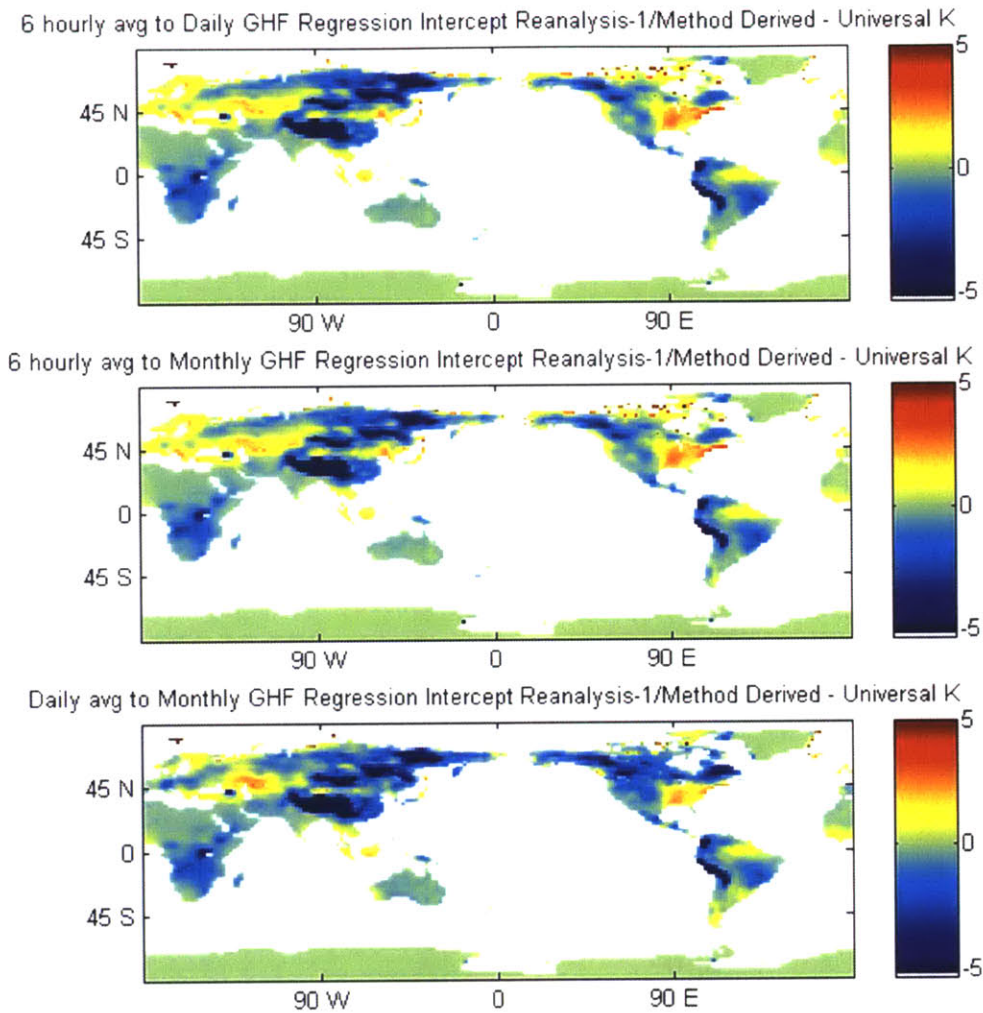


Figure 7-8: Regression Intercept of Reanalysis-1 GHF vs. Estimated GHF – Aggregated Data

Figures 7-7 and 7-8 show that the regression intercepts on the order of  $5 \text{ Wm}^{-2}$  are relatively small. Greater intercepts are observed in the Northern regions and in Central Asia, which will be regions of focus as we attempt to improve the estimated ground heat flux.

Problems associated with the thermal inertia parameter become more prevalent at certain locations. Theoretically, the thermal inertia could be estimated by regressing the estimated against measured ground heat flux for that region. Four points were selected over the globe to illustrate the differences between the estimated ground heat flux and the Reanalysis-1 ground heat flux using the uniform thermal inertia. These points are located in South America, Central Canada, Sahara Desert, and Tibet. At two locations (Central Canada and Sahara Desert), field experimental data are available and will be used in the following section to estimate an effective thermal inertia for the two biggest problem regions (deserts and high latitudes).

Over the South America point, estimated ground heat flux agrees well with the Reanalysis-1 ground heat flux on all time scales. The agreement is best for the 6-hourly time scale, as shown in Figure 7-9. There is a very little variation in the daily cycle. The regularity of this cycle can be seen in the correlation plot in the top left of Figure 7-15. The cycle is so regular that clusters of points are seen for different times during the daily cycle. Such features do not exist at the other two time scales. Daily and monthly time-series of the estimated and the observed ground heat flux are plotted in Figures 7-11 and 7-13. It is evident that the uniform thermal inertia leads to good estimates of ground heat flux compared with that from the reanalysis over the South America point.

Over the Sahara Desert, the estimated ground heat flux does not agree with the Reanalysis-1 ground heat flux. The bottom left panel of Figure 7-15 clearly shows the problem of the estimated ground heat flux over the Sahara Desert on the 6-hourly time scale. Later, observations from the HAPEX-SAHEL field experiment (3 degrees N latitude and 14 degrees E longitude) will be utilized to obtain an estimate of the thermal inertia parameter. Figure 7-9 indicates that there is a large difference between amplitude of the diurnal cycle in the estimated 6-hourly ground heat flux and the Reanalysis-1 ground heat flux. We found that the assumed value of the uniform thermal inertia is too large, causing an over estimate of the ground heat flux relative to the reanalysis data assumed to be reasonably accurate.

Time series of the ground heat fluxes at a location in Central Canada are shown in Figure 7-10, 7-12, 7-14 with the corresponding scatter plot in Figure 7-16. This location is near the BOREAS field experiment site (53 degrees N latitude and 106 degrees W longitude). On top of Figure 7-10 the 6-hourly data is presented over the period of mid March to mid June. Unlike the other points, ground heat flux at this location does not have a regular diurnal cycle. For the first half of this period (wintertime), estimated ground heat flux fluctuates wildly while the reanalysis flux is close to zero. For the middle portion of the year the reanalysis and estimated flux agree very well. It indicates that the diffusion coefficient varies seasonally. The same feature can be seen in the daily data over the entire year on the top portion of Figure 7-12. In Figure 7-16, the data points in the scatter plot are also grouped where the points at the beginning and the end of the 6-hourly and daily time series stay in one group with the data points in the middle of the time series in another. The monthly data does not show this pattern. Larger records are needed to ascertain if this behavior persists in the monthly data.

An additional location in Central Asia (Tibetan Plain, 34°N 86°E) is also studied. This location has already shown up on many of the early plots as a source of major difference between the estimated and reanalysis ground heat flux. Judging by the scatter plots in Figure 7-16, this location is similar to the South America with a strong seasonal cycle on the 6-hourly time scale using most of the year. The assumed uniform thermal inertia works well for most of the year across all time scales. Biases exist for this point as shown in Figure 7-12 at the daily time scale in the middle portion of the year. Such errors can not be fixed by altering the thermal inertia.

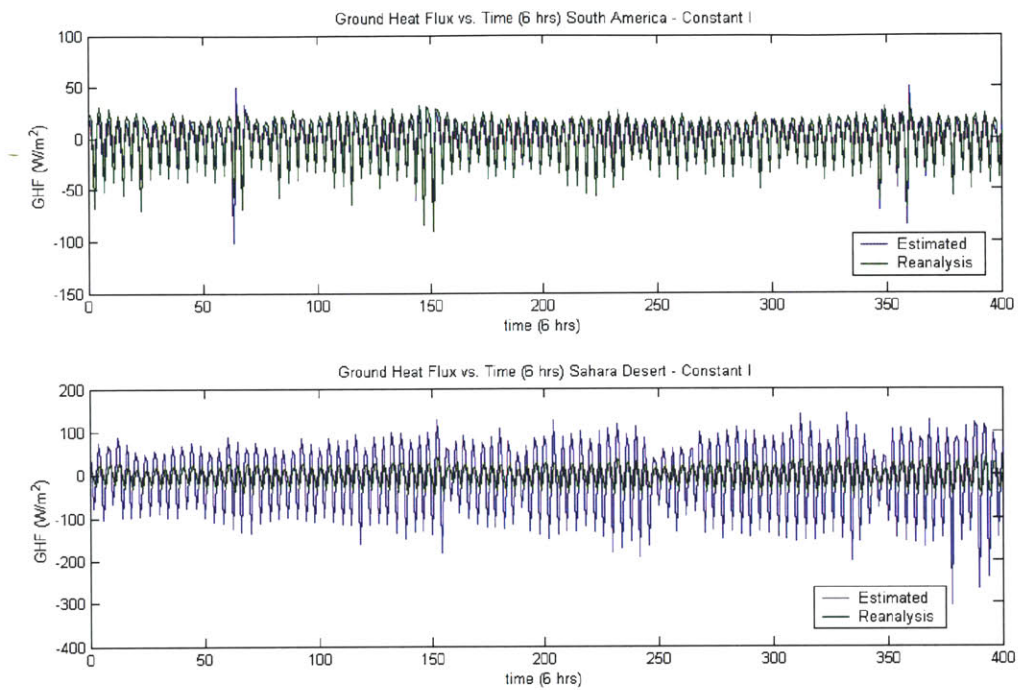


Figure 7-9: Comparison of Estimated and Reanalysis-1 Ground Heat Flux over South America and the Sahara Desert for the 6-hr time scale

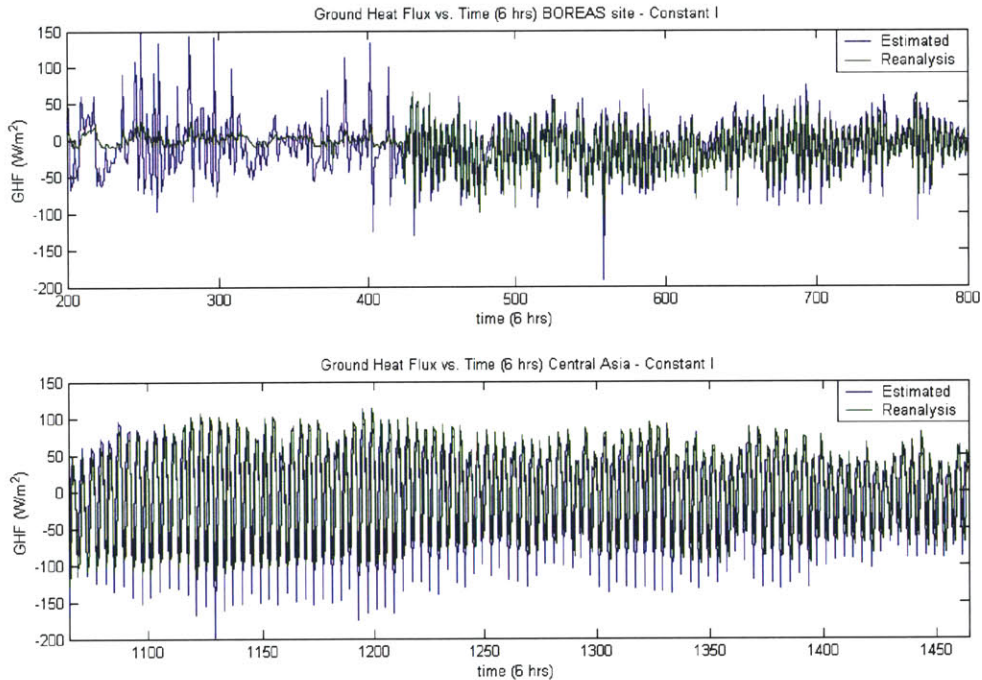


Figure 7-10: Comparison of Estimated and Reanalysis-1 Ground Heat Flux over the BOREAS site (Central Canada) and Central Asia for the 6-hr time scale

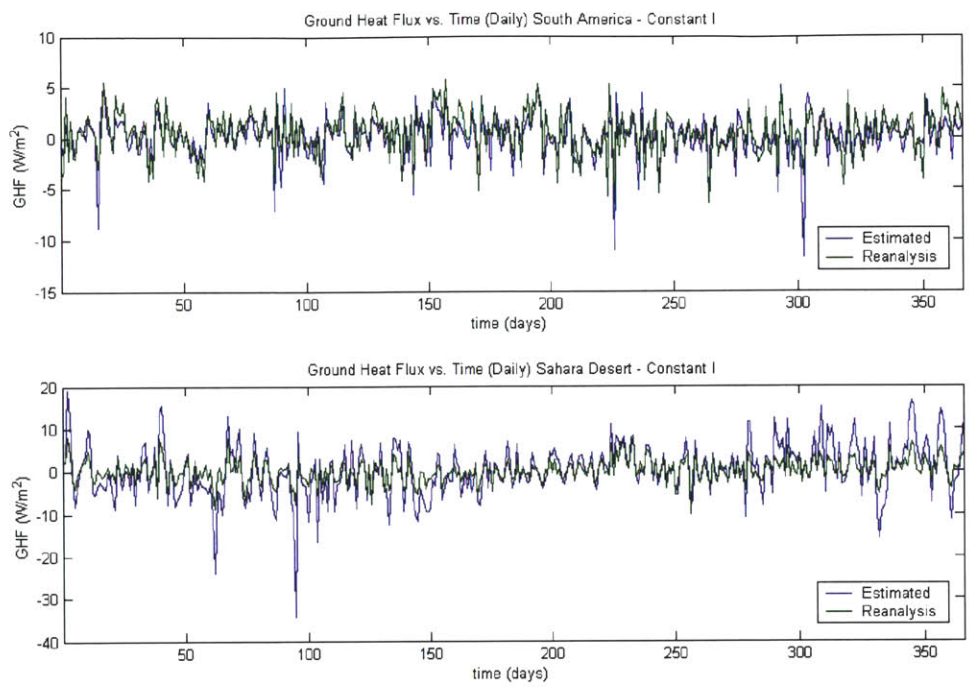


Figure 7-11: Comparison of Estimated and Reanalysis-1 Ground Heat Flux over South America and the Sahara Desert for the Daily time scale

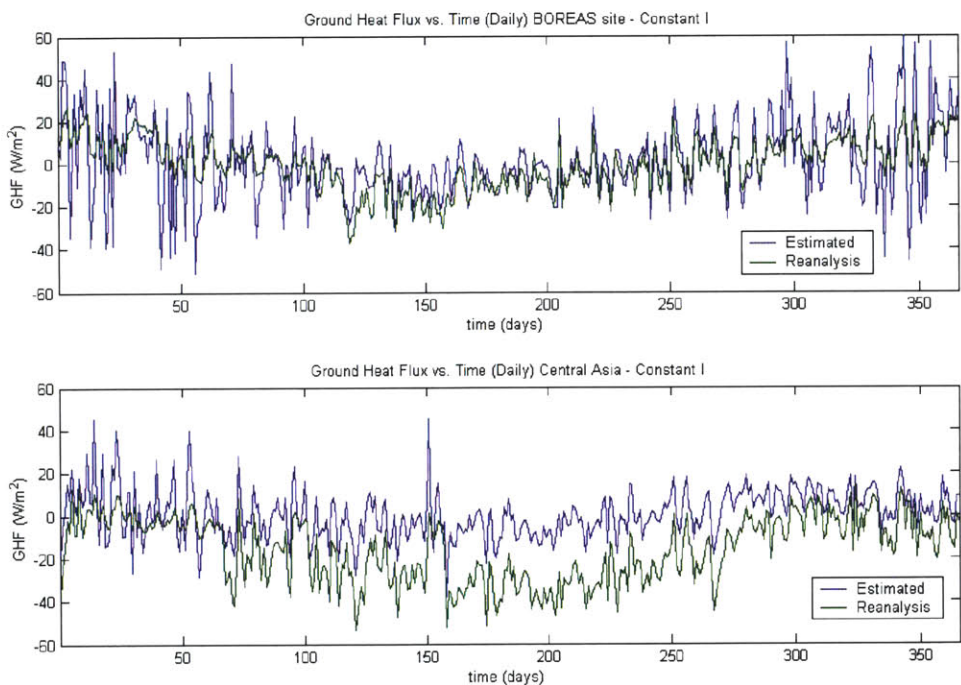


Figure 7-12: Comparison of Estimated and Reanalysis-1 Ground Heat Flux over the BOREAS site (Central Canada) and Central Asia for the Daily time scale

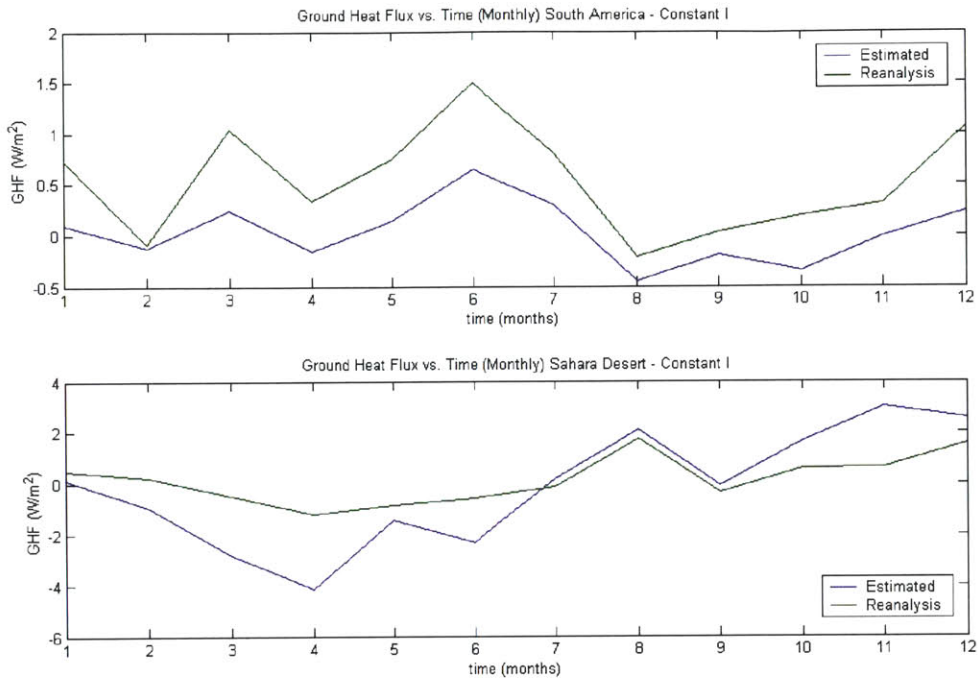


Figure 7-13: Comparison of Estimated and Reanalysis-1 Ground Heat Flux over South America and the Sahara Desert for the Monthly time scale

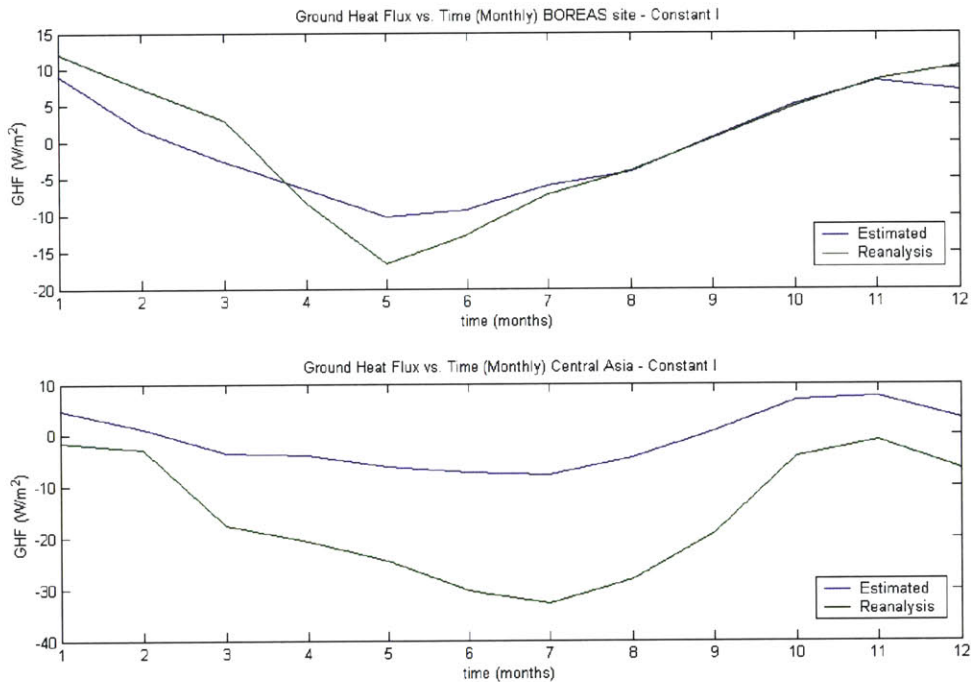


Figure 7-14: Comparison of Estimated and Reanalysis-1 Ground Heat Flux over the BOREAS site (Central Canada) and Central Asia for the Daily time scale

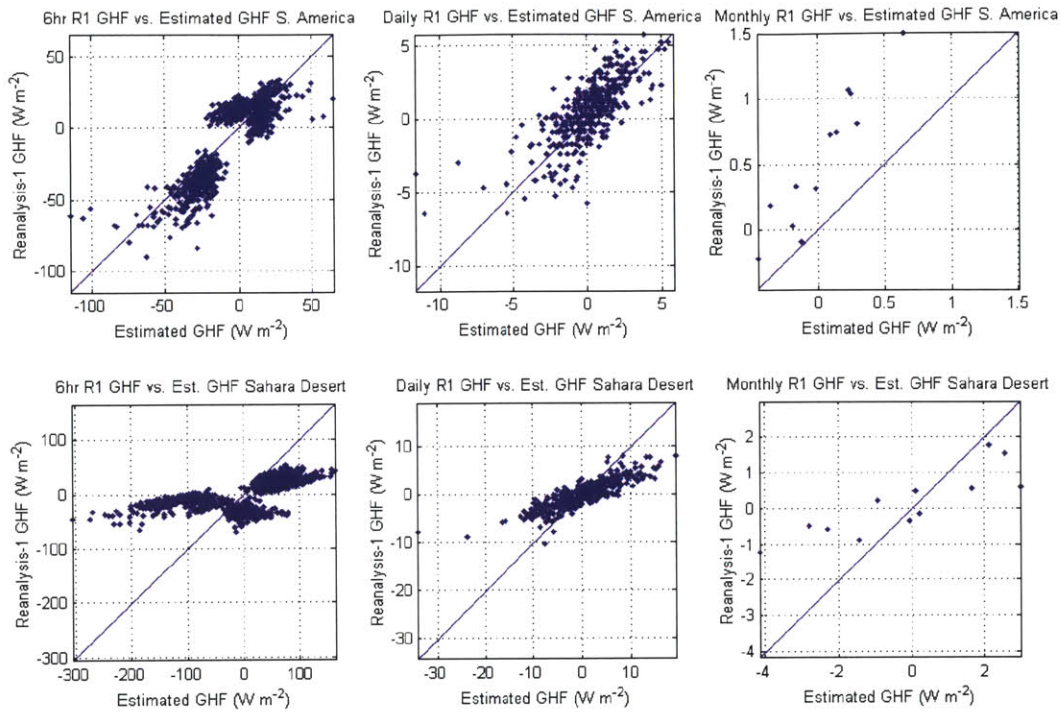


Figure 7-15: Comparison of Estimated and Reanalysis-1 Ground Heat Flux over South America and the Sahara Desert – Correlation Plots for All Time Scales

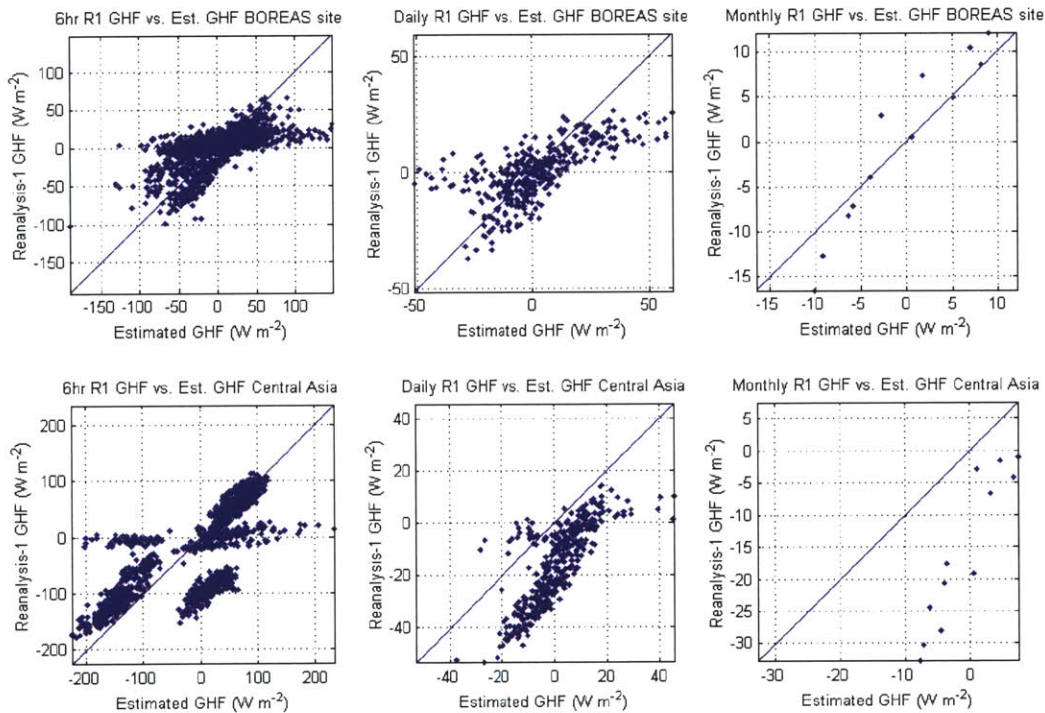


Figure 7-16: Comparison of Estimated and Reanalysis-1 Ground Heat Flux over the BOREAS site (Eastern Canada) and Central Asia – Correlation Plots for All Time Scales



## 7.2. Estimates of Thermal Inertia Parameter

For two of the most problematic regions, observed ground heat flux and skin temperature are available and will be used to obtain a realistic estimate of the thermal inertia parameter. The BOREAS field experiment is located in Central Canada. The HAPEX-SAHEL field experiment is located in the Sahara Desert.

### 7.2.1. BOREAS

It was first hypothesized that the larger differences between the reanalysis and the estimated ground heat flux are due to the frozen ground. Under this assumption the skin temperature could be used to indicate an alternative thermal inertia depending if the skin temperature were below freezing for a given time. The BOREAS field data used here consists of observations every thirty minutes from April 1 to December 30, 1996 when soil heat flux and was sampled at 3 cm depth. A soil temperature profile is also available, and can be used to extrapolate the soil surface skin temperature.

A reference ground heat flux was first calculated using Eq. (7-1) where the BOREAS derived 6-hourly skin temperature was used with the original default thermal inertia. Then, ten different thermal inertias are applied to frozen and non-frozen points separately to re-calculate the corresponding ground heat flux.

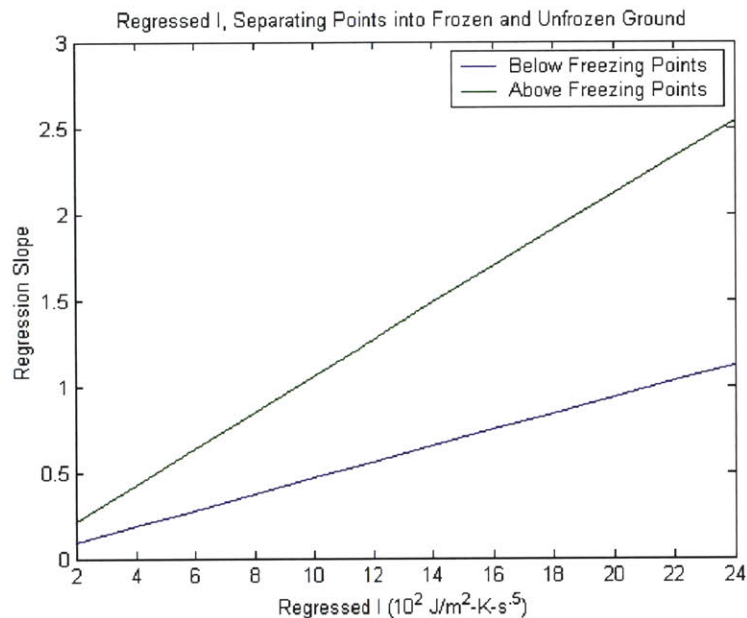


Figure 7-17: Regression Slope vs. Thermal Inertia Used to Obtain Slope For Two Different Types of Land

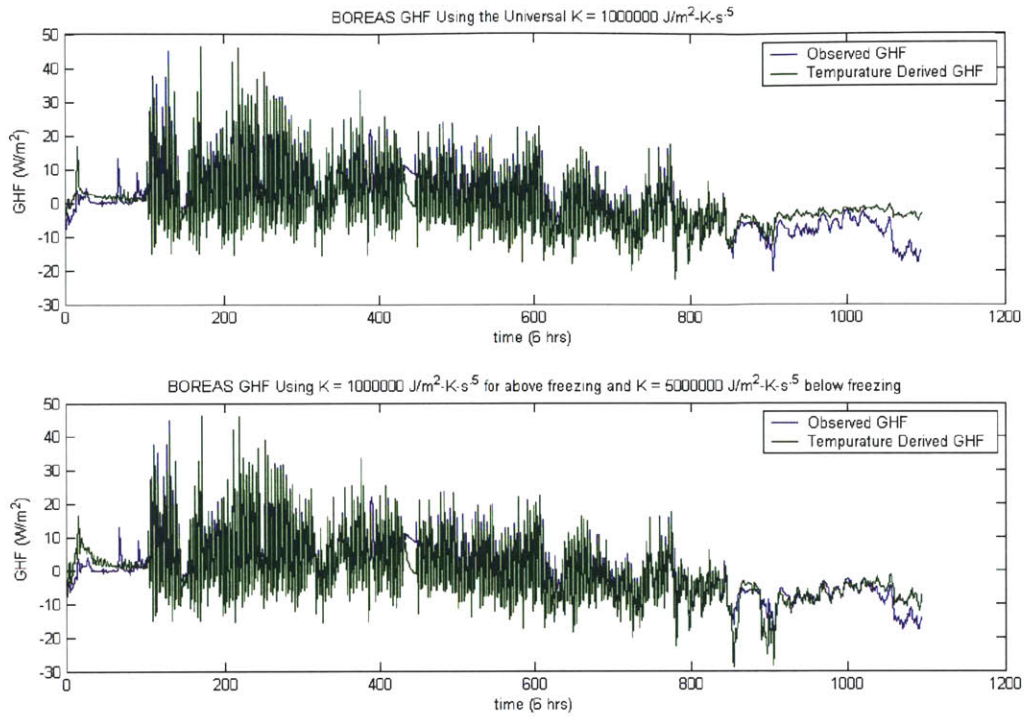


Figure 7-18: Estimated and Observed GHF vs. Time for 2 Sets of Scenarios for the Value of Thermal Inertia

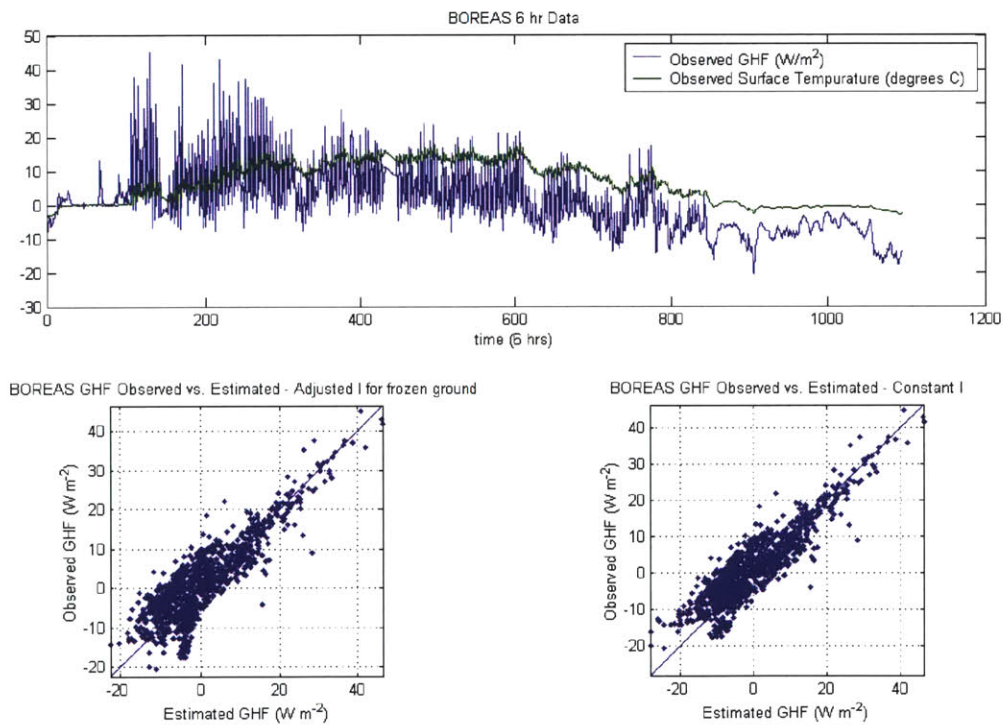


Figure 7-19: Top – Observed Data at the BOREAS site, Bottom – Scatter plots for 2 Sets of Scenarios for the Value of the Thermal Inertia

In Figure 7-17 the regression slopes are plotted for the comparison of the observed and the estimated ground heat flux with different thermal inertias applied to frozen and non-frozen points. The regression slope is closest to one for thermal inertia of  $1 \times 10^3 \text{ J m}^{-2} \text{ s}^{-1/2} \text{ K}^{-1}$  for the non-frozen points (default) and  $2.2 \times 10^3 \text{ J m}^{-2} \text{ s}^{-1/2} \text{ K}^{-1}$  for the frozen points. In Figure 7-18 the observed ground heat flux is compared to the estimated using the default thermal inertia and the newly estimated thermal inertia.

However, matching the reanalysis ground heat flux with that calculated using the BOREAS temperature record requires a very low thermal inertia under “frozen” conditions. This is not due to the frozen ground, and snow cover must be the cause.

Snow cover is an output parameter in the reanalysis data for each land point at every time scale, and can be used in place of temperature to separate points with different thermal inertias. For a full analysis of the impact of snow cover, the thermal inertia the Reanalysis-2 is introduced in addition to the Reanalysis-1 and BOREAS data. Relevant data and results using the default thermal inertia are presented in Figure 7-20 through 7-22.

Differences are evident between the surface temperature given by the Reanalysis-1 and Reanalysis-2 and the BOREAS field data. The middle panels of Figures 7-20, 7-21, and 7-22 illustrate the time series of skin temperature. The uniqueness of the BOREAS skin temperature record indicates that the skin temperature from both reanalysis datasets is not the same as the measured skin temperature at the BOREAS site. In fact, skin temperature from the BOREAS dataset is the surface soil temperature beneath the snow cover, and the reanalysis skin temperature is the surface snow temperature.

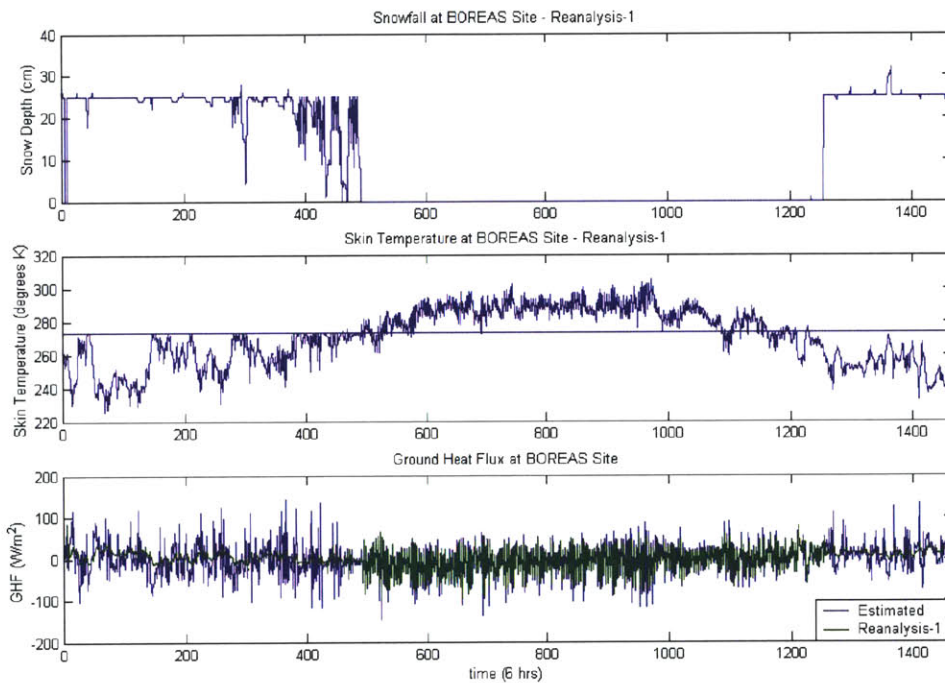


Figure 7-20: 6-hourly Time Series Data from the Reanalysis-1 for the BOREAS Site

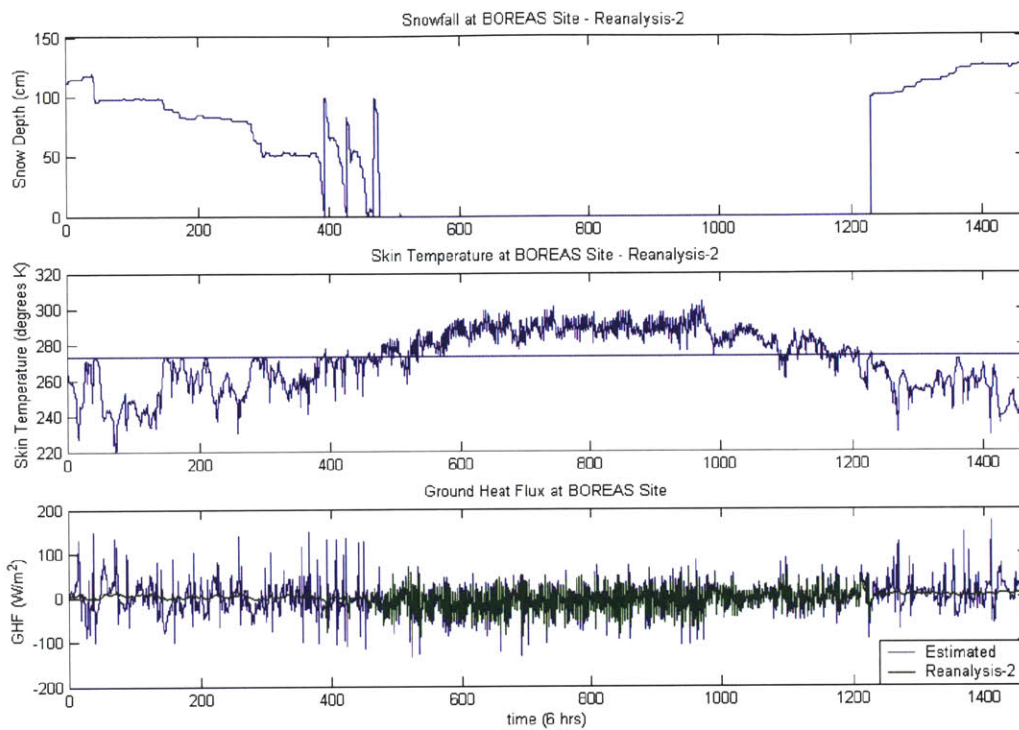


Figure 7-21: 6-hourly Time Series Data from the Reanalysis-2 for the BOREAS Site

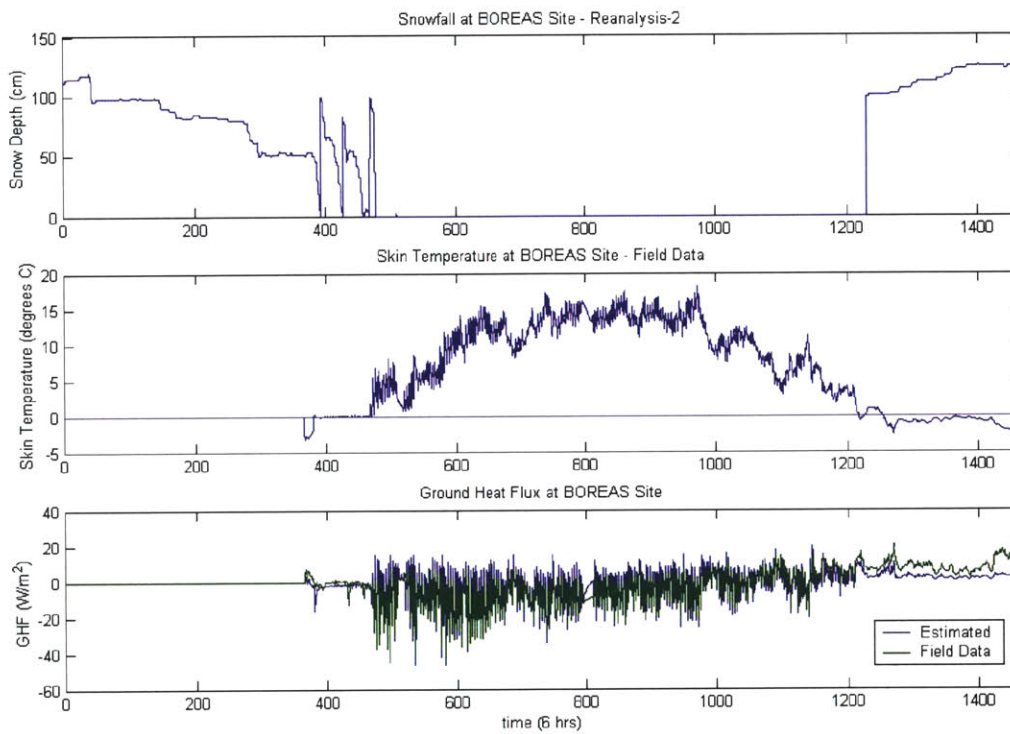


Figure 7-22: 6-hourly Time Series Data from the BOREAS experiment with the Reanalysis-2 snowfall for the BOREAS Site

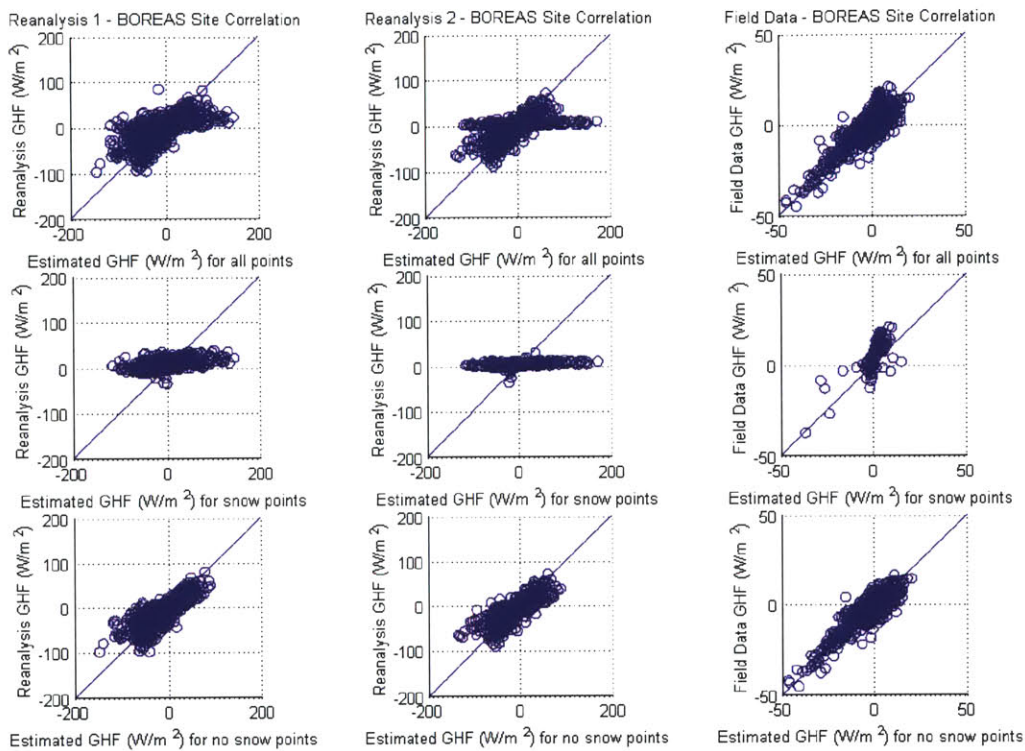


Figure 7-23: Correlation plots of Observed vs. Estimated Ground Heat Flux for the Reanalysis-1, Reanalysis-2, and BOREAS Experiment over the BOREAS site, distinguishing between snow points and non-snow points

In Figure 7-23 all three sources of data for this point are compared with the scatter plots of reanalysis vs. estimated ground heat flux with the default thermal inertia. The first row of graphs shows all data points. The second row shows the snow covered points only. The third row shows the snow free points only. As BOREAS data product does not have snow cover measurement, the Reanalysis-2 snow cover is used. In the case of both reanalyzes, there is a clear distinction between the scatter plots of snow points and non snow points. For snow points, the scatter plot reveals a linear regression with a very low slope, while non-snow points fit the 1:1 line well. In the scatter plot of BOREAS data, the snow points reveal a linear relationship with a slope of greater than 1. The reanalysis results correspond to substituting a thermal inertia much lower than the default thermal inertia into Equation 7-1 for all snow points in order to produce a regression between estimated and reanalysis ground heat flux which better fits the 1:1 line.

In the Reanalysis-1 and Reanalysis-2 datasets, sub-freezing temperature points are usually covered by snow or vice-versa. Therefore, it would be reasonable to assume that most points with sub-freezing temperature are covered with snow. In order to prove that snow is responsible for dampening the ground heat flux signal in the reanalysis due to a low thermal inertia, we examine two points where sub-freezing skin temperature persists without snow cover. In 1988, two locations in Central Asia and the Northern Reaches of Canada (much farther north than the BOREAS site) were experiencing sub-freezing temperatures and snow free.

In the top panel of Figure 7-24 the sub-freezing skin temperature is plotted for the duration of the data presented for the Tibetan Plain location. The middle panel of the figure shows the time series of the Reanalysis-1 ground heat flux and the estimated ground heat flux. It is clear from the scatter plot that the uniform thermal inertia is acceptable for this land type in terms of its magnitude although there is apparent bias. Nevertheless, the results are consistent with the reanalysis with a near 1:1 regression with temperatures below zero. Hence, it's the snow that causes the differences in the reanalysis and estimated ground heat flux. Further proof is provided for a point in the Northern Canadian Islands in Figure 7-25. The differences between the reanalysis and the estimated ground heat flux in northern land points are clearly corresponding to a much lower thermal inertia when snow is present.

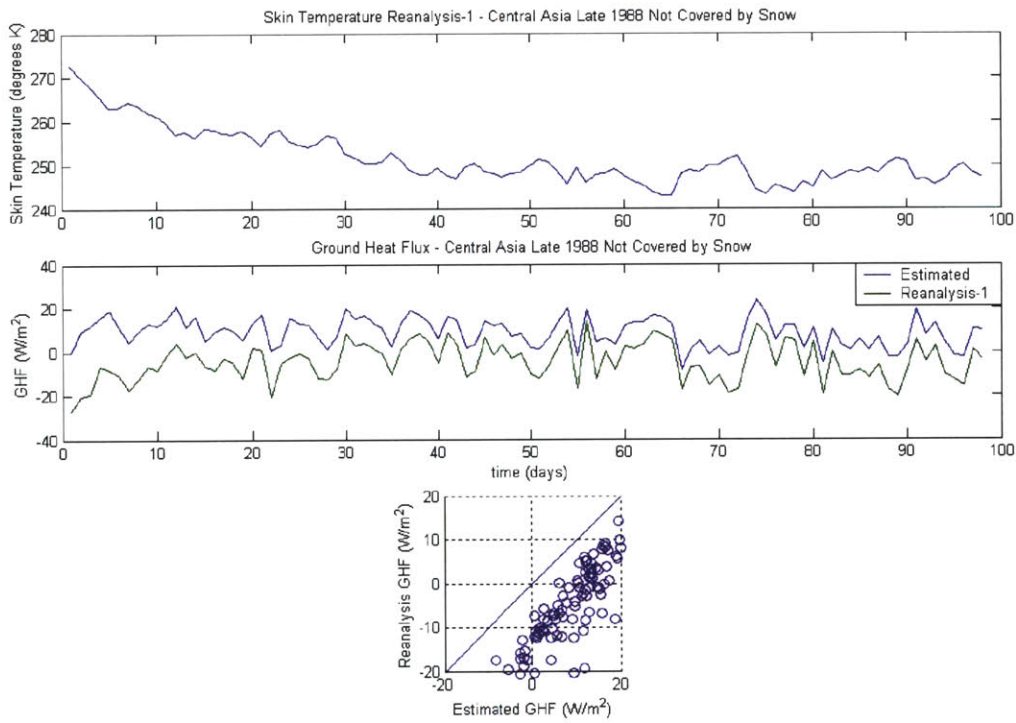


Figure 7-24: Plots for Central Asia in Late 1988 where skin temperature is below 0 degrees C and there is no snow cover, daily time step

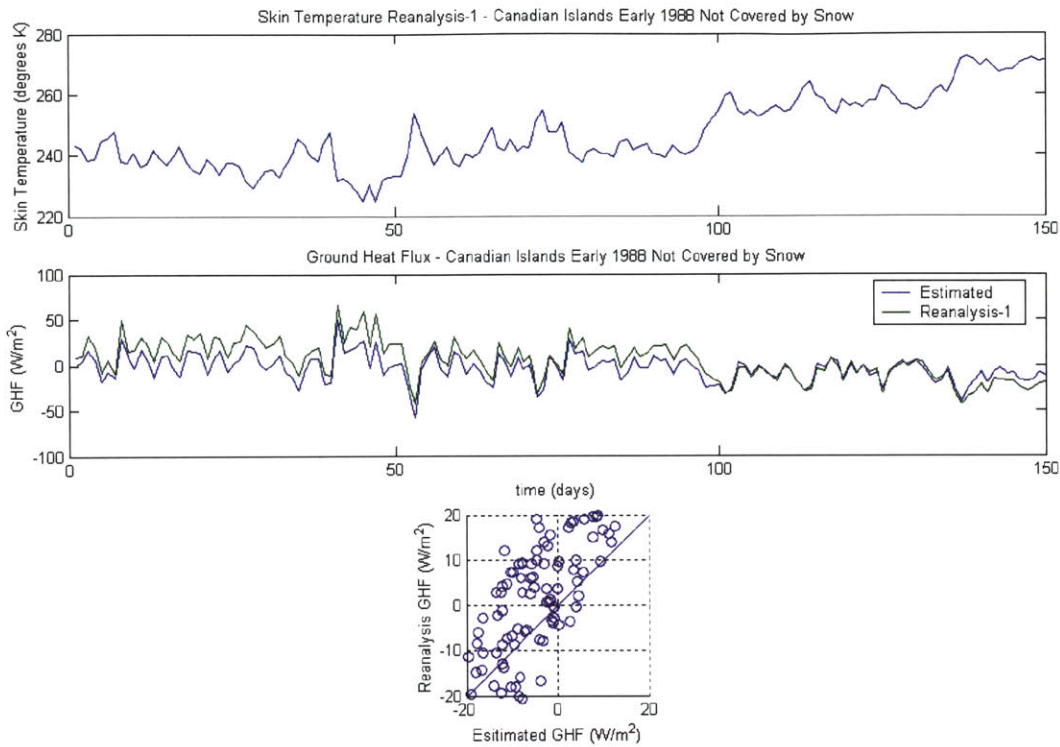


Figure 7-25: Plots for Canadian Islands in Late 1988 where skin temperature is below 0 degrees C and there is no snow cover, daily time step

### 7.2.2. HAPEX – SAHEL

Measurements of ground heat flux in a desert region are available from the HAPEX-SAHEL field experiment. Two datasets from this study are utilized, the ECSS\_HERB\_FLUX\_DATA and ECSS\_HERB\_SOIL\_1\_DAT prepared by Bruno Monteny are available at <http://www.ird.fr/hapex/>. The first dataset contains various surface heat fluxes including the ground heat flux, while the second one contains the surface temperature. The longest continuous record covers the period of 9/21/92 to 10/18/02. The time step of this data is 20 minutes. The observed surface temperature and ground heat flux are shown in Figure 7-26.

The true thermal inertia can be estimated graphically by plotting regression slope vs. thermal inertia. In order to estimate the thermal inertia, different values of thermal inertia are applied to calculate the ground heat flux. The best estimate of thermal inertia will be that corresponding to unity slope of the regression between the estimated and observed ground heat flux. The bottom panel of Figure 7-26 plots the regression slope versus the thermal inertia. The slope of one corresponds to the thermal inertia of about  $540 \text{ J m}^{-2} \text{ s}^{-1/2} \text{ K}^{-1}$ . In Figure 7-27 and 7-28 the estimated ground heat flux is compared to the observed ground heat flux for the default thermal inertia and the regression derived thermal inertia. Eq. (7-1) with the newly estimated thermal inertia can almost exactly duplicate the ground heat flux.

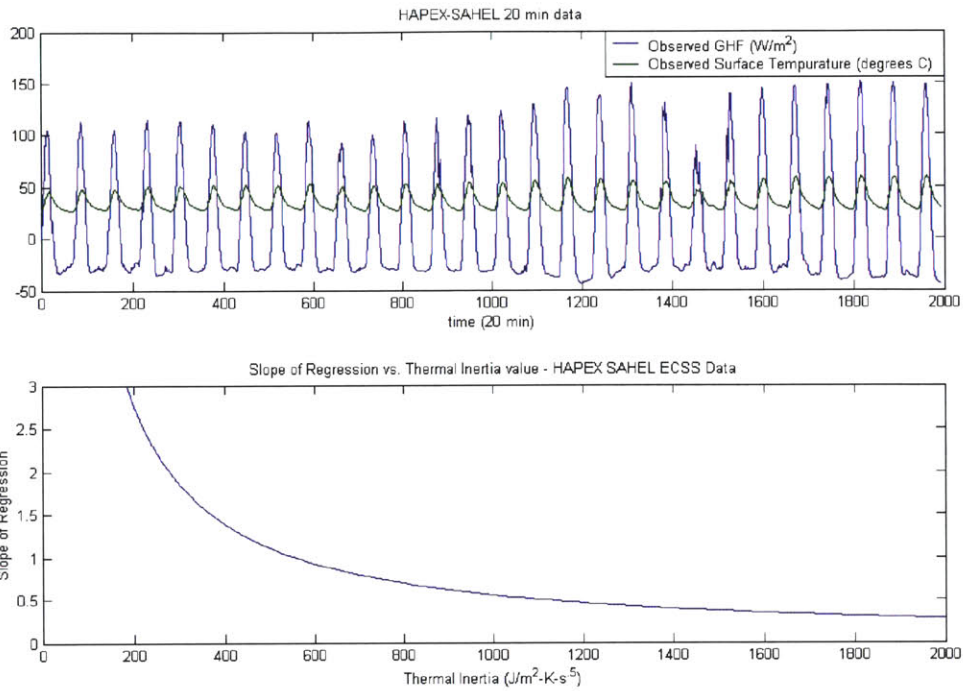


Figure 7-26: Top-HAPEX-SAHEL observed data, Bottom- Regression slope between the observed and method derived ground heat fluxes versus thermal inertia used

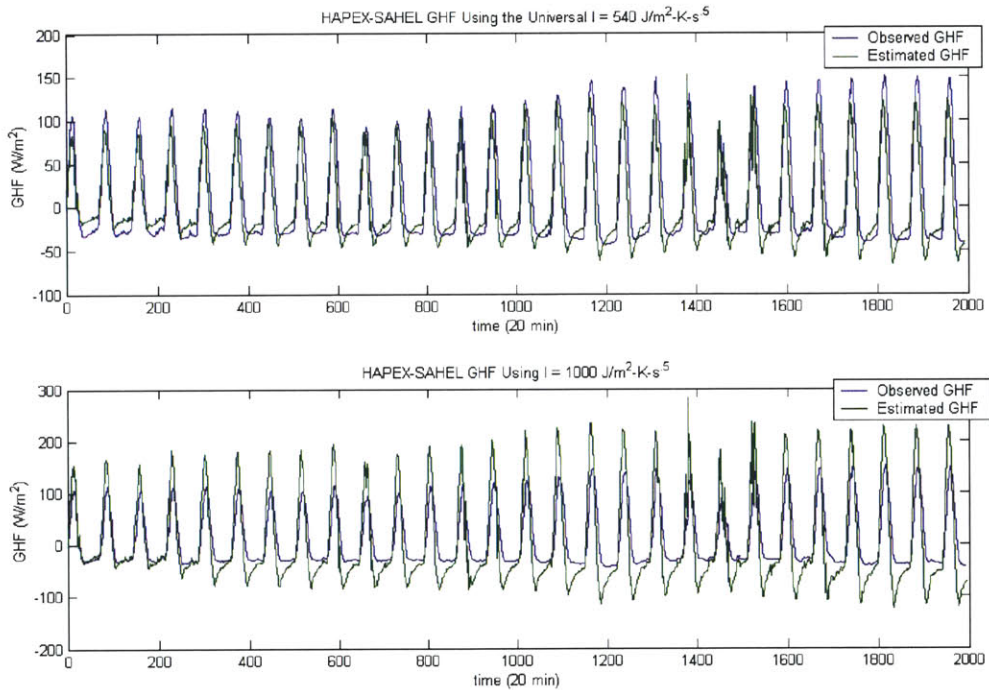


Figure 7-27: Time Series Plots of Observed Ground Heat Flux vs. Method Ground Heat Flux



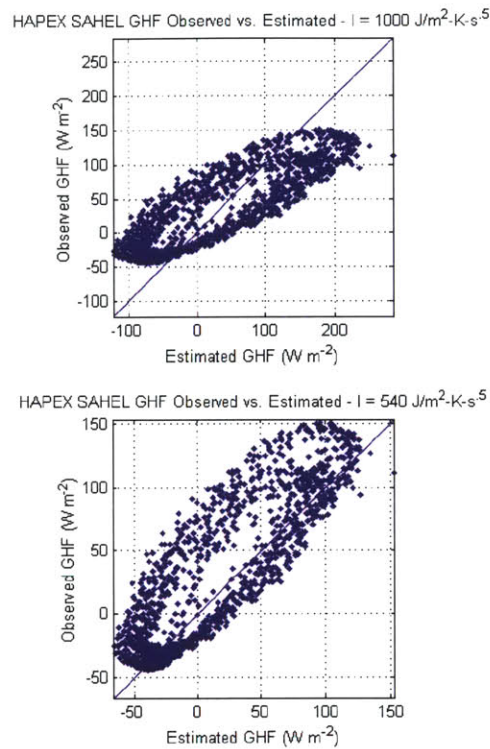


Figure 7-28: Scatter Plots of Observed Ground Heat Flux vs. Method Ground Heat Flux

### 7.3. Estimation of Thermal Inertia by Land Type

The thermal inertia is a function of land type which varies spatially. It would be unrealistic to estimate the thermal inertia at each of the 5914 land points over the entire globe by a regression approach as shown above. Such derived thermal inertia parameter would rely on the reanalysis ground heat flux data. Our objective is to create a unique ground heat flux dataset independent of existing model products. Each pixel of the reanalysis has an area of about 250 km x 250 km within which soil properties are rarely homogeneous. Thermal properties are known for individual soil materials, but an area-mean thermal inertia is difficult to calculate from the individual components. In addition, observations are only available at limited locations covering a small fraction of the earth's surface.

An alternative method is therefore suggested for the derivation of these thermal inertias. This method involves the use of a gridded land type dataset. The University of Maryland (UMD) Global Land Cover Classifications data set is a gridded dataset of broad land type classification available at three resolutions: 1-degree, .5-degree, and .25 degree. For this study, the 1-degree dataset is used to assign each reanalysis point with one of twelve basic land classifications. The 1 degree point nearest to the center of each reanalysis grid point is used. The results of this regridding are presented in Figure 7-29. The legend is given in Table 7-1, with the number 0 (blue) indicating water.

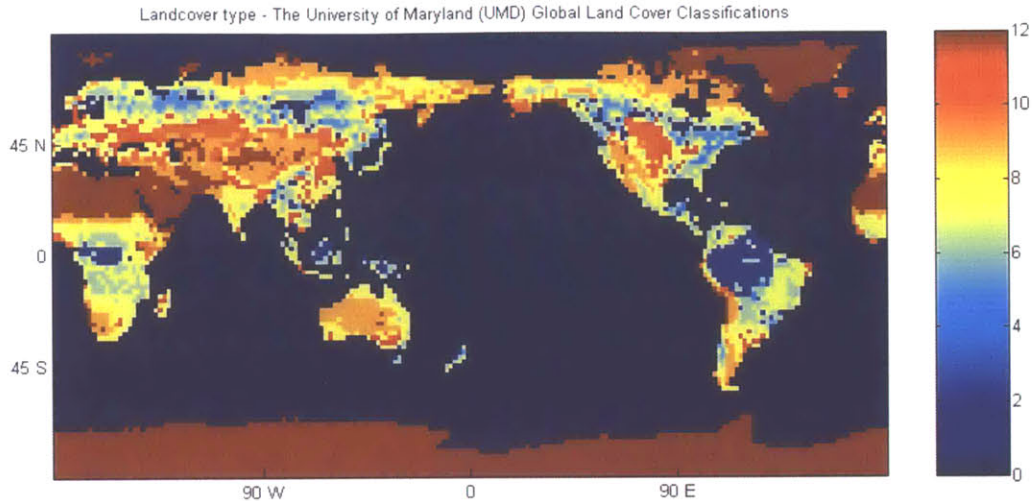


Figure 7-29: Land Type Classification

For all points pertaining to each land type across the globe, all estimated and reanalysis ground heat flux data are combined and plotted on one graph, and from this graph a regression slope is found. This slope is then used to find the thermal inertia for all the points of a given land type. By this methodology the reanalysis ground heat flux is used as a loose guideline to regress the thermal inertia for a large number of points, but is not used on a point by point basis. This approach keeps the ground heat flux methodology simple. The land classification is simple enough to be used globally.

Additionally, a separate classification is assigned to snow cover. It was shown in Section 7.2 that thermal inertia is highly sensitive to snow cover. Thus any point in each time series for each land point which contains snow is placed in a separate category. Thus, a representative thermal inertia for all snow points is obtained. This thermal inertia is applied to all data points where snow is present.

Now we use this methodology to derive the thermal inertia with the help of the Reanalysis-1 and Reanalysis-2 data at three resolutions. The results for each land type are given in Tables 7-1 through 7-6. Each table lists the regression slope and intercept for each land type. The regression intercept is included to check for biases. The correlation coefficient is provided in order to check the quality of correlation for a given land type. A good correlation assures the grouping of many different points into one land type, corresponds to one thermal inertia.

As observed in Section 7.1, the regression slopes vary with the time resolution of the input data for each land type. For example, the regression slope using monthly data is smaller than using daily data. The magnitude of this decrease is presented by an Increase Factor defined as the increase from daily to monthly in the monthly table, and increase from 6-hourly to daily in the daily table. There are improvements in the Increase Factor in the Reanalysis-2, but still for the Reanalysis-2 this factor is significantly greater than 1 across each time scale. In order to utilize this analysis to link the thermal inertias to these different land types, one set of thermal inertia coefficient must be chosen for each reanalysis.

Linear Regression of Ground Heat Flux Data By Land Type 1988 Monthly Time Scale - Reanalysis-1						
#	Land Type	Slope	Intercept	Correlation	Regressed I	Inc. Factor
0	Snow/water	0.4485	0.8134	0.4444	<b>449</b>	<b>4.32</b>
1	Evergreen Needleleaf Forests	1.5742	0.0514	0.8156	1574	1.73
2	Evergreen Broadleaf Forests	0.9316	-0.1625	0.3943	932	1.33
3	Deciduous Needleleaf Forests	2.2850	2.0035	0.8662	2285	2.14
4	Deciduous Broadleaf Forests	1.0697	-0.1565	0.7946	1070	1.34
5	Mixed Forest	1.4914	-0.4092	0.8362	1491	1.58
6	Woodlands	1.4160	-0.7832	0.8246	1416	1.59
7	Wooded Gasslands/Shrublands	1.2942	-0.6630	0.7905	1294	1.57
8	Closed Bushlands or Shrublands	1.3898	-0.9314	0.8328	1390	1.63
9	Open Shrublands	0.9639	-1.0933	0.6921	964	1.27
10	Grasslands	0.9918	-0.4550	0.7171	992	1.34
11	Croplands	1.0551	-0.0825	0.8294	1055	1.46
12	Barren	0.4905	-0.2419	0.6028	<b>491</b>	<b>0.85</b>

Table 7-1: Thermal Inertia Calculations Monthly Time Scale Reanalysis-1

Linear Regression of Ground Heat Flux Data By Land Type 1988 Daily Time Scale - Reanalysis-1							
#	Land Type	Slope	Intercept	Correlation	Regression K	Inc. Factor	Rounded Reg. K
0	Snow/water	0.1042	1.6341	0.3766	<b>104</b>	<b>3.03</b>	<b>100</b>
1	Evergreen Needleleaf Forests	0.9055	-4.2701	0.8495	906	1.30	<b>910</b>
2	Evergreen Broadleaf Forests	0.7028	-0.6177	0.6937	703	0.97	<b>700</b>
3	Deciduous Needleleaf Forests	1.0695	-9.6510	0.8613	1070	1.23	<b>1070</b>
4	Deciduous Broadleaf Forests	0.7955	0.4571	0.8164	796	1.27	<b>800</b>
5	Mixed Forest	0.9450	-3.2597	0.8516	945	1.31	<b>950</b>
6	Woodlands	0.8934	-2.2716	0.8131	893	1.63	<b>890</b>
7	Wooded Gasslands/Shrublands	0.8217	-1.7578	0.7953	822	1.89	<b>820</b>
8	Closed Bushlands or Shrublands	0.8521	-2.2885	0.7929	852	2.46	<b>850</b>
9	Open Shrublands	0.7575	-2.3280	0.7473	758	2.18	<b>760</b>
10	Grasslands	0.7417	-1.9820	0.7829	742	1.53	<b>740</b>
11	Croplands	0.7217	-0.8180	0.8329	722	1.48	<b>720</b>
12	Barren	0.5755	-0.5276	0.7559	<b>576</b>	<b>2.68</b>	<b>580</b>

Table 7-2: Thermal Inertia Calculations Daily Time Scale Reanalysis-1

Here, the daily time scale is used as a basis for finding the thermal inertia. The regressed thermal inertia in  $\text{J m}^{-2} \text{s}^{-1/2} \text{K}^{-1}$  is rounded in Tables 7.2 and 7.5 to the nearest  $10 \text{ J m}^{-2} \text{s}^{-1/2} \text{K}^{-1}$ . It is found that the daily time scale does not have the problem of inconsistency that exists in the 6-hourly and monthly time scale data. One problem with 6-hourly data is that the diurnal cycle is represented by only four points, leading to a grouping of data points (see Figure 7-15, the South American point on a 6 hourly time scale). The monthly data only has twelve points to represent the seasonal cycle and ground heat flux changes little on a monthly basis. This behavior is reflected in the correlation coefficients which, on the daily time scale, are highest for most land types, indicating a fairly strong linear relationship.

Linear Regression of Ground Heat Flux Data By Land Type 1988 6-hourly Time Scale - Reanalysis-1					
#	Land Type	Slope	Intercept	Correlation	Regressed K
0	Snow	0.0333	1.6773	0.2273	<b>33</b>
1	Evergreen Needleleaf Forests	0.6954	-5.2290	0.7988	695
2	Evergreen Broadleaf Forests	0.7231	-0.6797	0.7598	723
3	Deciduous Needleleaf Forests	0.8668	-12.1303	0.8326	867
4	Deciduous Broadleaf Forests	0.6281	0.6021	0.7820	628
5	Mixed Forest	0.7212	-4.0725	0.7997	721
6	Woodlands	0.5476	-2.8203	0.7173	548
7	Wooded Grasslands/Shrublands	0.4344	-2.1582	0.6694	434
8	Closed Bushlands or Shrublands	0.3459	-3.1758	0.6335	346
9	Open Shrublands	0.3475	-2.9603	0.6314	348
10	Grasslands	0.4848	-2.4880	0.7078	485
11	Croplands	0.4870	-0.8715	0.7179	487
12	Barren	0.2151	-0.4904	0.5853	<b>215</b>

Table 7-3: Thermal Inertia Calculations 6-hourly Time Scale Reanalysis-1

Linear Regression of Ground Heat Flux Data By Land Type 1988 Monthly Time Scale - Reanalysis-2						
#	Land Type	Slope	Intercept	Correlation	Regressed K	Inc. Factor
0	snow	0.3225	0.1036	0.4929	<b>323</b>	<b>7.51</b>
1	Evergreen Needleleaf Forests	1.0450	-1.2629	0.7615	1045	1.40
2	Evergreen Broadleaf Forests	0.7420	-0.8708	0.3956	742	0.95
3	Deciduous Needleleaf Forests	1.4538	-1.6900	0.8683	1454	1.50
4	Deciduous Broadleaf Forests	1.1115	0.1068	0.8999	1112	1.72
5	Mixed Forest	1.1130	-0.5935	0.8637	1113	1.45
6	Woodlands	1.1911	-0.9912	0.8154	1191	1.49
7	Wooded Grasslands/Shrublands	1.2201	-0.8437	0.7948	1220	1.56
8	Closed Bushlands or Shrublands	1.4961	-1.4572	0.8331	1496	1.83
9	Open Shrublands	1.1159	-1.4005	0.7759	1116	1.51
10	Grasslands	0.9377	-0.7924	0.7464	938	1.36
11	Croplands	0.9593	0.1430	0.8779	959	1.38
12	Barren	0.7074	-0.3437	0.8300	<b>707</b>	<b>1.10</b>

Table 7-4: Thermal Inertia Calculations Monthly Time Scale Reanalysis-2

There is a problem in the Reanalysis-2 where the thermal inertia for snow seems unrealistically low. This thermal inertia is therefore replaced by that from the Reanalysis-1 in the final product. The thermal inertia derived using daily data agrees well with that from HAPEX-SAHEL observational data. A thermal inertia of  $540 \text{ J m}^{-2} \text{ s}^{-1/2} \text{ K}^{-1}$  using the field data is well reproduced in the Reanalysis-1 and Reanalysis-2 over the barren land type. The thermal inertia found for snow points (at least in the Reanalysis-1) are also seems consistent with those observed in section 7.2.1, which indicated a nearly flat regression with a default value of thermal inertia.

Linear Regression of Ground Heat Flux Data By Land Type 1988 Daily Time Scale - Reanalysis-2							
#	Land Type	Slope	Intercept	Correlation	Regression I	Inc. Factor	Rounded Reg. I
0	Snow	0.0425	0.5042	0.2768	<b>43</b>	<b>2.69</b>	<b>40</b>
1	Evergreen Needleleaf Forests	0.7473	-3.0729	0.8490	747	1.26	<b>750</b>
2	Evergreen Broadleaf Forests	0.7801	-0.8525	0.7731	780	1.30	<b>780</b>
3	Deciduous Needleleaf Forests	0.9825	-8.4173	0.8652	983	1.31	<b>980</b>
4	Deciduous Broadleaf Forests	0.6481	-0.2584	0.8362	648	1.14	<b>650</b>
5	Mixed Forest	0.7689	-2.7090	0.8505	769	1.27	<b>770</b>
6	Woodlands	0.7993	-2.1442	0.8190	799	1.44	<b>800</b>
7	Wooded Gasslands/Shrublands	0.7802	-1.7803	0.8200	780	1.54	<b>780</b>
8	Closed Bushlands or Shrublands	0.8167	-2.8519	0.8163	817	1.75	<b>820</b>
9	Open Shrublands	0.7392	-2.5549	0.8139	739	1.63	<b>740</b>
10	Grasslands	0.6883	-2.0808	0.8204	688	1.36	<b>690</b>
11	Croplands	0.6959	-0.7033	0.8678	696	1.35	<b>700</b>
12	Barren	0.6399	-0.7676	0.8550	<b>640</b>	<b>1.67</b>	<b>640</b>

Table 7-5: Thermal Inertia Calculations Daily Time Scale Reanalysis-2

Linear Regression of Ground Heat Flux Data By Land Type 1988 6-hourly Time Scale - Reanalysis-2					
#	Land Type	Slope	Intercept	Correlation	Regressed I
0	snow	0.0158	0.6138	0.1827	<b>16</b>
1	Evergreen Needleleaf Forests	0.5947	-3.7037	0.7700	595
2	Evergreen Broadleaf Forests	0.6016	-0.7669	0.7178	602
3	Deciduous Needleleaf Forests	0.7493	-10.6310	0.7986	749
4	Deciduous Broadleaf Forests	0.5666	-0.1585	0.7606	567
5	Mixed Forest	0.6060	-3.3111	0.7675	606
6	Woodlands	0.5542	-2.4555	0.7235	554
7	Wooded Gasslands/Shrublands	0.5056	-1.9676	0.7032	506
8	Closed Bushlands or Shrublands	0.4657	-3.3991	0.6841	466
9	Open Shrublands	0.4521	-2.8283	0.6870	452
10	Grasslands	0.5064	-2.3047	0.7176	506
11	Croplands	0.5174	-0.6335	0.7305	517
12	Barren	0.3829	-0.6355	0.6558	<b>383</b>

Table 7-6: Thermal Inertia Calculations 6-hourly Time Scale Reanalysis-2

## 7.4. Presentation of the Ground Heat Flux Product

In this section the ground heat flux data products are presented. Example plots are shown for each season for the 6-hourly and daily time scales for both reanalyzes. All 12 months of monthly data are shown here. For each point in time, a plot of the ground heat flux product is accompanied by a plot of the ground heat flux at the same time from the reanalysis dataset, followed by the difference between these two. The difference is not meant as the accuracy of the product, but to indicate where the two data products differ. Lastly, each of the three time scales is aggregated to monthly data and presented for each month. A few general comments are provided.

The examples of the 6-hourly time data product are given in Figures 7-30 through 7-37. Ground heat fluxes are plotted every six hours starting from the indicated GMT. The diurnal cycles are shown in all 8 figures. As the day progresses, the negative (blue) ground heat flux propagates across the plot, corresponding to the sunrise and the increase in temperature, while the yellow/orange color represents sunset and a decrease in temperature. The main difference between the new product and the reanalysis in all cases is the amplitude of the maximum and minimum values. This could be predicted, as the thermal inertia from the daily time scale regression calculation is applied to the 6 hour method calculation, and these thermal inertias are larger for almost all land types than the thermal inertias found by regression the 6-hourly data. In terms of season, there are largest maximums and minimums occur during the summer (the location of these points shift during the time of day). Comparing the results of the Reanalysis-1 and the Reanalysis-2 for the 6-hourly time scale, there are smaller differences between the estimated and the reanalysis as shown the difference plots in Figures 7-34 and 7-37. As was seen earlier there were smaller Increase Factors in the regressed thermal inertia in the Reanalysis-2, and in general the Reanalysis-2 data are slightly better.

Examples of the daily data product are given in Figure 7-38 through 7-41. The first two days of the months of January, April, July, and October are shown for each reanalysis. The differences between the reanalysis and the new data are smaller here as the thermal inertia used for all products were derived from the daily reanalysis data. Overall, there is very good agreement between the reanalysis and the new estimates. The new data of daily ground heat flux is more negative in the northern regions than either reanalysis. Snow should not be a concern in July over these points and what causes the difference is unclear.

The monthly data are presented in Figures 7-42 to 7-47. In general the reanalysis ground heat flux has higher magnitude of minimums and maximums than the new estimates. The opposite was found for the 6-hourly datasets. This discrepancy is due to the use of the daily data to derive the thermal inertia. We believe that the new estimates of ground heat flux are more accurate than the reanalysis data. Differences between the estimated and reanalysis ground heat flux are reduced in the Reanalysis-2. Over all three time resolutions, the new data are better correlated to the Reanalysis-2 data. The seasonal cycle can clearly be seen in the new estimates. The signal is smoother in the new estimates of ground heat flux than in the reanalysis ground heat flux. In the final set of figures, all of the newly derived data are represented. The 6-hourly and daily products are aggregated to monthly data. Three plots are shown in sequence in Figures 7-48 to 7-53 for all months. All datasets are fairly consistent with one another, suggesting Eq. (7-1) is a promising tool in the estimation of ground heat flux using the skin temperature input.

Considerable differences are evident between the 6-hourly reanalysis and the new estimates of ground heat fluxes as shown in Figures 7-30 through 7-37. The difference is defined in all product figures as the estimated ground heat flux minus the reanalysis ground heat flux. Therefore a positive difference indicates estimated ground heat flux exceeds reanalysis ground heat flux, and a negative difference indicates reanalysis ground heat flux exceeds estimated ground heat flux. There is a negative difference over Central Asia and a positive difference over North America at 00:00 GMT. At 06:00 GMT there is a positive difference over Africa and a negative difference over North and South America. At 12:00 GMT there is a positive difference over Africa and a negative difference over North and South America, and at 18:00 GMT there

are no major differences between the two ground heat flux datasets. In general the estimated ground heat flux has greater diurnal variations. The positive and negative differences in ground heat flux propagate across the heat flux map as the day progresses and generally correspond to positive and negative peaks in the estimated ground heat flux.

The differences between the daily estimated and reanalysis ground heat flux are unique to the season being studied, but generally the same across reanalysis in Figures 7-38 through 7-41. In January there is a slight negative difference over the northern regions, and in July there is a slight positive difference over the northern regions. The reanalysis ground heat flux in northern regions is too high relative to the low estimates. In all four seasons there is a positive difference over Central Asia. This area was observed in Figure 7-24 where it is shown that the reanalysis ground heat flux is lower than the estimates at all times. Over a long period averaged ground heat flux should close, while the reanalysis ground heat flux does not have that feature for this region. Thus it appears that the reanalysis ground heat flux is underestimated.

There are noticeable positive differences between the monthly reanalysis and the new estimates of ground heat flux throughout the year over Central Asia. Again the reanalysis ground heat flux appears to be too small particularly during summer for this area. Differences between the estimated and reanalysis ground heat flux have similar features in the two reanalyses in Figure 7-42 through 7-47. During the summer months there is a significant positive difference between the reanalysis and estimated ground heat fluxes in both the northern and southern regions of the globe. The reanalysis ground heat flux appears to be unrealistically small over the summer in these areas.

The purpose of this investigation is to offer an alternative dataset of ground heat flux. This new product is developed with the help of the reanalysis modeled ground heat flux, land cover data, and the reliable skin temperature input from the reanalysis data. The key parameter is the thermal inertia of the land. The thermal inertia is time invariant by theory. However, using a regression to obtain this parameter over different time scales of the reanalysis, we find different values of the parameter. These problems are less severe in the Reanalysis-2. This research suggests that the discrepancies are probably due to flaws in the reanalysis model used to produce the 6-hourly and monthly ground heat flux.

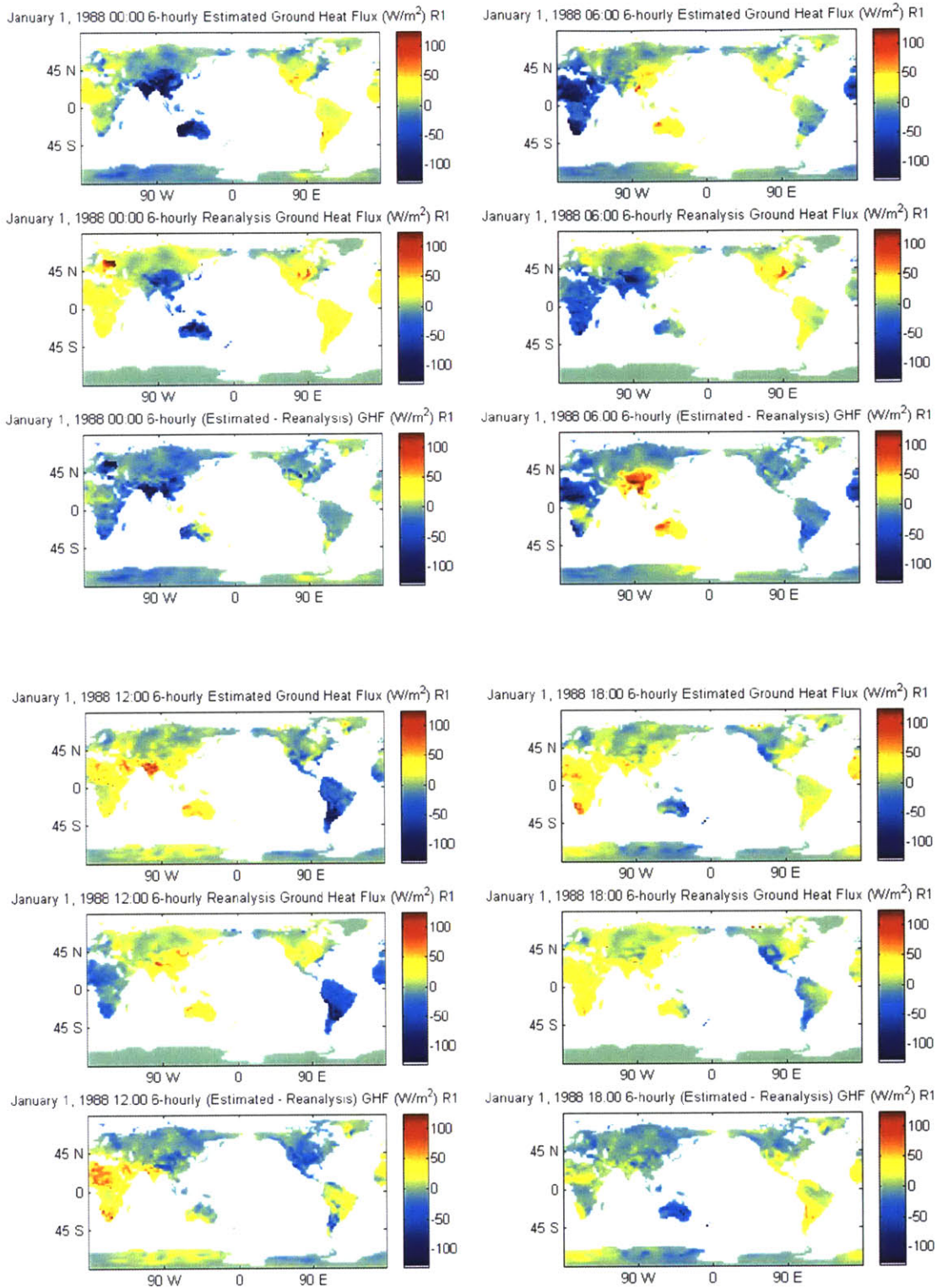


Figure 7-30: GHF Products Plots for 6-hourly time scale using the Reanalysis-1 – January 1, 1988 (winter)



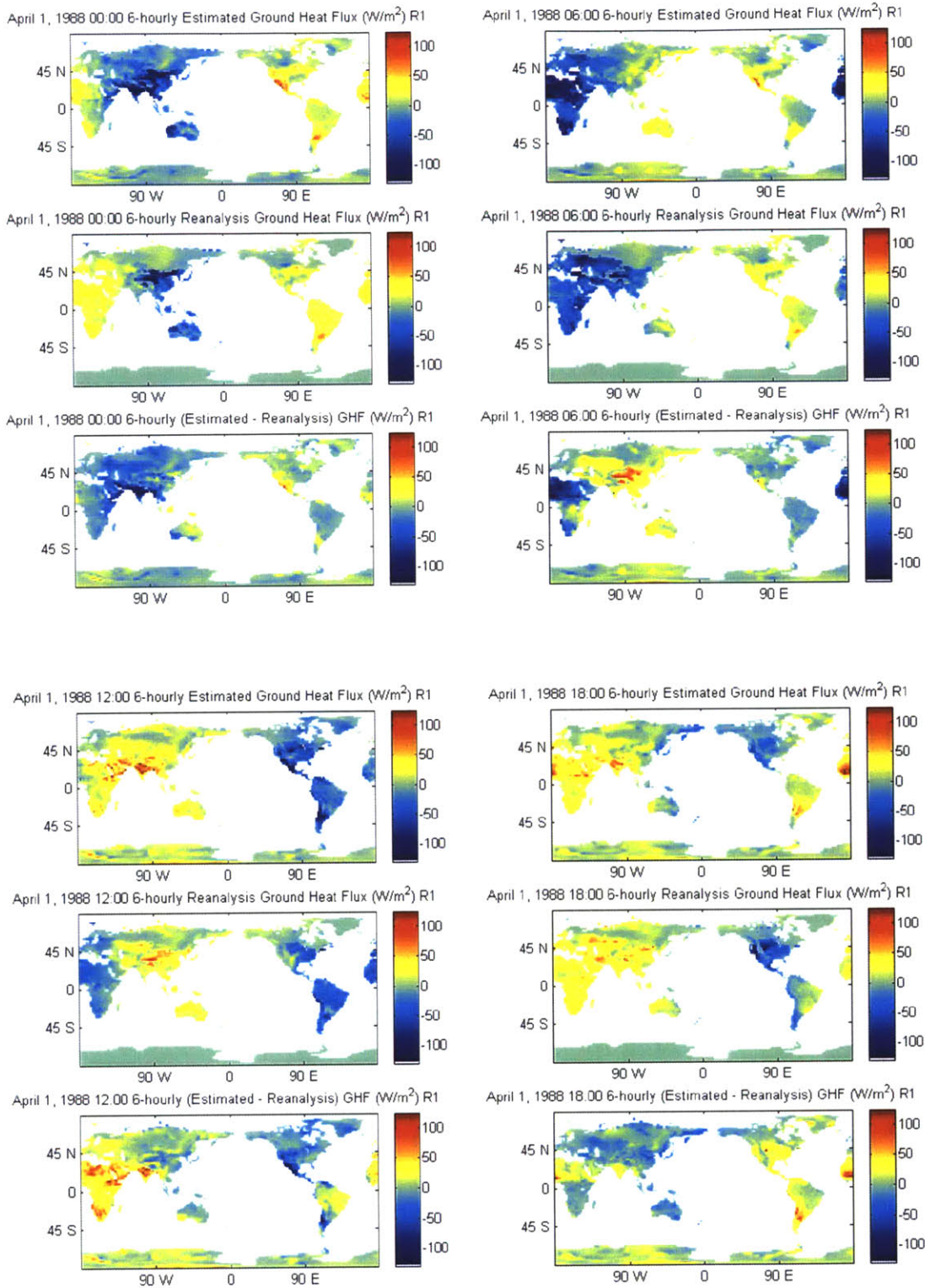


Figure 7-31: GHF Products Plots for 6-hourly time scale using the Reanalysis-1 – April 1, 1988 (spring)

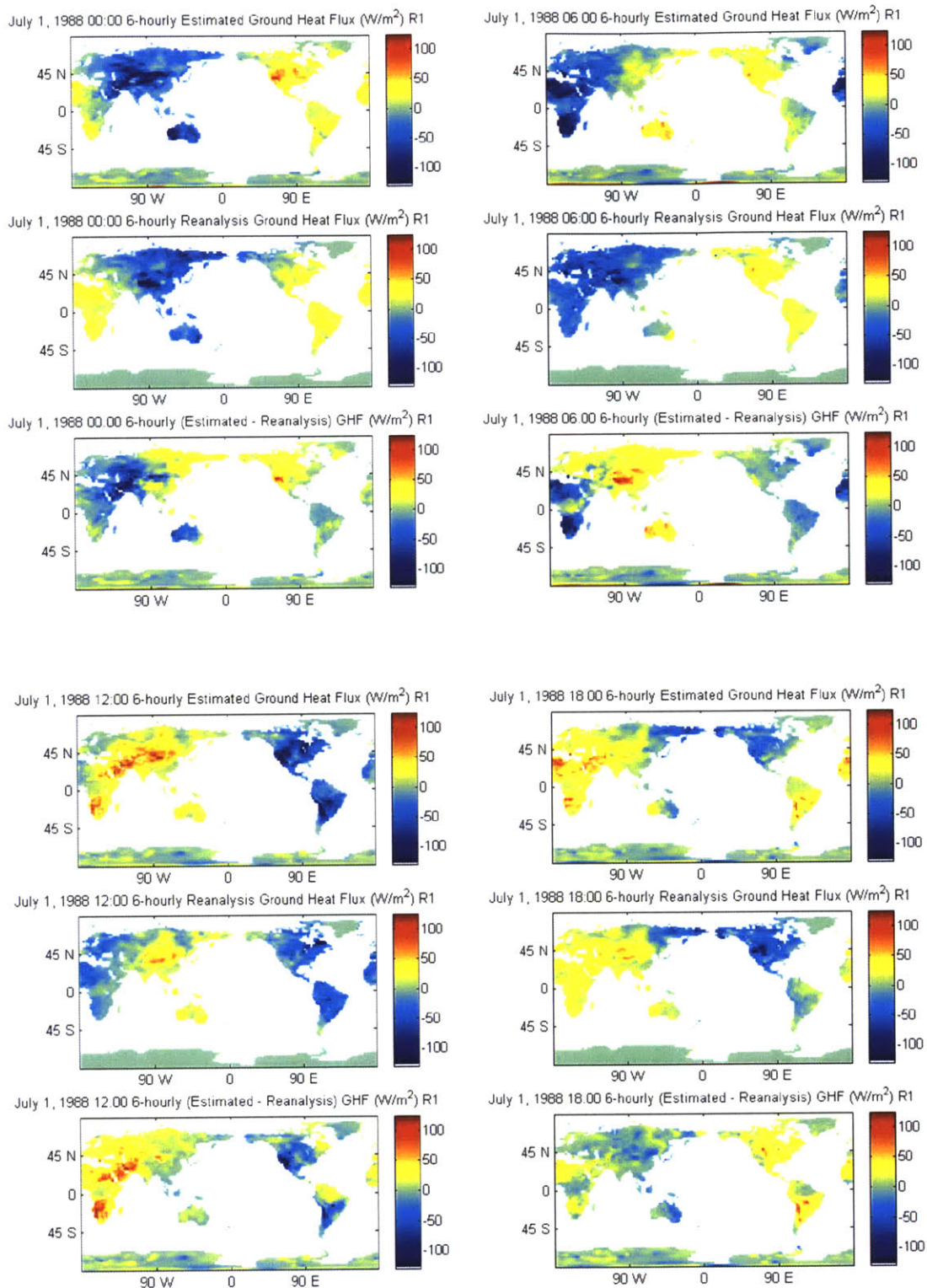


Figure 7-32: GHF Products Plots for 6-hourly time scale using the Reanalysis-1 – July 1, 1988 (summer)

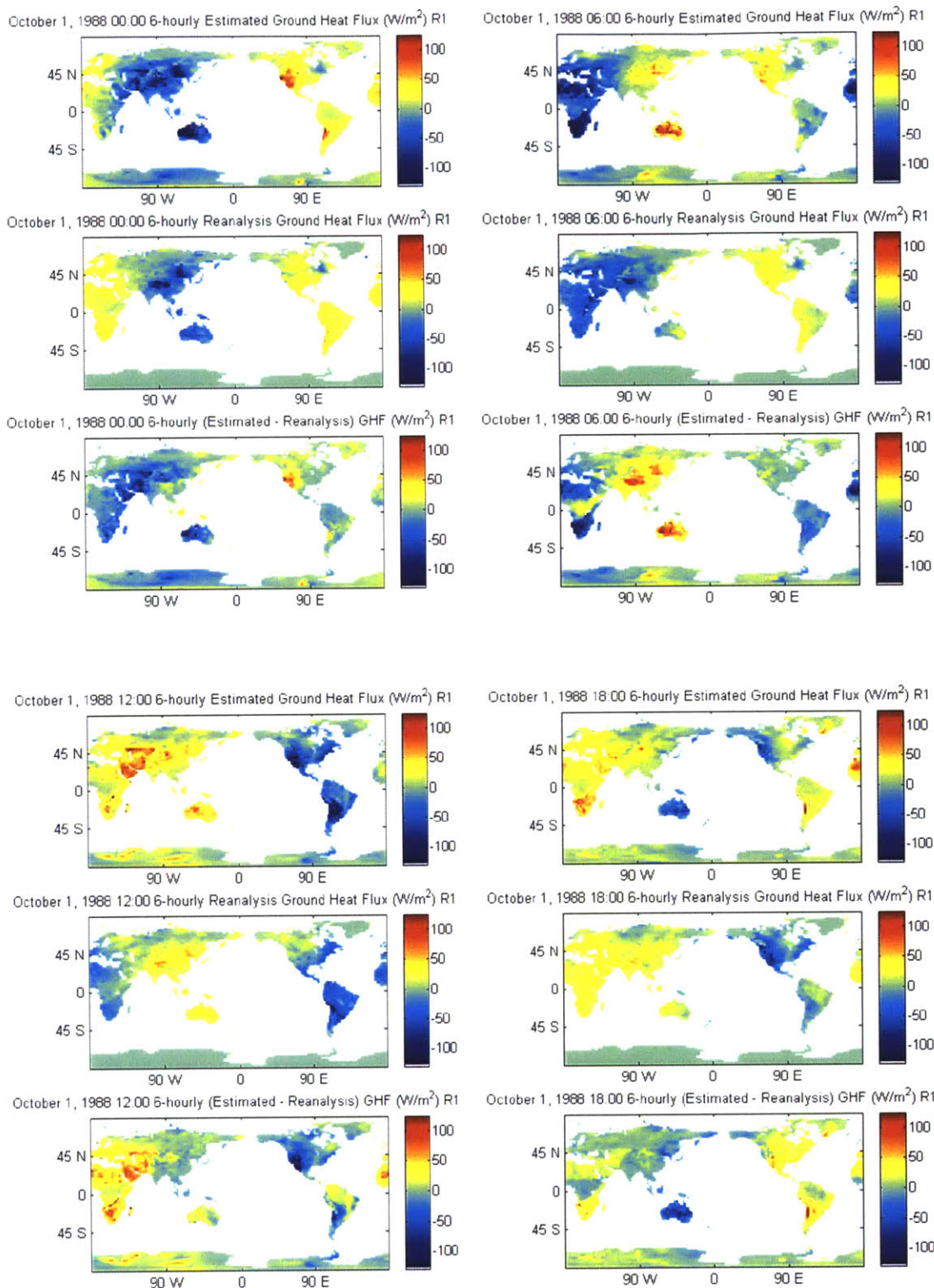


Figure 7-33: GHF Products Plots for 6-hourly time scale using the Reanalysis-1 – October 1, 1988 (fall)

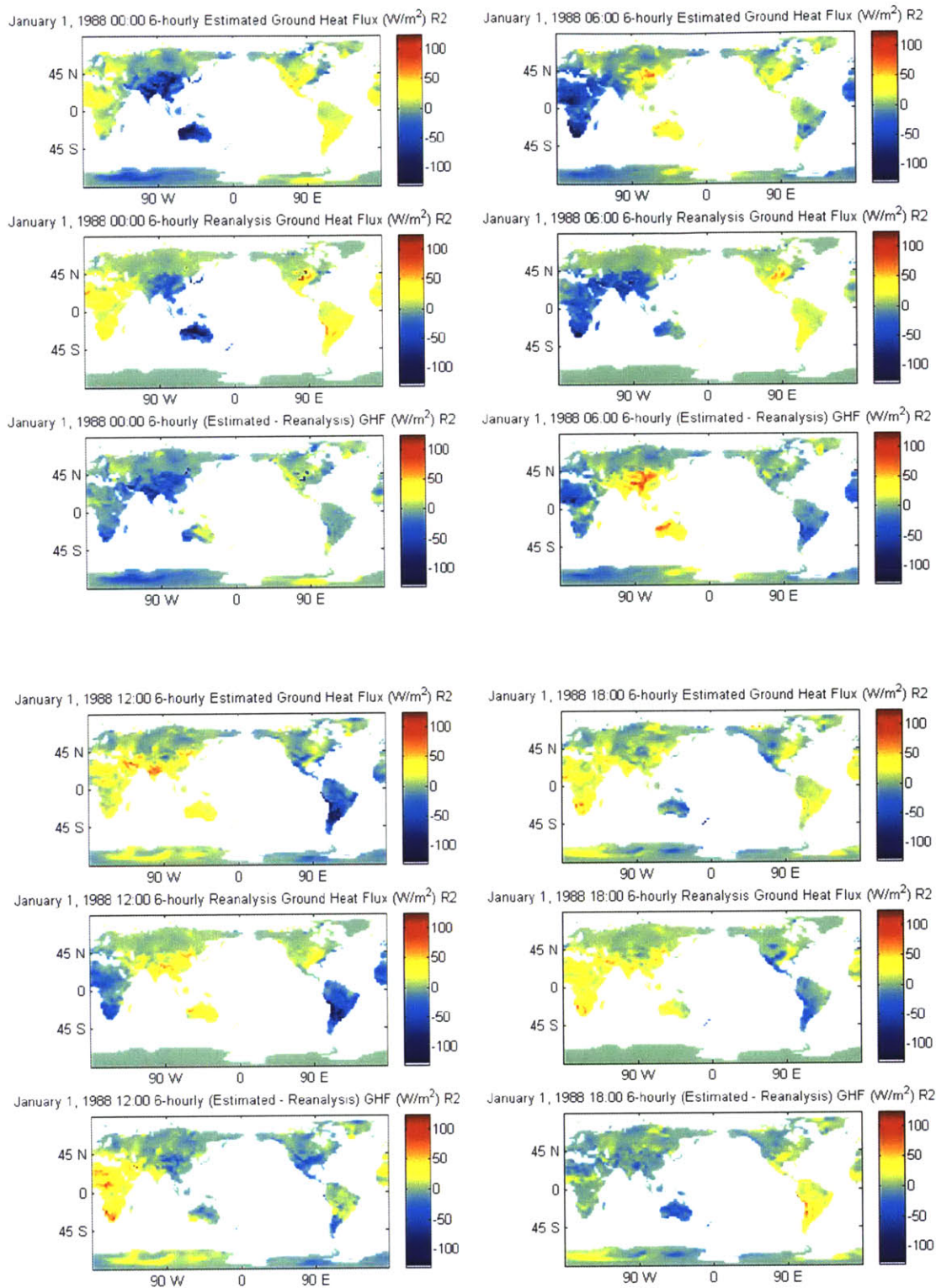


Figure 7-34: GHF Products Plots for 6-hourly time scale using the Reanalysis-2 – January 1, 1988 (winter)

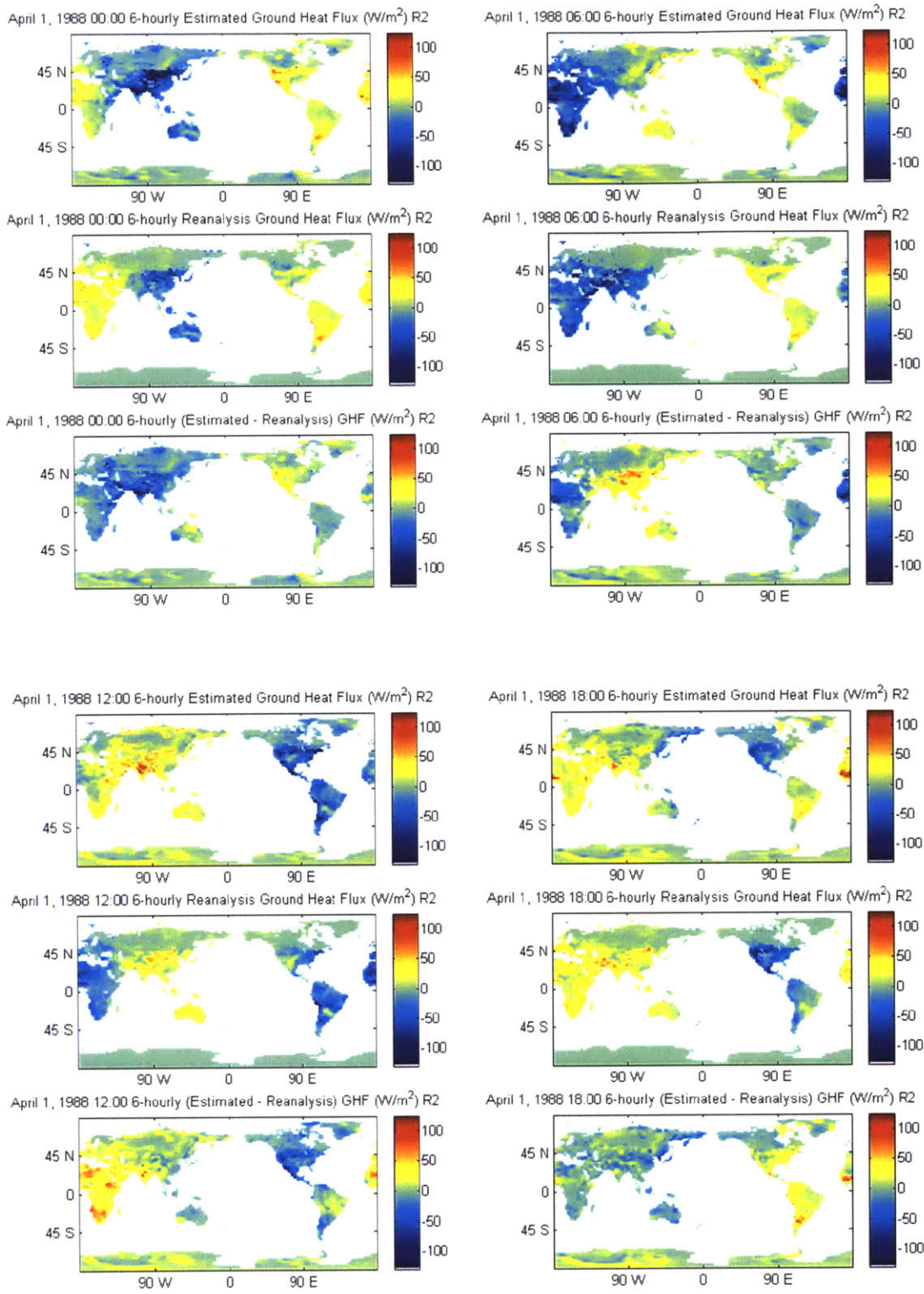


Figure 7-35: GHF Products Plots for 6-hourly time scale using the Reanalysis-2 – April 1, 1988 (spring)

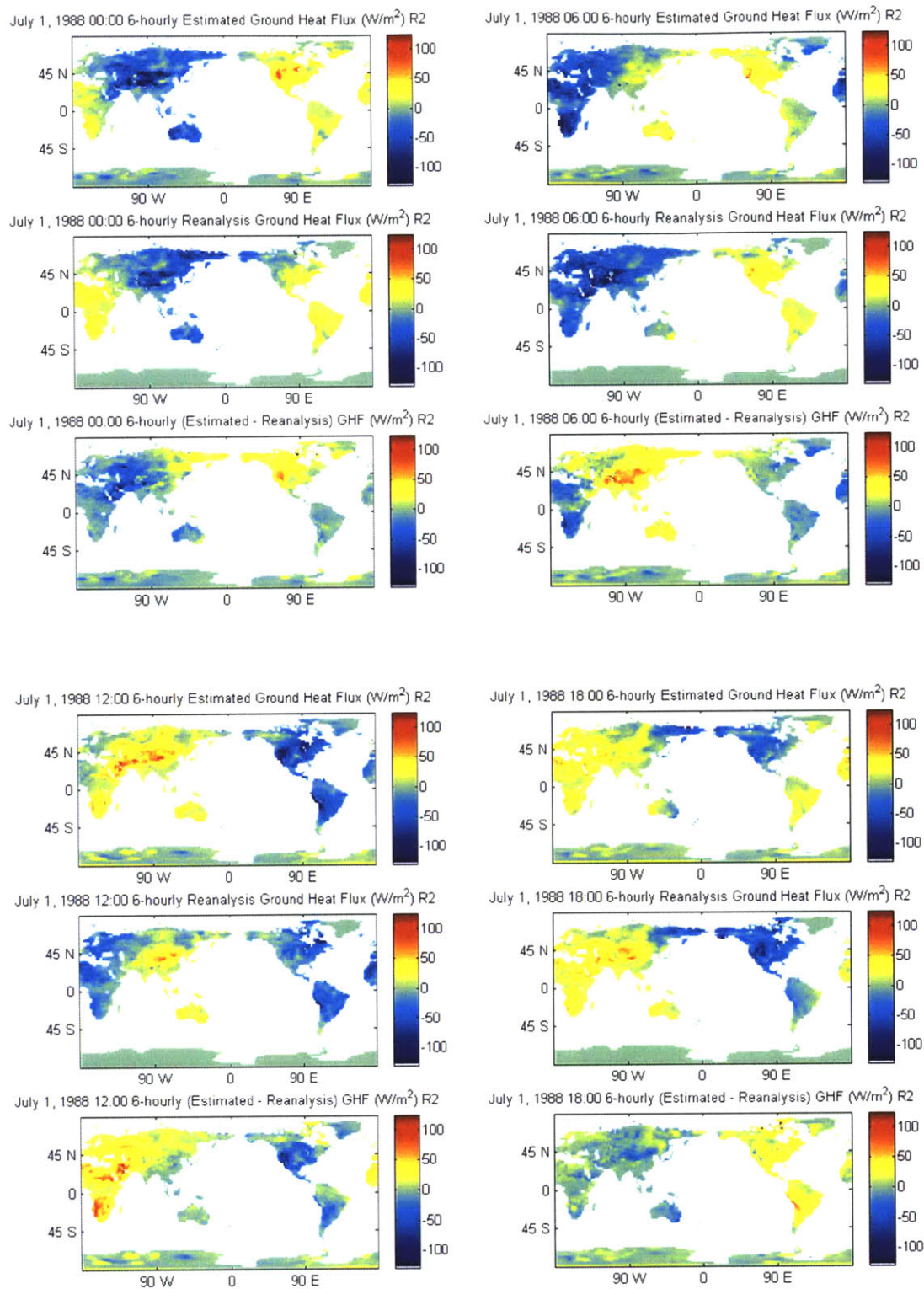


Figure 7-36: GHF Products Plots for 6-hourly time scale using the Reanalysis-2 – July 1, 1988 (summer)

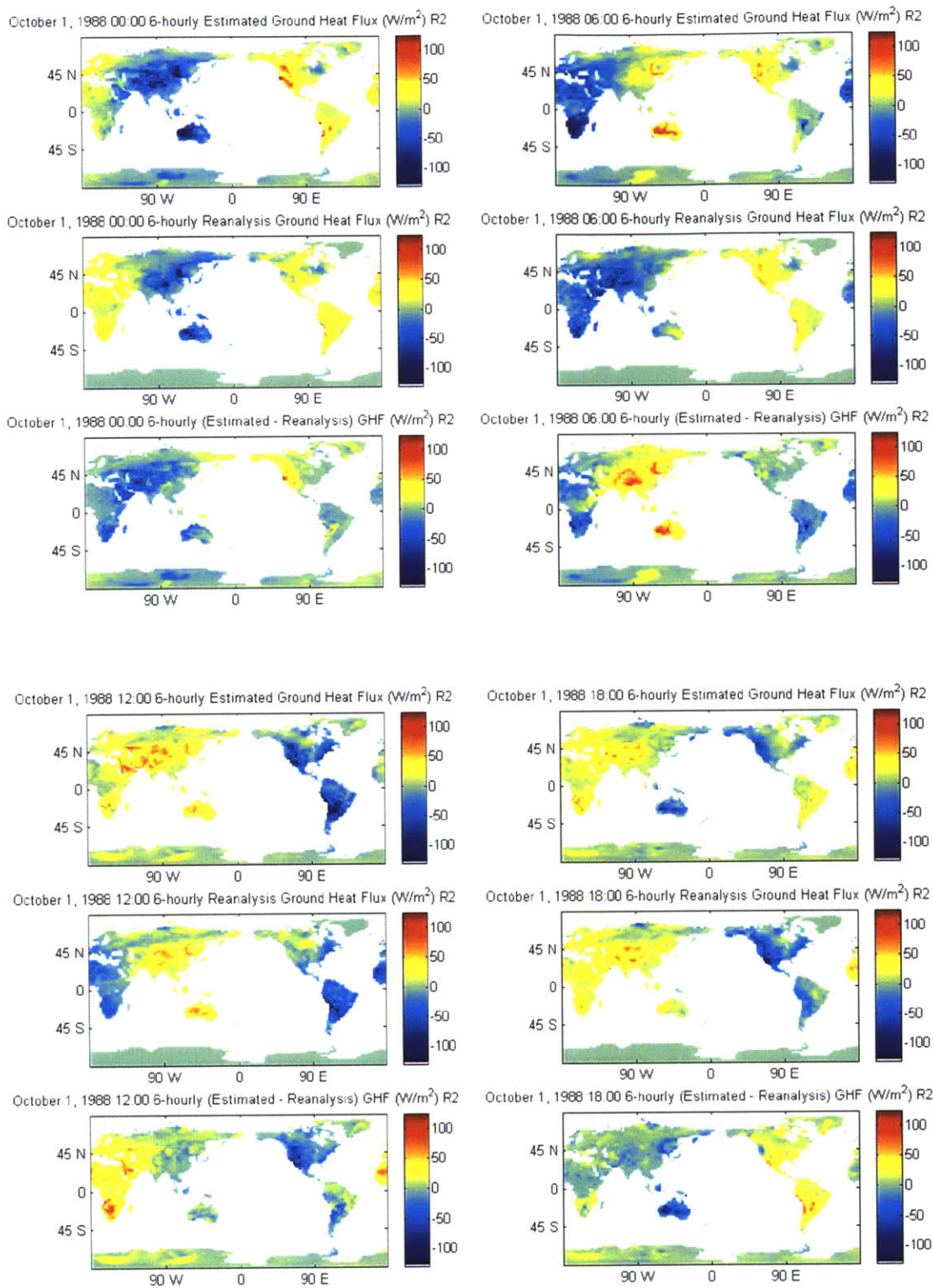


Figure 7-37: GHF Products Plots for 6-hourly time scale using the Reanalysis-1 – October 1, 1988 (fall)

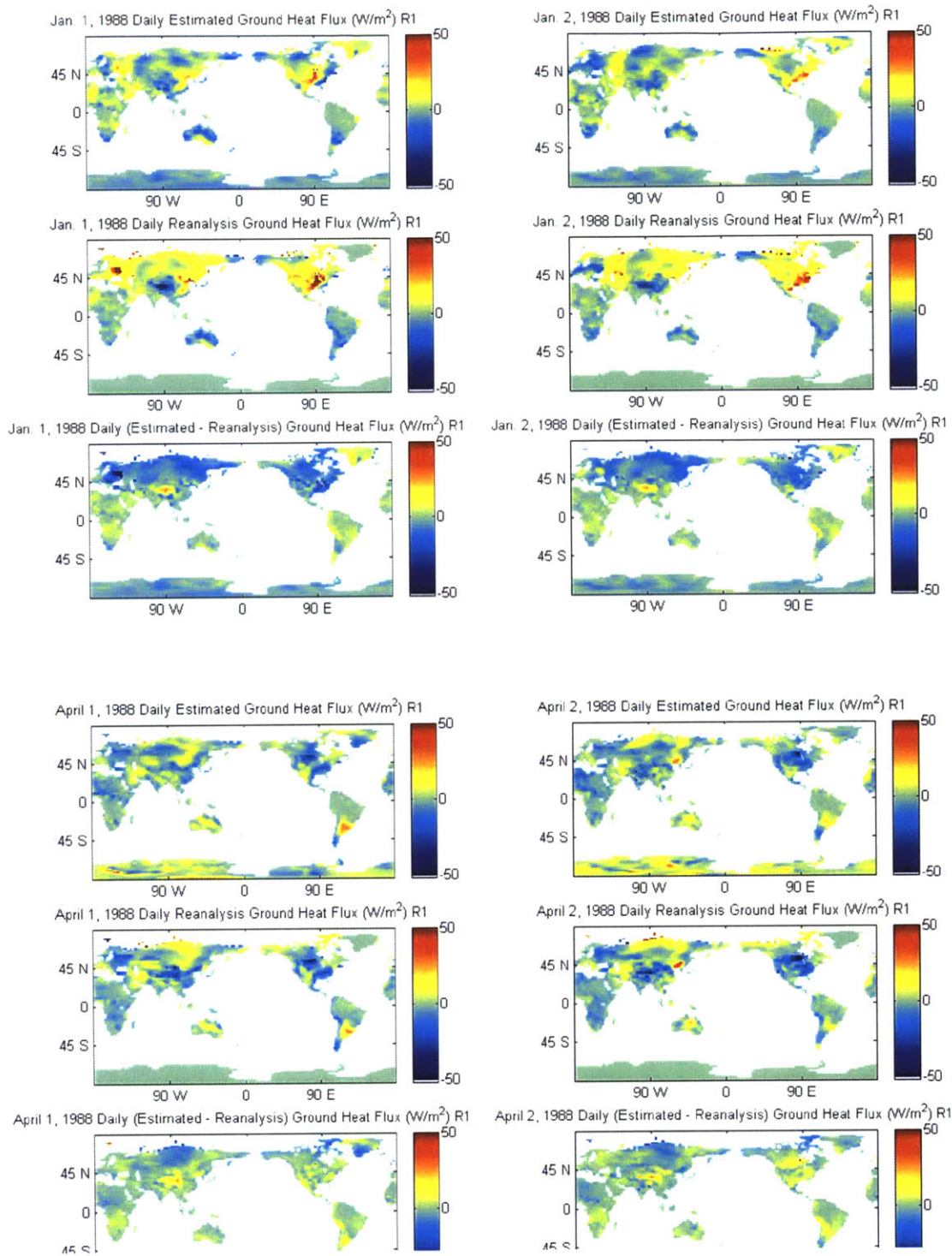


Figure 7-38: GHF Products Plots for Daily time scale using the Reanalysis-1 – January/April 1988 (winter/spring)



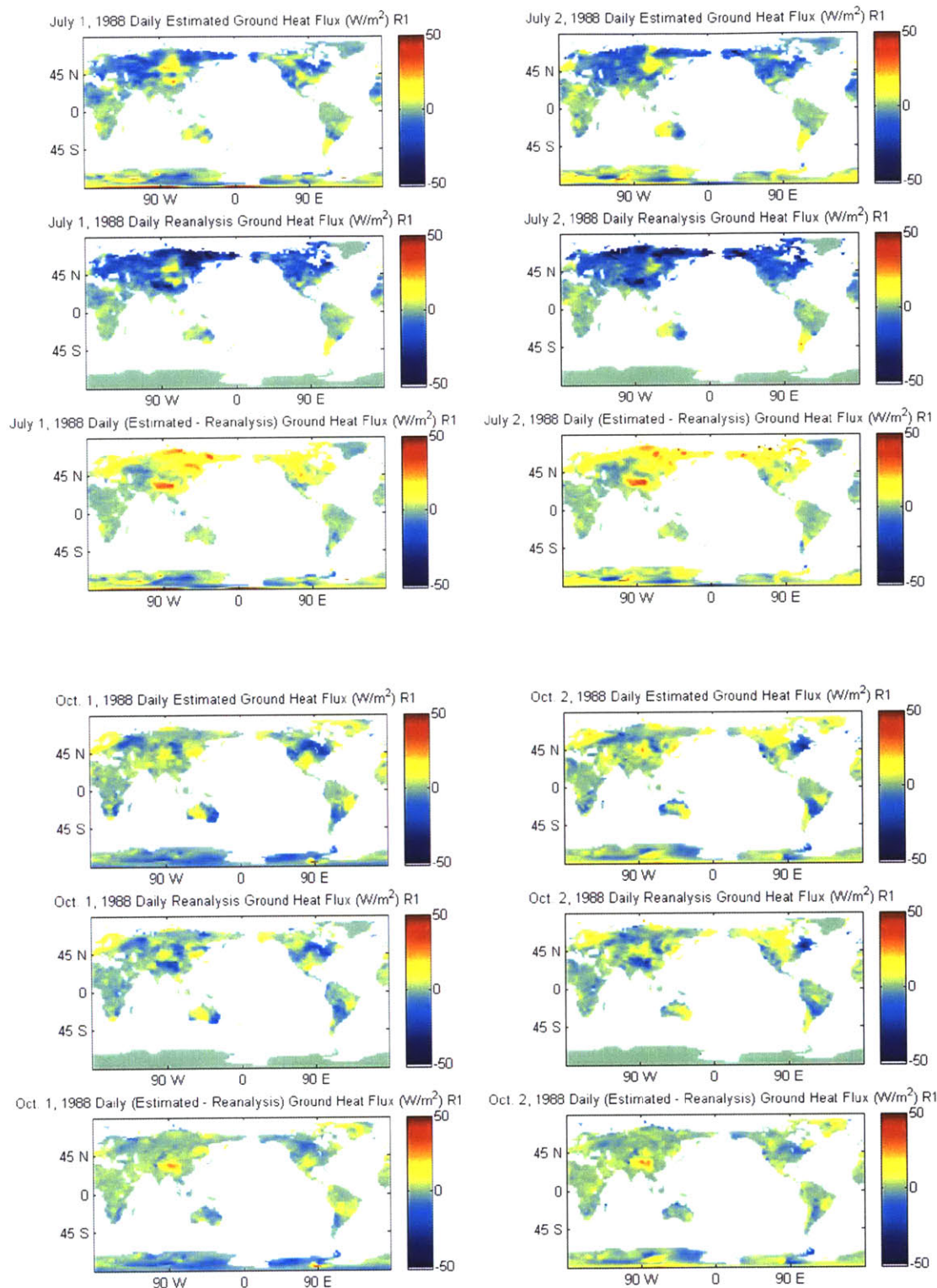


Figure 7-39: GHF Products Plots for Daily time scale using the Reanalysis-1 – July/October 1988 (summer/fall)

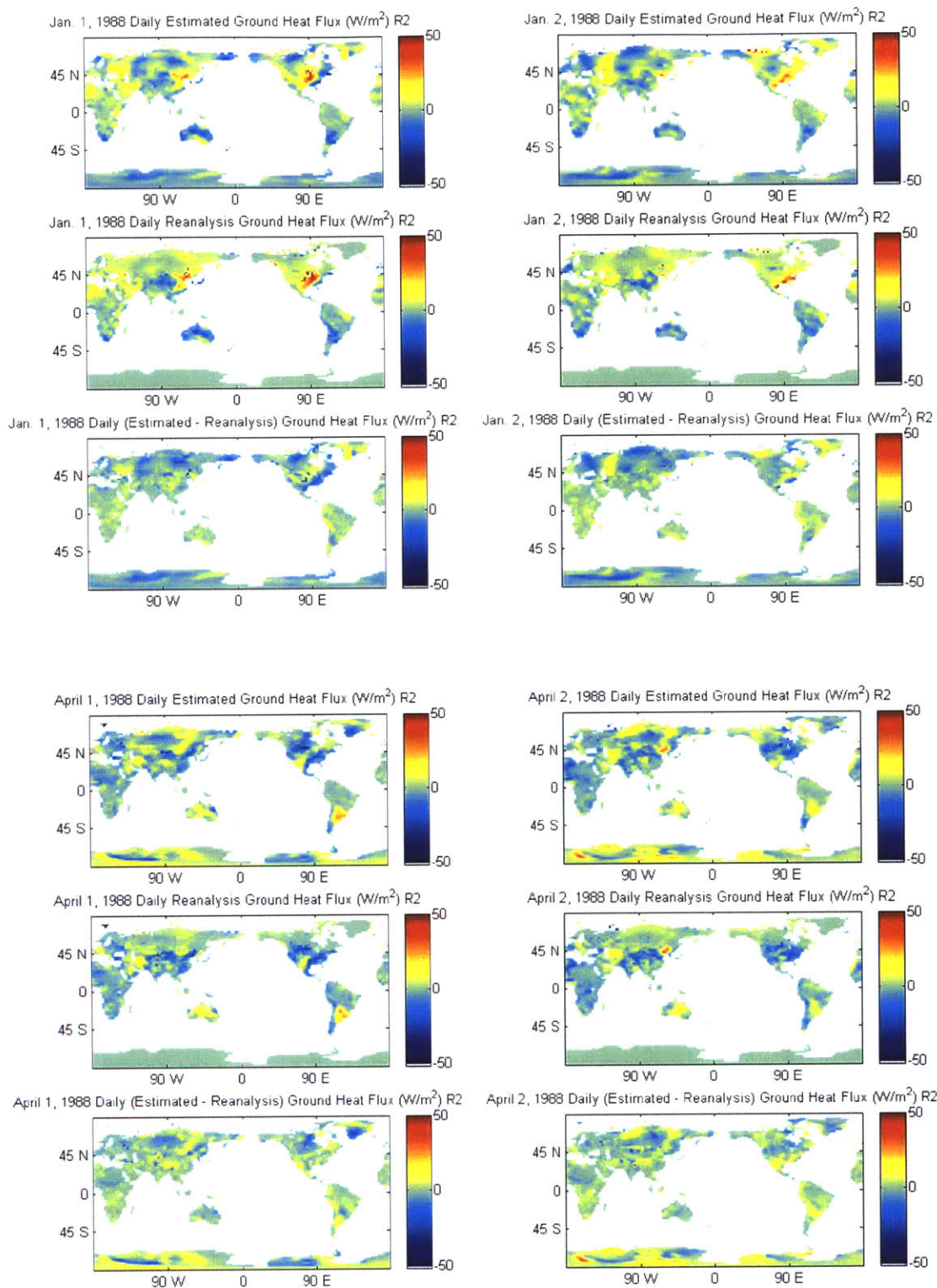


Figure 7-40: GHF Products Plots for Daily time scale using the Reanalysis-2 – January/April 1988 (winter/spring)

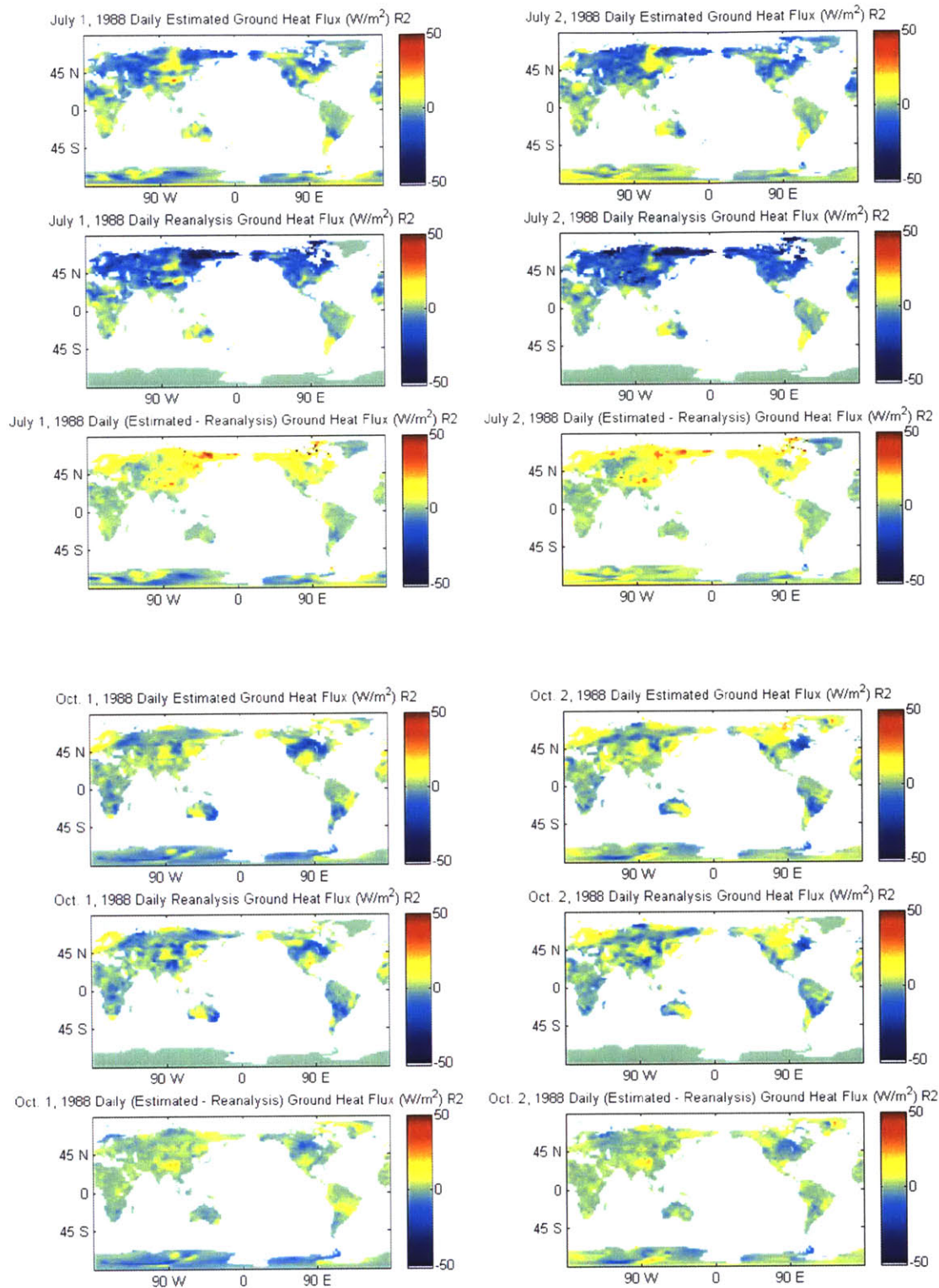


Figure 7-41: GHF Products Plots for Daily time scale using the Reanalysis-2 – July/October 1988 (summer/fall)

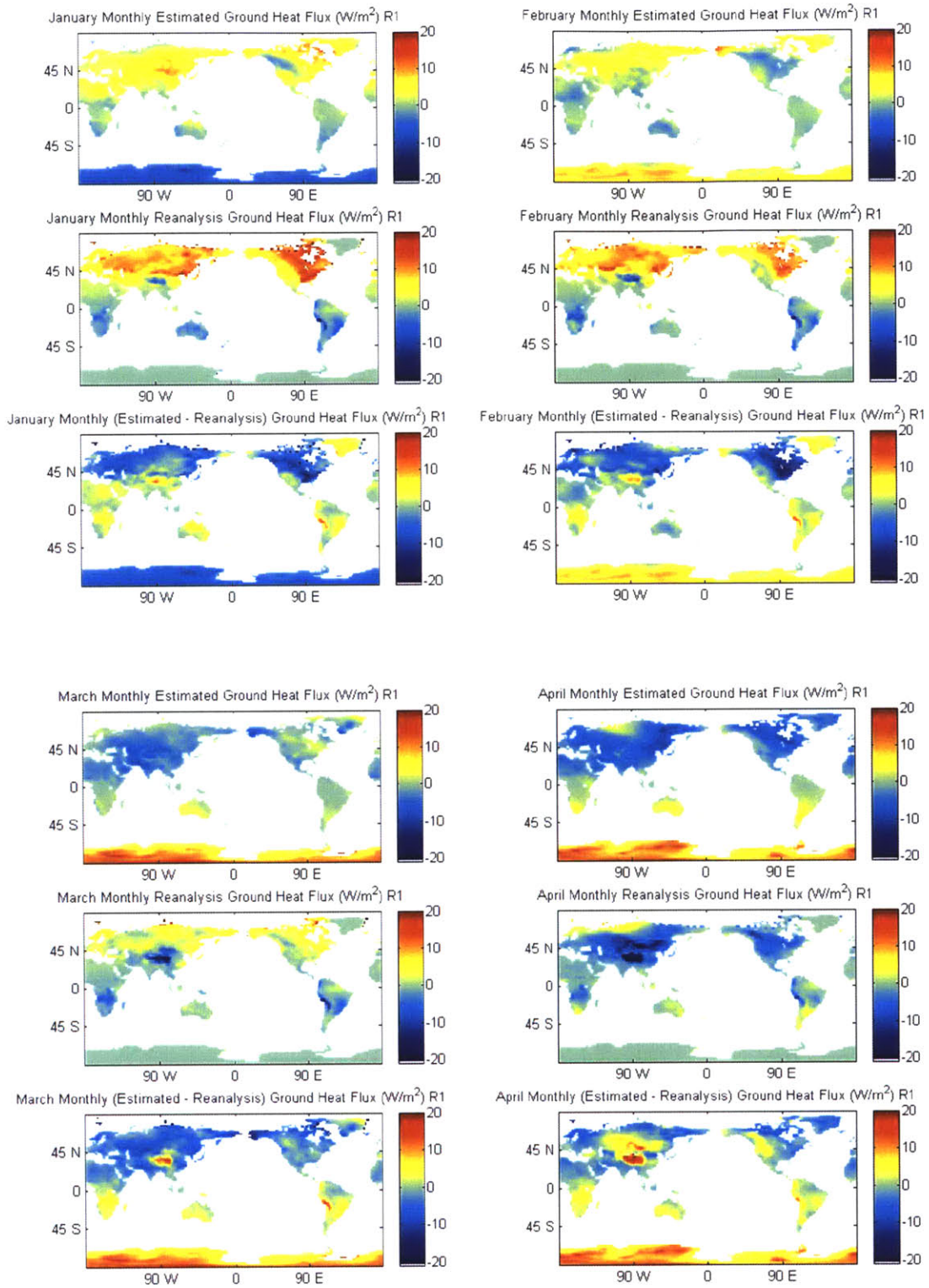


Figure 7-42: GHF Products Plots for Monthly time scale using the Reanalysis-1 – January, February, March, April 1988

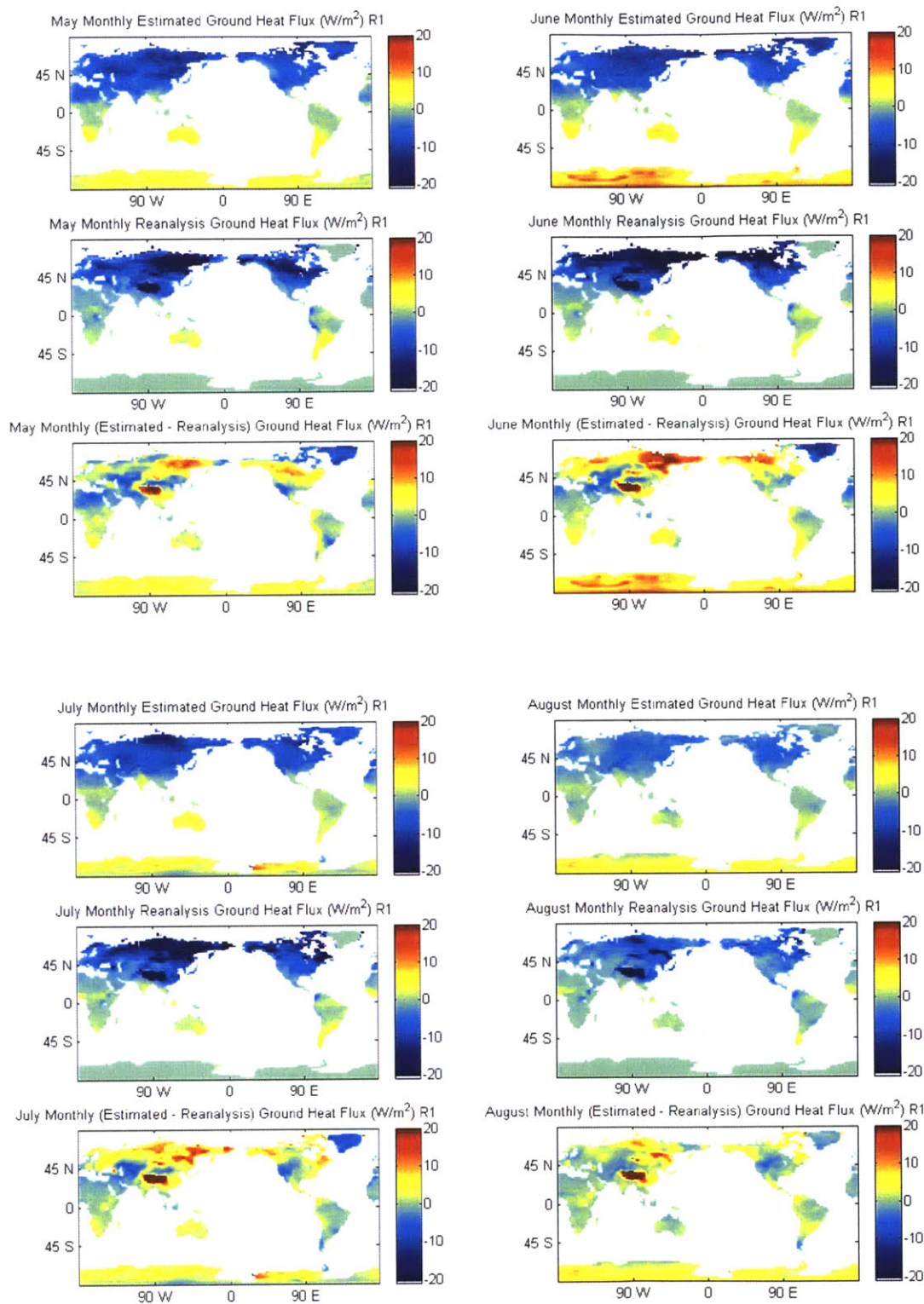


Figure 7-43: GHF Products Plots for Monthly time scale using the Reanalysis-1 – May, June, July, August 1988

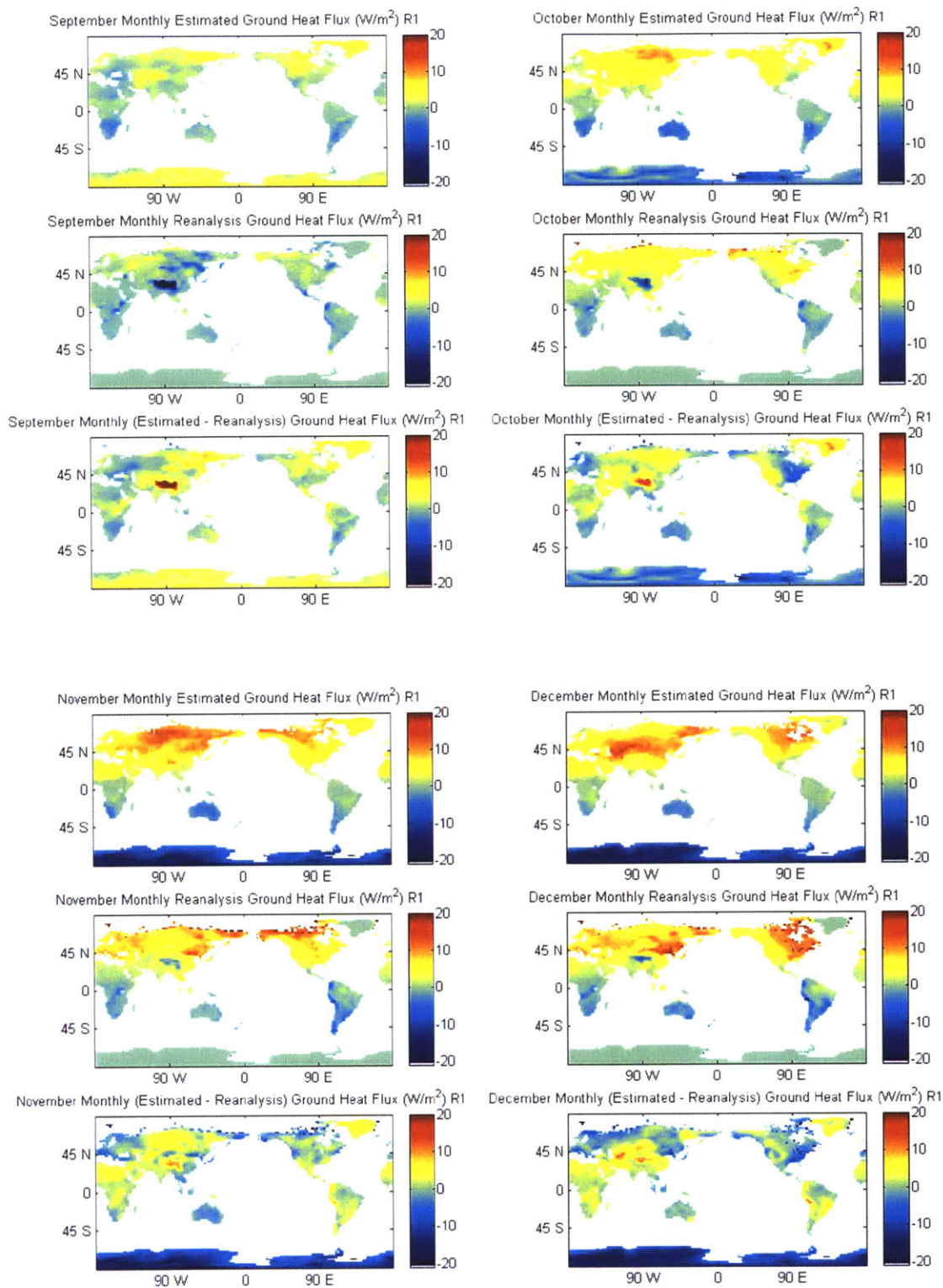


Figure 7-44: GHF Products Plots for Monthly time scale using the Reanalysis-1 – September, October, November, December 1988

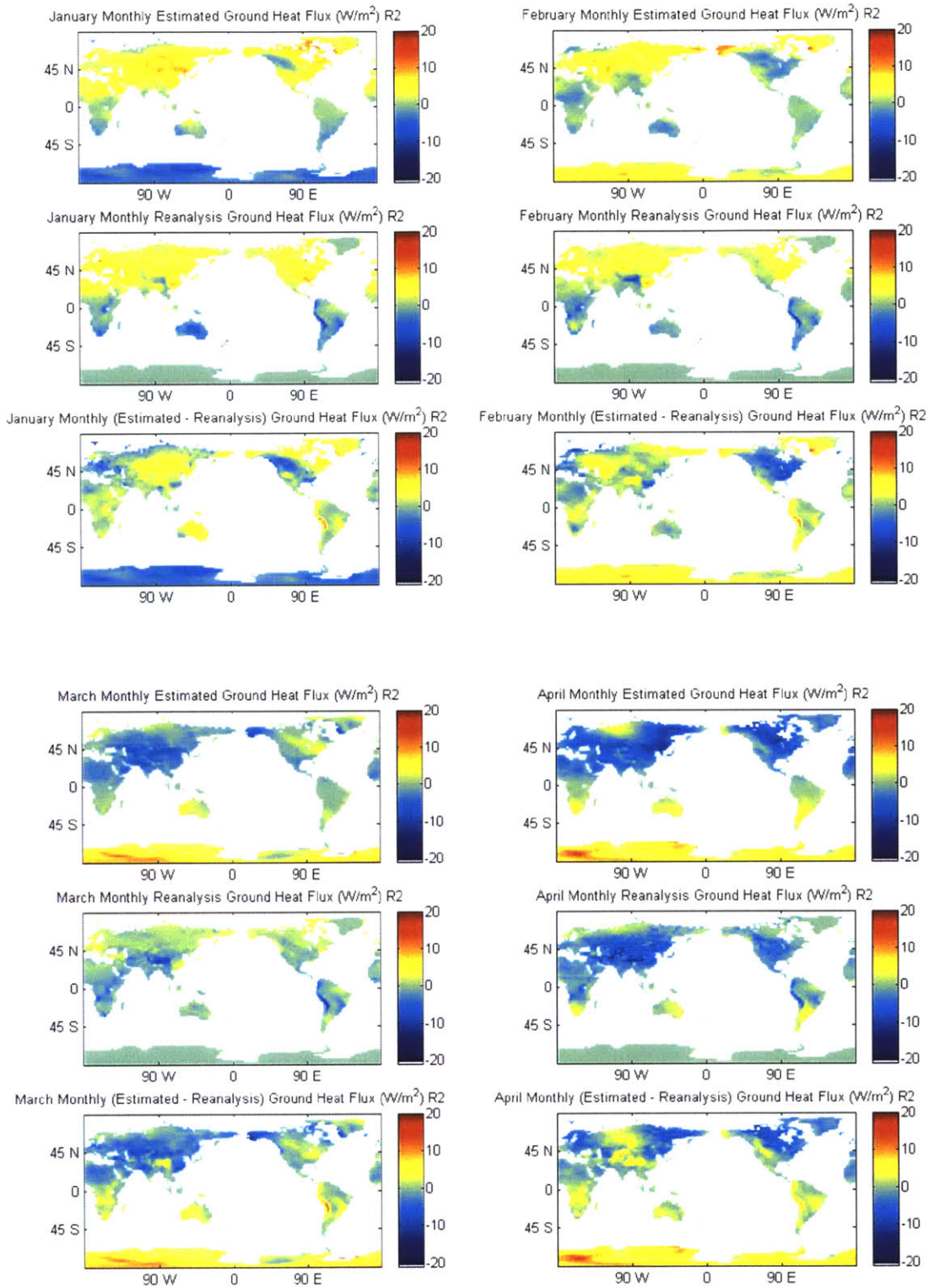


Figure 7-45: GHF Products Plots for Monthly time scale using the Reanalysis-2 – January, February, March, April 1988

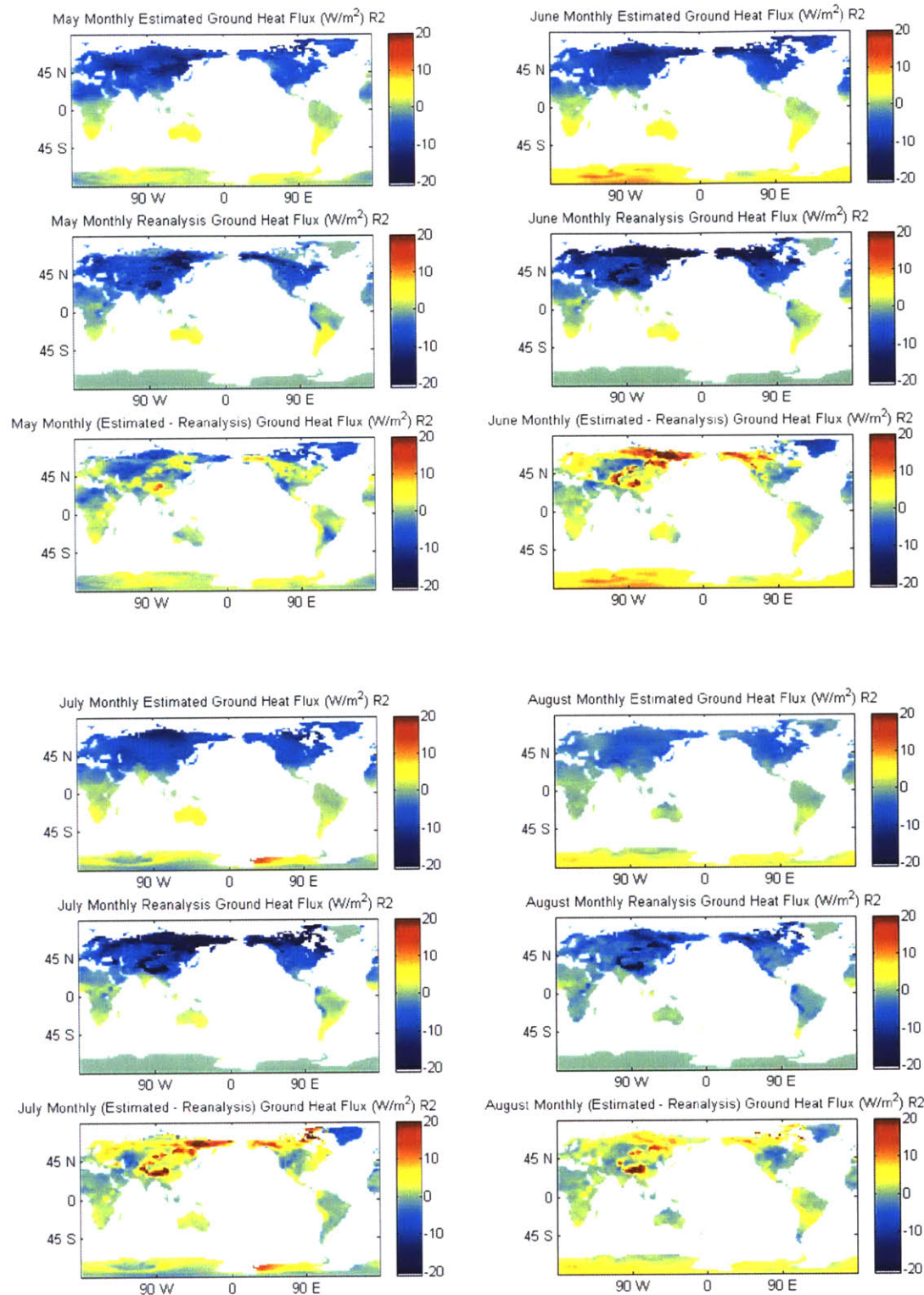


Figure 7-46: GHF Products Plots for Monthly time scale using the Reanalysis-2 – May, June, July, August 1988



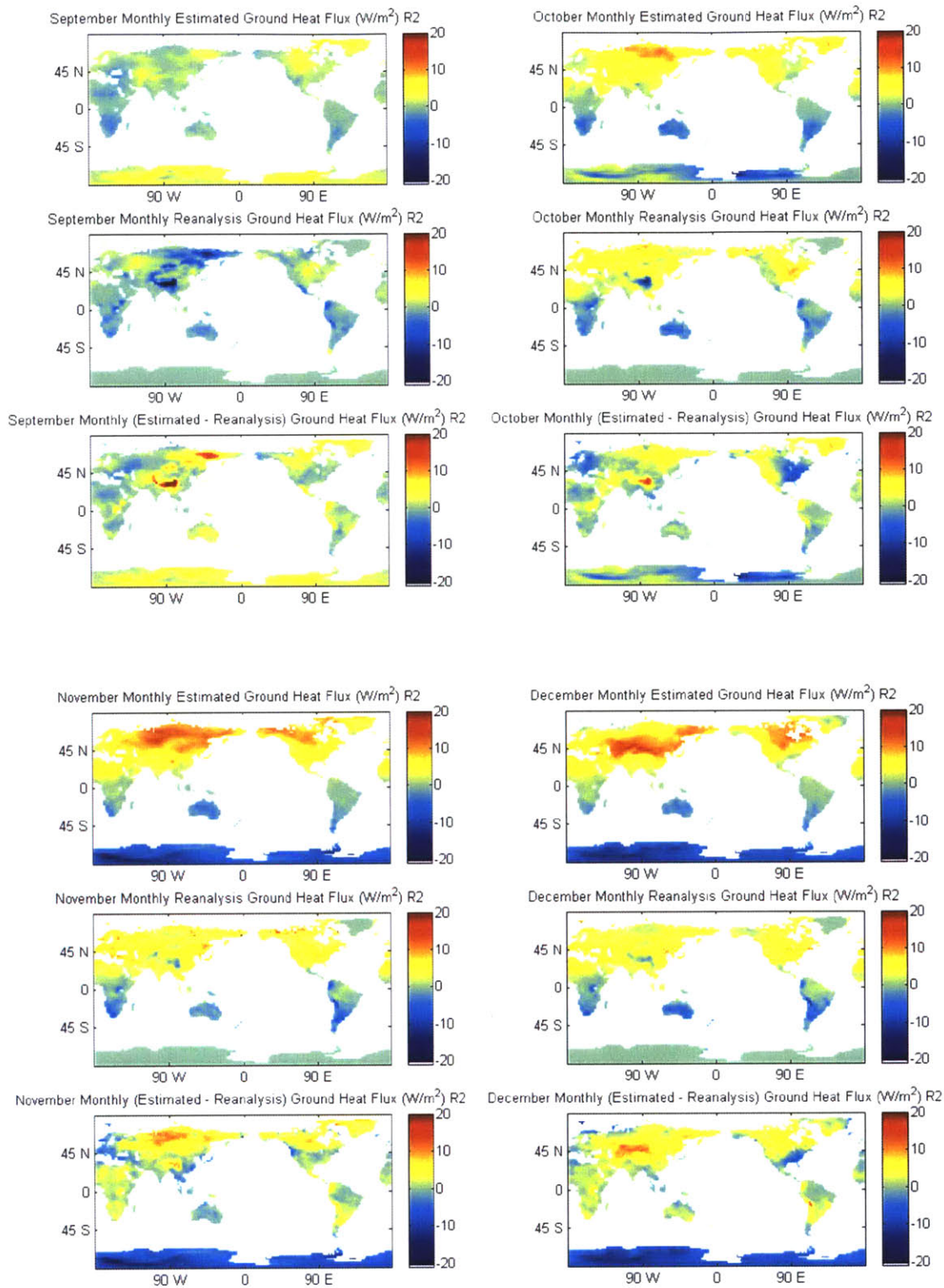


Figure 7-47: GHF Products Plots for Monthly time scale using the Reanalysis-2 – September, October, November, December 1988

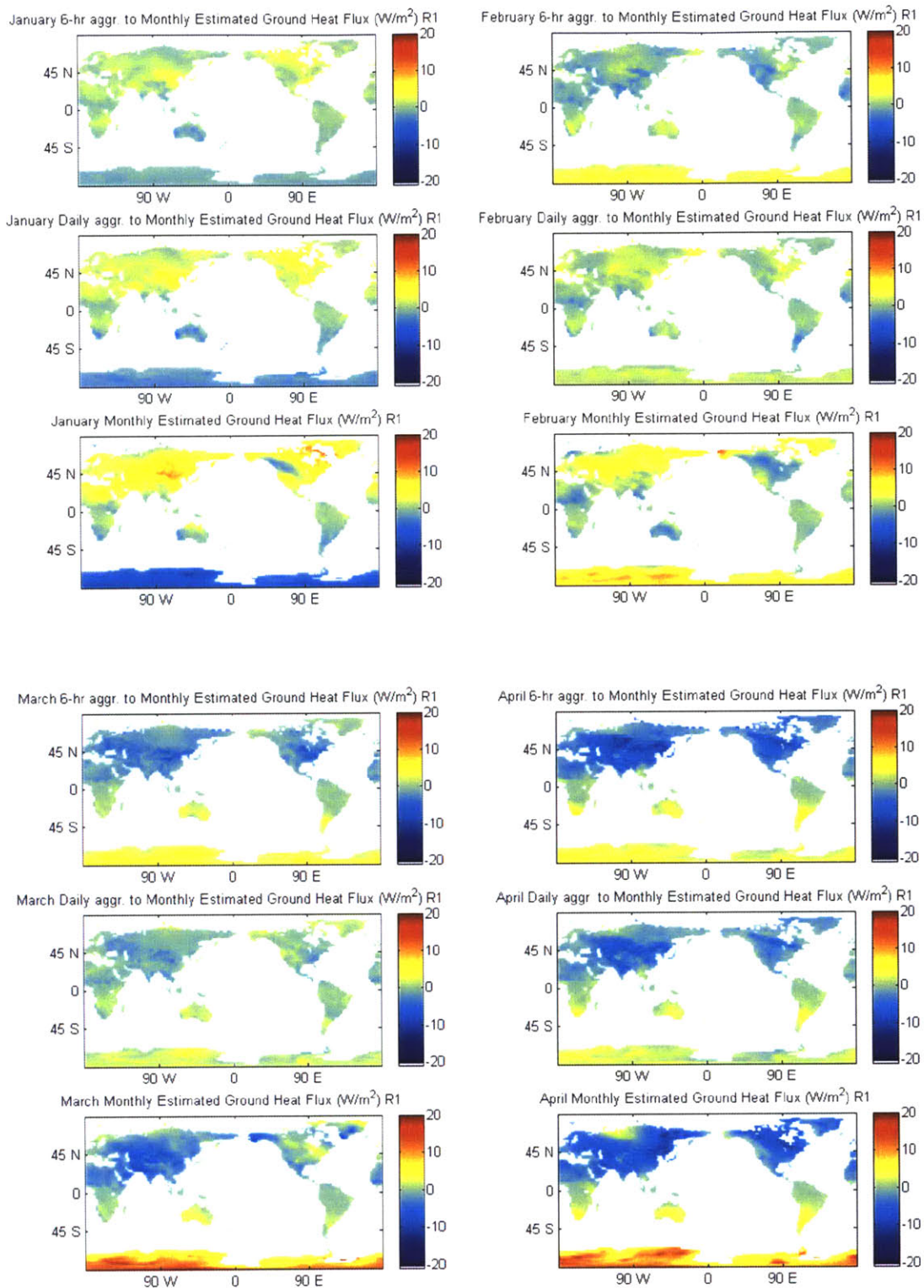


Figure 7-48: GHF Products Plots Aggregated to the Monthly time scale using the Reanalysis-1 – January, February, March, and April 1988

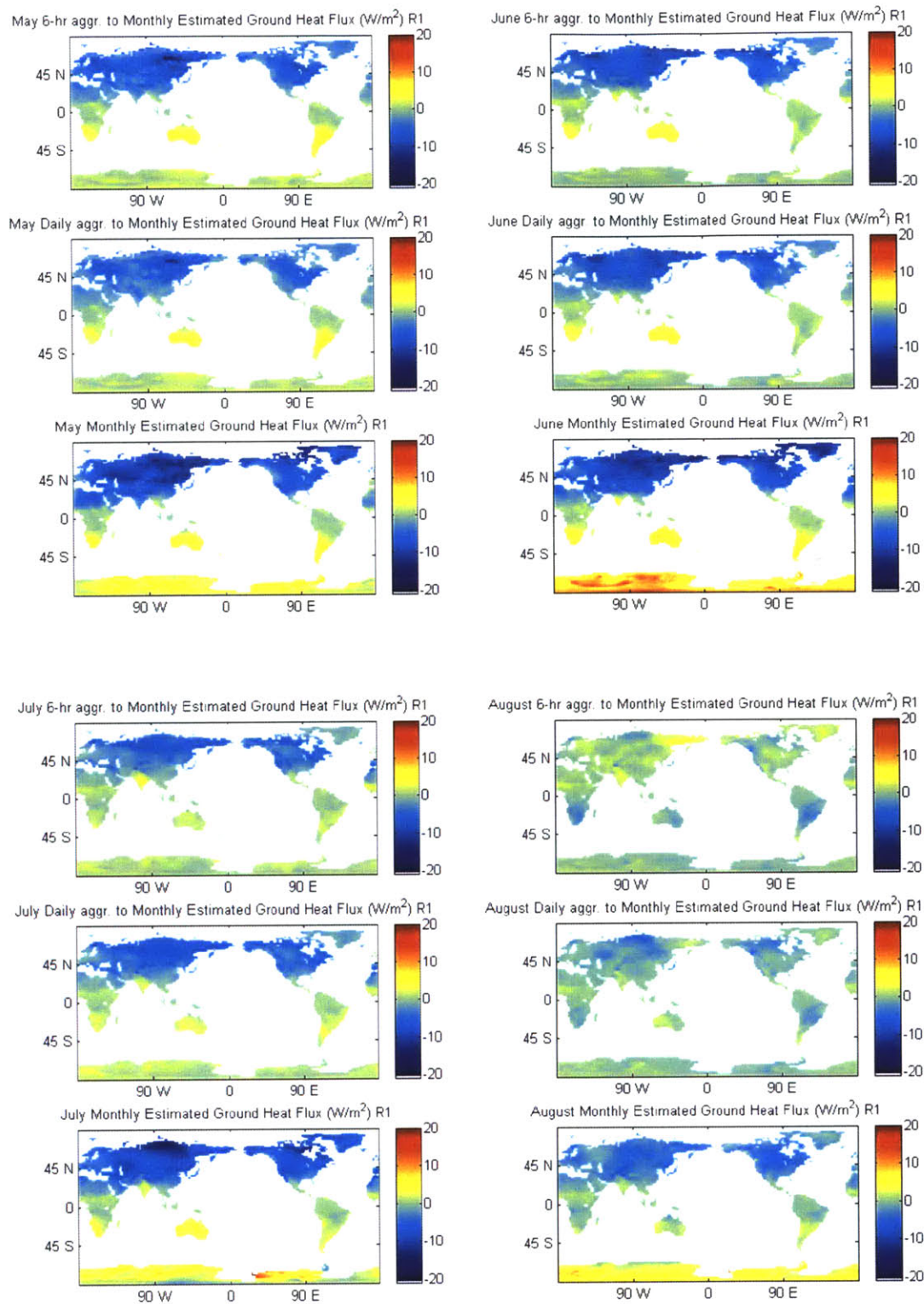


Figure 7-49: GHF Products Plots Aggregated to the Monthly time scale using the Reanalysis-1 – May, June, July, and August 1988

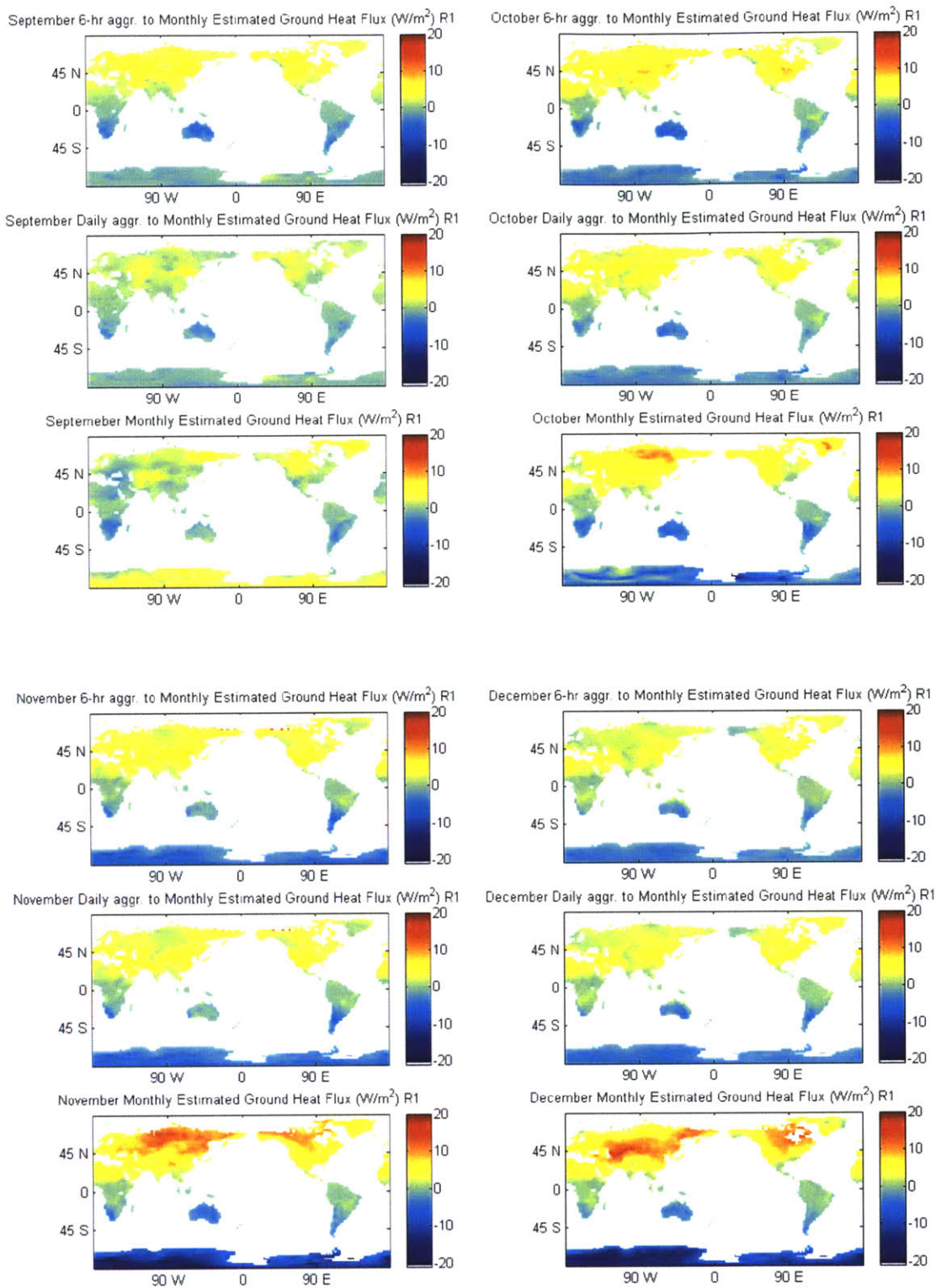


Figure 7-50: GHF Products Plots Aggregated to the Monthly time scale using the Reanalysis-1 – September, October, November, and December 1988

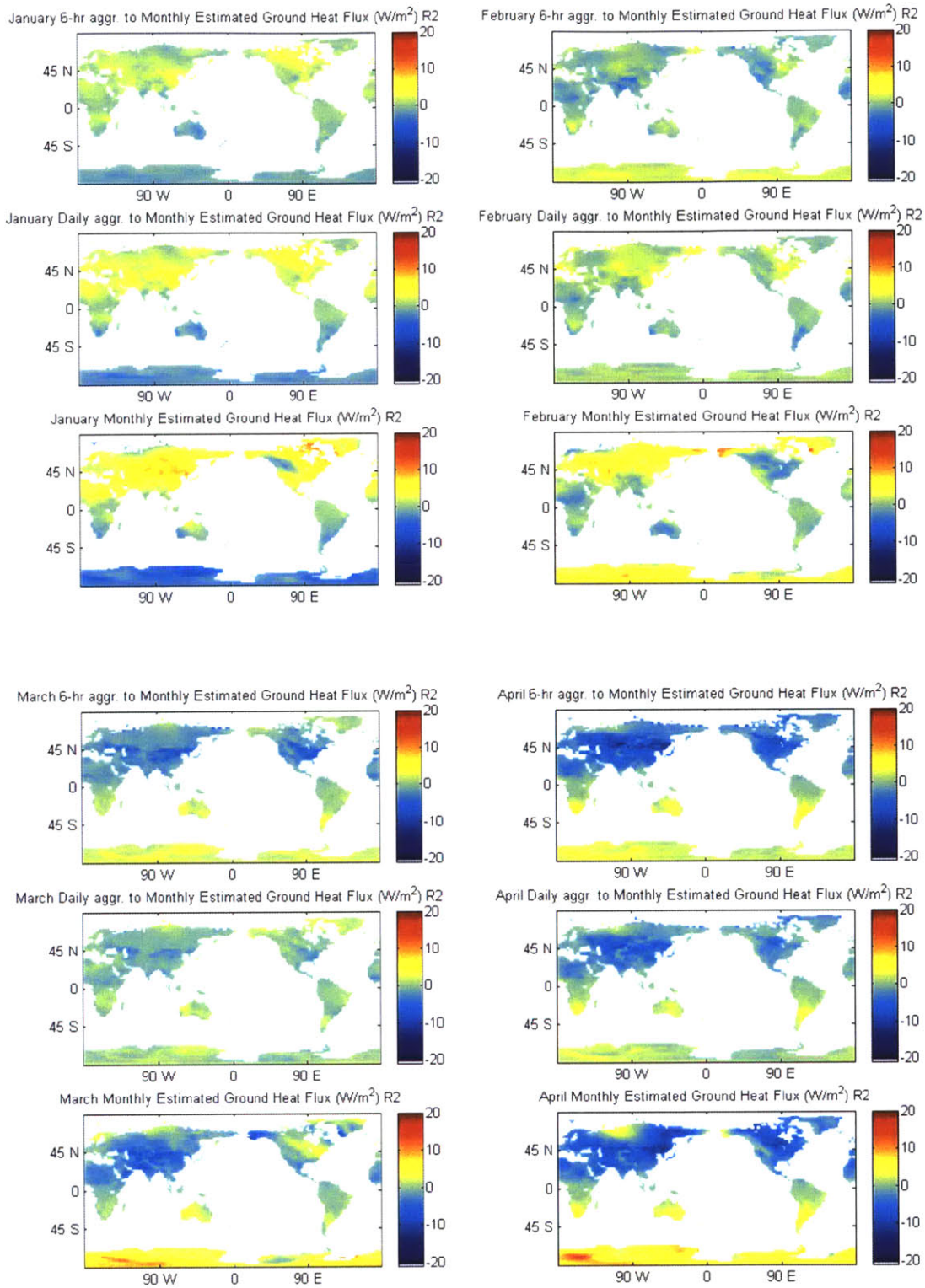


Figure 7-51: GHF Products Plots Aggregated to the Monthly time scale using the Reanalysis-2 – January, February, March, and April 1988

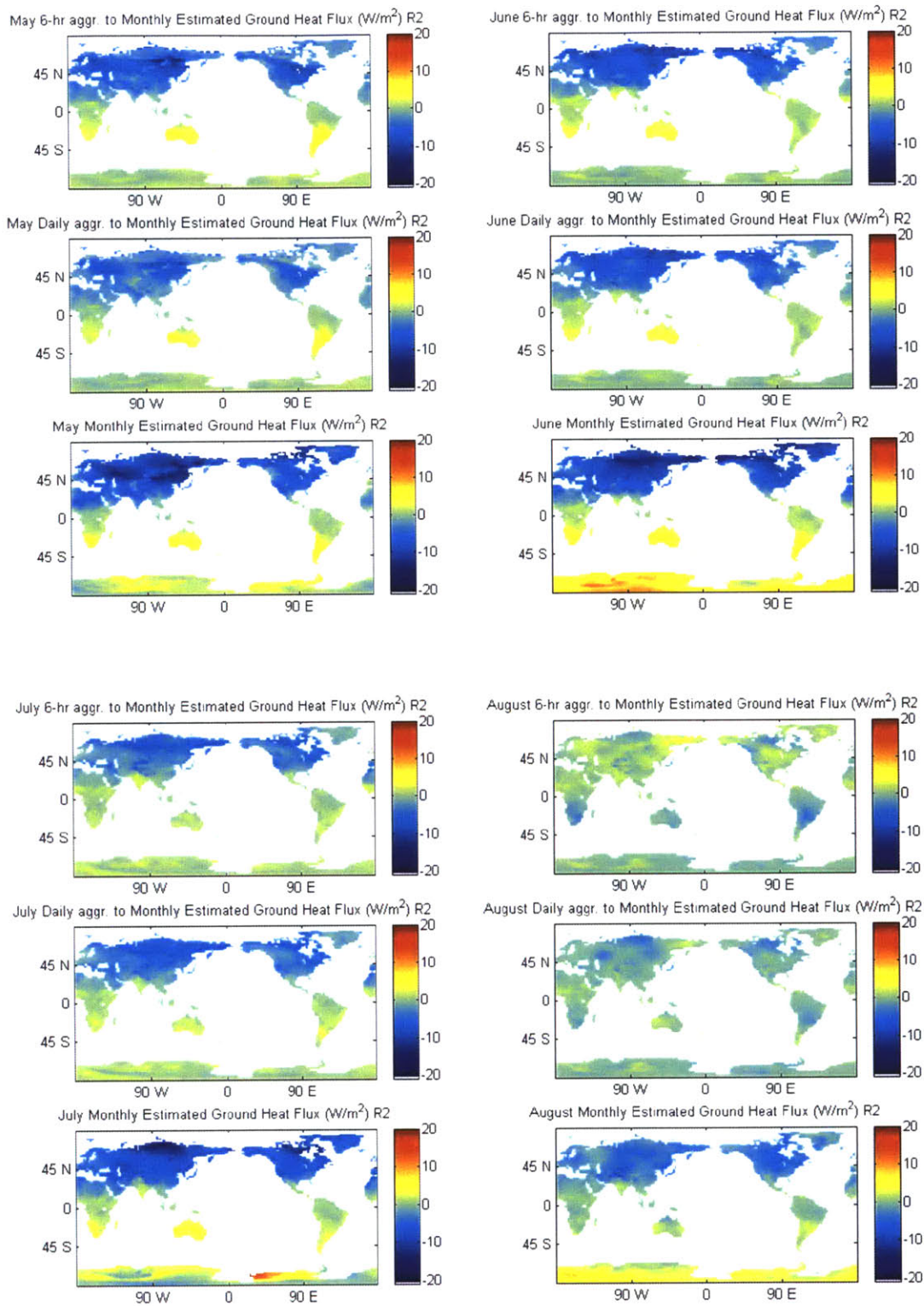


Figure 7-52: GHF Products Plots Aggregated to the Monthly time scale using the Reanalysis-2 – May, June, July, and August 1988

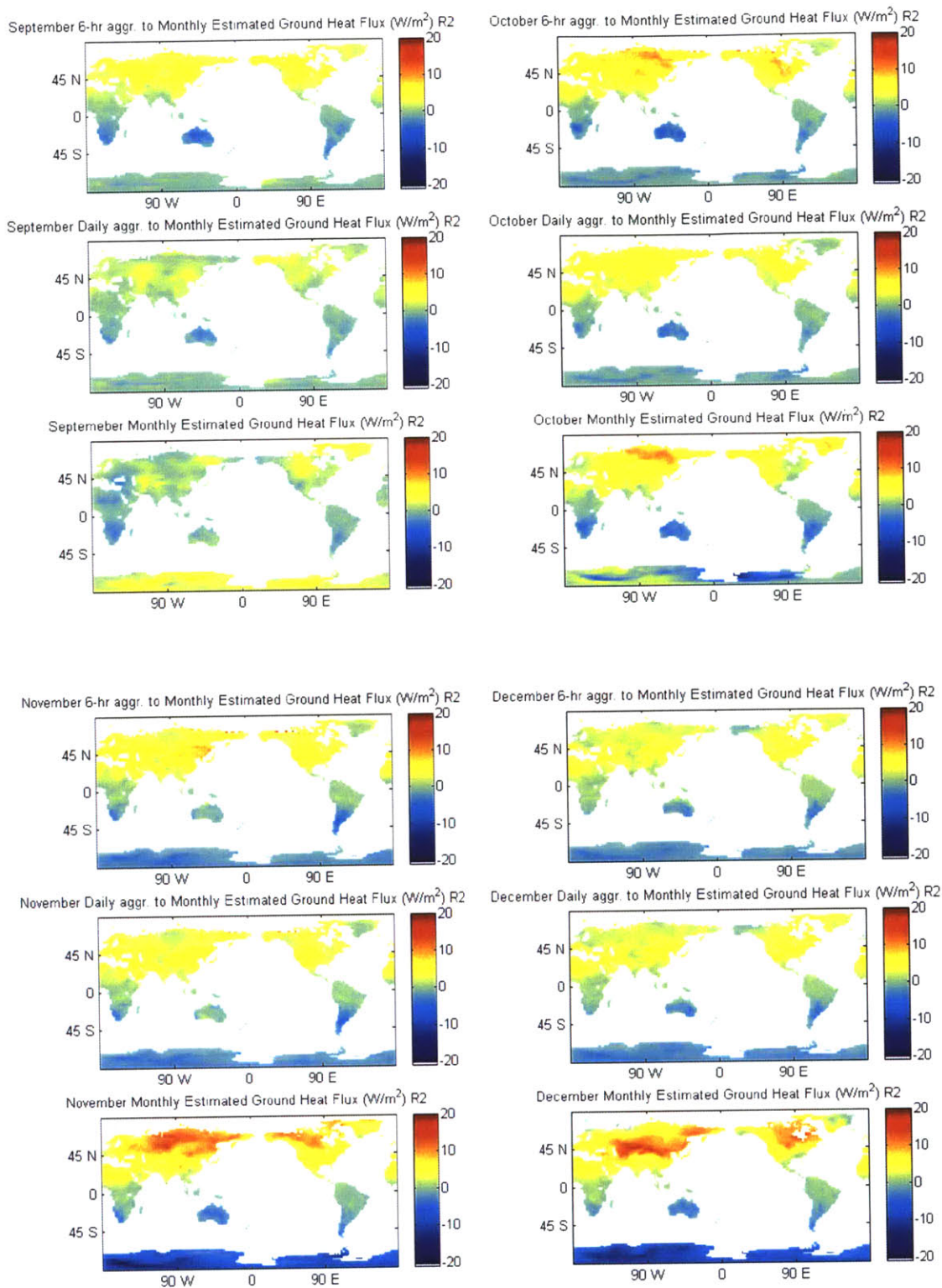


Figure 7-53: GHF Products Plots Aggregated to the Monthly time scale using the Reanalysis-2 – September, October, November, and December 1988

## 8. Conclusion

This work leads to a great deal of meaningful conclusions relative to the computation of water balances. The following is a brief summary of the major findings of this work.

### 8.1 Regional Water Balances

A number of evaluation methods are used in this work, and are outlined in Chapter 3. After the extensive use of these methodologies we get a general sense of which are most useful. The absolute error and bias are found to be very effective measures of the water balance residual. Also very useful is the regression slope on the monthly time scale where the adherence of different water balance variables to the seasonal cycle is evaluated.

The sensitivity of the water balance to different calculation methods, as well as domain scale and position is evaluated. The two moisture flux methodologies considered yield similar results for most control volumes. The three runoff methodologies evaluated also yield results which in general do not favor a specific method. The linear method is thought to have the best theoretical basis, but the point method produces slightly better results. The Reanalysis-2 gives slightly better results than the Reanalysis-1 for both atmospheric and surface balances.

A large amount of alternative precipitation datasets are used to compute both the atmospheric and surface water balances. The reanalysis precipitation completes both the atmospheric and surface balances consistently better than any other precipitation dataset. A specific look at TRMM precipitation datasets indicated these datasets are valuable in refining the water balance over the last five years (1998-2002). The use of the change in surface storage term derived from the reanalysis is the most effective method to account for change in surface storage. Nevertheless effective lag times can be found for all basin control volumes that eliminate the dependence on change in storage.

Control volumes which include a great deal of mountainous land (Mississippi, Arkansas, SWUS, and Larger US) yield poor atmospheric and surface balances. In the more successful South American balances the atmospheric balance is strengthened greatly by enlarging the control volume size, and the box control volumes compute a better atmospheric balance than the basin sized control volumes. The surface balance does not change much with basin size or shape.

The most important conclusion drawn from the evaluation of individual water balances is the general precipitation trend in all South American control volumes. The Reanalysis-1 precipitation in the NW Brazil, Larger Rondonia, East Brazil, Rondonia, Amazon, and Porto Velho control volumes increases in the second half of the yearly time series. This increase corresponds well with the moisture flux derived by the Reanalysis-1. However, the 55-year gauge-based precipitation dataset shows no precipitation increase for any of these control volumes. Other precipitation datasets over these regions are also inconsistent in their compliance to the reanalysis. Therefore, although some encouraging atmospheric water balances are computed for these control volumes it is difficult to decide whether these balances are the result of good data, or rather the reanalysis model forcing the data to balance.



## 8.2 Global Atmospheric Water Balance

In Chapter 6 the atmospheric water balance is computed over the entire globe between 60°N and 60°S. The balance is calculated over boxes of 30, 15, and 7.5 degrees on a side. As the resolution of the control volumes increases from 30 to 15 and 15 to 7.5 degrees on a side, the average residual of all control volumes across the globe increases by about a factor of two. Residuals are observed to visibly decrease from the monthly to yearly time scale. At the 30 degree resolution almost all control volumes perform well, and at all resolutions control volumes over the ocean perform. Biases over land at the 15 and 7.5 degree resolution are randomly largely positive and largely negative.

By varying precipitation in the global atmospheric water balance, it becomes apparent that the reanalysis precipitation is the most effective precipitation dataset at closing atmospheric water balances, as expected. Conducting separate balances for each month of the Reanalysis-2, a slightly better monthly balance is obtained in the winter months when all control volumes are taken into consideration in the Reanalysis-2.

By performing this analysis with the last five years of the Reanalysis-1, and replacing the precipitation with the TRMM 3B43 product for all control volumes inside of 30°N and 30°S latitude, results in biases that are closer to zero in the summertime. Breaking the Reanalysis-1 down and performing five decade long water balances, no large difference were found between the water balance properties calculated for the different decades.

When land and sea points are separated for various conditions, absolute errors are roughly twice as high over land. It is also observed that SSM/I precipitation is much more effective at closing the atmospheric water balance over land than over water.

## 8.3 Ground Heat Flux Research

A ground heat flux product is generated using the Wang and Bras (1999) method of estimation. The Reanalysis-1 and Reanalysis-2 skin temperatures at the 6-hourly, daily, and monthly time scales are used for this derivation for the year 1988. The key parameter in the derivation of ground heat flux is determined to be the thermal inertia of each land point. An original methodology is devised for obtaining the thermal inertia of all land points in the reanalysis, by dividing land points into twelve broad land cover classifications, including snow cover. Daily data is used to regress a thermal inertia for each of these twelve land classifications using all reanalysis ground heat flux data and estimated ground heat flux data for a given land type. These thermal inertias are then used to create a ground heat flux product at the 6-hourly, daily, and monthly time scales.

Samples of the finished product are provided in Chapter 7. There is a general increase in magnitude and sharpening of the diurnal cycle in the estimated 6-hourly ground heat flux when compared to the reanalysis ground heat flux. Significant areas of improvement include Central Asia and extreme northern and southern regions. The results of the new estimates are comparable to the reanalysis ground heat flux.

## 8.4 Concluding Remarks

This work includes an extensive evaluation of the Reanalysis-1, Reanalysis-2, and all other relevant and available datasets by application to various regional water balances. Of specific concern is the lack of ability to close regional water balances over land, specifically over a box of 7.5 degrees to a side. The integrity of the reanalysis data is simply not strong enough to complete such a balance reliably on either the atmospheric or surface levels.

The moisture flux and runoff variables are particularly uncertain. There is some success in calculating the moisture flux by known methodologies over regions of interest, but this is largely at the 15 and 30 degree resolutions. The runoff methodologies developed in this study were limited by the resolution of the data, and not terribly successful. Basin size control volumes were created as an alternative to box control volumes. Problems arose in the atmospheric balance of these control volumes, since in order to have a reasonably resolved basin a one-degree grid was necessary to form the boundaries of the control volume. However, the only data available to calculate moisture flux was gridded at a 2.5 degree resolution.

A major conclusion of this work is that global hydrometeorological datasets such as the reanalysis should take into account the water balance at finer resolutions when deriving its product. Current technologies are not adequate to close regional scale water balances over a number of years over North or South America at the 7.5 degree resolution. The theoretical framework for such a calculation is simple and well founded. It is the quality of available data that is lacking.

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# Appendix A

Eastern Brazil - Monthly Atmospheric																			
OI	Moist Flux	Re	Storage	Precipitation	Absolute Error	Bias	Slope	Intercept	R <sup>2</sup>	R <sup>2</sup> 1:1	Corr Coef	RMSE	AER	BR	SR	CCR	RMR	TR	BR
1	Method 1	1	Yes	reanalysis	92.49	-83.80	1.6812	80.04	0.7556	0.2477	0.8693	117.03	2	2	27	10	2	43	2
2	Method 1	1	Yes	gpcp	152.31	-151.86	1.4859	147.59	0.5801	-0.6984	0.7617	179.44	23	24	14	32	23	118	20
3	Method 1	1	Yes	gpcp	164.60	-164.16	1.6419	183.15	0.5906	-0.9289	0.7685	190.71	31	35	24	28	30	148	28
4	Method 1	1	Yes	gpcp trmm	149.36	-149.36	1.4653	148.47	0.8831	-1.2259	0.8265	164.27	22	23	13	12	16	86	13
5	Method 1	1	Yes	ghcn cams grid	115.28	-107.01	1.3097	103.75	0.4204	-0.2322	0.6484	149.76	7	7	6	52	9	81	10
6	Method 1	1	Yes	gpi	138.81	-135.92	1.1634	131.03	0.5296	-0.4988	0.7277	164.93	18	18	1	45	17	99	16
7	Method 1	1	Yes	trmm 3a46	129.00	-129.00	1.3000	123.17	0.7065	-0.7544	0.8405	143.23	12	15	5	11	6	49	4
8	Method 1	1	Yes	trmm 3b43	149.04	-149.04	1.3338	147.62	0.5967	-1.2667	0.7724	166.05	21	22	8	26	18	95	15
9	Method 1	1	Yes	ssmi	146.77	-145.06	1.3241	138.04	0.4872	-0.6696	0.6980	176.52	20	20	7	48	22	117	21
10	Method 1	1	Yes	opi2	156.32	-155.32	1.5373	153.18	0.4969	-0.8454	0.7049	186.38	28	28	20	47	29	152	31
11	Method 1	1	Yes	opi3	154.96	-154.41	1.4872	152.02	0.5831	-0.7460	0.7636	181.29	25	25	15	31	24	120	22
12	Method 1	1	Yes	cmap	155.39	-154.86	1.4972	152.29	0.5896	-0.7456	0.7678	181.55	28	28	17	29	25	123	23
13	Method 1	1	Yes	cmap2	155.74	-155.20	1.4905	152.83	0.5668	-0.7525	0.7660	181.91	27	27	16	30	28	126	24
14	Method 1	2	Yes	reanalysis	141.61	-137.73	1.4282	115.77	0.7702	0.0421	0.8776	166.08	19	19	11	8	19	78	8
15	Method 1	2	Yes	gpcp	202.36	-200.48	1.6935	208.44	0.6128	-0.8859	0.7828	233.03	43	43	28	22	44	180	42
16	Method 1	2	Yes	gpcp	216.98	-216.33	1.6484	228.71	0.5676	-1.3373	0.7534	245.35	50	50	26	34	52	212	51
17	Method 1	2	Yes	gpcp trmm	222.75	-222.75	1.9508	240.61	0.8344	-1.5204	0.9135	240.81	52	52	43	3	51	201	49
18	Method 1	2	Yes	ghcn cams grid	200.75	-199.47	1.7125	206.91	0.6495	-0.8447	0.8059	230.47	42	42	32	14	43	173	41
19	Method 1	2	Yes	gpi	195.57	-192.68	1.2558	190.28	0.5326	-0.9308	0.7298	223.01	41	41	3	44	42	171	40
20	Method 1	2	Yes	trmm 3a46	202.44	-202.44	1.8006	205.74	0.8469	-1.1037	0.9203	219.69	44	44	35	1	41	165	34
21	Method 1	2	Yes	trmm 3b43	220.54	-220.54	1.9691	238.59	0.8353	-1.4809	0.9140	238.92	51	51	44	2	49	197	47
22	Method 1	2	Yes	ssmi	208.96	-208.96	1.5198	208.81	0.5763	-1.1483	0.7591	237.92	49	49	19	33	48	198	48
23	Method 1	2	Yes	opi2	207.62	-204.58	1.7856	216.81	0.5552	-1.0058	0.7451	240.32	48	48	34	37	50	217	52
24	Method 1	2	Yes	opi3	205.59	-203.94	1.6994	214.37	0.6149	-0.9336	0.7842	235.96	46	46	29	21	46	188	45
25	Method 1	2	Yes	cmap	205.44	-203.87	1.7107	214.42	0.6200	-0.9305	0.7874	235.77	45	45	31	19	45	185	43
26	Method 1	2	Yes	cmap2	205.77	-204.21	1.7084	214.97	0.6199	-0.9350	0.7873	236.04	47	47	30	20	47	191	46
27	Method 2	7	Yes	reanalysis	88.08	-76.43	1.8384	50.19	0.7969	0.3263	0.8927	113.61	7	7	38	6	7	47	3
28	Method 2	1	Yes	gpcp	126.63	-120.61	1.6426	113.16	0.5515	-0.2336	0.7426	160.03	9	9	25	39	12	94	14
29	Method 2	1	Yes	gpcp	131.43	-125.01	1.8587	122.10	0.5907	-0.2974	0.7686	163.11	15	12	40	27	14	108	18
30	Method 2	1	Yes	gpcp trmm	128.26	-128.26	1.5509	126.25	0.6680	-0.6766	0.8173	147.94	11	14	21	13	7	66	6
31	Method 2	1	Yes	ghcn cams grid	106.44	-91.82	1.6189	82.28	0.4378	-0.0655	0.6616	143.01	4	3	23	50	5	85	12
32	Method 2	1	Yes	gpi	110.21	-98.85	1.3508	86.37	0.5111	-0.0161	0.7149	141.95	5	4	9	46	4	68	7
33	Method 2	7	Yes	trmm 3a46	100.45	-100.45	1.2099	93.87	0.6386	-0.1841	0.7991	121.84	3	5	2	16	3	29	1
34	Method 2	1	Yes	trmm 3b43	127.23	-127.23	1.3751	124.74	0.5524	-0.7277	0.7433	150.24	10	13	10	38	10	81	11
35	Method 2	1	Yes	ssmi	114.70	-106.28	1.2784	97.62	0.4345	-0.1256	0.6592	153.50	6	6	4	51	11	78	9
36	Method 2	1	Yes	opi2	138.43	-131.93	1.9851	133.04	0.4618	-0.4956	0.6795	175.96	17	17	45	49	21	149	29
37	Method 2	1	Yes	opi3	135.85	-129.79	1.9437	128.83	0.5370	-0.4061	0.7328	170.32	16	16	42	42	20	136	27
38	Method 2	1	Yes	cmap	129.52	-123.05	1.8382	115.98	0.5664	-0.2823	0.7526	162.98	13	10	37	35	13	108	17
39	Method 2	1	Yes	cmap2	129.99	-123.49	1.8301	116.86	0.5633	-0.2889	0.7505	163.39	14	11	36	36	15	112	19
40	Method 2	2	Yes	reanalysis	119.51	-109.87	1.5746	85.05	0.7654	0.2577	0.8749	146.60	8	8	22	9	8	55	5
41	Method 2	2	Yes	gpcp	167.06	-159.93	1.8746	165.94	0.6085	-0.3839	0.7800	202.89	33	32	41	23	33	162	33
42	Method 2	2	Yes	gpcp	171.64	-168.09	1.8402	180.01	0.6018	-0.6325	0.7758	203.95	36	36	39	25	34	170	37
43	Method 2	2	Yes	gpcp trmm	188.55	-188.55	2.1481	209.07	0.8088	-0.8471	0.8963	214.68	40	40	49	4	38	171	39
44	Method 2	2	Yes	ghcn cams grid	164.82	-158.91	2.0750	161.04	0.6443	-0.3562	0.8027	200.86	32	30	46	15	32	155	32
45	Method 2	2	Yes	gpi	153.11	-145.16	1.4568	137.40	0.5383	-0.3099	0.7337	186.31	24	21	12	41	28	128	25
46	Method 2	2	Yes	trmm 3a46	157.24	-157.24	1.7648	151.88	0.7752	-0.3638	0.8804	184.23	29	31	33	7	27	127	26
47	Method 2	2	Yes	trmm 3b43	184.74	-184.74	2.1513	200.93	0.8080	-0.7913	0.8889	211.41	39	39	50	5	37	170	38
48	Method 2	2	Yes	ssmi	159.08	-156.54	1.5015	150.94	0.5342	-0.4018	0.7309	197.98	30	29	18	43	31	151	30
49	Method 2	2	Yes	opi2	180.26	-172.01	2.2892	186.44	0.5410	-0.6252	0.7355	219.88	38	38	52	40	40	208	50
50	Method 2	2	Yes	opi3	177.06	-170.16	2.2265	191.12	0.6072	-0.5503	0.7793	214.75	37	37	51	24	39	188	44
51	Method 2	2	Yes	cmap	170.30	-162.94	2.0850	173.64	0.6221	-0.4388	0.7887	206.88	34	33	48	18	35	168	35
52	Method 2	2	Yes	cmap2	170.71	-163.38	2.0850	174.57	0.6226	-0.4433	0.7891	207.20	35	34	47	17	36	169	36

Eastern Brazil - Yearly Atmospheric																					
Of	Moist Flux	Re	Storage	Precipitation	Absolute Error	Bias	Slope	Oslope	Intercept	R <sup>2</sup>	R <sup>2</sup> 1:1	Corr Coef	RMSE	AER	BR	SR	CCR	RMR	TR	BR	
1	Method 1	1	Yes	reanalysis	85.09	-83.56	2.3669	2.3669	36.28	0.6833	-1.2476	0.8266	95.99	2	2	17	1	2	24	2	
2	Method 1	1	Yes	gpcp	151.91	-151.91	0.2785	3.5907	158.25	0.0150	-19.5640	0.1224	156.08	24	24	23	23	24	118	21	
3	Method 1	1	Yes	gpcc	164.04	-164.04	0.4885	2.0471	164.84	0.0342	-19.9624	0.1849	168.11	34	34	15	17	32	132	25	
4	Method 1	1	Yes	gpcc trmm	149.58	-149.58	0.6612	1.5124	151.13	0.3789	-129.5871	0.6155	150.00	23	23	7	6	21	80	14	
5	Method 1	1	Yes	ghcn cams grid	108.24	-106.74	-0.7556		125.68	0.0317	-2.8822	-0.1781	127.00	5	5	49	49	9	117	20	
6	Method 1	1	Yes	gpi	135.91	-135.91	0.7444	1.3434	143.56	0.0890	-15.3455	0.3147	139.87	18	18	5	12	18	71	8	
7	Method 1	1	Yes	trmm 3a46	132.26	-132.26	-0.2492		165.44	0.0120	-122.7783	-0.1096	132.95	16	16	41	43	14	130	23	
8	Method 1	1	Yes	trmm 3b43	149.30	-149.30	0.5621	1.7790	151.17	0.3579	-159.9959	0.5983	149.71	22	22	10	7	20	81	15	
9	Method 1	1	Yes	ssmi	147.55	-147.55	0.5157	1.9391	158.38	0.0248	-16.7985	0.1575	151.87	21	21	11	18	22	93	18	
10	Method 1	1	Yes	opi2	156.00	-156.00	0.2060	4.8544	159.74	0.0085	-20.6606	0.0922	160.19	28	28	25	24	28	133	26	
11	Method 1	1	Yes	opi3	155.36	-155.36	0.0629	15.8983	180.37	0.0010	-20.5898	0.0313	159.93	26	26	31	30	26	139	30	
12	Method 1	1	Yes	cmmap	155.29	-155.29	0.0690	14.4928	160.33	0.0012	-20.5656	0.0340	159.84	25	25	30	29	25	134	28	
13	Method 1	1	Yes	reanalysis	155.63	-155.63	0.0807	12.3916	160.29	0.0016	-20.6493	0.0399	160.15	27	27	29	28	27	138	29	
14	Method 1	2	Yes	reanalysis	137.81	-137.81	2.4636	2.4636	62.75	0.5834	-6.8154	0.7638	143.65	19	19	18	3	19	78	13	
15	Method 1	2	Yes	gpcc	200.57	-200.57	-0.3320		185.29	0.0121	-15.4178	-0.1101	208.21	43	43	44	44	43	217	48	
16	Method 1	2	Yes	gpcc	216.24	-216.24	0.0947	10.5597	198.96	0.0007	-14.0991	0.0284	224.28	50	50	28	31	52	111	46	
17	Method 1	2	Yes	gpcc trmm	223.22	-223.22	0.8689	1.1509	220.76	0.3450	-199.8318	0.5874	223.59	52	52	2	8	51	165	37	
18	Method 1	2	Yes	ghcn cams grid	199.55	-199.55	-0.1069		187.98	0.0015	-15.2416	-0.0389	207.09	42	42	34	34	42	194	44	
19	Method 1	2	Yes	gpi	192.72	-192.72	0.1521	6.5746	200.67	0.0025	-14.9489	0.0501	199.57	41	41	26	26	41	175	38	
20	Method 1	2	Yes	trmm 3a46	209.43	-209.43	0.5892	1.6972	210.55	0.0444	-485.0383	0.2108	209.65	48	48	9	16	44	165	36	
21	Method 1	2	Yes	trmm 3b43	221.01	-221.01	0.9003	1.1107	219.36	0.4088	-195.8171	0.6394	221.34	51	51	1	5	50	158	32	
22	Method 1	2	Yes	ssmi	214.51	-214.51	-0.1812		215.59	0.0024	-23.8250	-0.0492	219.43	49	49	37	37	49	221	49	
23	Method 1	2	Yes	opi2	204.66	-204.66	-0.1440		186.86	0.0023	-16.0059	-0.0479	211.90	47	47	36	36	47	213	47	
24	Method 1	2	Yes	opi3	204.02	-204.02	-0.3732		183.53	0.0190	-16.0016	-0.1378	211.88	45	45	47	47	48	230	51	
25	Method 1	2	Yes	cmmap	203.95	-203.95	-0.3637		183.70	0.0178	-15.9851	-0.1333	211.77	44	44	46	46	45	225	50	
26	Method 1	2	Yes	reanalysis	204.29	-204.29	-0.3434		183.88	0.0158	-16.0321	-0.1259	212.07	46	46	45	45	48	230	52	
27	Method 2	1	Yes	reanalysis	78.41	-78.37	1.9392	1.9392	46.97	0.6525	-1.5009	0.8078	85.39	1	1	12	2	1	17	1	Plot
28	Method 2	1	Yes	gpcp	120.67	-120.67	0.5043	1.9829	126.35	0.0477	-14.8550	0.2185	124.60	9	9	14	15	8	55	4	Plot
29	Method 2	1	Yes	gpcc	124.87	-124.87	0.6113	1.6359	126.19	0.0572	-18.7949	0.2391	128.03	12	12	8	14	11	57	6	
30	Method 2	1	Yes	gpcc trmm	131.91	-131.91	0.4775	2.0942	135.26	0.0878	-43.7273	0.2962	133.44	15	15	16	13	15	74	11	
31	Method 2	1	Yes	ghcn cams grid	92.56	-91.81	-1.2311		126.80	0.0950	-3.0568	-0.3082	109.73	3	3	51	50	4	111	19	
32	Method 2	1	Yes	gpi	96.81	-96.81	0.0396	25.2525	133.00	0.0003	-10.3574	0.0159	104.22	4	4	32	32	3	75	12	
33	Method 2	1	Yes	trmm 3a46	109.10	-109.10	-1.0402		186.14	0.3314	-50.7487	-0.5757	111.21	7	7	50	52	5	121	22	
34	Method 2	1	Yes	trmm 3b43	127.45	-127.45	0.6913	1.4465	129.50	0.2326	-41.5017	0.4823	128.68	13	13	6	11	13	56	5	
35	Method 2	1	Yes	ssmi	109.00	-109.00	0.2312	4.3253	133.20	0.0083	-12.1743	0.0909	113.78	6	6	24	25	6	67	7	
36	Method 2	1	Yes	opi2	132.69	-132.69	0.4047	2.4710	132.40	0.0245	-18.0103	0.1566	136.43	17	17	19	19	17	89	17	
37	Method 2	1	Yes	opi3	130.83	-130.83	0.3049	3.2798	131.78	0.0166	-17.5517	0.1289	134.78	14	14	22	22	16	88	16	
38	Method 2	1	Yes	cmmap	123.61	-123.61	0.3125	3.2000	129.52	0.0190	-15.6785	0.1379	127.79	10	10	21	21	10	72	9	
39	Method 2	1	Yes	reanalysis	124.05	-124.05	0.3297	3.0331	129.52	0.0214	-15.7846	0.1463	128.20	11	11	20	20	12	74	10	
40	Method 2	2	Yes	reanalysis	108.92	-108.92	1.9487	1.9487	68.95	0.4825	-5.2871	0.6946	115.90	8	8	13	4	7	40	3	Plot
41	Method 2	2	Yes	gpcp	159.98	-159.98	-0.1335		152.20	0.0020	-12.1195	-0.0445	167.43	30	30	35	35	31	161	35	
42	Method 2	2	Yes	gpcc	167.99	-167.99	0.1420	7.0423	155.81	0.0017	-13.5294	0.0410	174.47	35	35	27	27	38	160	34	
43	Method 2	2	Yes	gpcc trmm	188.94	-188.94	1.2606	1.2606	193.60	0.2495	-51.6613	0.4995	190.32	40	40	3	10	40	133	27	
44	Method 2	2	Yes	ghcn cams grid	156.96	-156.96	0.0062	161.2903	153.14	0.0000	-11.6418	0.0021	164.35	29	29	33	33	29	153	31	
45	Method 2	2	Yes	gpi	145.17	-145.17	-0.4733		170.23	0.0251	-11.4218	-0.1583	152.86	20	20	48	48	23	159	33	
46	Method 2	2	Yes	trmm 3a46	168.31	-168.31	-1.8084		203.05	0.2786	-80.5453	-0.5278	170.33	36	36	52	51	33	208	45	
47	Method 2	2	Yes	trmm 3b43	185.13	-185.13	1.3055	1.3055	189.43	0.3058	-49.5409	0.5530	186.45	39	39	4	9	39	130	24	
48	Method 2	2	Yes	ssmi	162.06	-162.06	-0.3087		177.03	0.0113	-17.9883	-0.1064	167.40	31	31	43	42	30	177	39	
49	Method 2	2	Yes	opi2	172.06	-172.06	-0.2207		148.93	0.0041	-13.9779	-0.0642	178.89	38	38	38	38	38	191	42	
50	Method 2	2	Yes	opi3	170.21	-170.21	-0.3010		147.97	0.0095	-13.7271	-0.0976	177.39	37	37	42	41	37	194	43	
51	Method 2	2	Yes	cmmap	162.99	-162.99	-0.2433		150.71	0.0067	-12.8003	-0.0817	170.47	32	32	40	40	34	178	40	
52	Method 2	2	Yes	reanalysis	163.43	-163.43	-0.2090		150.96	0.0049	-12.6608	-0.0703	170.85	33	33	38	39	35	178	41	



Eastern Brazil - Monthly Surface																							
Ol	Runoff Meth	Re	Storage	Precipitation	Absolute Error	Bias	Abias	Slope	Intercept	R <sup>2</sup>	R <sup>2</sup> 1.1	Corr Coef	RMSE	AER	BR	SR	CCR	RMR	TR	BR	Plot		
1	Linear	1	Yes	reanalysis	81.89	-35.38	35.38	0.6430	-24.20	0.2035	0.0045	0.4511	95.90	428	419	43	70	409	1369	314	Plot		
2	Linear	1	Yes	gpcp	80.34	-15.11	15.11	0.4925	-9.23	0.1080	-0.0306	0.3286	99.12	421	230	64	130	422	1267	270			
3	Linear	1	Yes	gpcp	73.39	-2.87	2.87	0.6768	-1.78	0.1698	0.1302	0.4120	90.71	375	61	37	89	377	939	161			
4	Linear	1	Yes	gpcp trmm	77.72	-17.73	17.73	0.6434	-16.42	0.1521	0.0736	0.3901	95.59	404	252	42	103	408	1209	250			
5	Linear	1	Yes	ghcn cams grid	75.58	-20.00	20.00	0.7465	-16.08	0.1954	0.1291	0.4420	89.24	389	277	24	73	364	1127	225			
6	Linear	1	Yes	gpi	86.62	-40.19	40.19	0.5682	-24.82	0.1874	-0.0897	0.4330	102.10	445	442	56	79	432	1454	343			
7	Linear	1	Yes	trmm 3a46	87.29	-56.97	56.97	0.5849	-43.97	0.2028	-0.2507	0.4504	107.49	447	464	52	71	445	1479	351			
8	Linear	1	Yes	trmm 3b43	80.08	-20.57	20.57	0.5708	-17.72	0.1282	0.0121	0.3580	97.90	420	282	54	113	418	1287	278			
9	Linear	1	Yes	ssmi	82.53	-37.34	37.34	0.5787	-24.24	0.1911	-0.0532	0.4372	101.34	433	426	53	75	429	1416	327			
10	Linear	1	Yes	opi2	76.81	-4.15	4.15	0.6348	-4.57	0.1033	0.0673	0.3214	93.79	398	110	45	132	399	1084	213			
11	Linear	1	Yes	opi3	76.56	-6.30	6.30	0.5930	-5.88	0.1094	0.0536	0.3307	94.47	397	160	51	128	406	1142	230			
12	Linear	1	Yes	cmmap	78.68	-13.38	13.38	0.5701	-9.76	0.1191	0.0325	0.3451	95.75	411	217	55	119	410	1212	251			
13	Linear	1	Yes	cmmap2	78.63	-12.94	12.94	0.5634	-9.45	0.1167	0.0290	0.3416	95.93	410	215	57	121	411	1214	252			
14	Linear	1	No	reanalysis	62.34	-35.21	35.21	-0.0861	-1.22	0.2432	-0.4932	-0.4932	81.69	239	417	400	440	300	1796	408			
15	Linear	1	No	gpcp	54.24	-15.23	15.23	-0.0872	-2.63	0.1604	-0.2594	-0.4005	73.59	160	234	401	431	206	1432	332			
16	Linear	1	No	gpcp	47.45	-3.11	3.11	-0.0204	0.35	0.0656	-0.1647	-0.2562	60.67	109	73	331	399	107	1019	192			
17	Linear	1	No	gpcp trmm	60.69	-15.30	15.30	-0.2197	-10.85	0.3274	-0.1020	-0.5722	77.37	220	235	444	451	248	1598	379			
18	Linear	1	No	ghcn cams grid	48.39	-19.37	19.37	-0.0797	-2.69	0.1474	-0.2919	-0.3839	65.04	125	268	392	426	138	1349	307			
19	Linear	1	No	gpi	67.51	-39.57	39.57	-0.0823	-1.04	0.1465	-0.31340	-0.3827	91.05	309	438	395	425	384	1951	431			
20	Linear	1	No	trmm 3a46	84.63	-46.14	46.14	-0.2563	-6.77	0.2569	-0.15244	-0.7259	105.35	443	445	450	465	440	2243	460			
21	Linear	1	No	trmm 3b43	61.94	-18.19	18.19	-0.2304	-10.00	0.3851	-0.11275	-0.6206	80.21	232	255	447	457	283	1874	392			
22	Linear	1	No	ssmi	71.80	-35.99	35.99	-0.1069	-1.58	0.2101	-0.265700	-0.4583	91.40	359	423	412	438	387	2019	443			
23	Linear	1	No	opi2	43.38	-2.50	2.50	-0.1180	-3.76	0.1713	-0.15283	-0.4138	56.49	53	53	424	434	58	1022	194			
24	Linear	1	No	opi3	46.57	-4.65	4.65	-0.0999	-3.53	0.1490	-0.18058	-0.3860	61.13	96	129	406	427	110	1168	236			
25	Linear	1	No	cmmap	50.17	-12.09	12.09	-0.0935	-2.87	0.1533	-0.2738	-0.3916	66.82	143	202	403	429	150	1327	297			
26	Linear	1	No	cmmap2	50.13	-11.64	11.64	-0.0932	-2.91	0.1530	-0.27511	-0.3912	66.83	141	198	402	428	151	1320	294			
27	Linear	1	mon	reanalysis	62.02	-35.11	35.11	-0.0654	-1.86	0.1406	-0.461437	-0.3748	80.60	234	415	379	424	287	1739	399			
28	Linear	1	mon	gpcp	53.42	-14.93	14.93	-0.0759	-2.75	0.1213	-0.253447	-0.3483	72.85	156	226	389	419	195	1385	319			
29	Linear	1	mon	gpcp	46.58	-2.28	2.28	-0.0138	0.38	0.0302	-0.1628480	-0.1739	59.82	95	47	326	389	96	953	168			
30	Linear	1	mon	gpcp trmm	59.71	-14.95	14.95	-0.1991	-10.46	0.2751	-0.101194	-0.5245	76.88	209	227	443	446	239	1564	370			
31	Linear	1	mon	ghcn cams grid	48.09	-19.38	19.38	-0.0707	-2.85	0.1161	-0.292439	-0.3407	64.64	120	269	380	417	136	1322	295			
32	Linear	1	mon	gpi	66.34	-39.18	39.18	-0.0618	-1.78	0.0822	-0.301637	-0.2867	89.62	296	434	377	405	368	1880	422			
33	Linear	1	mon	trmm 3a46	83.05	-47.16	47.16	-0.2280	-6.82	0.4095	-0.148752	-0.6399	104.41	437	446	445	459	438	2275	457			
34	Linear	1	mon	trmm 3b43	59.50	-16.40	16.40	-0.1863	-10.10	0.2502	-0.104110	-0.5002	77.27	204	249	442	442	245	1582	376			
35	Linear	1	mon	ssmi	70.89	-35.62	35.62	-0.0964	-1.89	0.1697	-0.259476	-0.4120	90.74	349	420	404	432	379	1984	439			
36	Linear	1	mon	opi2	43.28	-2.54	2.54	-0.1012	-3.71	0.1262	-0.148389	-0.3552	55.85	51	56	408	422	52	989	179			
37	Linear	1	mon	opi3	46.11	-4.69	4.69	-0.0849	-3.51	0.1077	-0.175844	-0.3282	60.50	94	130	398	411	106	1139	229			
38	Linear	1	mon	cmmap	49.30	-12.26	12.26	-0.0801	-2.93	0.1125	-0.210818	-0.3353	66.16	133	203	393	415	148	1292	282			
39	Linear	1	mon	cmmap2	49.23	-11.82	11.82	-0.0802	-2.97	0.1131	-0.21007	-0.3364	66.19	131	200	394	416	149	1290	281			
40	Linear	1	mon	reanalysis	61.58	-35.06	35.06	-0.0356	-2.80	0.0415	-0.441897	-0.2037	78.96	228	414	349	393	271	1655	389			
41	Linear	1	mon	gpcp	52.37	-14.55	14.55	-0.0367	-3.14	0.0282	-0.253939	-0.1680	70.35	152	223	351	386	176	1288	279			
42	Linear	1	mon	gpcp	45.27	-1.58	1.58	0.0064	0.42	0.0077	-0.1629739	0.0816	58.42	84	26	310	268	82	770	111			
43	Linear	1	mon	gpcp trmm	58.53	-14.63	14.63	-0.1251	-9.57	0.1267	-0.102527	-0.3503	73.06	197	224	428	421	200	1470	347			
44	Linear	1	mon	ghcn cams grid	47.73	-19.42	19.42	-0.0406	-3.32	0.0384	-0.278773	-0.1959	63.18	113	270	359	391	129	1262	268			
45	Linear	1	mon	gpi	65.75	-38.91	38.91	-0.0236	-3.09	0.0120	-0.284039	-0.1095	87.22	285	432	334	360	351	1762	405			
46	Linear	1	mon	trmm 3a46	82.05	-47.75	47.75	-0.1344	-8.15	0.1594	-0.148810	-0.3992	99.18	430	447	432	430	423	2162	452			
47	Linear	1	mon	trmm 3b43	59.78	-15.15	15.15	-0.1150	-9.45	0.1067	-0.106023	-0.3297	73.58	210	231	421	412	205	1479	350			
48	Linear	1	mon	ssmi	71.01	-35.67	35.67	-0.0600	-3.02	0.0657	-0.246063	-0.2564	88.71	351	421	376	400	359	1907	425			
49	Linear	1	mon	opi2	42.51	-2.50	2.50	-0.0474	-3.59	0.0278	-0.135745	-0.1666	53.54	37	52	368	384	28	869	136			
50	Linear	1	mon	opi3	45.42	-4.72	4.72	-0.0396	-3.49	0.0235	-0.162936	-0.1532	58.33	85	133	356	376	81	1031	198			
51	Linear	1	mon	cmmap	48.50	-12.35	12.35	-0.0372	-3.24	0.0244	-0.184685	-0.1581	83.94	127	205	352	379	133	1196	245			
52	Linear	1	mon	cmmap2	48.42	-11.91	11.91	-0.0375	-3.25	0.0248	-0.18653	-0.1573	83.97	126	201	353	381	134	1195	244			
53	Linear	1	mon	reanalysis	60.78	-35.12	35.12	0.0091	-4.21	0.0027	-0.112550	0.0523	76.41	221	416	303	289	233	1462	348			
54	Linear	1	mon	gpcp	51.39	-14.54	14.54	0.0047	-3.59	0.0005	-0.218540	0.0214	67.94	150	222	311	309	158	1150	231			
55	Linear	1	mon	gpcp	45.06	-1.76	1.76	0.0213	0.41	0.0750	-0.1582970	0.2738	57.72	83	30	273	185	73	624	71			
56	Linear	1	mon	gpcp trmm	57.11	-14.14	14.14	-0.0439	-9.44	0.0155	-0.2075	-0.1246	69.18	182	219	364	365	168	1298	289			
57	Linear	1	mon	ghcn cams grid	46.83	-19.49	19.49	-0.0087	-3.86	0.0010	-0.263128	-0.0321	61.48	101	274	323	325	114	1137	226			
58	Linear	1	mon	gpi	66.01	-39.20	39.20	0.0022	-4.00	0.0011	-0.272906	0.0102	85.76	290	435	318	318	339	1700	395			
59	Linear	1	mon	trmm 3a46	82.86	-48.34	48.34	-0.0746	-9.52	0.0498	-0.140899	-0.2232	96.57	435	448	387	396	415	2081	448			
60	Linear	1	mon	trmm 3b43	58.39	-15.52	15.52	-0.0494	-9.25	0.0205	-0.97246	-0.1431	70.32	194	237	372	369	175	1347	303			
61	Linear	1	mon	ssmi	70.97	-35.97	35.97	-0.0312	-3.95	0.0178	-0.235975	-0.1333	87.15	350	422	343	368	350	1833	415			
62	Linear	1	mon	opi2	41.06	-2.65	2.65	0.0114	-3.52	0.0016	-0.120928	0.0401	50.84	24	58	296	294	17	689	86			
63	Linear	1	mon	opi3	43.89	-4.89	4.89	0.0098	-3.55	0.0014	-0.147785	0.0378	55.81	58	136	299	298	51	842	124			
64	Linear	1																					

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OI	Runoff Meth	Re	Storage	Precipitation	Absolute Error	Bias	Abias	Slope	Intercept	R <sup>2</sup>	R <sup>2</sup> 1.1	Corr Coef	RMSE	AER	BR	SR	CCR	AMR	TR	BR
102	Linear	2	No	op3	66.10	2.91	2.91	-0.1118	-16.09	0.0222	-2.1851	-0.1490	80.68	294	64	417	373	289	1437	334
103	Linear	2	No	cm3p	68.66	-4.31	4.31	-0.1089	-15.25	0.0247	-2.5456	-0.1571	85.12	333	120	415	380	328	1576	374
104	Linear	2	No	cm2p	68.52	-3.86	3.86	-0.1079	-15.29	0.0243	-2.5419	-0.1558	85.08	326	99	413	378	327	1543	364
105	Linear	2	1 mon	reanalysis	95.84	-57.11	57.11	-0.1338	-8.45	0.0802	-7.2730	-0.2832	130.26	462	465	431	404	464	2226	458
106	Linear	2	1 mon	gpcp	71.40	-6.97	6.97	-0.1067	-14.97	0.0285	-3.0605	-0.1688	91.26	355	169	411	387	386	1708	396
107	Linear	2	1 mon	gpcp	63.95	19.50	19.50	0.0923	5.98	0.0298	-3.1489	0.1727	73.02	258	275	139	224	197	1093	218
108	Linear	2	1 mon	gpcp trmm	89.82	-23.85	23.85	-0.5993	-56.98	0.4847	-3.1551	-0.6962	112.06	449	325	467	461	450	2152	451
109	Linear	2	1 mon	ghcn cams grid	68.25	-10.01	10.01	-0.1041	-14.63	0.0234	-2.8625	-0.1531	86.67	318	185	410	375	347	1635	387
110	Linear	2	1 mon	gpi	77.95	-25.89	25.89	-0.0387	-8.65	0.0047	-3.6644	-0.0683	100.38	405	339	354	342	426	1866	420
111	Linear	2	1 mon	trmm 3a46	111.02	-55.64	55.64	-0.5721	-41.43	0.6164	-4.9841	-0.7851	139.96	468	459	466	468	468	2329	468
112	Linear	2	1 mon	trmm 3b43	91.69	-27.99	27.99	-0.6177	-54.80	0.5235	-3.3267	-0.7236	114.35	451	356	468	464	456	2195	456
113	Linear	2	1 mon	ssmi	82.52	-22.91	22.91	-0.1309	-9.93	0.0486	-3.8037	-0.2205	106.15	432	315	430	395	442	2014	442
114	Linear	2	1 mon	opi2	62.31	5.00	5.00	-0.1195	-16.49	0.0214	-1.8639	-0.1461	76.64	238	138	425	371	236	1408	322
115	Linear	2	1 mon	opi3	65.21	3.29	3.29	-0.1146	-16.20	0.0232	-2.1713	-0.1522	80.65	279	78	420	374	288	1439	337
116	Linear	2	1 mon	cm3p	67.46	-3.92	3.92	-0.1130	-15.35	0.0263	-2.5346	-0.1622	85.14	308	100	419	383	329	1539	360
117	Linear	2	1 mon	cm2p	67.31	-3.48	3.48	-0.1114	-15.39	0.0256	-2.5285	-0.1600	85.07	306	85	416	382	326	1515	356
118	Linear	2	2 mon	reanalysis	94.33	-56.64	56.64	-0.1020	-9.88	0.0463	-6.9189	-0.2152	127.68	459	462	409	394	463	2187	455
119	Linear	2	2 mon	gpcp	88.59	-8.58	8.58	-0.0757	-14.78	0.0142	-2.8794	-0.1193	89.36	329	167	388	362	365	1611	383
120	Linear	2	2 mon	gpcp	60.01	20.51	20.51	0.1485	7.23	0.0768	-2.7781	0.2771	69.60	212	280	105	162	172	931	156
121	Linear	2	2 mon	gpcp trmm	82.58	-18.65	18.65	-0.4429	-52.20	0.2752	-2.7715	-0.5246	101.91	434	259	461	447	431	2032	445
122	Linear	2	2 mon	ghcn cams grid	65.58	-9.73	9.73	-0.0856	-14.59	0.0158	-2.5660	-0.1256	85.68	281	183	399	366	335	1564	371
123	Linear	2	2 mon	gpi	75.66	-25.34	25.34	0.0073	-9.28	0.0002	-3.3542	0.0128	97.15	390	337	307	316	416	1766	406
124	Linear	2	2 mon	trmm 3a46	106.47	-54.75	54.75	-0.4250	-38.95	0.3708	-4.8005	-0.6090	131.70	466	457	459	454	465	2301	466
125	Linear	2	2 mon	trmm 3b43	84.63	-22.92	22.92	-0.4240	-49.95	0.2588	-2.8508	-0.5087	102.98	442	316	458	443	434	2093	449
126	Linear	2	2 mon	ssmi	81.81	-23.47	23.47	-0.1118	-11.26	0.0384	-4.0127	-0.1959	104.26	427	324	418	392	437	1998	441
127	Linear	2	2 mon	opi2	61.27	5.22	5.22	-0.0970	-16.09	0.0140	-1.7894	-0.1184	75.78	225	140	405	361	224	1355	308
128	Linear	2	2 mon	opi3	62.54	3.57	3.57	-0.0768	-15.57	0.0103	-2.0260	-0.1017	78.82	243	91	391	353	268	1346	301
129	Linear	2	2 mon	cm3p	64.52	-3.57	3.57	-0.0721	-14.97	0.0106	-2.3451	-0.1031	82.98	267	90	385	359	312	1413	325
130	Linear	2	2 mon	cm2p	64.36	-3.13	3.13	-0.0720	-15.00	0.0106	-2.3453	-0.1030	82.98	265	74	384	357	313	1393	320
131	Linear	2	3 mon	reanalysis	93.84	-56.58	56.58	-0.0298	-12.85	0.0039	-6.2738	-0.0628	122.52	457	461	342	341	462	2063	446
132	Linear	2	3 mon	gpcp	68.60	-6.45	6.45	-0.0440	-14.45	0.0048	-2.7241	-0.0694	87.66	331	165	365	343	353	1557	366
133	Linear	2	3 mon	gpcp	61.05	20.95	20.95	0.1143	7.37	0.0478	-3.1763	0.2185	71.66	223	286	126	194	189	1018	191
134	Linear	2	3 mon	gpcp trmm	75.82	-15.95	15.95	-0.2369	-45.17	0.0863	-2.3635	-0.2937	92.91	391	242	449	407	395	1884	423
135	Linear	2	3 mon	ghcn cams grid	65.05	-9.63	9.63	-0.0269	-14.23	0.0016	-2.3148	-0.0394	82.71	275	182	338	331	311	1437	336
136	Linear	2	3 mon	gpi	76.14	-24.87	24.87	0.0502	-9.56	0.0079	-3.1184	0.0891	94.08	379	300	213	266	402	1605	381
137	Linear	2	3 mon	trmm 3a46	101.30	-53.10	53.10	-0.2311	-38.32	0.1196	-4.2843	-0.3458	121.50	465	456	448	418	481	2248	462
138	Linear	2	3 mon	trmm 3b43	77.60	-20.19	20.19	-0.2291	-44.02	0.0827	-2.4572	-0.2876	94.19	403	278	446	406	405	1938	430
139	Linear	2	3 mon	ssmi	81.50	-24.06	24.06	-0.0714	-12.56	0.0160	-3.8587	-0.1265	101.80	426	326	382	367	430	1931	427
140	Linear	2	3 mon	opi2	59.37	5.33	5.33	-0.0289	-14.67	0.0012	-1.5890	-0.0353	73.09	203	143	340	327	202	1215	253
141	Linear	2	3 mon	opi3	61.36	3.72	3.72	-0.0312	-14.67	0.0017	-1.8684	-0.0413	76.94	226	95	344	332	241	1238	263
142	Linear	2	3 mon	cm3p	63.72	-3.42	3.42	-0.0263	-14.39	0.0014	-2.1596	-0.0377	80.75	255	83	337	320	292	1297	288
143	Linear	2	3 mon	cm2p	63.62	-2.98	2.98	-0.0259	-14.40	0.0014	-2.1581	-0.0370	80.73	253	68	336	329	290	1276	272
144	Linear	2	4 mon	reanalysis	93.43	-56.85	56.85	0.0162	-14.79	0.0012	-5.8632	0.0342	119.23	456	463	291	300	460	1970	437
145	Linear	2	4 mon	gpcp	68.23	-6.55	6.55	-0.0103	-14.18	0.0003	-2.5553	-0.0163	85.81	317	166	325	322	340	1470	348
146	Linear	2	4 mon	gpcp	64.42	20.77	20.77	0.0310	6.31	0.0035	-3.7803	0.0593	76.74	266	284	255	283	237	1325	296
147	Linear	2	4 mon	gpcp trmm	67.65	-16.79	16.79	0.0236	-39.49	0.0009	-1.5746	0.0293	82.05	312	251	267	301	306	1437	335
148	Linear	2	4 mon	ghcn cams grid	64.34	-9.74	9.74	0.0116	-14.05	0.0003	-2.1480	0.0171	80.75	264	184	295	313	291	1347	304
149	Linear	2	4 mon	gpi	75.03	-24.97	24.97	0.0526	-9.52	0.0087	-3.1033	0.0933	94.13	385	331	203	263	403	1585	377
150	Linear	2	4 mon	trmm 3a46	95.59	-56.00	56.00	-0.0390	-41.09	0.0033	-3.4424	-0.0576	112.57	461	460	355	340	452	2068	447
151	Linear	2	4 mon	trmm 3b43	70.18	-21.23	21.23	0.0229	-39.81	0.0008	-1.6765	0.0288	83.66	341	289	288	303	316	1517	357
152	Linear	2	4 mon	ssmi	80.76	-24.69	24.69	-0.0249	-13.21	0.0019	-3.5906	-0.0441	98.90	425	328	335	335	421	1844	418
153	Linear	2	4 mon	opi2	58.07	5.27	5.27	0.0182	-13.73	0.0006	-1.4447	0.0235	71.16	190	142	279	306	186	1103	222
154	Linear	2	4 mon	opi3	60.67	3.60	3.60	0.0077	-13.96	0.0001	-1.7292	0.0102	75.18	219	93	305	317	217	1151	232
155	Linear	2	4 mon	cm3p	63.39	-3.53	3.53	0.0105	-13.99	0.0002	-2.0078	0.0150	78.93	250	87	297	314	269	1217	257
156	Linear	2	4 mon	cm2p	63.26	-3.08	3.08	0.0103	-13.99	0.0002	-2.0085	0.0148	78.94	249	71	298	315	270	1203	246
157	Point	1	Yes	reanalysis	75.89	-28.44	28.44	0.7582	-20.87	0.2825	0.1658	0.5315	87.57	392	364	22	39	352	1189	238
158	Point	1	Yes	gpcp	74.90	-6.16	6.16	0.5967	-1.49	0.1584	0.0821	0.3980	93.56	384	156	50	98	397	1085	214
159	Point	1	Yes	gpcp	70.37	3.22	3.22	0.7586	4.05	0.2083	0.1857	0.4564	88.57	343	76	23	67	357	866	135
160	Point	1	Yes	gpcp trmm	66.12	-1.20	1.20	0.8581	-0.68	0.2778	0.2701	0.5271	83.74	295	16	6	40	317	674	84
161	Point	1	Yes	ghcn cams grid	70.76	-13.04	13.04	0.8441	-10.63	0.2493	0.2222	0.4993	84.43	348	216	9	50	320	943	164
162	Point	1	Yes	gpi	78.53	-30.21	30.21	0.6716	-18.52	0.2826	0.1041	0.5124	92.45	409	386	36	46	393	1272	271
163	Point	1	Yes	trmm 3a46	69.08	-36.25	36.25	0.8335	-31.03	0.4172	0.2565	0.6459	82.36	336	425	11	19	310	1101	221
164	Point	1	Yes	trmm 3b43	69.15	-4.25	4.25	0.7783	-2.78	0.2448	0.2230	0.4947	85.68	337	118	18	53	336	862	133
165	Point	1	Yes	ssmi	73.96	-26.52	26.52	0.6945	-17.03	0.2772	0.1509	0.5265	90.68	380	348	34	41			

Eastern Brazil - Monthly Surface																				
OI	Runoff Meth	Re	Storage	Precipitation	Absolute Error	Bias	Abias	Slope	Intercept	R <sup>2</sup>	R <sup>2</sup> 1.1	Corr Coef	RMSE	AFR	BR	SR	CCR	RMR	TR	BR
203	Point	1	2 mon	trmm 3b43	50.71	-0.84	0.84	0.0493	4.02	0.0527	-19.5132	0.2296	80.29	144	8	215	188	103	658	82
204	Point	1	2 mon	ssmi	62.24	-24.99	24.99	0.0618	3.92	0.1375	-35.9032	0.3708	75.52	236	332	184	109	222	1083	211
205	Point	1	2 mon	opi2	39.38	6.35	6.35	0.0651	5.38	0.0936	-19.5389	0.3059	47.61	11	162	176	141	7	497	40
206	Point	1	2 mon	opi3	41.97	4.13	4.13	0.0541	5.25	0.0783	-23.9841	0.2797	52.51	30	108	200	161	22	521	47
207	Point	1	2 mon	cmcp	44.46	-3.44	3.44	0.0524	4.89	0.0861	-28.1715	0.2934	56.89	69	84	204	148	64	569	58
208	Point	1	2 mon	cmcp2	44.48	-2.99	2.99	0.0519	4.92	0.0846	-28.2762	0.2909	57.00	70	70	208	151	66	565	57
209	Point	1	3 mon	reanalysis	55.93	-28.25	28.25	-0.0347	4.14	0.0714	-73.8516	-0.2672	75.69	169	359	347	401	223	1499	355
210	Point	1	3 mon	gpcc	48.34	-5.61	5.61	0.0447	4.90	0.0753	-34.5201	0.2744	63.29	122	147	226	164	131	790	115
211	Point	1	3 mon	gpcc	45.45	4.09	4.09	0.0488	6.19	0.1159	-44.2212	0.3404	56.62	86	107	217	123	60	593	62
212	Point	1	3 mon	gpcc trmm	51.26	-0.29	0.29	0.0430	4.02	0.0386	-19.0280	0.1964	60.34	149	2	228	213	104	696	89
213	Point	1	3 mon	ghcn cams grid	43.27	-12.59	12.59	0.0351	2.39	0.0536	-42.4358	0.2314	56.79	50	210	246	186	62	754	108
214	Point	1	3 mon	gpi	58.51	-29.32	29.32	0.0507	4.17	0.1018	-41.7038	0.3191	77.57	196	373	211	136	252	1168	237
215	Point	1	3 mon	trmm 3a46	66.73	-31.05	31.05	0.0382	3.69	0.0349	-26.2200	0.1868	79.47	299	398	239	217	276	1429	331
216	Point	1	3 mon	trmm 3b43	50.95	-1.84	1.84	0.0379	3.90	0.0310	-19.9502	0.1759	61.24	146	32	240	221	113	752	107
217	Point	1	3 mon	ssmi	63.01	-25.31	25.31	0.0547	4.05	0.1084	-36.3382	0.3293	76.29	246	335	198	129	229	1137	227
218	Point	1	3 mon	opi2	39.44	6.20	6.20	0.0613	5.37	0.0825	-19.6413	0.2872	47.82	12	157	185	153	9	517	45
219	Point	1	3 mon	opi3	42.07	3.96	3.96	0.0495	5.25	0.0652	-24.1465	0.2553	52.78	32	102	214	175	24	547	53
220	Point	1	3 mon	cmcp	44.57	-3.84	3.84	0.0467	4.92	0.0674	-28.1819	0.2596	57.01	72	98	223	174	67	634	77
221	Point	1	3 mon	cmcp2	44.51	-3.39	3.39	0.0468	4.94	0.0679	-28.2423	0.2605	57.07	71	81	222	173	68	615	69
222	Point	1	4 mon	reanalysis	57.05	-28.34	28.34	-0.0426	4.40	0.1070	-74.5986	-0.3272	76.12	181	363	360	410	227	1541	362
223	Point	1	4 mon	gpcc	50.04	-5.58	5.58	0.0171	5.20	0.0109	-36.6020	0.1046	85.24	140	146	287	251	141	965	171
224	Point	1	4 mon	gpcc	47.98	3.98	3.98	0.0132	6.25	0.0085	-47.7118	0.0921	58.96	118	103	283	265	90	869	137
225	Point	1	4 mon	gpcc trmm	50.15	-0.81	0.81	0.0546	3.71	0.0831	-18.8417	0.2513	60.07	142	9	199	178	98	826	72
226	Point	1	4 mon	ghcn cams grid	44.16	-12.69	12.69	0.0148	2.70	0.0095	-44.1797	0.0972	57.94	62	211	282	259	76	900	141
227	Point	1	4 mon	gpi	60.38	-29.35	29.35	0.0168	5.35	0.0112	-44.3894	0.1057	80.16	216	374	289	248	282	1409	323
228	Point	1	4 mon	trmm 3a46	67.92	-33.83	33.83	0.0417	3.15	0.0405	-26.2168	0.2012	79.80	314	413	230	210	279	1446	339
229	Point	1	4 mon	trmm 3b43	49.65	-1.88	1.88	0.0491	3.61	0.0530	-19.8056	0.2301	61.03	135	36	216	187	108	682	85
230	Point	1	4 mon	ssmi	64.67	-25.31	25.31	0.0309	4.67	0.0350	-38.5790	0.1871	78.26	269	336	256	216	261	1338	299
231	Point	1	4 mon	opi2	41.13	5.99	5.99	0.0209	5.32	0.0096	-21.2944	0.0978	49.78	26	155	277	256	12	728	100
232	Point	1	4 mon	opi3	43.29	3.81	3.81	0.0203	5.28	0.0110	-25.6486	0.1047	54.42	52	97	278	249	32	708	94
233	Point	1	4 mon	cmcp	45.90	-3.93	3.93	0.0184	5.18	0.0104	-29.9271	0.1021	58.79	90	101	281	255	88	815	119
234	Point	1	4 mon	cmcp2	45.84	-3.48	3.48	0.0184	5.18	0.0105	-29.9940	0.1027	58.85	89	86	282	253	89	799	117
235	Point	2	Yes	reanalysis	54.40	-32.01	32.01	0.0481	-9.73	0.0380	-0.2132	0.5900	86.60	162	410	66	28	346	1012	187
236	Point	2	Yes	gpcc	49.74	18.05	18.05	0.6532	15.66	0.3555	0.2027	0.5962	70.21	137	254	40	26	174	631	75
237	Point	2	Yes	gpcc	59.69	51.05	51.05	0.8919	49.51	0.5074	0.1329	0.7123	78.47	208	454	4	14	265	945	166
238	Point	2	Yes	gpcc trmm	44.90	-12.91	12.91	0.3707	-24.16	0.2571	-0.5552	0.5071	60.29	82	214	74	47	102	519	46
239	Point	2	Yes	ghcn cams grid	43.18	15.02	15.02	0.7925	14.22	0.4523	0.3848	0.6725	81.67	49	228	15	17	115	424	26
240	Point	2	Yes	gpi	48.07	0.92	0.92	0.6393	7.05	0.4219	0.2875	0.6496	68.11	119	10	44	18	160	351	10
241	Point	2	Yes	trmm 3a46	56.51	-40.63	40.63	0.2942	-35.69	0.2298	-1.8001	0.4793	80.84	176	443	77	63	293	1052	203
242	Point	2	Yes	trmm 3b43	45.48	-16.72	16.72	0.3543	-25.80	0.2340	-0.6627	0.4837	62.34	87	250	75	61	122	595	63
243	Point	2	Yes	ssmi	53.38	4.33	4.33	0.5973	8.83	0.3544	0.1905	0.5954	73.48	155	121	49	27	204	556	55
244	Point	2	Yes	opi2	51.10	30.13	30.13	0.8465	27.22	0.3560	0.1975	0.5966	70.44	147	385	7	25	179	743	104
245	Point	2	Yes	opi3	49.67	28.27	28.27	0.7935	24.74	0.3712	0.2168	0.6092	69.58	136	361	14	22	171	704	93
246	Point	2	Yes	cmcp	47.44	21.05	21.05	0.7458	18.54	0.3830	0.2669	0.6189	67.32	108	287	25	20	155	595	64
247	Point	2	Yes	cmcp2	47.75	21.49	21.49	0.7417	18.83	0.3791	0.2585	0.6157	67.71	114	292	27	21	157	611	67
248	Point	2	No	reanalysis	82.29	-31.62	31.62	0.0540	9.23	0.0086	-2.9591	0.0929	110.89	431	408	201	264	448	1752	403
249	Point	2	No	gpcc	69.18	18.44	18.44	0.1423	12.55	0.0336	-1.2957	0.1833	84.44	338	257	110	218	321	1244	264
250	Point	2	No	gpcc	74.06	49.43	49.43	0.3120	39.67	0.1336	-1.2567	0.3655	86.29	381	449	76	112	344	1362	311
251	Point	2	No	gpcc trmm	80.60	-11.25	11.25	-0.2861	-34.23	0.2334	-4.5653	-0.4832	92.39	424	197	451	439	392	1903	424
252	Point	2	No	ghcn cams grid	63.43	15.41	15.41	0.2091	12.37	0.0627	-0.8104	0.2504	77.03	251	236	83	179	242	991	180
253	Point	2	No	gpi	74.81	0.68	0.68	0.1401	15.31	0.0380	-1.3958	0.1950	91.13	383	5	111	214	385	1088	220
254	Point	2	No	trmm 3a46	93.30	-37.71	37.71	-0.3179	-28.48	0.3751	-6.9235	-0.6125	115.00	455	427	456	455	457	2250	463
255	Point	2	No	trmm 3b43	80.35	-15.05	15.05	-0.3130	-33.52	0.2782	-4.7654	-0.5274	94.04	422	229	454	448	401	1954	432
256	Point	2	No	ssmi	79.74	4.58	4.58	0.0733	14.93	0.0099	-1.5738	0.0993	96.34	417	127	156	256	413	1369	315
257	Point	2	No	opi2	64.87	30.52	30.52	0.1793	14.97	0.0318	-0.9341	0.1783	77.50	271	388	91	219	251	1220	258
258	Point	2	No	opi3	65.17	28.66	28.66	0.1873	14.77	0.0412	-0.9982	0.2029	78.78	278	367	89	207	267	1208	248
259	Point	2	No	cmcp	65.78	21.44	21.44	0.1733	13.28	0.0412	-1.0435	0.2029	79.67	286	291	93	208	278	1156	233
260	Point	2	No	cmcp2	65.79	21.88	21.88	0.1725	13.35	0.0408	-1.0528	0.2020	79.85	287	295					

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Ol	Runoff Meth	Re	Storage	Precipitation	Absolute Error	Bias	Abias	Slope	Intercept	R <sup>2</sup>	R <sup>2</sup> 1:1	Corr Coef	RMSE	AER	BR	SR	CCR	RMR	TR	BR
304	Point	2	4 mon	ghcn cams grid	71.69	16.29	16.29	-0.0364	11.77	0.0019	-1.6075	-0.0434	90.49	356	247	350	334	372	1659	391
305	Point	2	4 mon	gpi	79.96	1.51	1.51	0.0503	17.00	0.0049	-1.7463	0.0701	98.03	419	22	212	276	419	1348	306
306	Point	2	4 mon	trmm 3a46	90.21	-41.72	41.72	-0.0768	-26.27	0.0237	-5.7554	-0.1538	102.17	450	444	390	377	433	2094	450
307	Point	2	4 mon	trmm 3b43	67.02	-7.21	7.21	-0.0405	-26.78	0.0048	-3.2151	-0.0695	76.88	303	170	358	344	240	1415	326
308	Point	2	4 mon	ssmi	83.86	0.69	0.69	-0.0349	12.28	0.0028	-2.4798	-0.0532	100.35	440	6	349	339	425	1558	368
309	Point	2	4 mon	opi2	68.59	31.30	31.30	-0.0346	11.26	0.0012	-1.3618	-0.0343	86.12	328	406	346	326	342	1748	402
310	Point	2	4 mon	opi3	71.08	29.63	29.63	-0.0488	11.07	0.0027	-1.5464	-0.0524	89.42	353	381	369	338	366	1807	410
311	Point	2	4 mon	cmmap	72.14	22.50	22.50	-0.0430	11.48	0.0025	-1.6257	-0.0499	90.80	362	306	362	337	380	1747	401
312	Point	2	4 mon	cmmap2	72.23	22.95	22.95	-0.0427	11.46	0.0025	-1.6326	-0.0496	90.92	363	317	361	336	381	1758	404
313	Area	1	Yes	reanalysis	75.37	-26.47	26.47	0.7978	-20.14	0.3032	0.2099	0.5507	86.56	387	347	13	34	345	1126	224
314	Area	1	Yes	gpcp	74.67	-4.64	4.64	0.6438	-0.51	0.1791	0.1220	0.4232	92.86	382	128	41	83	394	1028	197
315	Area	1	Yes	gpcp	70.74	2.51	2.51	0.7689	3.29	0.2132	0.1933	0.4618	88.56	347	54	20	66	356	843	125
316	Area	1	Yes	gpcp trmm	65.75	4.22	4.22	0.9733	4.32	0.3327	0.3307	0.5768	83.12	284	114	1	29	314	742	103
317	Area	1	Yes	ghcn cams grid	70.70	-11.07	11.07	0.8860	-9.31	0.2662	0.2488	0.5159	84.28	346	196	5	45	319	911	146
318	Area	1	Yes	gpi	78.23	-28.60	28.60	0.7091	-18.24	0.2832	0.1526	0.5322	91.42	406	366	33	36	388	1231	261
319	Area	1	Yes	trmm 3a46	65.45	-29.60	29.60	0.9410	-27.75	0.4843	0.3949	0.6959	77.85	280	380	2	15	256	933	157
320	Area	1	Yes	trmm 3b43	68.34	1.17	1.17	0.8939	1.88	0.3005	0.2961	0.5482	84.52	321	14	3	36	323	697	90
321	Area	1	Yes	ssmi	73.03	-24.44	24.44	0.7440	-16.48	0.3061	0.2105	0.5532	89.15	372	327	26	32	363	1120	223
322	Area	1	Yes	opi2	72.57	6.21	6.21	0.8394	6.03	0.1748	0.1645	0.4181	90.22	367	158	10	87	370	992	181
323	Area	1	Yes	opi3	73.04	4.07	4.07	0.7640	4.31	0.1757	0.1573	0.4192	90.61	373	106	21	86	374	960	169
324	Area	1	Yes	cmmap	73.64	-2.95	2.95	0.7342	-0.71	0.1913	0.1653	0.4374	90.36	379	65	30	74	371	919	150
325	Area	1	Yes	cmmap2	73.62	-2.50	2.50	0.7268	-0.32	0.1881	0.1609	0.4338	90.60	378	51	31	78	373	911	147
326	Area	1	No	reanalysis	54.08	-26.30	26.30	0.0687	2.85	0.3030	-65.2616	0.5505	68.26	159	345	164	35	162	865	134
327	Area	1	No	gpcp	46.72	-4.76	4.76	0.0641	6.09	0.1583	-33.7894	0.3979	61.90	98	134	181	100	118	631	74
328	Area	1	No	gpcp	44.14	2.27	2.27	0.0717	5.42	0.3710	-61.8820	0.6091	55.29	61	45	159	23	42	330	7
329	Area	1	No	gpcp trmm	44.36	6.64	6.64	0.1102	9.89	0.1850	-12.0701	0.4301	55.75	66	168	128	80	50	492	38
330	Area	1	No	ghcn cams grid	41.49	-10.43	10.43	0.0598	4.09	0.1627	-41.5420	0.4033	54.79	27	188	188	95	35	533	50
331	Area	1	No	gpi	57.05	-27.97	27.97	0.0587	5.53	0.1416	-42.1218	0.3763	76.29	180	355	190	106	230	1061	204
332	Area	1	No	trmm 3a46	59.18	-18.77	18.77	0.0998	9.44	0.2004	-17.4080	0.4476	70.77	198	260	133	72	183	846	127
333	Area	1	No	trmm 3b43	45.66	3.55	3.55	0.0927	9.59	0.1400	-13.3253	0.3742	57.93	88	88	138	108	75	497	41
334	Area	1	No	ssmi	61.43	-23.09	23.09	0.0583	6.19	0.1242	-35.7394	0.3524	74.85	227	321	191	115	215	1069	208
335	Area	1	No	opi2	37.81	7.86	7.86	0.0685	6.83	0.1877	-19.0846	0.4096	46.57	5	180	143	92	5	425	27
336	Area	1	No	opi3	40.85	5.72	5.72	0.0711	6.86	0.1374	-23.6090	0.3707	51.54	19	150	160	110	20	459	31
337	Area	1	No	cmmap	42.83	-1.65	1.65	0.0706	6.19	0.1598	-27.5420	0.3994	55.82	43	28	182	97	47	377	17
338	Area	1	No	cmmap2	42.84	-1.21	1.21	0.0702	6.22	0.1583	-27.6479	0.3979	55.72	44	17	163	99	49	372	15
339	Area	1	1 mon	reanalysis	53.58	-26.23	26.23	0.0374	3.82	0.0898	-69.2834	0.2966	70.35	157	343	244	144	177	1065	207
340	Area	1	1 mon	gpcp	46.75	-4.49	4.49	0.0646	6.10	0.1599	-33.5753	0.3999	61.82	99	125	179	96	116	615	70
341	Area	1	1 mon	gpcp	44.10	2.97	2.97	0.0655	5.43	0.3052	-62.0897	0.5524	55.32	60	88	172	33	43	374	16
342	Area	1	1 mon	gpcp trmm	44.88	5.88	5.88	0.1273	9.15	0.2772	-12.9026	0.5265	54.75	81	152	119	42	34	428	28
343	Area	1	1 mon	ghcn cams grid	41.04	-10.47	10.47	0.0689	3.93	0.2067	-41.6317	0.4547	54.41	22	189	168	68	31	478	35
344	Area	1	1 mon	gpi	55.97	-27.83	27.83	0.0770	4.88	0.2432	-40.3166	0.4931	74.81	170	350	152	55	214	941	162
345	Area	1	1 mon	trmm 3a46	57.98	-21.20	21.20	0.1265	7.54	0.3577	-18.5146	0.5981	69.32	189	288	120	24	169	790	116
346	Area	1	1 mon	trmm 3b43	44.83	4.24	4.24	0.1289	8.87	0.2954	-13.2693	0.5435	55.05	79	116	116	37	39	387	18
347	Area	1	1 mon	ssmi	59.93	-22.76	22.76	0.0819	5.49	0.2430	-33.6568	0.4930	73.08	211	312	149	56	201	929	154
348	Area	1	1 mon	opi2	37.87	7.79	7.79	0.0892	6.83	0.1781	-18.9559	0.4220	46.49	6	179	141	84	4	414	21
349	Area	1	1 mon	opi3	40.41	5.64	5.64	0.0741	6.85	0.1491	-23.4270	0.3862	51.44	16	149	155	104	19	443	30
350	Area	1	1 mon	cmmap	42.79	-1.86	1.86	0.0726	6.15	0.1684	-27.3237	0.4104	55.50	41	34	157	91	45	368	14
351	Area	1	1 mon	cmmap2	42.81	-1.41	1.41	0.0720	6.19	0.1659	-27.4453	0.4074	55.62	42	20	158	93	48	361	13
352	Area	1	2 mon	reanalysis	53.95	-26.20	26.20	0.0034	4.90	0.0007	-73.6845	0.0270	72.56	158	342	315	305	191	1311	291
353	Area	1	2 mon	gpcp	47.76	-4.16	4.16	0.0553	6.23	0.1167	-34.0377	0.3416	62.34	115	111	196	122	123	667	83
354	Area	1	2 mon	gpcp	44.79	3.56	3.56	0.0509	5.48	0.1824	-63.4113	0.4271	56.08	76	89	210	81	53	509	43
355	Area	1	2 mon	gpcp trmm	48.35	4.94	4.94	0.0828	9.07	0.1173	-14.3811	0.3425	57.83	123	137	146	120	74	600	65
356	Area	1	2 mon	ghcn cams grid	41.85	-10.53	10.53	0.0548	4.09	0.1401	-43.1761	0.3743	55.16	29	190	197	107	40	563	56
357	Area	1	2 mon	gpi	56.48	-27.41	27.41	0.0682	5.20	0.1904	-40.8579	0.4363	75.49	175	349	165	76	221	988	178
358	Area	1	2 mon	trmm 3a46	60.16	-23.42	23.42	0.1066	7.77	0.2482	-19.3628	0.4982	71.42	214	323	130	51	187	905	144
359	Area	1	2 mon	trmm 3b43	48.75	4.25	4.25	0.0833	8.94	0.1242	-15.0213	0.3525	58.56	130	117	145	114	83	589	61
360	Area	1	2 mon	ssmi	60.41	-22.85	22.85	0.0819	5.44	0.2436	-33.7314	0.4936	73.30	217	314	148	54	203	936	159
361	Area	1	2 mon	opi2	39.24	7.78	7.78	0.0711	6.81	0.1131	-19.7851	0.3363	47.50	10	178	161	125	6	480	36
362	Area	1	2 mon	opi3	41.98	5.56	5.56	0.0580	6.87	0.0914	-24.2752	0.3023	52.41	31	145	182	143	21	532	49
363	Area	1	2 mon	cmmap	44.34	-2.00	2.00	0.0562	6.29	0.1008	-28.3299	0.3176	56.57	65	39	194	137	59	484	39
364	Area	1	2 mon	cmmap2	44.34	-1.55	1.55	0.0559	6.32	0.0998	-28.4372	0.3160	56.68	64	25	195	138	61	483	37
365	Area	1	3 mon	reanalysis	56.05	-26.27	26.27	-0.0284	5.90	0.0517	-77.7574	-0.2274	74.55	171	344	339	397	210	1461	345
366	Area	1	3 mon	gpcp	48.68	-4.17	4.17	0.0378	6.42	0.0548	-35.3738	0.2336	83.63	129	112	242	185	132	800	118
367	Area	1	3 mon	gpcp	46.08	3.31	3.31	0.0338	5.45	0.0809	-66.1861	0.2843	57.28	93	79	250	157	69	648	80
368	Area	1	3 mon	gpcp trmm	49.98	4.51	4.51	0.0642	8.72	0.0740	-15.7464	0.2721	59.41	138	126	180	166	91	701	91

Eastern Brazil - Monthly Surface																											
ID	Runoff Meth	Re	Storage	Precipitation	Absolute Error	Bias	Abias	Slope	Intercept	R <sup>2</sup>	R <sup>2</sup> 1.1	Corr Coef	RMSE	AER	BR	SR	CCR	RMR	TR	BR							
405	Area	2	No	gpcc	60.54	18.63	18.63	0.1442	12.75	0.1117	-4.1808	0.3342	70.51	218	258	108	126	180	890	140							
406	Area	2	No	gpcc	63.20	38.31	38.31	0.2271	27.34	0.2157	-3.6395	0.4644	70.86	248	428	82	65	185	1008	186							
407	Area	2	No	gpcc trmm	68.32	7.38	7.38	-0.1006	-12.28	0.1378	-16.5286	-0.3712	75.02	320	172	407	423	216	1538	359							
408	Area	2	No	ghcn cams grid	54.38	15.61	15.61	0.1878	12.49	0.1636	-3.1507	0.4045	63.11	161	239	88	94	128	710	95							
409	Area	2	No	gpi	64.59	-2.27	2.27	0.1231	12.64	0.0959	-4.7778	0.3097	78.31	268	46	124	139	262	839	123							
410	Area	2	No	trmm 3a46	79.09	-17.85	17.85	-0.1266	-9.95	0.2806	-22.8490	-0.5297	91.85	415	253	429	449	391	1937	429							
411	Area	2	No	trmm 3b43	68.01	3.58	3.58	-0.1157	-12.11	0.1815	-16.7441	-0.4261	75.48	315	92	423	435	220	1485	352							
412	Area	2	No	ssmi	68.40	2.46	2.46	0.0791	12.75	0.0388	-5.2281	0.1969	81.56	322	49	150	212	299	1032	200							
413	Area	2	No	opi2	55.32	30.71	30.71	0.1791	15.16	0.1026	-3.0373	0.3204	62.24	164	390	92	135	120	901	142							
414	Area	2	No	opi3	56.19	28.86	28.86	0.1795	14.83	0.1224	-3.3017	0.3498	64.25	174	368	90	118	135	885	139							
415	Area	2	No	cmap	56.81	21.63	21.63	0.1667	13.41	0.1233	-3.4451	0.3511	65.31	177	293	98	116	142	826	121							
416	Area	2	No	cmap2	56.92	22.08	22.08	0.1664	13.48	0.1229	-3.4706	0.3506	65.50	179	298	99	117	144	837	122							
417	Area	2	1 mon	reanalysis	76.17	-31.19	31.19	0.0657	8.91	0.0413	-9.3058	0.2031	99.61	395	403	170	206	424	1598	380							
418	Area	2	1 mon	gpcc	60.90	18.96	18.96	0.1141	12.56	0.0695	-4.4881	0.2635	72.69	222	263	127	171	193	976	175							
419	Area	2	1 mon	gpcc	61.08	38.85	38.85	0.2579	27.80	0.2752	-3.3908	0.5246	69.09	224	430	80	44	167	945	165							
420	Area	2	1 mon	gpcc trmm	67.33	10.58	10.58	-0.1691	-13.64	0.3621	-17.2978	-0.6018	76.75	307	191	437	452	238	1625	385							
421	Area	2	1 mon	ghcn cams grid	55.56	15.91	15.91	0.1504	12.36	0.1042	-3.4855	0.3228	65.71	167	241	103	131	145	787	114							
422	Area	2	1 mon	gpi	63.63	-2.03	2.03	0.1465	12.14	0.1358	-4.4810	0.3685	76.30	254	40	106	111	231	742	102							
423	Area	2	1 mon	trmm 3a46	79.53	-19.49	19.49	-0.1547	-9.05	0.4146	-23.7640	-0.6439	93.85	416	273	435	460	400	1984	440							
424	Area	2	1 mon	trmm 3b43	66.92	6.44	6.44	-0.1764	-13.06	0.4010	-17.5541	-0.6332	77.29	301	164	439	458	246	1608	382							
425	Area	2	1 mon	ssmi	67.62	1.76	1.76	0.0942	12.15	0.0560	-5.1321	0.2367	80.36	310	31	137	184	286	948	167							
426	Area	2	1 mon	opi2	55.40	30.93	30.93	0.1729	15.05	0.0952	-3.0762	0.3086	62.64	166	396	94	140	124	920	151							
427	Area	2	1 mon	opi3	57.15	29.22	29.22	0.1463	14.29	0.0803	-3.5423	0.2834	66.13	183	371	107	159	147	967	172							
428	Area	2	1 mon	cmap	57.33	22.00	22.00	0.1356	13.12	0.0808	-3.7032	0.2842	67.29	185	297	113	158	154	907	145							
429	Area	2	1 mon	cmap2	57.41	22.45	22.45	0.1356	13.18	0.0808	-3.7265	0.2843	67.46	186	305	112	156	156	915	148							
430	Area	2	2 mon	reanalysis	78.39	-30.84	30.84	0.0464	9.63	0.0205	-9.6405	0.1431	101.13	408	394	224	233	428	1687	393							
431	Area	2	2 mon	gpcc	63.43	18.21	18.21	0.0859	12.25	0.0393	-4.7864	0.1982	74.58	252	264	144	211	211	1082	210							
432	Area	2	2 mon	gpcc	62.73	39.61	39.61	0.2440	27.82	0.2429	-3.5220	0.4928	70.37	244	440	81	57	178	1000	183							
433	Area	2	2 mon	gpcc trmm	64.00	13.77	13.77	-0.1225	-12.34	0.1872	-16.1463	-0.4327	72.86	260	218	427	436	196	1537	358							
434	Area	2	2 mon	ghcn cams grid	58.28	16.07	16.07	0.0990	12.04	0.0452	-3.9637	0.2126	69.07	193	244	134	197	166	834	158							
435	Area	2	2 mon	gpi	63.95	-1.62	1.62	0.1318	12.44	0.1095	-4.6418	0.3309	77.61	259	27	115	127	253	781	112							
436	Area	2	2 mon	trmm 3a46	78.76	-20.83	20.83	-0.1222	-8.39	0.2673	-23.5344	-0.5170	91.74	413	285	428	445	390	1959	433							
437	Area	2	2 mon	trmm 3b43	64.28	9.49	9.49	-0.1156	-11.68	0.1710	-16.0545	-0.4135	72.66	263	181	422	433	192	1491	353							
438	Area	2	2 mon	ssmi	68.91	1.09	1.09	0.0617	11.40	0.0291	-6.7218	0.1707	81.92	335	13	185	225	305	1063	206							
439	Area	2	2 mon	opi2	58.50	31.02	31.02	0.1151	13.83	0.0422	-3.4536	0.2054	65.43	195	397	125	202	143	1062	205							
440	Area	2	2 mon	opi3	59.29	29.37	29.37	0.1050	13.46	0.0413	-3.8596	0.2033	68.35	201	375	131	205	163	1075	209							
441	Area	2	2 mon	cmap	59.56	22.23	22.23	0.0989	12.65	0.0428	-4.0261	0.2068	69.51	205	299	135	199	170	1008	185							
442	Area	2	2 mon	cmap2	59.66	22.67	22.67	0.0988	12.69	0.0428	-4.0501	0.2068	69.67	207	309	136	200	173	1025	195							
443	Area	2	3 mon	reanalysis	82.87	-30.82	30.82	0.0175	10.91	0.0029	-10.1960	0.0539	103.88	436	393	285	287	436	1837	418							
444	Area	2	3 mon	gpcc	66.52	19.31	19.31	0.0171	11.78	0.0015	-5.5262	0.0394	79.31	298	267	286	295	275	1421	328							
445	Area	2	3 mon	gpcc	66.97	39.60	39.60	0.1255	26.18	0.0645	-4.5015	0.2541	77.65	302	439	123	176	254	1294	286							
446	Area	2	3 mon	gpcc trmm	62.35	15.19	15.19	-0.0722	-10.14	0.0774	-17.8632	-0.2783	70.77	240	232	386	403	182	1443	338							
447	Area	2	3 mon	ghcn cams grid	61.62	16.13	16.13	0.0378	11.82	0.0066	-4.5336	0.0812	73.03	229	245	241	270	198	1183	242							
448	Area	2	3 mon	gpi	68.12	-1.44	1.44	0.0883	13.26	0.0492	-5.2016	0.2218	81.49	316	21	142	191	298	968	173							
449	Area	2	3 mon	trmm 3a46	78.30	-20.57	20.57	-0.0720	-7.70	0.1101	-25.7256	-0.3318	88.75	407	281	383	413	360	1844	417							
450	Area	2	3 mon	trmm 3b43	62.52	10.95	10.95	-0.0708	-9.81	0.0763	-17.8421	-0.2762	70.73	242	195	381	402	181	1401	321							
451	Area	2	3 mon	ssmi	71.78	1.00	1.00	0.0254	11.46	0.0051	-7.4756	0.0712	84.90	358	11	265	275	325	1234	262							
452	Area	2	3 mon	opi2	60.11	31.09	31.09	0.0384	12.40	0.0047	-3.9470	0.0685	69.05	213	401	238	277	165	1294	285							
453	Area	2	3 mon	opi3	61.70	29.49	29.49	0.0315	12.22	0.0037	-4.4175	0.0611	72.26	230	378	252	280	190	1330	298							
454	Area	2	3 mon	cmap	62.28	22.34	22.34	0.0312	11.99	0.0043	-4.6258	0.0652	73.63	237	303	254	279	207	1280	274							
455	Area	2	3 mon	cmap2	62.39	22.78	22.78	0.0313	12.00	0.0043	-4.6486	0.0656	73.78	241	313	253	278	208	1293	283							
456	Area	2	4 mon	reanalysis	87.23	-31.05	31.05	-0.0293	12.94	0.0082	-11.0819	-0.0904	108.08	446	399	341	350	446	1982	438							
457	Area	2	4 mon	gpcc	69.24	19.24	19.24	-0.0532	11.29	0.0151	-6.2690	-0.1228	83.84	339	265	374	363	318	1659	390							
458	Area	2	4 mon	gpcc	72.37	39.39	39.39	-0.0206	24.16	0.0017	-5.8822	-0.0417	85.69	365	437	332	333	337	1804	409							
459	Area	2	4 mon	gpcc trmm	59.25	14.46	14.46	-0.0189	-9.23	0.0054	-16.4968	-0.0734	68.21	199	220	330	346	161	1256	266							
460	Area	2	4 mon	ghcn cams grid	64.88	16.05	16.05	-0.0473	11.49	0.0103	-5.3126	-0.1015	78.13	272	24												

Eastern Brazil - Yearly Surface																												
OI	Runoff Meth	Re	Storage	Precipitation	Absolute Error	Bias	Abias	Slope	Oslope	Intercept	R <sup>2</sup>	R <sup>2</sup> 1:1	Corr Coef	RMSE	AER	BR	SR	CCR	RMR	TR	BR	Plot						
1	Linear	1	Yes	reanalysis	37.02	-34.45	34.45	-0.0942		-0.20	0.0288	-11.4648	-0.1696	44.09	131	137	139	129	136	672	145	Plot						
2	Linear	1	Yes	gpcp	16.39	-15.48	15.48	0.0803	12.4533	-4.82	0.0098	-3.2482	0.0988	22.70	57	75	76	90	59	357	72							
3	Linear	1	Yes	gpcp	8.60	-1.80	1.80	0.1530	6.5359	1.08	0.1476	-4.5376	0.3841	10.55	3	12	69	23	2	109	4							
4	Linear	1	Yes	gpcp trmm	24.23	-24.23	24.23	0.0475	21.0526	-18.13	0.0237	-49.7181	0.1540	27.21	75	101	89	76	68	409	96							
5	Linear	1	Yes	ghcn cams grid	21.09	-18.85	18.85	0.2015	4.9628	-6.32	0.0475	-2.9353	0.2180	25.00	64	84	66	63	62	339	64							
6	Linear	1	Yes	gpi	40.24	-40.24	40.24	0.2790	3.5842	-14.58	0.1175	-16.6033	0.3427	42.30	142	145	54	31	130	502	127							
7	Linear	1	Yes	trmm 3a46	57.42	-57.42	57.42	-0.2150		-11.54	0.5676	-570.6146	-0.7534	58.38	152	154	144	151	151	752	154							
8	Linear	1	Yes	trmm 3b43	23.41	-21.45	21.45	-0.2295		-13.27	0.2061	-15.2074	-0.4540	28.02	74	92	146	146	71	529	133							
9	Linear	1	Yes	ssmi	37.97	-37.97	37.97	-0.1532		-1.67	0.0335	-15.4405	-0.1831	41.79	135	142	142	131	127	677	146							
10	Linear	1	Yes	opi2	13.85	-3.40	3.40	-0.0262		-3.90	0.0008	-1.3670	-0.0288	16.94	44	27	113	106	43	333	57							
11	Linear	1	Yes	opi3	13.27	-5.28	5.28	0.0762	13.1234	-3.99	0.0084	-1.4511	0.0915	17.24	41	46	78	92	44	301	40							
12	Linear	1	Yes	cmmap	15.58	-12.48	12.48	0.0691	14.4718	-4.48	0.0075	-2.6391	0.0867	21.01	52	70	81	94	56	353	70							
13	Linear	1	Yes	cmmap2	15.46	-12.03	12.03	0.0689	14.5138	-4.45	0.0075	-2.5655	0.0868	20.79	50	68	83	93	55	349	69							
14	Linear	1	No	reanalysis	36.63	-35.21	35.21	-0.0846						43.05	129	138	136	142	133	678	147							
15	Linear	1	No	gpcp	17.95	-15.23	15.23	-0.0496		-3.06	0.0140	-13.4617	-0.1184	21.59	58	73	124	119	57	431	102							
16	Linear	1	No	gpcp	9.35	-3.11	3.11	-0.0097		0.31	0.0047	-53.7817	-0.0682	11.89	7	24	107	111	9	258	23							
17	Linear	1	No	gpcp trmm	20.16	-20.16	20.16	-0.0367		-13.52	0.0225	-61.9974	-0.1498	24.08	63	90	121	126	61	461	114							
18	Linear	1	No	ghcn cams grid	21.18	-19.62	19.62	0.0411	24.3309	-4.58	0.0137	-24.2803	0.1171	24.05	65	89	91	88	60	393	89							
19	Linear	1	No	gpi	39.57	-39.57	39.57	0.2194	4.5579	-11.78	0.1724	-38.5907	0.4152	41.16	140	144	61	19	125	489	122							
20	Linear	1	No	trmm 3a46	53.23	-53.23	53.23	-0.0622		-13.12	0.6310	-999.0000	-0.7943	54.01	150	150	131	153	148	732	151							
21	Linear	1	No	trmm 3b43	20.15	-18.19	18.19	-0.2150		-10.10	0.3242	-22.2722	-0.5693	25.08	62	80	145	147	64	498	125							
22	Linear	1	No	ssmi	36.83	-36.83	36.83	-0.0649		-3.31	0.0136	-32.5815	-0.1167	39.68	130	140	133	118	121	642	142							
23	Linear	1	No	opi2	11.50	-3.15	3.15	-0.0535		-3.66	0.0130	-5.3405	-0.1140	14.29	26	25	125	117	17	310	43							
24	Linear	1	No	opi3	11.76	-5.00	5.00	-0.0344		-3.59	0.0064	-6.5806	-0.0802	15.63	31	42	116	112	36	337	60							
25	Linear	1	No	cmmap	16.03	-12.22	12.22	-0.0363		-3.32	0.0078	-10.9838	-0.0883	19.65	54	69	119	113	53	408	95							
26	Linear	1	No	cmmap2	15.77	-11.78	11.78	-0.0402		-3.31	0.0096	-10.7712	-0.0982	19.48	53	66	122	115	52	408	94							
27	Linear	2	Yes	reanalysis	59.08	-56.53	56.53	-0.3148		0.26	0.0200	-2.6980	-0.1413	70.62	154	152	148	124	156	734	152	Plot						
28	Linear	2	Yes	gpcp	30.62	-6.47	6.47	0.7036	1.4213	-8.51	0.0871	0.0406	0.2952	35.97	109	53	21	10	110	337	61							
29	Linear	2	Yes	gpcp	24.33	24.30	24.30	1.0412	1.0412	24.89	0.4233	-0.8961	0.6506	29.14	78	103	4	7	79	271	27							
30	Linear	2	Yes	gpcp trmm	32.56	-32.56	32.56	0.6216	1.6088	-39.32	0.3911	-9.6895	0.6254	33.77	120	135	31	10	95	991	88							
31	Linear	2	Yes	ghcn cams grid	30.41	-9.50	9.50	0.6711	1.4901	-10.76	0.0807	-0.0055	0.2842	36.82	105	61	26	48	113	353	71							
32	Linear	2	Yes	gpi	28.24	-24.43	24.43	1.6480	1.6480	-35.42	0.5505	-0.1107	0.7420	33.92	89	105	32	3	97	326	54							
33	Linear	2	Yes	trmm 3a46	67.94	-67.94	67.94	-0.5033		-47.96	0.1767	-65.6011	-0.4203	69.21	156	156	153	144	155	764	156							
34	Linear	2	Yes	trmm 3b43	36.37	-36.37	36.37	0.4274	2.3397	-44.42	0.2113	-12.5626	0.4597	38.04	127	139	38	16	118	438	104							
35	Linear	2	Yes	ssmi	29.24	-19.06	19.06	0.6662	1.5011	-15.24	0.0573	-0.2249	0.2395	40.75	95	85	27	58	123	388	86							
36	Linear	2	Yes	opi2	30.47	5.61	5.61	0.1188	1.0188	5.96	0.1389	0.1155	0.3727	34.54	106	49	2	26	101	284	33							
37	Linear	2	Yes	opi3	30.85	3.75	3.75	0.7768	1.2873	-0.06	0.1005	0.0818	0.3171	35.19	111	34	13	35	103	296	38							
38	Linear	2	Yes	cmmap	30.56	-3.47	3.47	0.7148	1.3990	-6.28	0.0913	0.0678	0.3022	35.45	108	28	20	42	106	304	42							
39	Linear	2	Yes	cmmap2	30.50	-3.03	3.03	0.7320	1.3661	-5.79	0.0961	0.0764	0.3100	35.29	107	21	18	38	104	288	36							
40	Linear	2	No	reanalysis	57.68	-57.37	57.37	-0.6519		13.98	0.1635	-5.5504	-0.4044	67.99	153	153	154	143	154	757	155							
41	Linear	2	No	gpcp	22.83	-7.31	7.31	0.2511	3.9825	-12.45	0.0212	-0.2432	0.1457	29.62	70	58	57	80	82	347	67							
42	Linear	2	No	gpcp	20.04	18.84	18.84	0.2790	3.5842	8.61	0.0364	-1.1565	0.1908	28.39	61	83	53	69	73	339	65							
43	Linear	2	No	gpcp trmm	28.26	-28.26	28.26	0.3166	3.1586	-40.48	0.3127	-24.2105	0.5592	29.55	90	115	48	14	81	348	68							
44	Linear	2	No	ghcn cams grid	23.08	-10.34	10.34	0.3956	2.5278	-12.66	0.0536	-0.2229	0.2316	29.38	72	63	40	62	80	317	49							
45	Linear	2	No	gpi	30.90	-26.19	26.19	1.1764	1.1764	-29.19	0.3937	-0.5424	0.6275	33.78	112	110	8	9	96	335	59							
46	Linear	2	No	trmm 3a46	62.21	-62.21	62.21	0.3362	2.9744	-53.37	0.3663	-248.1002	0.6053	62.47	155	155	45	11	153	519	131							
47	Linear	2	No	trmm 3b43	32.07	-32.07	32.07	0.2399	4.1684	-42.76	0.2050	-31.5530	0.4528	33.58	119	134	59	17	94	423	98							
48	Linear	2	No	ssmi	30.21	-25.26	25.26	0.7003	1.4280	-21.84	0.1087	-0.7186	0.3297	36.86	102	106	22	33	114	377	81							
49	Linear	2	No	opi2	21.90	4.77	4.77	0.4708	2.1240	-5.26	0.0567	-0.0472	0.2382	27.18	67	40	37	59	67	270	25							
50	Linear	2	No	opi3	21.83	2.91	2.9																					

Eastern Brazil - Yearly Surface																													
Off	Runoff Meth	Re	Storage	Precipitation	Absolute Error	Bias	Abias	Slope	Oslope	Intercept	R <sup>2</sup>	R <sup>2</sup> 1:1	Corr Coef	RMSE	AER	BR	SR	CCR	RMR	TR	BR								
102	Point	2	No	opi3	34.19	28.66	28.66	0.3258	3.0713	17.13	0.0219	-0.8265	0.1479	44.59	123	118	47	78	139	505	129								
103	Point	2	No	cmmap	31.72	21.44	21.44	0.2829	3.5348	14.36	0.0177	-0.5183	0.1331	40.66	117	91	52	83	122	465	118								
104	Point	2	No	cmmap2	31.99	21.88	21.88	0.2888	3.4626	14.55	0.0185	-0.5337	0.1361	40.86	118	94	50	82	124	468	118								
105	Area	1	Yes	reanalysis	30.00	-25.54	25.54	0.0343	29.1545	4.88	0.0047	-8.9281	0.0687	35.35	99	107	94	97	105	502	126								
106	Area	1	Yes	gpcc	8.85	-5.01	5.01	0.0941	10.6270	5.49	0.0276	-2.9548	0.1660	15.27	8	43	75	74	31	231	18								
107	Area	1	Yes	gpcc	8.99	3.59	3.59	0.1762	5.6754	6.38	0.1697	-4.0923	0.4119	10.87	5	32	67	20	4	128	6								
108	Area	1	Yes	gpcc trmm	10.68	1.63	1.63	-0.0063		8.07	0.0009	-24.1308	-0.0305	12.82	14	10	106	107	11	248	21								
109	Area	1	Yes	ghcn cams grid	14.94	-9.87	9.87	0.1449	6.9013	3.54	0.0305	-1.7936	0.1747	18.91	47	62	71	72	48	300	39								
110	Area	1	Yes	gpi	28.88	-28.65	28.65	-0.0160		7.52	0.0022	-54.6015	-0.0467	31.57	93	117	110	108	89	517	130								
111	Area	1	Yes	trmm 3a46	30.33	-30.33	30.33	-0.2867		18.25	0.8280	-142.4420	-0.9099	32.29	104	125	147	155	91	622	141								
112	Area	1	Yes	trmm 3b43	10.58	0.29	0.29	0.0570	17.5439	6.57	0.0621	-16.9596	0.2491	13.34	11	1	85	54	12	163	9								
113	Area	1	Yes	ssmi	24.87	-24.43	24.43	-0.0720		9.31	0.0780	-76.4978	-0.2793	27.94	81	104	134	134	70	523	132								
114	Area	1	Yes	opi2	13.34	7.07	7.07	0.0324	30.8642	6.60	0.0026	-3.1710	0.0510	15.68	42	57	95	98	38	330	56								
115	Area	1	Yes	opi3	12.80	5.21	5.21	0.0699	14.3062	6.48	0.0145	-3.0126	0.1205	15.38	39	45	80	87	33	284	32								
116	Area	1	Yes	cmmap	10.81	-2.01	2.01	0.0690	14.4928	5.99	0.0154	-2.8554	0.1240	15.08	16	14	82	86	25	223	16								
117	Area	1	Yes	cmmap2	10.75	-1.57	1.57	0.0763	13.1062	5.96	0.0190	-2.8121	0.1379	14.99	15	9	77	81	24	206	12								
118	Area	1	No	reanalysis	28.80	-26.30	26.30	0.0439	22.7790	3.62	0.0984	-116.3603	0.3136	34.11	92	111	90	36	98	427	101								
119	Area	1	No	gpcc	11.63	-4.76	4.76	-0.0358		7.25	0.0189	-17.6465	-0.1375	15.23	29	39	118	122	30	338	62								
120	Area	1	No	gpcc	8.23	2.27	2.27	0.0135	74.0741	5.61	0.0174	-96.1297	0.1318	11.39	2	15	101	85	6	209	13								
121	Area	1	No	gpcc trmm	13.77	5.70	5.70	-0.0906		12.68	0.5488	-93.2181	-0.7408	14.69	43	50	138	150	20	401	92								
122	Area	1	No	ghcn cams grid	14.97	-10.65	10.65	-0.0155		5.28	0.0045	-30.5450	-0.0669	17.75	48	64	109	110	45	376	80								
123	Area	1	No	gpi	28.57	-27.97	27.97	-0.0756		10.32	0.0604	-66.0116	-0.2457	31.20	91	114	135	133	88	561	138								
124	Area	1	No	trmm 3a46	26.14	-26.14	26.14	-0.1339		16.67	0.8213	-367.4412	-0.7882	27.90	85	109	141	152	69	556	136								
125	Area	1	No	trmm 3b43	12.21	3.55	3.55	0.0715	13.9860	9.73	0.0598	-10.8028	0.2444	13.83	34	30	79	55	15	213	14								
126	Area	1	No	ssmi	24.31	-23.29	23.29	0.0183	61.3497	7.67	0.0023	-39.8234	0.0484	26.58	77	100	98	99	66	440	105								
127	Area	1	No	opi2	12.32	7.32	7.32	0.0051	198.0784	6.84	0.0003	-15.9698	0.0174	14.53	37	59	102	102	19	319	50								
128	Area	1	No	opi3	12.44	5.47	5.47	-0.0406		6.89	0.0232	-17.8118	-0.1524	15.22	38	47	123	127	29	364	77								
129	Area	1	No	cmmap	11.44	-1.76	1.76	-0.0365		7.15	0.0204	-18.6936	-0.1427	14.84	24	11	120	125	23	303	41								
130	Area	1	No	cmmap2	11.51	-1.31	1.31	-0.0328		7.10	0.0186	-18.6543	-0.1290	14.82	27	8	115	120	22	292	37								
131	Area	2	Yes	reanalysis	38.21	-30.59	30.59	0.0048	208.3333	12.39	0.0000	-1.2378	0.0026	46.67	136	127	103	104	143	613	139								
132	Area	2	Yes	gpcc	30.21	19.47	19.47	0.6576	1.5207	17.12	0.1055	-0.3127	0.3248	35.74	100	88	28	34	108	358	74								
133	Area	2	Yes	gpcc	43.77	43.77	43.77	1.0956	1.0956	45.12	0.5245	-4.2668	0.7242	45.91	146	147	6	5	142	446	109								
134	Area	2	Yes	gpcc trmm	8.21	3.09	3.09	0.3410	2.9326	-8.69	0.2737	-0.9567	0.5232	9.47	1	23	44	15	1	84	1								
135	Area	2	Yes	ghcn cams grid	28.88	16.45	16.45	0.6237	1.6033	15.00	0.0966	-0.2165	0.3108	34.41	94	79	30	37	99	339	66								
136	Area	2	Yes	gpi	14.66	-0.51	0.51	1.2972	1.2972	-5.57	0.5278	0.4997	0.7265	18.32	46	2	14	4	47	113	5								
137	Area	2	Yes	trmm 3a46	29.99	-29.99	29.99	-0.6912		-7.46	0.5170	-21.7497	-0.7191	32.67	98	123	156	149	93	619	140								
138	Area	2	Yes	trmm 3b43	10.34	-0.72	0.72	0.2181	4.5851	-11.71	0.1280	-1.5279	0.3578	10.77	9	4	62	28	3	106	3								
139	Area	2	Yes	ssmi	25.20	6.95	6.95	0.5279	1.8943	12.36	0.0564	-0.0445	0.2375	30.08	83	56	34	60	83	316	48								
140	Area	2	Yes	opi2	35.32	31.55	31.55	0.9184	1.0889	30.00	0.1565	-0.8677	0.3956	42.63	125	132	5	22	131	415	97								
141	Area	2	Yes	opi3	34.29	29.69	29.69	0.7243	1.3806	24.98	0.1211	-0.8025	0.3480	41.88	124	122	19	29	129	423	99								
142	Area	2	Yes	cmmap	31.31	22.47	22.47	0.6771	1.4769	19.29	0.1135	-0.4313	0.3369	37.32	115	97	25	32	116	385	83								
143	Area	2	Yes	cmmap2	31.40	22.92	22.92	0.6905	1.4482	19.72	0.1185	-0.4449	0.3443	37.50	116	99	23	30	117	385	84								
144	Area	2	No	reanalysis	33.78	-31.43	31.43	-0.3323		26.12	0.0958	-4.5991	-0.3095	41.86	122	131	149	136	128	666	144								
145	Area	2	No	gpcc	23.38	18.63	18.63	0.2052	4.8733	13.17	0.0319	-1.5562	0.1787	28.29	73	82	65	71	72	363	76								
146	Area	2	No	gpcc	38.31	38.31	38.31	0.3334	2.9994	28.85	0.0942	-7.3925	0.3068	41.62	137	143	46	40	126	492	124								
147	Area	2	No	gpcc trmm	11.98	7.38	7.38	0.0360	27.7778	-9.84	0.0958	-105.7981	0.3095	12.50	33	60	93	39	10	235	19								
148	Area	2	No	ghcn cams grid	19.08	15.61	15.61	0.3482	2.8719	13.10	0.0936	-1.0127	0.3060	25.10	59	77	43	41	65	285	34								
149	Area	2	No	gpi	11.42	-2.27	2.27	0.8276	1.2083	0.66	0.3981	0.3666	0.6309	15.15	22	16	10	8	28	84	2	Plot							
150	Area	2	No	trmm 3a46	24.26	-24.26	24.26	0.1453	6.8823	-12.88	0.9583	-558.4187	0.9789	25.02	76	102	70	1	63	312	46								
151	Area	2	No	trmm 3b43	10.93	3.58	3.58	0.0306	32.6797	-10.06	0.0789	-87.9688	0.2809	11.41	17	31	96	50	7	201	11								
152	Area	2	No	ssmi	13.88	0.75	0.75	0.5820	1.7794	5.76	0.1409	0.0539	0.3754	19.28	45	6	33	25	50	159	8								
153	Area	2	No	opi2	30.78	30.71	30.71	0.3706	2.6983	18.79	0.0792	-3.1627	0.2815	36.10	110	128	42	49	111	440	106								
154	Area	2	No	opi3	29.41	28.86	28.86	0.2385	4.1929	15.84	0.0408	-3.0354	0.2020	35.54	96	119	60	65	107	447	110								
155	Area	2	No	cmmap	24.67	21.63	21.63	0.2151	4.6490	13.89	0.0356	-1.9339	0.1887	30.30	80	93	64	70	84	391	87								
156	Area	2	No	cmmap2	24.97	22.08	22.08	0.2178	4.5914	14.01	0.0367	-1.9932	0.1915	30.61	82	95	63	68	86										

Larger Rondonia - Monthly Atmospheric																							
Ol	Moist Flux	Re	Storage	Precipitation	Absolute Error	Bias	Abias	Slope	Oslope	Intercept	R <sup>2</sup>	R <sup>2</sup> 1.1	Corr Coef	RMSE	AER	BR	SR	CCR	RMR	TR	BR	BR	
1	Method 1	1	Yes	reanalysis	18.07	-1.85	1.85	0.9826	1.0177	2.81	0.7671	0.7654	0.8758	23.20	1	6	1	4	1	13	1		
2	Method 1	1	Yes	gpcp	69.42	-49.87	49.67	0.2509	3.9857	73.91	0.2197	-3.2437	0.4688	83.43	42	36	47	47	42	214	44		
3	Method 1	1	Yes	gpcp	76.04	-56.71	56.71	0.1546	6.4683	78.36	0.1626	-8.6425	0.4033	88.68	47	43	52	50	45	237	50		
4	Method 1	1	Yes	gpcp trmm	44.81	-19.77	19.77	0.5826	1.7775	38.09	0.5475	0.0879	0.7400	52.65	6	20	29	28	6	89	15		
5	Method 1	1	Yes	ghcn cams grid	61.44	-14.75	14.75	0.2263	4.4189	47.30	0.1229	-1.4093	0.3506	74.22	36	15	50	52	33	186	40		
6	Method 1	1	Yes	gpi	68.75	9.14	9.14	0.2474	4.0420	53.41	0.3229	-2.7111	0.5682	81.46	40	12	48	37	39	176	37		
7	Method 1	1	Yes	trmm 3a46	46.34	-0.18	0.18	0.4904	2.0392	34.04	0.4938	-0.0395	0.7027	54.21	7	1	32	32	7	79	10		
8	Method 1	1	Yes	trmm 3b43	46.83	-20.29	20.29	0.5124	1.9516	40.57	0.4652	-0.0938	0.6821	57.20	8	21	31	34	9	103	19		
9	Method 1	1	Yes	ssmi	50.20	-15.08	15.08	0.3654	2.7367	50.34	0.3589	-0.8363	0.5991	60.93	18	17	35	35	19	124	30		
10	Method 1	1	Yes	opi2	70.90	-57.92	57.92	0.2877	3.4758	74.76	0.2424	-3.2354	0.4924	84.45	44	44	41	39	44	212	43		
11	Method 1	1	Yes	opi3	69.12	-51.32	51.32	0.2667	3.7495	73.50	0.2312	-3.0813	0.4808	82.90	41	39	44	42	41	207	42		
12	Method 1	1	Yes	cmapp	68.33	-50.99	50.99	0.2681	3.7300	73.76	0.2308	-3.0620	0.4805	81.94	39	38	43	43	40	203	41		
13	Method 1	1	Yes	cmapp2	69.68	-52.50	52.50	0.2621	3.8153	74.35	0.2250	-3.2261	0.4743	83.57	43	41	45	45	43	217	46		
14	Method 1	2	Yes	reanalysis	22.86	6.73	6.73	1.1380	1.1380	-13.91	0.9183	0.8992	0.9583	28.47	3	8	10	2	3	26	3	Plot	
15	Method 1	2	Yes	gpcp	48.38	1.16	1.16	0.8357	1.1966	6.48	0.5569	0.5352	0.7463	61.13	14	2	18	26	20	80	12		
16	Method 1	2	Yes	gpcp	38.67	-1.65	1.65	0.7779	1.9895	10.94	0.6419	0.5892	0.8012	48.89	5	4	25	14	5	53	5	Plot	
17	Method 1	2	Yes	gpcp trmm	53.63	50.54	50.54	0.8321	1.2018	-41.71	0.7477	0.2558	0.8647	64.18	23	37	19	7	22	108	23		
18	Method 1	2	Yes	ghcn cams grid	47.31	6.75	6.75	0.8525	1.1730	0.94	0.5885	0.5652	0.7671	59.12	10	9	16	19	10	64	7		
19	Method 1	2	Yes	gpi	75.04	69.69	69.69	0.6351	1.5746	-33.52	0.6878	-0.3074	0.8172	91.71	45	50	28	12	49	184	38		
20	Method 1	2	Yes	trmm 3a46	68.04	67.53	67.53	0.7857	1.2728	-51.38	0.7391	-0.1389	0.8597	79.44	38	48	24	9	37	156	35		
21	Method 1	2	Yes	trmm 3b43	54.02	51.38	51.38	0.8318	1.2022	-42.39	0.7574	0.2495	0.8703	64.45	25	40	20	6	23	114	25		
22	Method 1	2	Yes	ssmi	53.82	47.55	47.55	0.8252	1.2118	-35.33	0.6762	0.2917	0.8223	67.25	24	35	22	10	25	116	26		
23	Method 1	2	Yes	opi2	50.22	-7.66	7.66	0.8699	1.1496	12.58	0.5268	0.5078	0.7258	62.91	20	11	12	30	21	84	17		
24	Method 1	2	Yes	opi3	48.52	-1.65	1.65	0.8547	1.1700	8.00	0.5564	0.5400	0.7459	60.82	15	5	15	27	18	80	11		
25	Method 1	2	Yes	cmapp	47.70	-1.40	1.40	0.8737	1.1446	6.95	0.5836	0.5516	0.7508	60.04	12	3	11	23	15	64	6		
26	Method 1	2	Yes	cmapp2	48.25	-2.92	2.92	0.8832	1.1585	8.72	0.5810	0.5459	0.7490	60.43	13	7	13	24	16	73	9		
27	Method 2	1	Yes	reanalysis	21.59	-12.60	12.60	1.0332	1.0332	10.93	0.7804	0.7192	0.8834	27.17	2	14	2	3	2	23	2	Plot	
28	Method 2	1	Yes	gpcp	78.87	-58.97	58.97	0.2391	4.1824	83.92	0.2162	-3.9724	0.4650	92.99	50	45	49	48	50	242	51		
29	Method 2	1	Yes	gpcp	85.16	-66.97	66.97	0.1828	5.4705	88.81	0.2047	-8.1318	0.4524	98.21	52	47	51	49	52	251	52		
30	Method 2	1	Yes	gpcp trmm	49.58	-31.15	31.15	0.5302	1.8861	49.42	0.6240	-0.2454	0.7900	56.44	17	30	30	16	8	101	18		
31	Method 2	1	Yes	ghcn cams grid	59.69	-26.85	26.85	0.2739	3.8510	53.14	0.1311	-1.0649	0.3621	73.62	34	25	42	51	32	184	39		
32	Method 2	1	Yes	gpi	64.45	-11.52	11.52	0.2556	3.9124	65.22	0.3072	-2.3727	0.5543	77.03	37	13	46	38	36	170	36		
33	Method 2	1	Yes	trmm 3a46	52.35	-14.87	14.87	0.4034	2.4789	52.20	0.4773	-0.8673	0.6909	60.56	22	16	34	33	17	122	27		
34	Method 2	1	Yes	trmm 3b43	50.20	-30.21	30.21	0.4889	2.0454	50.64	0.5307	-0.4122	0.7285	59.61	19	29	33	29	13	123	28		
35	Method 2	1	Yes	ssmi	55.25	-25.72	25.72	0.3275	3.0534	62.30	0.3330	-1.4117	0.5770	68.39	27	23	36	36	27	149	34		
36	Method 2	1	Yes	opi2	82.12	-76.31	76.31	0.3119	3.2062	86.60	0.2204	-4.1207	0.4695	95.52	51	52	37	46	51	237	49		
37	Method 2	1	Yes	opi3	77.88	-69.76	69.76	0.3047	3.2819	84.72	0.2322	-3.7094	0.4819	91.60	49	51	38	41	48	227	47		
38	Method 2	1	Yes	cmapp	75.62	-66.61	66.61	0.3027	3.3036	84.21	0.2367	-3.5601	0.4865	89.23	46	46	39	40	48	217	45		
39	Method 2	1	Yes	cmapp2	76.98	-67.85	67.85	0.2949	3.3910	84.77	0.2296	-3.7199	0.4792	90.78	48	49	40	44	47	228	48		
40	Method 2	2	Yes	reanalysis	25.20	-6.82	6.82	1.1745	1.1745	-2.50	0.9244	0.8996	0.9614	32.75	4	10	17	7	4	36	4	Plot	
41	Method 2	2	Yes	gpcp	56.42	-17.08	17.08	0.8974	1.1143	21.50	0.5646	0.5299	0.7514	70.87	28	18	8	22	28	104	20		
42	Method 2	2	Yes	gpcp	47.40	-23.66	23.66	0.8918	1.1213	27.85	0.6618	0.5880	0.8135	60.01	11	22	9	13	14	69	8		
43	Method 2	2	Yes	gpcp trmm	47.00	41.77	41.77	0.8205	1.2188	-33.18	0.7434	0.4173	0.8622	59.15	9	33	23	8	11	84	13		
44	Method 2	2	Yes	ghcn cams grid	54.63	-17.25	17.25	1.0467	1.0467	15.25	0.5941	0.5650	0.7708	68.17	26	19	3	18	26	92	16		
45	Method 2	2	Yes	gpi	58.23	40.71	40.71	0.7685	1.3012	-21.16	0.6408	0.3947	0.8005	73.05	31	32	26	15	30	134	31		
46	Method 2	2	Yes	trmm 3a46	59.22	55.22	55.22	0.7281	1.3734	-36.42	0.6726	0.0640	0.8201	74.46	33	42	27	11	34	147	33		
47	Method 2	2	Yes	trmm 3b43	49.15	43.82	43.82	0.8310	1.2034	-35.39	0.7598	0.4086	0.8716	59.59	16	34	21	5	12	88	14		
48	Method 2	2	Yes	ssmi	51.17	30.02	30.02	0.8568	1.1671	-20.68	0.6124	0.4909	0.7826	66.27	21	28	14	17	24	104	21		
49	Method 2	2	Yes	opi2	60.61	-34.94	34.94	1.0850	1.0850	32.80	0.5211	0.4036	0.7219	79.83	35	31	7	31	38	142	32		
50	Method 2	2	Yes	opi3	58.95	-28.98	28.98	1.0767	1.0767	26.58	0.5594	0.4779	0.7479	74.69	32	27	5	25	35	124	29		
51	Method 2	2	Yes	cmapp	57.52	-25.79	25.79	1.0815	1.0815	22.98	0.5697	0.5042	0.7548	72.78	29	24	6	20	29	108	22		
52	Method 2	2	Yes	cmapp2	58.02	-27.03	27.03	1.0671	1.0671	24.81	0.5864	0.4957	0.7526	73.40	30	26	4	21	31	112	24		



Larger Rondonia - Yearly Atmospheric

Ol	Moist Flux	Re	Storage	Precipitation	Absolute Error	Bias	Abias	Slope	Oslope	Intercept	R <sup>2</sup>	R <sup>2</sup> 1:1	Corr Coef	RMSE	AER	BR	SR	CCR	RMF	TR	BR	
1	Method 1	1	Yes	reanalysis	11.54	-1.80	1.80	1.1224	1.1224	-4.90	0.8606	0.8475	0.9277	13.23	1	6	5	11	1	24	1	
2	Method 1	1	Yes	gpcp	49.73	-49.73	49.73	0.0016	625.0000	82.03	0.0000	-6.9490	0.0014	56.50	33	35	31	31	36	166	36	
3	Method 1	1	Yes	gpcp	56.65	-56.65	56.65	-0.2627		88.99	0.1529	-28.2595	-0.3911	61.26	42	42	37	52	42	215	42	
4	Method 1	1	Yes	gpcp trmm	21.54	-21.54	21.54	1.3233	1.3233	7.37	0.7372	0.0091	0.8586	25.93	11	21	11	15	11	69	11	
5	Method 1	1	Yes	ghcn cams grid	32.63	-14.85	14.85	0.1363	7.3368	51.03	0.0025	-0.2882	0.0504	38.78	16	17	25	25	16	99	16	
6	Method 1	1	Yes	gpr	16.56	9.17	9.17	0.6527	1.5321	19.70	0.3647	0.0507	0.6039	19.46	8	12	18	18	6	62	6	Plot
7	Method 1	1	Yes	trmm 3a46	14.34	-0.21	0.21	2.0580	2.0580	-68.79	0.9441	0.6945	0.9716	16.62	4	1	23	5	3	36	3	
8	Method 1	1	Yes	trmm 3b43	20.28	-20.28	20.28	1.5458	1.5458	-2.41	0.8980	0.0918	0.9476	23.20	10	20	19	9	10	68	10	
9	Method 1	1	Yes	ssmi	16.29	-13.06	13.06	0.7029	1.4227	29.73	0.3581	-0.1461	0.5985	21.07	7	15	15	19	8	64	8	
10	Method 1	1	Yes	opi2	58.55	-58.55	58.55	-0.0209		82.58	0.0004	-9.4937	-0.0201	64.92	43	43	33	33	44	196	44	
11	Method 1	1	Yes	opi3	52.54	-52.54	52.54	-0.0104		82.39	0.0001	-7.7088	-0.0094	59.14	40	40	32	32	39	183	39	
12	Method 1	1	Yes	cmcp	52.29	-52.29	52.29	0.0433	23.0947	80.80	0.0015	-7.5269	0.0384	58.52	39	39	27	26	38	169	38	
13	Method 1	1	Yes	cmcp2	53.81	-53.81	53.81	0.0192	52.0833	81.54	0.0003	-7.9851	0.0172	60.07	41	41	29	29	40	180	40	
14	Method 1	2	Yes	reanalysis	11.82	6.64	6.64	1.3092	1.3092	-22.74	0.9189	0.8395	0.9586	15.85	2	8	9	7	2	28	2	
15	Method 1	2	Yes	gpcp	38.75	1.07	1.07	-0.4026		64.15	0.0208	-0.2320	-0.1441	43.91	20	2	42	41	22	127	22	
16	Method 1	2	Yes	gpcp	23.17	-1.73	1.73	-0.6285		69.85	0.1197	-0.6895	-0.3459	29.84	12	4	47	51	13	127	13	
17	Method 1	2	Yes	gpcp trmm	50.58	50.58	50.58	1.4475	1.4475	-74.13	0.8439	-3.9743	0.9186	51.83	35	37	17	12	30	131	30	
18	Method 1	2	Yes	ghcn cams grid	39.54	6.66	6.66	-0.6236		77.91	0.0315	-0.2103	-0.1774	43.52	23	9	46	45	20	143	20	
19	Method 1	2	Yes	gpi	69.62	69.62	69.62	0.9360	1.0684	-63.28	0.2895	-4.8325	0.5381	74.30	50	50	3	22	49	174	49	
20	Method 1	2	Yes	trmm 3a46	69.15	69.15	69.15	2.1630	2.1630	-155.57	0.9977	-6.3132	0.9989	70.57	49	49	24	1	46	169	46	
21	Method 1	2	Yes	trmm 3b43	51.42	51.42	51.42	1.7340	1.7340	-90.65	0.9794	-4.0920	0.9897	52.44	36	38	20	3	31	128	31	
22	Method 1	2	Yes	ssmi	51.83	50.57	50.57	0.9959	1.0041	-50.28	0.2890	-3.0708	0.5376	55.67	37	38	1	23	34	131	34	
23	Method 1	2	Yes	opi2	42.65	-7.74	7.74	-0.6626		70.39	0.0811	-0.4680	-0.2849	47.93	26	11	50	47	26	160	26	
24	Method 1	2	Yes	opi3	38.81	-1.74	1.74	-0.4437		64.81	0.0262	-0.2530	-0.1618	44.28	21	5	44	42	25	137	25	
25	Method 1	2	Yes	cmcp	37.97	-1.49	1.49	-0.2808		57.76	0.0101	-0.2013	-0.1004	43.35	18	3	39	37	19	116	19	
26	Method 1	2	Yes	cmcp2	38.26	-3.01	3.01	-0.3481		60.19	0.0159	-0.2284	-0.1261	43.84	19	7	40	40	21	127	21	
27	Method 2	1	Yes	reanalysis	15.17	-12.54	12.54	1.3156	1.3156	-3.32	0.8983	0.7406	0.9478	19.61	5	14	10	8	7	44	7	Plot
28	Method 2	1	Yes	gpcp	59.02	-59.02	59.02	-0.1054		95.28	0.0062	-7.2351	-0.0790	68.16	44	44	35	35	45	203	45	
29	Method 2	1	Yes	gpcp	66.91	-66.91	66.91	-0.2723		100.90	0.0940	-22.3974	-0.3066	71.59	46	46	38	48	47	225	47	
30	Method 2	1	Yes	gpcp trmm	32.40	-32.40	32.40	0.9897	1.0312	33.85	0.5859	-1.9641	0.7854	34.93	15	29	2	16	15	77	15	
31	Method 2	1	Yes	ghcn cams grid	39.20	-26.87	26.87	0.0632	15.8228	60.66	0.0003	-0.5516	0.0183	48.39	22	25	26	28	27	128	27	
32	Method 2	1	Yes	gpi	18.66	-11.49	11.49	0.7281	1.3734	31.11	0.3211	0.0001	0.5667	21.87	9	13	14	20	9	65	9	
33	Method 2	1	Yes	trmm 3a46	13.37	-13.37	13.37	1.7801	1.7801	-34.66	0.8720	0.3757	0.9338	18.42	3	16	21	10	4	54	4	
34	Method 2	1	Yes	trmm 3b43	30.21	-30.21	30.21	1.2154	1.2154	21.60	0.8005	-1.6141	0.8947	31.59	14	28	7	13	14	76	14	
35	Method 2	1	Yes	ssmi	23.21	-23.21	23.21	0.8750	1.1429	30.08	0.3789	-0.8171	0.6155	28.70	13	22	6	17	12	70	12	
36	Method 2	1	Yes	opi2	76.89	-76.89	76.89	-0.1172		93.57	0.0076	-11.8088	-0.0874	82.51	52	52	36	36	52	228	52	
37	Method 2	1	Yes	opi3	70.92	-70.92	70.92	-0.0399		92.66	0.0008	-9.9798	-0.0276	76.40	51	51	34	34	51	221	51	
38	Method 2	1	Yes	cmcp	67.73	-67.73	67.73	0.0276	36.2319	91.18	0.0004	-9.0882	0.0192	73.23	47	47	28	27	48	197	48	
39	Method 2	1	Yes	cmcp2	68.98	-68.98	68.98	0.0072	138.8889	91.66	0.0000	-9.4436	0.0051	74.51	48	48	30	30	50	206	50	
40	Method 2	2	Yes	reanalysis	15.64	-6.91	6.91	1.4361	1.4361	-16.36	0.9487	0.8387	0.9740	18.50	6	10	16	4	5	41	5	Plot
41	Method 2	2	Yes	gpcp	48.31	-17.17	17.17	-0.6579		88.65	0.0421	-0.3642	-0.2052	53.80	30	18	49	46	32	175	32	
42	Method 2	2	Yes	gpcp	36.07	-23.73	23.73	-0.6337		86.92	0.0972	-1.3470	-0.3117	40.89	17	23	48	50	17	155	17	
43	Method 2	2	Yes	gpcp trmm	41.77	41.77	41.77	1.0820	1.0820	-45.69	0.7932	-5.1540	0.8906	42.50	25	33	4	14	18	94	18	
44	Method 2	2	Yes	ghcn cams grid	45.08	-17.35	17.35	-0.8331		96.06	0.0314	-0.2625	-0.1773	51.76	29	19	51	44	29	172	29	
45	Method 2	2	Yes	gpi	42.94	40.83	40.83	1.2903	1.2903	-65.13	0.3079	-0.9446	0.5549	50.94	27	32	8	21	28	116	28	
46	Method 2	2	Yes	trmm 3a46	59.67	59.67	59.67	1.8184	1.8184	-115.55	0.9927	-8.9503	0.9963	60.31	45	45	22	2	41	155	41	
47	Method 2	2	Yes	trmm 3b43	43.82	43.82	43.82	1.3359	1.3359	-60.57	0.9326	-5.6658	0.9657	44.24	28	34	12	6	24	104	24	
48	Method 2	2	Yes	ssmi	39.93	33.99	33.99	1.3659	1.3659	-57.94	0.2642	-0.8025	0.5331	44.19	24	30	13	24	23	114	23	
49	Method 2	2	Yes	opi2	52.27	-35.04	35.04	-0.9366		83.93	0.0944	-0.8877	-0.3072	63.29	38	31	52	49	43	213	43	
50	Method 2	2	Yes	opi3	50.31	-29.07	29.07	-0.6233		79.74	0.0303	-0.5732	-0.1739	57.78	34	27	45	43	37	186	37	
51	Method 2	2	Yes	cmcp	48.70	-25.88	25.88	-0.3869		73.59	0.0119	-0.4568	-0.1092	55.60	31	24	41	38	33	167	33	
52	Method 2	2	Yes	cmcp2	49.32	-27.12	27.12	-0.4358		74.74	0.0157	-0.5009	-0.1251	56.44	32	26	43	39	35	175	35	

Larger Rondonia - Monthly Surface																				
Ol	Runoff Meth	Re	Storage	Precipitation	Absolute Error	Bias	Slope	Intercept	R <sup>2</sup>	R <sup>2</sup> 1:1	Corr Coef	RMSE	AER	BR	SR	CCR	RMR	TR	BR	Plot
1	Linear	1	Yes	reanalysis	73.38	-58.55	0.5956	-38.22	0.1404	-0.6290	0.3743	89.03	159	243	3	45	140	590	63	Plot
2	Linear	1	Yes	gpcp	92.08	-44.13	0.0851	-14.12	0.1001	-1.6141	0.1019	109.51	345	130	60	83	295	913	176	
3	Linear	1	Yes	gpcp	85.32	-33.31	0.1336	-10.16	0.0291	-1.4720	0.1705	99.11	274	43	57	65	220	659	86	
4	Linear	1	Yes	gpcp trmm	110.38	-65.92	-0.0005	-27.02	0.0000	-1.6198	-0.0004	127.36	419	291	139	137	397	1383	305	
5	Linear	1	Yes	ghcn cams grid	88.94	-44.83	0.1531	-14.16	0.1222	-1.0702	0.1489	100.21	313	152	54	69	232	820	134	
6	Linear	1	Yes	gpi	102.38	-85.74	0.2568	-32.12	0.1140	-2.3762	0.3376	127.12	384	403	38	51	395	1271	282	
7	Linear	1	Yes	trmm 3a46	110.85	-100.87	0.3539	-60.44	0.1324	-1.9752	0.3638	134.78	420	444	14	48	426	1352	301	
8	Linear	1	Yes	trmm 3b43	111.17	-66.68	-0.0567	-24.44	0.0029	-1.7501	-0.0543	129.47	422	297	254	175	409	1557	341	
9	Linear	1	Yes	ssmi	84.53	-69.50	0.4002	-36.88	0.1919	-1.2007	0.4380	105.15	265	327	6	33	262	893	169	
10	Linear	1	Yes	opi2	81.64	-27.39	0.0276	-12.84	0.0007	-1.0017	0.0259	95.51	242	17	88	108	185	640	77	
11	Linear	1	Yes	opi3	86.57	-33.93	0.0133	-12.71	0.0002	-1.2049	0.0132	100.24	284	58	103	123	233	801	126	
12	Linear	1	Yes	cmap	88.36	-37.25	0.0108	-12.29	0.0001	-1.2717	0.0107	101.79	305	95	109	128	239	876	161	
13	Linear	1	Yes	cmap2	88.55	-36.01	0.0060	-12.16	0.0000	-1.2832	0.0060	102.04	309	83	114	134	242	882	183	
14	Linear	1	No	reanalysis	63.31	-58.42	-0.1030	-2.97	0.1684	-47.3344	-0.4104	76.50	39	239	367	326	30	1001	209	
15	Linear	1	No	gpcp	78.54	-44.17	-0.0847	-8.59	0.2087	-42.6131	-0.4569	99.34	215	132	326	367	224	1264	280	
16	Linear	1	No	gpcp	75.19	-33.18	-0.0391	-5.41	0.5345	-437.3965	-0.7311	89.96	179	38	218	455	151	1041	226	
17	Linear	1	No	gpcp trmm	87.50	-65.02	-0.2241	-17.41	0.4970	-21.6955	-0.7050	114.12	296	287	456	450	331	1820	400	
18	Linear	1	No	ghcn cams grid	66.05	-44.38	-0.0735	-5.51	0.2040	-59.5133	-0.4517	85.75	117	138	297	363	111	1026	217	
19	Linear	1	No	gpi	106.79	-85.28	-0.0608	-8.75	0.1061	-57.3677	-0.3257	129.76	399	400	268	269	411	1747	384	
20	Linear	1	No	trmm 3a46	111.22	-95.38	-0.1823	-21.40	0.3602	-30.0588	-0.8002	136.01	423	427	441	423	429	2143	452	
21	Linear	1	No	trmm 3b43	86.42	-65.79	-0.2286	-16.68	0.5122	-21.8735	-0.7157	114.19	282	290	459	451	334	1816	399	
22	Linear	1	No	ssmi	91.71	-68.91	-0.0762	-10.36	0.1036	-34.5811	-0.3218	109.65	341	320	301	268	296	1526	336	
23	Linear	1	No	opi2	62.91	-26.24	-0.1258	-9.39	0.2892	-25.9996	-0.5377	77.21	35	7	410	410	34	896	170	
24	Linear	1	No	opi3	66.38	-32.78	-0.1108	-8.89	0.2478	-29.5293	-0.4978	82.10	85	32	392	397	74	980	198	
25	Linear	1	No	cmmap	67.43	-36.58	-0.1072	-8.64	0.2339	-30.7516	-0.4837	83.88	102	87	385	389	88	1051	230	
26	Linear	1	No	cmmap2	67.64	-35.34	-0.1058	-8.81	0.2330	-30.8459	-0.4827	84.01	108	75	381	385	92	1041	225	
27	Linear	1	mon	reanalysis	63.74	-58.35	-0.0524	-5.52	0.0435	-45.7063	-0.2087	75.25	47	238	244	215	23	767	112	
28	Linear	1	mon	gpcp	79.33	-43.87	-0.0906	-8.44	0.2377	-42.6726	-0.4875	99.57	223	127	344	391	226	1311	292	
29	Linear	1	mon	gpcp	75.97	-32.99	-0.0456	-5.22	0.7110	-443.2172	-0.8550	90.64	190	35	229	465	156	1075	240	
30	Linear	1	mon	gpcp trmm	89.64	-65.93	-0.2318	-16.86	0.5267	-21.7967	-0.7257	115.35	321	292	460	454	343	1870	416	
31	Linear	1	mon	ghcn cams grid	68.78	-44.48	-0.0802	-5.25	0.2425	-60.0078	-0.4924	86.16	127	140	312	394	118	1091	246	
32	Linear	1	mon	gpi	108.17	-85.18	-0.0773	-7.58	0.1715	-58.1243	-0.4141	130.91	410	397	302	330	416	1855	411	
33	Linear	1	mon	trmm 3a46	117.26	-98.10	-0.2636	-15.74	0.7199	-31.6207	-0.8485	140.97	443	435	465	463	447	2253	463	
34	Linear	1	mon	trmm 3b43	86.24	-64.58	-0.2434	-16.33	0.5723	-21.5580	-0.7565	114.35	281	286	462	457	335	1821	401	
35	Linear	1	mon	ssmi	93.66	-68.77	-0.1056	-8.80	0.1968	-35.4807	-0.4457	111.35	360	319	380	361	313	1733	378	
36	Linear	1	mon	opi2	63.99	-28.49	-0.1278	-9.34	0.2973	-28.0362	-0.5452	77.39	41	8	413	415	37	914	177	
37	Linear	1	mon	opi3	66.81	-33.05	-0.1125	-8.84	0.2546	-29.5640	-0.5046	82.29	92	36	396	401	76	1001	210	
38	Linear	1	mon	cmmap	68.03	-36.68	-0.1077	-8.63	0.2360	-30.7769	-0.4858	84.06	116	90	387	390	94	1077	241	
39	Linear	1	mon	cmmap2	68.16	-35.44	-0.1060	-8.81	0.2339	-30.8607	-0.4837	84.17	119	77	382	388	98	1064	234	
40	Linear	1	mon	reanalysis	63.81	-58.31	-0.0033	-7.99	0.0002	-44.1204	-0.0130	74.01	49	236	147	147	16	595	65	
41	Linear	1	mon	gpcp	78.90	-43.85	-0.0807	-8.78	0.1885	-42.0678	-0.4342	99.05	219	126	313	353	219	1230	273	
42	Linear	1	mon	gpcp	75.36	-32.65	-0.0391	-5.36	0.5442	-441.4968	-0.7377	90.28	183	31	217	456	153	1040	223	
43	Linear	1	mon	gpcp trmm	89.79	-66.77	-0.2007	-17.79	0.3919	-21.1506	-0.6261	114.62	324	298	447	428	336	1833	409	
44	Linear	1	mon	ghcn cams grid	68.15	-44.58	-0.0679	-5.67	0.1737	-59.1044	-0.4168	85.54	118	143	282	336	108	987	204	
45	Linear	1	mon	gpi	108.67	-85.37	-0.0775	-7.56	0.1721	-58.0969	-0.4148	131.21	415	401	303	332	418	1869	415	
46	Linear	1	mon	trmm 3a46	120.02	-100.27	-0.2825	-13.90	0.8016	-31.8310	-0.8953	143.14	453	442	467	467	453	2282	466	
47	Linear	1	mon	trmm 3b43	85.56	-65.33	-0.2048	-17.53	0.4026	-20.7792	-0.6345	113.24	276	288	450	432	327	1773	388	
48	Linear	1	mon	ssmi	94.77	-68.92	-0.1140	-8.35	0.2315	-35.7507	-0.4811	112.10	368	322	397	383	319	1789	392	
49	Linear	1	mon	opi2	62.05	-26.72	-0.1078	-9.64	0.2108	-25.2823	-0.4592	76.44	23	10	389	368	29	819	132	
50	Linear	1	mon	opi3	65.59	-33.30	-0.0945	-9.22	0.1789	-28.8160	-0.4230	81.41	72	42	352	344	64	874	159	
51	Linear	1	mon	cmmap	67.07	-36.65	-0.0893	-9.11	0.1624	-29.9988	-0.4029	83.16	96	89	342	317	83	927	183	
52	Linear	1	mon	cmmap2	67.16	-35.40	-0.0879	-9.26	0.1606	-30.0745	-0.4008	83.26	97	76	339	314	84	910	173	
53	Linear	1	mon	reanalysis	63.51	-58.29	0.0341	-9.89	0.0185	-42.9366	0.1361	73.07	44	235	78	73	11	441	30	
54	Linear	1	mon	gpcp	77.33	-44.00	-0.0485	-9.81	0.0681	-40.2025	-0.2609	97.06	204	129	236	251	200	1020	214	
55	Linear	1	mon	gpcp	74.53	-32.80	-0.0212	-5.79	0.1611	-433.5681	-0.4014	89.29	169	33	195	315	142	854	152	
56	Linear	1	mon	gpcp trmm	86.97	-68.17	-0.1114	-21.19	0.1183	-19.3537	-0.3439	110.82	289	315	394	273	308	1579	343	
57	Linear	1	mon	ghcn cams grid	66.68	-44.70	-0.0373	-6.77	0.0524	-56.8506	-0.2289	83.91	90	149	215	226	89	769	116	
58	Linear	1	mon	gpi	107.74	-85.83	-0.0485	-9.64	0.0671	-56.4291	-0.2591	129.66	406	405	237	247	410	1705	373	
59	Linear	1	mon	trmm 3a46	119.69	-101.86	-0.2005	-19.18	0.3980	-30.0937	-0.6308	141.01	452	447	446	431	448	2224	458	
60	Linear																			

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Ol	Runoff Meth	File	Storage	Precipitation	Absolute Error	Bias	Slope	Intercept	R <sup>2</sup>	R <sup>2</sup> 1:1	Corr Coef	RMSE	AER	BR	SR	CCR	RMR	TR	BR
102	Linear	2	No	opi3	76.10	-51.27	-0.1069	-16.72	0.1969	-29.6947	-0.4437	95.84	194	174	384	360	189	1301	291
103	Linear	2	No	cmmap	77.62	-54.46	-0.1053	-16.44	0.1928	-30.9636	-0.4391	97.80	208	211	376	358	206	1359	302
104	Linear	2	No	cmmap2	77.36	-53.22	-0.1036	-16.63	0.1905	-30.9011	-0.4364	97.71	205	199	370	355	205	1334	298
105	Linear	2	1 mon	reanalysis	95.46	-73.50	-0.0855	-15.50	0.1750	-46.0047	-0.4183	118.81	371	352	330	339	366	1758	385
106	Linear	2	1 mon	gpcp	89.35	-62.84	-0.0717	-17.00	0.1283	-41.6397	-0.3581	113.16	316	255	290	283	326	1470	327
107	Linear	2	1 mon	gpcp	88.15	-55.80	-0.0722	-14.44	0.1826	-55.0272	-0.4273	108.19	302	223	291	349	288	1453	318
108	Linear	2	1 mon	gpcp trmm	93.67	-70.50	-0.1682	-16.79	0.1482	-30.8762	-0.6467	119.37	356	337	432	437	370	1932	425
109	Linear	2	1 mon	ghcn cams grid	83.77	-62.81	-0.0834	-16.51	0.1344	-35.6813	-0.3667	104.96	256	253	323	293	261	1386	307
110	Linear	2	1 mon	gpi	123.80	-103.05	-0.0852	-11.95	0.2427	-76.2733	-0.4926	148.69	484	454	329	395	464	2106	449
111	Linear	2	1 mon	trmm 3a46	118.89	-97.81	-0.1639	-13.84	0.4109	-40.3278	-0.6410	140.53	449	434	429	434	446	2192	456
112	Linear	2	1 mon	trmm 3b43	93.24	-72.79	-0.1672	-16.46	0.4142	-31.6248	-0.6436	120.76	354	348	430	435	375	1942	426
113	Linear	2	1 mon	ssmi	108.57	-83.93	-0.1144	-12.07	0.3014	-50.9969	-0.5490	127.10	414	388	399	417	394	2012	437
114	Linear	2	1 mon	opi2	72.62	-45.10	-0.1039	-17.47	0.1701	-25.8094	-0.4125	89.73	150	156	372	327	147	1152	255
115	Linear	2	1 mon	opi3	75.33	-51.02	-0.0883	-17.34	0.1338	-28.8592	-0.3658	94.70	182	169	340	292	180	1163	257
116	Linear	2	1 mon	cmmap	77.05	-54.20	-0.0872	-17.09	0.1319	-30.1258	-0.3631	96.69	200	205	335	288	196	1224	270
117	Linear	2	1 mon	cmmap2	76.67	-52.96	-0.0848	-17.28	0.1273	-30.0292	-0.3568	96.53	197	193	327	281	194	1192	263
118	Linear	2	2 mon	reanalysis	94.08	-73.65	-0.0730	-16.20	0.1276	-45.4649	-0.3572	118.26	363	355	293	282	361	1654	360
119	Linear	2	2 mon	gpcp	87.65	-62.83	-0.0501	-17.97	0.0626	-40.5666	-0.2502	111.85	298	254	240	240	315	1347	299
120	Linear	2	2 mon	gpcp	86.00	-55.57	-0.0444	-15.49	0.0911	-52.8545	-0.2628	106.45	278	220	226	252	272	1248	277
121	Linear	2	2 mon	gpcp trmm	92.69	-70.58	-0.1486	-17.91	0.3276	-30.2024	-0.5723	119.14	351	338	425	419	369	1902	420
122	Linear	2	2 mon	ghcn cams grid	81.62	-62.80	-0.0548	-17.77	0.0580	-34.5720	-0.2409	103.47	241	252	248	233	253	1227	271
123	Linear	2	2 mon	gpi	122.77	-102.91	-0.0739	-12.90	0.1825	-75.2235	-0.4272	148.06	461	452	298	348	463	2022	438
124	Linear	2	2 mon	trmm 3a46	121.10	-100.30	-0.1780	-12.67	0.4735	-40.9205	-0.6881	142.88	456	443	439	446	452	2236	460
125	Linear	2	2 mon	trmm 3b43	92.22	-72.66	-0.1496	-17.55	0.3328	-30.9193	-0.5769	120.50	347	347	426	420	374	1914	423
126	Linear	2	2 mon	ssmi	107.84	-83.95	-0.1072	-12.68	0.2662	-50.8224	-0.5159	126.92	408	389	386	404	393	1980	433
127	Linear	2	2 mon	opi2	70.64	-45.15	-0.0621	-18.55	0.0609	-24.5197	-0.2468	87.64	142	157	272	237	129	937	186
128	Linear	2	2 mon	opi3	73.29	-51.04	-0.0580	-18.32	0.0577	-27.8338	-0.2403	93.16	157	170	260	232	169	968	205
129	Linear	2	2 mon	cmmap	74.84	-54.24	-0.0568	-18.17	0.0560	-29.0918	-0.2367	95.17	173	206	255	231	182	1047	229
130	Linear	2	2 mon	cmmap2	74.58	-52.99	-0.0545	-18.32	0.0526	-28.9753	-0.2294	94.98	171	194	247	227	181	1020	213
131	Linear	2	3 mon	reanalysis	91.85	-73.84	-0.0261	-18.66	0.0163	-43.2888	-0.1275	115.53	342	358	205	204	344	1453	319
132	Linear	2	3 mon	gpcp	84.81	-62.92	-0.0156	-19.39	0.0661	-38.9053	-0.0779	109.67	267	261	181	193	298	1200	265
133	Linear	2	3 mon	gpcp	83.44	-55.81	-0.0091	-16.82	0.0029	-50.3633	-0.0536	104.34	252	224	164	174	259	1073	239
134	Linear	2	3 mon	gpcp trmm	91.07	-71.61	-0.0878	-20.53	0.1136	-28.3486	-0.3371	116.80	335	344	337	271	353	1640	354
135	Linear	2	3 mon	ghcn cams grid	78.59	-62.90	-0.0115	-19.57	0.0026	-32.9454	-0.0506	101.15	217	259	171	170	236	1053	231
136	Linear	2	3 mon	gpi	120.48	-103.02	-0.0371	-15.97	0.0480	-72.6367	-0.2145	145.91	454	453	214	221	459	1801	396
137	Linear	2	3 mon	trmm 3a46	120.48	-101.90	-0.1293	-16.04	0.2460	-39.3352	-0.4960	141.88	455	449	414	396	449	2163	454
138	Linear	2	3 mon	trmm 3b43	90.64	-73.95	-0.0911	-20.16	0.1220	-29.1321	-0.3493	118.35	331	359	345	276	363	1674	364
139	Linear	2	3 mon	ssmi	105.44	-84.18	-0.0570	-16.13	0.0766	-49.2781	-0.2767	124.40	392	390	256	256	386	1860	365
140	Linear	2	3 mon	opi2	67.48	-45.30	-0.0132	-19.73	0.0027	-23.0104	-0.0523	85.07	104	160	174	173	105	716	99
141	Linear	2	3 mon	opi3	70.06	-51.17	-0.0114	-19.70	0.0022	-26.2772	-0.0474	90.87	140	173	169	169	157	808	128
142	Linear	2	3 mon	cmmap	71.77	-54.33	-0.0097	-19.73	0.0016	-27.4989	-0.0404	92.66	146	208	165	163	168	850	149
143	Linear	2	3 mon	cmmap2	71.45	-53.08	-0.0101	-19.73	0.0018	-27.4460	-0.0426	92.59	145	196	166	165	167	839	145
144	Linear	2	4 mon	reanalysis	89.63	-74.24	0.0195	-21.13	0.0091	-41.1136	0.0954	112.86	320	366	98	88	322	1194	264
145	Linear	2	4 mon	gpcp	81.92	-63.14	0.0290	-21.32	0.0210	-36.6916	0.1450	106.77	244	269	85	70	276	944	189
146	Linear	2	4 mon	gpcp	81.39	-56.42	0.0370	-18.56	0.0478	-47.2212	0.2186	101.35	238	228	73	52	237	828	138
147	Linear	2	4 mon	gpcp trmm	88.75	-73.33	-0.0153	-23.97	0.0034	-26.2317	-0.0582	113.76	311	350	178	179	330	1348	300
148	Linear	2	4 mon	ghcn cams grid	75.86	-63.17	0.0420	-21.88	0.0340	-30.8912	0.1845	98.21	188	270	66	57	213	794	122
149	Linear	2	4 mon	gpi	118.60	-103.30	0.0058	-19.54	0.0011	-69.7695	0.0336	143.35	448	455	116	104	454	1577	342
150	Linear	2	4 mon	trmm 3a46	116.74	-101.73	-0.0494	-22.22	0.0360	-36.4172	-0.1898	138.34	441	446	239	212	438	1776	389
151	Linear	2	4 mon	trmm 3b43	88.48	-75.61	-0.0134	-24.03	0.0026	-26.8709	-0.0511	115.09	307	371	175	171	340	1364	303
152	Linear	2	4 mon	ssmi	102.78	-84.66	-0.0033	-19.95	0.0003	-47.7752	-0.0163	121.74	387	399	148	149	378	1455	321
153	Linear	2	4 mon	opi2	64.20	-45.48	0.0444	-21.20	0.0311	-21.2031	0.1764	81.94	56	165	65	59	73	418	24
154	Linear	2	4 mon	opi3	67.42	-51.40	0.0419	-21.39	0.0301	-24.4571	0.1734	87.74	101	177	67	61	130	536	44
155	Linear	2	4 mon	cmmap	69.23	-54.54	0.0415	-21.51	0.0299	-25.7326	0.1730	89.92	133	212	68	62	149	624	75
156	Linear	2	4 mon	cmmap2	68.86	-53.29	0.0413	-21.45	0.0302	-25.6347	0.1738	89.75	131	200	70	60	148	609	70
157	Point	1	Yes	reanalysis	72.07	-57.21	0.6191	-38.06	0.1524	-0.5823	0.3904	87.46	147	234	1	44	127	553	52
158	Point	1	Yes	gpcp	90.39	-42.47	0.1065	-13.17	0.0165	-1.5395	0.1283	107.29	329	111	58	75	280	853	150
159	Point	1	Yes	gpcp	84.41	-32.33	0.1443	-9.46	0.0340	-1.4266	0.1845	98.01	264	29	55	56	211	615	72
160	Point	1	Yes	gpcp trmm	107.03	-63.41	0.0451	-26.28	0.0019	-1.5276	0.0438	123.30	401	277	64	99	382	1223	269
161	Point	1	Yes	ghcn cams grid	87.56	-43.48	0.1748	-13.80	0.0291	-1.0112	0.1705	98.48	297	124	52	64	215	752	110
162	Point	1	Yes	gpi	100.77	-84.20	0.2708	-31.57	0.1286	-2.3122	0.3586	124.86	382	391	30	49	388	1240	274
163	Point	1	Yes	trmm 3a46	107.39	-97.53	0.3939	-59.61	0.1679	-1.8249	0.4057	129.79	404	433	7	43	412	1299	290
164	Point	1	Yes	trmm 3b43	107.94	-64.21	-0.0109	-23.80	0.0001	-1.6610	-0.0106	125.50	409	284	188	145	390	1396	310
165	Point	1	Yes	ssmi	82.80	-67.89	0.4174	-36.20	0.2127	-1.1368	0.4612	102.64	250	313	4	31	247	845	148
166	Point	1	Yes	opi2	80.16	-25.74	0.0543	-11.59	0.0026	-0.9452	0.0514	93.64	231	6	61	96	173	567	55
167	Point	1	Yes	opi3	85.04	-32.28	0.0396	-11.62	0.0015	-1.1418	0.0394	98.26	270	28	72	101	214	685	92
168	Point	1	Yes	cmmap	86.73	-35.59	0.0367	-11.28	0.0013	-1.2073	0.0367	99.78	285	79	74				

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QI	Runoff Meth	Re	Storage	Precipitation	Absolute Error	Bias	Slope	Intercept	R <sup>2</sup>	R <sup>2</sup> 1:1	Corr Coef	RMSE	AER	BR	SR	CCR	RMR	TR	BR
203	Point	1	2 mon	trmm 3b43	83.23	-63.09	-0.1746	-16.48	0.3636	-24.4924	-0.6031	109.91	251	268	437	425	301	1682	366
204	Point	1	2 mon	ssmi	93.06	-67.32	-0.0981	-7.62	0.2146	-43.2761	-0.4632	109.93	352	303	357	371	302	1685	369
205	Point	1	2 mon	opi2	60.74	-25.07	-0.0914	-8.25	0.1872	-30.0080	-0.4326	74.72	12	3	347	352	19	733	102
206	Point	1	2 mon	opi3	64.24	-31.65	-0.0795	-7.91	0.1564	-34.2433	-0.3954	79.66	57	21	308	311	52	749	108
207	Point	1	2 mon	cmap	65.72	-34.99	-0.0746	-7.83	0.1398	-35.6291	-0.3739	81.37	76	71	299	297	63	806	127
208	Point	1	2 mon	cmmap2	65.84	-33.75	-0.0734	-7.95	0.1383	-35.7421	-0.3719	81.50	80	55	296	296	66	793	121
209	Point	1	3 mon	reanalysis	62.14	-56.96	0.0244	-8.05	0.0120	-53.0836	0.1096	72.07	26	229	91	80	6	432	27
210	Point	1	3 mon	gpcp	76.12	-42.36	-0.0435	-8.33	0.0677	-48.4918	-0.2601	95.76	195	110	224	248	188	965	194
211	Point	1	3 mon	gpcp	73.99	-31.88	-0.0177	-4.95	0.1544	-587.5696	-0.3929	88.67	166	25	184	308	138	821	135
212	Point	1	3 mon	gpcp trmm	84.14	-66.09	-0.0951	-19.80	0.1073	-23.1157	-0.3276	108.15	262	294	353	270	287	1466	325
213	Point	1	3 mon	ghcn cams grid	65.66	-43.37	-0.0317	-5.64	0.0478	-70.0372	-0.2187	82.75	73	121	209	223	81	707	96
214	Point	1	3 mon	gpi	106.08	-84.32	-0.0394	-8.79	0.0552	-68.7216	-0.2349	127.84	395	392	219	229	401	1636	353
215	Point	1	3 mon	trmm 3a46	116.22	-99.07	-0.1704	-18.46	0.3612	-35.9824	-0.6010	137.15	438	440	434	424	434	2170	455
216	Point	1	3 mon	trmm 3b43	80.96	-64.18	-0.1044	-19.28	0.1289	-22.8590	-0.3590	107.17	234	283	374	286	279	1456	322
217	Point	1	3 mon	ssmi	91.97	-67.39	-0.0722	-9.07	0.1163	-42.0961	-0.3411	108.73	344	305	292	272	292	1505	332
218	Point	1	3 mon	opi2	59.20	-25.29	-0.0542	-8.82	0.0658	-28.3238	-0.2564	72.78	9	4	246	245	9	513	40
219	Point	1	3 mon	opi3	62.84	-31.75	-0.0447	-8.68	0.0495	-32.5294	-0.2224	77.83	33	24	227	225	41	550	51
220	Point	1	3 mon	cmmap	63.96	-34.75	-0.0428	-8.86	0.0457	-33.8107	-0.2139	79.46	52	67	223	219	50	611	71
221	Point	1	3 mon	cmmap2	64.07	-33.50	-0.0423	-8.72	0.0458	-33.9223	-0.2141	79.59	54	49	222	220	51	596	66
222	Point	1	4 mon	reanalysis	61.74	-56.89	0.0318	-8.43	0.0204	-52.8094	0.1427	71.92	18	231	83	72	5	409	22
223	Point	1	4 mon	gpcp	74.20	-42.52	-0.0116	-9.38	0.0048	-46.2233	-0.0696	93.71	167	113	172	186	174	812	130
224	Point	1	4 mon	gpcp	72.92	-32.49	0.0000	-5.39	0.0000	-573.5738	-0.0011	87.56	153	30	137	138	128	586	60
225	Point	1	4 mon	gpcp trmm	79.47	-67.87	-0.0006	-23.79	0.0000	-20.8355	-0.0020	103.84	225	312	141	141	254	1073	238
226	Point	1	4 mon	ghcn cams grid	64.24	-43.43	-0.0035	-6.66	0.0006	-67.3798	-0.0243	81.23	58	123	149	151	62	543	47
227	Point	1	4 mon	gpi	104.43	-84.69	-0.0066	-11.18	0.0015	-66.4002	-0.0393	126.00	390	394	158	182	391	1495	329
228	Point	1	4 mon	trmm 3a46	111.10	-98.92	-0.0668	-25.77	0.0557	-32.9242	-0.2360	132.90	421	439	279	230	423	1792	394
229	Point	1	4 mon	trmm 3b43	77.20	-65.52	-0.0089	-23.15	0.0009	-20.5883	-0.0303	102.85	201	289	163	158	249	1060	233
230	Point	1	4 mon	ssmi	90.07	-67.66	-0.0333	-11.21	0.0248	-40.3806	-0.1574	106.84	325	310	210	207	277	1329	297
231	Point	1	4 mon	opi2	57.32	-25.31	-0.0115	-9.49	0.0030	-26.4108	-0.0545	70.49	3	5	170	176	2	356	12
232	Point	1	4 mon	opi3	61.00	-31.69	-0.0054	-9.55	0.0007	-30.5410	-0.0270	75.62	15	22	152	153	25	367	13
233	Point	1	4 mon	cmmap	61.94	-34.52	-0.0049	-9.61	0.0006	-31.7229	-0.0247	77.18	21	65	151	152	33	422	25
234	Point	1	4 mon	cmmap2	62.07	-33.27	-0.0054	-9.61	0.0007	-31.8462	-0.0272	77.32	24	41	153	154	36	408	21
235	Point	2	Yes	reanalysis	67.90	-62.89	0.3626	-28.87	0.4696	-2.9555	0.6853	89.04	112	258	13	6	141	530	43
236	Point	2	Yes	gpcp	65.42	-52.63	0.2955	-22.26	0.3262	-2.9108	0.5712	88.53	70	192	25	13	136	436	29
237	Point	2	Yes	gpcp	60.95	-44.53	0.3117	-17.91	0.3960	-2.6664	0.6316	80.58	14	141	22	10	57	244	5
238	Point	2	Yes	gpcp trmm	73.34	-68.75	0.2987	-35.20	0.3037	-3.7974	0.5511	96.66	158	318	24	15	195	710	97
239	Point	2	Yes	ghcn cams grid	61.77	-52.46	0.3443	-24.31	0.3427	-2.2730	0.5854	80.99	19	191	15	12	61	298	6
240	Point	2	Yes	gpi	97.31	-95.39	0.3076	-36.94	0.4917	-6.9419	0.7012	120.91	379	428	23	4	377	1211	267
241	Point	2	Yes	trmm 3a46	90.09	-89.81	0.3908	-47.68	0.5470	-4.6266	0.7396	108.65	326	415	8	2	291	1042	227
242	Point	2	Yes	trmm 3b43	75.26	-70.79	0.2891	-35.34	0.2836	-4.0036	0.5325	98.72	181	340	26	19	216	782	119
243	Point	2	Yes	ssmi	79.63	-77.37	0.3757	-36.66	0.5532	-4.2345	0.7438	98.03	227	380	10	1	212	830	139
244	Point	2	Yes	opi2	52.48	-34.77	0.3248	-17.72	0.2489	-1.4302	0.4889	69.79	1	68	20	25	1	115	1
245	Point	2	Yes	opi3	56.49	-40.73	0.3267	-19.72	0.2745	-1.7195	0.5240	73.83	2	102	19	21	14	158	2
246	Point	2	Yes	cmmap	57.90	-43.93	0.3270	-20.77	0.2777	-1.8610	0.5270	75.72	6	128	18	20	27	199	4
247	Point	2	Yes	cmmap2	57.78	-42.88	0.3216	-20.18	0.2742	-1.8552	0.5236	75.65	5	114	21	22	26	188	3
248	Point	2	No	reanalysis	84.89	-62.32	0.0127	-9.62	0.0026	-23.9675	0.0505	106.37	268	248	104	97	271	988	206
249	Point	2	No	gpcp	82.00	-52.05	-0.0142	-8.33	0.0033	-22.9768	-0.0578	104.24	245	188	176	177	257	1043	228
250	Point	2	No	gpcp	80.34	-44.30	-0.0196	-4.87	0.0090	-30.8209	-0.0950	99.14	232	137	190	200	221	980	199
251	Point	2	No	gpcp trmm	91.51	-69.03	-0.1242	-15.26	0.0145	-27.3676	-0.4631	116.27	339	324	407	370	349	1789	393
252	Point	2	No	ghcn cams grid	75.91	-51.88	-0.0163	-8.24	0.0034	-19.1406	-0.0583	95.53	189	184	183	180	186	922	180
253	Point	2	No	gpi	115.81	-94.57	-0.0345	-12.54	0.0290	-48.8128	-0.1703	139.87	435	424	211	209	445	1724	377
254	Point	2	No	trmm 3a46	111.27	-91.05	-0.0922	-15.51	0.1237	-33.2775	-0.3518	133.07	424	416	348	277	424	1889	417
255	Point	2	No	trmm 3b43	91.47	-71.07	-0.1274	-14.84	0.2251	-27.9953	-0.4744	117.55	338	341	412	378	357	1826	404
256	Point	2	No	ssmi	101.21	-77.07	-0.0599	-7.96	0.0634	-34.3931	-0.2518	120.02	383	378	266	241	373	1641	355
257	Point	2	No	opi2	65.96	-34.19	-0.0478	-7.74	0.0238	-14.0136	-0.1544	82.48	81	61	235	206	79	662	87
258	Point	2	No	opi3	68.38	-40.16	-0.0221	-8.25	0.0056	-15.4383	-0.0746	86.31	123	99	197	190	119	728	100
259	Point	2	No	cmmap	69.71	-43.35	-0.0201	-8.25	0.0046	-16.0924	-0.0681	88.01	137	120	193	184	133	767	113
260	Point	2	No	cmmap2	69.58	-42.10	-0.0188	-8.32	0.0041	-16.0807	-0.0644	87.98	135	105	185	181	132	738	104
261	Point	2	1 mon	reanalysis	86.08	-62.39	0.0122	-9.60	0.0024	-23.9697	0.0486	106.56	279	249	107	98	274	1007	211
262	Point	2	1 mon	gpcp	81.55	-51.72	-0.0017	-8.88	0.0000	-22.3929	-0.0070	103.14	239	180	143	143	251	956	190
263	Point	2	1 mon	gpcp	80.58	-44.30	0.0016	-5.79	0.0001	-29.8891	0.0080	97.85	233	136	133	131	208	841	147
264	Point	2	1 mon	gpcp trmm	91.28	-67.59	-0.1352	-15.40	0.0254	-27.1036	-0.5024	115.94	337	308	418	400	347	1810	397
265	Point	2	1 mon	ghcn cams grid	75.72	-51.69	-0.0006	-8.93	0.0000	-18.6495	-0.0020	94.53	187	179	140	140	179	825	136
266	Point	2	1 mon	gpi	116.62	-94.19	-0.0283	-7.79	0.0209	-48.3395	-0.1446	139.28	439	419	207	205	443	1713	376
267	Point	2	1 mon	trmm 3a46	115.81	-94.52	-0.1413	-12.18	0.2812	-35.2939	-0.5303	137.21	436	423	422	408	435	2124	450
268	Point	2	1 mon	trmm 3b43	90.81	-69.88	-0.1318	-15.27	0.2399	-27.7028	-0.4898	117.17	332	334	415	393	358	1830	407
269	Point	2	1 mon	ssmi	102.45	-76.02	-0.0585	-7.75	0.0607	-34.0840	-0.2464	119.09	385	372	281	235	368	1621	351
270	Point	2	1 mon	opi2	65.77	-33.98	-0.0228	-8.38	0.0054	-13.4214	-0.0734	80.99	78	59	201	189	60		

Larger Rondonia - Monthly Surface

QI	Runoff Meth	Re	Storage	Precipitation	Absolute Error	Bias	Slope	Intercept	R <sup>2</sup>	R <sup>2</sup> 1:1	Corr Coef	RMSE	AER	BR	SR	CCR	RMR	TR	BR	
304	Point	2	4	mon	gpcn cams grid	73.07	-52.24	0.0281	-10.36	0.0102	-18.1371	0.1008	93.26	156	190	87	84	170	667	93
305	Point	2	4	mon	gpi	112.31	-94.78	0.0237	-12.52	0.0138	-46.2803	0.1175	136.45	427	425	92	79	431	1454	320
306	Point	2	4	mon	trmm 3a46	113.79	-98.66	-0.0199	-21.39	0.0054	-31.7246	-0.0734	134.61	430	438	192	188	425	1673	363
307	Point	2	4	mon	trmm 3b43	86.13	-72.92	0.0049	-22.27	0.0003	-23.8917	0.0179	112.36	280	349	117	113	320	1179	260
308	Point	2	4	mon	ssmi	97.79	-77.53	0.0189	-14.25	0.0074	-37.1124	0.0862	115.70	380	381	99	89	346	1295	289
309	Point	2	4	mon	opi2	62.62	-34.55	0.0248	-9.77	0.0064	-12.5979	0.0802	78.61	32	66	90	90	44	322	10
310	Point	2	4	mon	opi3	66.04	-40.47	0.0196	-9.76	0.0044	-14.5734	0.0663	84.13	83	101	96	91	96	467	34
311	Point	2	4	mon	cmmap	67.38	-43.61	0.0195	-9.82	0.0044	-15.2716	0.0663	86.00	99	125	97	92	115	528	42
312	Point	2	4	mon	cmmap2	67.27	-42.36	0.0187	-9.77	0.0041	-15.2902	0.0643	86.04	98	109	100	93	117	517	41
313	Area	1	Yes		reanalysis	73.66	-58.76	0.5979	-38.54	0.1390	-0.6223	0.3729	89.55	164	246	2	46	145	603	68
314	Area	1	Yes		gpcp	92.47	-44.55	0.0874	-14.61	0.0107	-1.5771	0.1033	110.05	349	142	59	81	305	936	185
315	Area	1	Yes		gpcn	85.52	-33.54	0.1353	-10.44	0.0295	-1.4552	0.1717	99.29	275	50	56	63	223	667	88
316	Area	1	Yes		gpcn trmm	111.56	-67.34	0.0022	-28.54	0.0000	-1.5753	0.0021	129.05	425	304	130	136	407	1402	311
317	Area	1	Yes		gpcn cams grid	89.15	-45.04	0.1568	-14.50	0.0229	-1.0505	0.1512	100.53	315	155	53	68	234	825	137
318	Area	1	Yes		gpi	102.89	-86.34	0.2602	-32.97	0.1139	-2.3235	0.3376	127.81	388	411	37	50	400	1286	287
319	Area	1	Yes		trmm 3a46	112.46	-102.73	0.3649	-63.00	0.1336	-1.9116	0.3655	136.85	429	451	12	47	432	1371	304
320	Area	1	Yes		trmm 3b43	112.36	-68.07	-0.0555	-25.88	0.0027	-1.7028	-0.0520	131.18	428	314	251	172	417	1582	344
321	Area	1	Yes		ssmi	85.13	-70.22	0.4052	-37.86	0.1912	-1.1745	0.4372	106.03	271	336	5	34	268	914	178
322	Area	1	Yes		opi2	82.16	-27.81	0.0285	-13.27	0.0007	-0.9815	0.0265	96.16	248	18	86	107	192	651	82
323	Area	1	Yes		opi3	87.06	-34.35	0.0149	-13.16	0.0002	-1.1796	0.0146	100.85	290	63	101	120	235	809	129
324	Area	1	Yes		cmmap	88.80	-37.67	0.0126	-12.75	0.0002	-1.2449	0.0123	102.40	312	96	105	125	244	882	164
325	Area	1	Yes		cmmap2	89.01	-36.43	0.0077	-12.62	0.0001	-1.2559	0.0077	102.65	314	86	111	132	248	891	167
326	Area	1	No		reanalysis	63.43	-58.62	-0.1007	-3.29	0.1487	-43.8777	-0.3857	76.66	43	245	361	303	31	983	200
327	Area	1	No		gpcp	78.50	-44.59	-0.0825	-9.08	0.1810	-39.0449	-0.4254	99.48	214	145	320	347	225	1251	278
328	Area	1	No		gpcn	75.09	-33.41	-0.0374	-5.69	0.4732	-422.1661	-0.6879	89.93	178	47	216	445	150	1036	221
329	Area	1	No		gpcn trmm	87.95	-66.43	-0.2215	-18.93	0.4454	-20.1620	-0.6674	115.03	299	296	454	440	339	1828	405
330	Area	1	No		gpcn cams grid	67.97	-44.59	-0.0699	-5.84	0.1705	-54.9087	-0.4130	85.72	114	146	285	328	110	983	201
331	Area	1	No		gpi	106.94	-85.89	-0.0574	-9.60	0.0868	-52.8102	-0.2947	130.04	400	406	259	264	413	1742	380
332	Area	1	No		trmm 3a46	112.02	-97.24	-0.1713	-23.95	0.2928	-28.0136	-0.5411	136.99	426	432	435	413	433	2139	451
333	Area	1	No		trmm 3b43	86.96	-67.17	-0.2275	-18.11	0.4651	-20.3384	-0.6820	115.17	288	301	458	442	341	1830	406
334	Area	1	No		ssmi	91.92	-69.62	-0.0712	-11.35	0.0631	-31.8993	-0.2883	109.96	343	328	287	280	303	1521	335
335	Area	1	No		opi2	62.87	-26.66	-0.1248	-9.82	0.2608	-23.8724	-0.5107	77.44	34	9	408	403	38	892	168
336	Area	1	No		opi3	66.39	-33.20	-0.1092	-9.34	0.2203	-27.1005	-0.4694	82.31	86	39	390	373	77	965	193
337	Area	1	No		cmmap	67.45	-37.00	-0.1054	-9.10	0.2072	-28.2225	-0.4552	84.09	103	92	378	365	95	1033	220
338	Area	1	No		cmmap2	67.66	-35.76	-0.1041	-9.27	0.2065	-28.3052	-0.4545	84.21	111	80	373	364	99	1027	218
339	Area	1	1	mon	reanalysis	63.89	-58.56	-0.0458	-6.05	0.0308	-42.2484	-0.1754	75.31	51	244	230	211	24	760	111
340	Area	1	1	mon	gpcp	79.49	-44.29	-0.0931	-8.78	0.2298	-39.3546	-0.4794	100.03	226	135	349	380	230	1320	295
341	Area	1	1	mon	gpcn	76.03	-33.24	-0.0464	-5.44	0.7289	-427.7367	-0.8538	90.79	192	40	232	464	159	1087	245
342	Area	1	1	mon	gpcn trmm	90.55	-67.42	-0.2415	-17.97	0.5248	-20.4972	-0.7245	116.92	330	307	461	453	355	1906	422
343	Area	1	1	mon	gpcn cams grid	68.83	-44.69	-0.0812	-5.43	0.2298	-55.6816	-0.4794	86.38	129	148	315	379	120	1091	247
344	Area	1	1	mon	gpi	108.52	-85.79	-0.0779	-8.16	0.1596	-53.7272	-0.3995	131.44	413	404	305	313	421	1856	413
345	Area	1	1	mon	trmm 3a46	118.47	-100.09	-0.2659	-17.58	0.6764	-29.8412	-0.8224	142.64	447	441	466	462	451	2267	464
346	Area	1	1	mon	trmm 3b43	87.14	-66.05	-0.2547	-17.36	0.5748	-20.2770	-0.7581	115.96	293	294	464	458	348	1856	412
347	Area	1	1	mon	ssmi	94.19	-69.48	-0.1062	-9.48	0.1845	-32.8916	-0.4295	111.96	364	326	383	350	316	1739	379
348	Area	1	1	mon	opi2	63.62	-26.90	-0.1324	-9.69	0.2926	-24.0968	-0.5409	77.91	45	11	418	412	42	926	182
349	Area	1	1	mon	opi3	67.04	-33.47	-0.1167	-9.17	0.2508	-27.3479	-0.5008	82.80	94	48	401	399	82	1024	215
350	Area	1	1	mon	cmmap	68.25	-37.10	-0.1116	-8.95	0.2322	-28.4600	-0.4819	84.57	121	94	395	384	101	1095	248
351	Area	1	1	mon	cmmap2	68.39	-35.86	-0.1100	-9.13	0.2303	-28.5351	-0.4799	84.68	124	82	391	381	102	1080	243
352	Area	1	2	mon	reanalysis	63.98	-58.52	0.0045	-8.59	0.0003	-40.7483	0.0172	74.04	53	241	115	17	547	50	
353	Area	1	2	mon	gpcp	79.27	-44.27	-0.0873	-8.98	0.2023	-39.0204	-0.4498	99.79	222	133	336	362	228	1281	284
354	Area	1	2	mon	gpcn	75.57	-32.90	-0.0422	-5.53	0.6070	-426.4320	-0.7791	90.60	185	34	221	460	155	1055	232
355	Area	1	2	mon	gpcn trmm	91.00	-68.30	-0.2189	-18.57	0.4274	-20.0448	-0.6538	116.65	333	316	453	438	351	1891	418
356	Area	1	2	mon	gpcn cams grid	68.36	-44.80	-0.0731	-5.70	0.1860	-55.1165	-0.4313	85.99	122	151	294	351	114	1032	219
357	Area	1	2	mon	gpi	109.13	-85.98	-0.0818	-7.87	0.1760	-53.9046	-0.4195	131.98	416	408	319	340	422	1905	421
358	Area	1	2	mon	trmm 3a46	121.74	-102.31	-0.2983	-14.88	0.2544	-30.3034	-0.9085	145.44	459	450	468	468	458	2303	468
359	Area	1	2	mon	trmm 3b43	86.88	-66.83	-0.2247	-18.24	0.4442	-19.7044	-0.6665	115.32	286	298	457	439	342	1823	402
360	Area	1	2	mon	ssmi	95.28	-69.63	-0.1194	-8.77	0.2331	-33.2991	-0.4828	112.96	369	329	405	386	323	1812	398
361	Area	1	2	mon	opi2	62.53	-27.13	-0.1178	-9.90	0.2307	-23.5837	-0.4803	77.24	31	14	402	382	35	864	157
362	Area	1	2	mon	opi3	66.07	-33.72	-0.1035	-9.44	0.1968	-26.8421	-0.4436	82.20	84	53	369	359	75	940	188
363	Area	1	2	mon	cmmap	67.55	-37.07	-0.0981	-9.31	0.1792	-27.9256	-0.4233	83.94	105	93	356	345	91	990	207
364	Area	1	2	mon	cmmap2	67.84	-35.82	-0.0965	-9.47	0.1775	-27.9948	-0.4213	84.04	107	81	355	342	93	978	196
365	Area	1	3	mon	reanalysis	63.71	-58.50	0.0399	-10.38	0.0234	-39.7097	0.1531	73.15	46	240	71	67	12	436	28
366	Area	1	3	mon	gpcp	77.94	-44.43	-0.0572	-9.95	0.0868	-37.4234	-0.2946	97.95	210	139	257	263	210	1079	242
367	Area	1	3	mon	gpcn	74.87	-33.05	-0.0255	-5.92	0.2232	-418.8135	-0.4725	89.71	175	37	204	377	146	939	187
368	Area	1	3	mon	gpcn trmm	88.49	-69.70	-0.1330	-21.81	0.1546	-18.4505	-0.3932	113.11	308	332	417	310	325	1692	371
369	Area	1	3	mon	gpcn cams grid	67.07	-44.92	-0.0450	-6.70	0.0702	-53.1813	-0.2650	84.50	9						

Larger Rondonia - Monthly Surface																			
Ol	Runoff Meth	Re	Storage	Precipitation	Absolute Error	Bias	Slope	Intercept	R <sup>2</sup>	R <sup>2</sup> 1:1	Corr Coef	RMSE	AER	BR	SR	CCR	RMR	TR	BR
405	Area	2	No	gpcp	89.47	-63.25	-0.0807	-16.66	0.1453	-37.8480	-0.3812	114.17	317	273	314	300	333	1537	338
406	Area	2	No	gpcp	87.47	-55.55	-0.0826	-13.68	0.2186	-50.9797	-0.4676	108.56	295	219	321	372	289	1496	330
407	Area	2	No	gpcp trmm	96.84	-75.56	-0.1612	-20.03	0.3123	-26.2491	-0.5589	122.60	375	370	428	418	380	1971	431
408	Area	2	No	ghcn cams grid	83.96	-63.08	-0.0952	-16.05	0.1564	-32.4107	-0.3954	105.88	258	267	354	312	267	1458	324
409	Area	2	No	gpi	123.25	-104.20	-0.0789	-13.13	0.1761	-64.8954	-0.4196	149.23	463	463	307	341	467	2041	443
410	Area	2	No	trmm 3a46	117.30	-98.33	-0.1166	-21.11	0.1743	-32.3114	-0.4175	139.72	444	436	400	337	444	2061	445
411	Area	2	No	trmm 3b43	96.84	-77.61	-0.1678	-19.37	0.3373	-26.9172	-0.5808	124.09	376	382	431	421	385	1995	435
412	Area	2	No	ssmi	108.18	-86.18	-0.1036	-14.22	0.2076	-43.3594	-0.4557	128.39	411	409	371	366	403	1960	429
413	Area	2	No	opi2	72.85	-45.39	-0.1257	-16.97	0.2227	-23.7787	-0.4719	91.18	152	163	409	376	161	1261	279
414	Area	2	No	opi3	76.00	-51.35	-0.1055	-16.85	0.1711	-26.4686	-0.4137	96.00	191	176	379	329	191	1266	281
415	Area	2	No	cmep	77.54	-54.55	-0.1031	-16.59	0.1650	-27.5771	-0.4062	97.92	207	213	368	322	209	1319	294
416	Area	2	No	cmep2	77.28	-53.30	-0.1014	-16.78	0.1629	-27.5197	-0.4036	97.82	202	201	363	318	207	1291	288
417	Area	2	1 mon	reanalysis	95.55	-73.59	-0.0785	-15.96	0.1316	-40.7739	-0.3628	118.60	372	354	306	287	364	1683	367
418	Area	2	1 mon	gpcp	89.50	-62.92	-0.0733	-17.02	0.1193	-37.2413	-0.3454	113.47	318	282	295	274	329	1478	328
419	Area	2	1 mon	gpcp	88.26	-55.46	-0.0714	-14.13	0.1634	-50.1091	-0.4042	108.06	304	218	288	320	286	1416	312
420	Area	2	1 mon	gpcp trmm	96.90	-74.17	-0.2034	-18.85	0.4941	-26.7551	-0.7029	123.90	377	364	449	449	384	2023	439
421	Area	2	1 mon	ghcn cams grid	84.00	-62.89	-0.0850	-16.52	0.1247	-31.9152	-0.3531	105.27	259	257	328	279	264	1387	308
422	Area	2	1 mon	gpi	124.57	-103.75	-0.0866	-12.53	0.2116	-64.7631	-0.4600	149.45	467	459	332	368	468	2095	448
423	Area	2	1 mon	trmm 3a46	122.34	-101.87	-0.1934	-15.77	0.4658	-34.9531	-0.6825	145.28	460	448	444	443	457	2252	462
424	Area	2	1 mon	trmm 3b43	96.42	-76.46	-0.2024	-18.43	0.4908	-27.3944	-0.7006	125.32	374	375	448	448	389	2034	442
425	Area	2	1 mon	ssmi	109.75	-85.15	-0.1221	-12.79	0.2668	-43.3945	-0.5354	128.58	416	396	406	409	404	2033	441
426	Area	2	1 mon	opi2	72.84	-45.18	-0.1077	-17.46	0.1631	-23.1517	-0.4039	90.18	151	158	388	319	152	1168	258
427	Area	2	1 mon	opi3	75.56	-51.10	-0.0934	-17.26	0.1334	-25.9226	-0.3653	95.21	184	171	350	291	183	1179	259
428	Area	2	1 mon	cmep	77.32	-54.28	-0.0912	-17.04	0.1285	-27.0209	-0.3585	97.13	203	207	346	285	201	1242	275
429	Area	2	1 mon	cmep2	76.92	-53.04	-0.0887	-17.24	0.1242	-26.9354	-0.3525	96.98	198	195	341	278	198	1210	266
430	Area	2	2 mon	reanalysis	94.33	-73.76	-0.0893	-16.51	0.1029	-40.5402	-0.3208	118.26	365	357	283	267	360	1632	352
431	Area	2	2 mon	gpcp	88.02	-62.94	-0.0558	-17.84	0.0695	-36.5667	-0.2637	112.46	301	263	252	253	321	1390	309
432	Area	2	2 mon	gpcp	86.50	-55.32	-0.0515	-14.97	0.0852	-48.7486	-0.2918	106.90	283	217	242	262	278	1282	285
433	Area	2	2 mon	gpcp trmm	95.90	-74.24	-0.1743	-20.39	0.3641	-25.9397	-0.6034	123.21	373	365	436	426	381	1981	434
434	Area	2	2 mon	ghcn cams grid	82.08	-62.91	-0.0614	-17.60	0.0652	-31.1810	-0.2554	104.09	246	260	271	244	255	1276	283
435	Area	2	2 mon	gpi	123.66	-103.69	-0.0800	-13.16	0.1808	-64.2960	-0.4252	149.17	465	458	310	346	466	2045	444
436	Area	2	2 mon	trmm 3a46	124.70	-104.35	-0.2054	-14.69	0.5125	-35.3337	-0.7159	147.49	468	464	451	452	462	2297	467
437	Area	2	2 mon	trmm 3b43	95.43	-76.32	-0.1773	-19.88	0.3773	-26.5925	-0.6142	124.69	370	373	438	427	387	1995	436
438	Area	2	2 mon	ssmi	109.39	-85.26	-0.1193	-13.21	0.2760	-43.6105	-0.5254	128.71	417	398	404	406	405	2030	440
439	Area	2	2 mon	opi2	71.15	-45.27	-0.0703	-18.46	0.0698	-22.1834	-0.2641	88.34	144	159	286	254	135	978	197
440	Area	2	2 mon	opi3	73.67	-51.15	-0.0665	-18.17	0.0680	-25.1812	-0.2607	93.88	165	172	278	250	175	1040	224
441	Area	2	2 mon	cmep	75.23	-54.35	-0.0648	-18.01	0.0651	-26.2932	-0.2552	95.86	180	209	276	243	190	1098	249
442	Area	2	2 mon	cmep2	74.99	-53.10	-0.0624	-18.17	0.0617	-26.1919	-0.2484	95.68	176	197	273	239	187	1072	237
443	Area	2	3 mon	reanalysis	92.11	-73.96	-0.0232	-18.94	0.0115	-38.5940	-0.1072	115.59	346	361	202	202	345	1456	323
444	Area	2	3 mon	gpcp	85.21	-63.04	-0.0221	-19.24	0.0109	-35.0844	-0.1043	110.35	273	266	198	201	307	1245	276
445	Area	2	3 mon	gpcp	84.00	-55.60	-0.0126	-16.47	0.0051	-46.2500	-0.0716	104.58	260	221	173	187	260	1101	250
446	Area	2	3 mon	gpcp trmm	94.01	-75.30	-0.1113	-23.13	0.1473	-24.3781	-0.3838	120.90	361	369	393	302	376	1801	395
447	Area	2	3 mon	ghcn cams grid	79.03	-63.02	-0.0190	-19.37	0.0063	-29.7316	-0.0791	101.83	220	265	186	195	240	1106	252
448	Area	2	3 mon	gpi	121.86	-103.82	-0.0435	-16.24	0.0535	-62.1650	-0.2312	147.08	458	460	225	228	461	1832	408
449	Area	2	3 mon	trmm 3a46	123.99	-106.01	-0.1571	-18.03	0.2954	-34.0457	-0.5435	146.65	466	468	427	414	460	2235	459
450	Area	2	3 mon	trmm 3b43	93.63	-77.85	-0.1142	-22.72	0.1546	-25.0291	-0.3932	122.44	355	383	398	309	379	1824	403
451	Area	2	3 mon	ssmi	107.06	-85.44	-0.0677	-16.70	0.0902	-42.1718	-0.3003	126.11	403	402	281	285	392	1743	381
452	Area	2	3 mon	opi2	67.94	-45.42	-0.0222	-19.62	0.0069	-20.8394	-0.0833	85.85	113	164	199	197	112	785	120
453	Area	2	3 mon	opi3	70.68	-51.29	-0.0215	-19.51	0.0071	-23.8155	-0.0842	91.51	143	175	196	198	164	876	160
454	Area	2	3 mon	cmep	72.34	-54.45	-0.0192	-19.52	0.0057	-24.8967	-0.0758	93.48	149	210	188	191	172	910	174
455	Area	2	3 mon	cmep2	72.08	-53.20	-0.0195	-19.54	0.0060	-24.8487	-0.0775	93.39	148	198	189	192	171	898	172
456	Area	2	4 mon	reanalysis	89.52	-74.34	0.0214	-21.33	0.0098	-36.6792	0.0988	112.96	319	367	93	86	324	1189	262
457	Area	2	4 mon	gpcp	82.58	-63.24	0.0207	-21.07	0.0096	-33.1731	0.0979	107.58	249	272	94	87	283	985	203
458	Area	2	4 mon	gpcp	81.60	-58.29	0.0329	-18.27	0.0345	-43.3262	0.1857	101.68	240	227	79	55	238	839	146
459	Area	2	4 mon	gpcp trmm	91.20	-76.93	-0.0199	-27.35	0.0246	-22.1905	-0.0682	116.88	336	377	191	185	354	1443	316
460	Area	2	4 mon	ghcn cams grid	76.34	-63.27	0.0344	-21.66	0.0204	-27.8843	0.1430	98.90	196	274	77	71	217	835	142
461	Area	2	4 mon	gpi	119.52	-104.17	0.0025	-20.13	0.0002	-59.5765	0.0133	144.39	450	462	129	122	456	1619	350
462	Area	2	4 mon	trmm 3a46	119.67	-105.74	-0.0609	-25.36	0.0444	-31.1881	-0.2107	142.34	451	467	270	218	450	1856	414
463	Area	2	4 mon	trmm 3b43	91.02	-79.22	-0.0190	-27.35	0.0042	-22.7558	-0.0649	118.29	334	384	187	182	362	1449	317
464	Area	2	4 mon	ssmi	104.08	-85.96	-0.0105	-20.79	0.0022	-40.6968	-0.0468	123.32	389	407	167	168	383	1514	334
465	Area	2	4 mon	opi2	64.74	-45.58	0.0358	-21.08	0.0180	-19.2048	0.1343	82.72	60	166	76	74	80	456	32
466	Area	2	4 mon	opi3	68.22	-51.50	0.0323	-21.19	0.0160	-22.1642	0.1264	88.57	120	178	81	78	137	594	64
467	Area	2	4 mon	cmep	69.91	-54.64	0.0324	-21.29	0.0162	-23.2962	0.1274	90.71	139	214	80	77	158	668	89
468	Area	2	4 mon	cmep2	69.62	-53.39	0.0321	-21.24	0.0163	-23.2158	0.1276	90.56	136	202	82	76	154	650	81

Larger Rondonia - Yearly Surface																			
Ol	Runoff Meth	Re	Storage	Precipitation	Absolute Error	Bias	Slope	Intercpt	R <sup>2</sup>	R <sup>2</sup> 1.1	Corr Coef	RMSE	AER	BR	SR	CCR	RMR	TR	BR
1	Linear	1	Yes	reanalysis	57.60	-57.51	-0.0881	-2.81	0.0558	-39.3236	-0.2362	65.72	77	78	50	76	87	368	76
2	Linear	1	Yes	gpcp	45.61	-44.66	-0.2074	-5.08	0.1120	-21.0760	-0.3347	50.34	50	49	92	97	49	337	63
3	Linear	1	Yes	gpcp	35.61	-35.61	0.1745	-13.56	0.2375	-40.6846	0.4874	38.53	23	27	17	3	13	83	2
4	Linear	1	Yes	gpcp trmm	66.30	-66.30	-0.5988	-0.11	0.9442	-50.9236	-0.9717	71.11	105	105	136	153	107	606	137
5	Linear	1	Yes	ghcn cams grid	43.41	-43.32	-0.1079	-3.37	0.0135	-18.6128	-0.1161	46.26	39	40	58	59	36	232	16
6	Linear	1	Yes	gpi	86.31	-86.31	0.1768	-26.92	0.0723	-60.9552	0.2689	88.10	142	142	16	14	141	455	114
7	Linear	1	Yes	trmm 3a46	84.44	-84.44	-0.8250	27.92	0.8976	-66.3514	-0.9474	87.41	135	135	153	146	139	708	149
8	Linear	1	Yes	trmm 3b43	81.16	-81.16	-0.7453	8.61	0.8638	-31.9756	-0.9294	66.25	87	87	149	141	91	555	129
9	Linear	1	Yes	ssmi	69.80	-69.80	0.1300	-22.02	0.0242	-32.1652	0.1556	72.07	112	112	27	29	109	389	86
10	Linear	1	Yes	opi2	29.90	-26.82	-0.2859	-7.61	0.2106	-10.3153	-0.4590	36.04	5	5	122	119	5	256	30
11	Linear	1	Yes	opi3	34.13	-32.78	-0.1571	-8.60	0.0547	-12.2772	-0.2340	39.94	15	13	74	74	16	192	8
12	Linear	1	Yes	cmapp	37.12	-35.97	-0.1560	-8.12	0.0546	-14.2188	-0.2337	41.80	30	31	72	73	29	235	17
13	Linear	1	Yes	cmapp2	35.98	-34.73	-0.1616	-8.19	0.0805	-13.5745	-0.2460	40.90	27	24	76	79	25	231	15
14	Linear	1	No	reanalysis	58.42	-58.42	-0.0824	-4.01	0.0985	-81.1961	-0.3139	66.05	80	80	48	89	90	387	85
15	Linear	1	No	gpcp	44.36	-44.17	-0.2220	-4.09	0.1493	-24.1548	-0.3864	49.81	44	44	102	108	47	345	66
16	Linear	1	No	gpcp	33.18	-33.18	0.0081	-6.68	0.0258	-999.0000	0.1607	37.08	11	15	35	28	8	97	4
17	Linear	1	No	gpcp trmm	72.09	-72.09	-0.3410	-16.57	0.6712	-126.6829	-0.8193	75.32	118	118	127	130	118	611	138
18	Linear	1	No	ghcn cams grid	44.25	-44.25	-0.1123	-4.13	0.0295	-39.1310	-0.1718	46.55	42	45	60	67	39	253	28
19	Linear	1	No	gpi	85.28	-85.28	0.1700	-25.41	0.0693	-61.7092	0.2632	87.10	139	139	19	16	138	451	111
20	Linear	1	No	trmm 3a46	92.50	-92.50	-0.6029	6.19	0.9216	-150.6182	-0.9600	94.58	147	147	137	150	147	728	152
21	Linear	1	No	trmm 3b43	65.79	-65.79	-0.6607	0.59	0.6923	-36.8538	-0.8321	70.28	102	104	144	131	102	583	132
22	Linear	1	No	ssmi	69.98	-69.98	0.0090	-15.55	0.0001	-35.9765	0.0112	72.53	113	113	34	36	111	407	94
23	Linear	1	No	opi2	28.82	-26.30	-0.2751	-7.26	0.2270	-11.6643	-0.4764	35.34	3	3	120	122	3	251	24
24	Linear	1	No	opi3	32.89	-32.27	-0.2485	-6.18	0.1594	-14.4221	-0.3992	39.00	10	10	113	112	15	260	32
25	Linear	1	No	cmapp	35.87	-35.46	-0.2141	-6.21	0.1197	-16.4803	-0.3460	41.52	26	26	96	100	28	276	40
26	Linear	1	No	cmapp2	34.73	-34.22	-0.2170	-6.41	0.1270	-15.7388	-0.3564	40.63	20	20	97	103	24	264	35
27	Linear	2	Yes	reanalysis	75.15	-75.15	0.1144	-27.89	0.0276	-13.8365	0.1662	82.85	124	124	30	26	130	434	107
28	Linear	2	Yes	gpcp	66.07	-64.89	-0.1187	-16.66	0.0063	-9.6546	-0.0793	70.21	103	101	63	47	101	415	98
29	Linear	2	Yes	gpcp	60.54	-60.54	0.3305	-34.65	0.0817	-16.3143	0.2858	62.86	86	86	9	13	78	272	37
30	Linear	2	Yes	gpcp trmm	59.97	-59.97	-0.0851	-8.08	0.0081	-21.6381	-0.0902	63.29	84	84	49	49	79	345	67
31	Linear	2	Yes	ghcn cams grid	65.19	-64.72	-0.3579	-6.41	0.0266	-9.4105	-0.1831	69.40	99	99	128	64	99	489	122
32	Linear	2	Yes	gpi	106.60	-106.60	0.2858	-46.32	0.0582	-33.1441	0.2413	108.70	155	155	12	22	155	499	125
33	Linear	2	Yes	trmm 3a46	77.63	-77.63	0.4396	-39.37	0.1035	-29.4812	0.3217	79.03	130	130	6	10	124	400	89
34	Linear	2	Yes	trmm 3b43	62.02	-62.02	0.1658	-20.41	0.0238	-22.3136	0.1544	64.23	88	88	21	30	81	308	49
35	Linear	2	Yes	ssmi	83.75	-83.75	0.7058	-64.49	0.2209	-18.6692	0.4700	85.55	134	134	2	5	133	408	96
36	Linear	2	Yes	opi2	49.86	-47.03	-0.0592	-20.29	0.0017	-5.3327	-0.0416	54.13	55	54	44	42	56	251	26
37	Linear	2	Yes	opi3	54.60	-52.99	-0.0287	-20.88	0.0003	-6.4484	-0.0172	58.70	65	61	42	40	67	275	39
38	Linear	2	Yes	cmapp	57.58	-56.18	-0.1052	-18.16	0.0040	-7.2659	-0.0836	61.84	76	74	57	44	75	326	57
39	Linear	2	Yes	cmapp2	56.45	-54.94	-0.1036	-18.35	0.0041	-6.9813	-0.0637	60.76	72	69	56	45	70	312	53
40	Linear	2	No	reanalysis	73.43	-73.43	-0.0482	-17.49	0.0572	-162.8807	-0.2392	80.64	119	119	43	77	128	486	120
41	Linear	2	No	gpcp	63.17	-63.17	-0.0948	-15.97	0.0468	-108.7629	-0.2163	65.39	93	93	53	71	86	396	88
42	Linear	2	No	gpcp	55.97	-55.97	-0.1746	-10.54	0.2419	-156.3035	-0.4919	58.18	69	72	81	126	66	414	97
43	Linear	2	No	gpcp trmm	71.91	-71.91	-0.5119	0.40	0.8576	-91.7030	-0.9261	75.06	117	117	133	139	117	623	141
44	Linear	2	No	ghcn cams grid	63.00	-63.00	-0.1689	-12.81	0.0690	-103.2641	-0.2628	64.32	91	91	79	84	82	427	104
45	Linear	2	No	gpi	103.55	-103.55	-0.0205	-17.41	0.0027	-285.4641	-0.0519	104.97	153	153	41	43	153	453	128
46	Linear	2	No	trmm 3a46	93.56	-93.56	-0.7928	28.85	0.9154	-119.9281	-0.9567	95.45	148	148	151	148	148	743	154
47	Linear	2	No	trmm 3b43	73.96	-73.96	-0.8033	6.00	0.9188	-95.5630	-0.9585	76.61	122	122	138	149	120	651	146
48	Linear	2	No	ssmi	85.22	-85.22	-0.2061	-6.26	0.1530	-163.6364	-0.3911	86.84	138	138	91	109	137	613	139
49	Linear	2	No	opi2	45.31	-45.31	-0.1642	-15.91	0.1553	-59.3806	-0.3940	48.95	48	52	78	111	45	334	60
50	Linear	2	No	opi3	51.27	-51.27	-0.1575	-15.15	0.1032	-71.7336	-0.3213	53.72	57	57	75	91	55	335	61
51	Linear	2	No	cmapp	54.46	-54.46	-0.1569	-14.66	0.1049	-80.3554	-0.3238	56.82	63	67	73	93	63	359	72
52	Linear	2	No	cmapp2	53.22	-53.22	-0.1465	-15.20	0.0946	-77.0839	-0.3076	55.66	61	63	68	88	61	341	64
53	Point	1	Yes	reanalysis	56.31	-56.17	-0.0754	-2.11	0.0495	-40.1680	-0.2085	64.35	71	73	46	70	83	343	65
54	Point	1	Yes	gpcp	44.10	-43.02	-0.2035	-3.54	0.1162	-21.3217	-0.3410	48.76	41	39	90	96	44	312	52
55	Point	1	Yes	gpcp	34.63	-34.63	0.1664	-12.35	0.2112	-38.0095	0.4596	37.69	19	22	20	6	12	79	1
56	Point	1	Yes	gpcp trmm	63.42	-63.42	-0.6196	3.63	0.9585	-44.7453	-0.9791	68.55	96	96	140	156	97	585	133
57	Point	1	Yes	ghcn cams grid	42.13	-41.97	-0.0965	-2.42	0.0115	-18.6674	-0.1072	44.89	35	35	54	51	33	208	11
58	Point	1	Yes	gpi	84.76	-84.76	0.1515	-23.55	0.0590	-65.5182	0.2428	86.60	136	136	24	21	135	452	112
59	Point	1	Yes	trmm 3a46	81.73	-81.73	-0.8496	32.15	0.8765	-57.5043	-0.9362	84.90	131	131	154	144	132	692	148
60	Point	1	Yes	trmm 3b43	58.68	-58.68	-0.7343	10.64	0.8859	-31.3942	-0.9412	63.88	82	82	148	145	80	537	127
61	Point	1	Yes	ssmi	68.15	-68.15	0.1091	-19.22	0.0191	-34.4589	0.1382	70.42	108	108	31	32	103	382	83
62	Point	1	Yes	opi2	28.51	-26.16	-0.2837	-5.98	0.2235	-10.2929	-0.4727	34.68	2	2	121	121	2	248	23
63	Point	1	Yes	opi3	32.63	-31.12	-0.1491	-7.10	0.0532	-12.1951	-0.2306	37.49	9	8	69	72	11	169	7
64	Point	1	Yes	cmapp	35.61	-34.31	-0.1509	-6.58	0.0551	-14.2014	-0.2347	40.24	24	21	70	75	22	212	12
65	Point	1	Yes	cmapp2	34.48	-33.07	-0.1560	-6.66	0.0608	-13.5423	-0.2485	39.36	17	14	71	80	18	200	10
66	Point	1	No	reanalysis	57.07	-57.07	-0.0697	-3.30	0.0812	-89.9503	-0.2850	64.85	74	76	45	85	85	365	75
67	Point	1	No	gpcp	42.86	-42.51	-0.2181	-2.56	0.1625	-25.5511	-0.4031	48.21	36	37	99	114	43	329	58
68	Point	1	No	gpcp	32.19	-32.19	0.0000	-5.47	0.0000	-999.0000	0.0005	36.26	8	9	38	38	6	99	5
69	Point	1	No	gpcp trmm	69.21	-69.21	-0.3618	-12.83	0.7183	-111.9237	-0.8475	72.66	111	111	129	134	112	597	134
70	Point	1	No	ghcn cams grid	42.90	-42.90	-0.1010	-3.19	0.0275	-42.5821	-0.1659	45.15	37	38					

Larger Rondonia - Yearly Surface																			
OT	Runoff Meth	Re	Storage	Precipitation	Absolute Error	Bias	Slope	Intercept	R <sup>2</sup>	R <sup>2</sup> 1.1	Corr Coef	RMSE	AER	BR	SR	CCR	RMR	TR	BR
102	Point	2	No	opi3	40.16	-40.16	-0.2746	-0.37	0.1316	-19.7597	-0.3630	44.31	32	33	119	106	32	322	56
103	Point	2	No	cmmap	43.35	-43.35	-0.2491	-0.37	0.1108	-22.5467	-0.3329	47.19	38	41	114	96	41	330	59
104	Point	2	No	cmmap2	42.10	-42.10	-0.2414	-0.94	0.1078	-21.4893	-0.3283	46.11	34	36	112	95	35	312	51
105	Area	1	Yes	reanalysis	57.81	-57.71	-0.0938	-2.73	0.0600	-37.6064	-0.2449	68.01	79	79	52	78	89	377	79
106	Area	1	Yes	gpcc	46.01	-45.10	-0.2244	-4.94	0.1200	-19.6782	-0.3465	50.91	51	51	104	101	51	358	71
107	Area	1	Yes	gpcc	35.84	-35.84	0.1743	-13.78	0.2393	-41.5838	<b>0.4892</b>	<b>38.74</b>	25	28	18	2	14	87	3
108	Area	1	Yes	gpcc trmm	67.97	-67.97	-0.6381	-0.15	0.9570	-47.6885	-0.9782	72.89	107	107	142	155	113	624	142
109	Area	1	Yes	ghcn cams grid	43.61	-43.53	-0.1105	-3.48	0.0134	-17.8245	-0.1159	46.52	40	42	59	58	37	236	18
110	Area	1	Yes	gpi	86.91	-86.91	<b>0.1836</b>	-28.01	0.0714	-56.5427	<b>0.2672</b>	88.73	144	144	14	15	143	460	116
111	Area	1	Yes	trmm 3a46	86.16	-86.16	-0.8873	30.04	0.9077	-60.4082	-0.9527	89.26	141	141	155	147	144	728	151
112	Area	1	Yes	trmm 3b43	62.54	-62.54	-0.8002	9.41	0.8763	-29.4269	-0.9361	67.83	90	90	152	143	95	570	131
113	Area	1	Yes	ssmi	70.54	-70.54	0.1247	-22.46	0.0204	-30.1435	0.1429	72.90	114	114	29	31	114	402	92
114	Area	1	Yes	opi2	<b>30.28</b>	<b>-27.23</b>	-0.3067	-7.72	0.2220	-9.7237	-0.4712	<b>36.66</b>	6	6	125	120	7	264	34
115	Area	1	Yes	opi3	34.54	-33.20	-0.1748	-8.64	0.0620	-11.5332	-0.2491	39.64	18	16	82	82	20	218	13
116	Area	1	Yes	cmmap	37.52	-36.39	-0.1720	-8.15	0.0608	-13.3262	-0.2465	42.38	31	32	80	81	31	255	29
117	Area	1	Yes	cmmap2	36.39	-35.15	-0.1777	-8.24	0.0671	-12.7308	-0.2589	41.49	29	25	83	83	27	247	21
118	Area	1	No	reanalysis	58.62	-58.62	-0.0881	-3.93	0.1003	-72.9307	-0.3167	66.34	81	81	51	90	92	395	87
119	Area	1	No	gpcc	44.77	-44.59	-0.2390	-3.95	0.1331	-21.7879	-0.3912	50.41	47	48	110	110	50	365	74
120	Area	1	No	gpcc	33.41	-33.41	0.0079	-6.90	0.0267	-999.0000	0.1633	<b>37.28</b>	14	17	36	27	9	103	6
121	Area	1	No	gpcc trmm	73.76	-73.76	-0.3803	-16.61	0.6974	-110.7851	-0.8351	77.11	121	121	130	132	121	625	144
122	Area	1	No	ghcn cams grid	44.45	-44.45	-0.1150	-4.24	0.0276	-35.1874	-0.1660	46.82	45	47	82	66	40	260	33
123	Area	1	No	gpi	85.89	-85.89	<b>0.1788</b>	-26.50	0.0667	-56.6743	0.2583	87.75	140	140	15	17	140	452	113
124	Area	1	No	trmm 3a46	84.21	-84.21	-0.8651	8.31	0.9376	-130.6803	-0.9683	96.41	149	149	145	152	150	745	155
125	Area	1	No	trmm 3b43	67.17	-67.17	-0.7156	1.40	0.7119	-33.6963	-0.8437	71.86	106	106	147	133	108	600	135
126	Area	1	No	ssmi	70.72	-70.72	0.0036	-16.00	0.0000	-32.7757	0.0043	73.37	115	115	37	37	115	419	101
127	Area	1	No	opi2	<b>29.21</b>	<b>-26.72</b>	-0.2960	-7.37	0.2323	-10.6261	-0.4820	<b>36.01</b>	4	4	123	124	4	259	31
128	Area	1	No	opi3	<b>33.29</b>	<b>-32.69</b>	-0.2662	-6.22	0.1618	-13.0801	-0.4022	39.63	13	12	117	113	19	274	38
129	Area	1	No	cmmap	36.28	-35.88	-0.2301	-6.24	0.1223	-14.9177	-0.3497	42.13	28	29	106	102	30	295	45
130	Area	1	No	cmmap2	35.14	-34.64	-0.2332	-6.46	0.1297	-14.2556	-0.3602	41.25	22	23	107	105	26	283	42
131	Area	2	Yes	reanalysis	75.23	-75.23	0.1560	-30.19	0.0514	-13.6909	0.2267	62.43	125	125	23	25	129	427	105
132	Area	2	Yes	gpcc	66.28	-64.97	-0.1880	-13.76	0.0158	-9.7419	-0.1256	70.48	104	102	86	80	104	456	115
133	Area	2	Yes	gpcc	60.12	-60.12	<b>0.2877</b>	-32.57	0.0634	-16.5441	0.2517	62.53	85	85	11	19	77	277	41
134	Area	2	Yes	gpcc trmm	63.62	-63.62	-0.1343	-9.37	0.0225	-26.9463	-0.1500	66.78	97	97	68	83	93	416	99
135	Area	2	Yes	ghcn cams grid	65.39	-64.80	-0.4452	-2.74	0.0412	-9.4720	-0.2029	69.59	100	100	131	68	100	499	124
136	Area	2	Yes	gpi	107.25	-107.25	<b>0.3318</b>	-50.85	0.0837	-35.6661	<b>0.2892</b>	109.14	156	156	8	12	156	488	121
137	Area	2	Yes	trmm 3a46	81.95	-81.95	<b>0.3394</b>	-36.84	0.0649	-34.6647	0.2548	83.34	132	132	7	18	131	420	102
138	Area	2	Yes	trmm 3b43	65.67	-65.67	0.0932	-20.44	0.0084	-27.8058	0.0914	67.80	101	103	32	34	94	364	73
139	Area	2	Yes	ssmi	85.06	-85.06	<b>0.7012</b>	-65.49	0.2279	-20.1364	<b>0.4773</b>	86.74	137	137	3	4	136	417	100
140	Area	2	Yes	opi2	50.07	-47.11	-0.1303	-18.57	0.0084	-5.4208	-0.0915	54.49	58	55	65	50	59	285	43
141	Area	2	Yes	opi3	54.81	-53.07	-0.1206	-18.10	0.0052	-6.5344	-0.0721	59.03	66	62	64	46	68	306	48
142	Area	2	Yes	cmmap	57.79	-56.27	-0.1822	-15.59	0.0121	-7.3437	-0.1101	62.12	78	75	85	55	76	369	77
143	Area	2	Yes	cmmap2	56.65	-55.02	-0.1815	-15.84	0.0125	-7.0613	-0.1116	61.06	73	70	84	56	72	355	69
144	Area	2	No	reanalysis	73.51	-73.51	-0.0066	-19.79	0.0008	-118.8690	-0.0282	80.29	120	120	39	41	126	446	109
145	Area	2	No	gpcc	63.25	-63.25	-0.1641	-13.07	0.1034	-79.4815	-0.3215	65.79	94	94	77	92	88	445	108
146	Area	2	No	gpcc	55.55	-55.55	-0.2174	-8.47	0.2727	-112.6222	-0.5223	57.97	68	71	98	129	65	431	106
147	Area	2	No	gpcc trmm	75.56	-75.56	-0.5612	-0.90	0.7776	-76.1253	-0.8818	78.82	126	126	135	136	123	646	145
148	Area	2	No	ghcn cams grid	63.08	-63.08	-0.2562	-9.14	0.1172	-76.6799	-0.3424	64.64	92	92	116	99	84	483	119
149	Area	2	No	gpi	104.20	-104.20	0.0255	-21.94	0.0026	-178.6276	0.0509	105.61	154	154	33	35	154	530	126
150	Area	2	No	trmm 3a46	97.87	-97.87	-0.8930	31.38	0.9350	-105.6115	-0.9669	99.89	152	152	156	151	152	763	156
151	Area	2	No	trmm 3b43	77.61	-77.61	-0.6759	5.98	0.8703	-79.2571	-0.9329	80.40	129	129	146	142	127	673	147
152	Area	2	No	ssmi	86.53	-86.53	-0.2107	-7.26	0.1053	-111.0249	-0.3245	88.27	143	143	94	94	142	616	140
153	Area	2	No	opi2	45.39	-45.39	-0.2353	-14.20	0.2351	-44.5449	-0.4848	49.49	49	53	109	125	48	382	82
154	Area	2	No	opi3	51.35	-51.35	-0.2434	-12.36	0.1811	-53.8373	-0.4371	54.21	58	58	115	118	57	406	93
155	Area	2	No	cmmap	54.55	-54.55	-0.2340	-12.09	0.1719	-59.9261	-0.4146	57.24	84	88	108	117	64	421	103
156	Area	2	No	cmmap2	53.30	-53.30	-0.2244	-12.70	0.1638	-57.5318	-0.4047	56.11	82	84	105	115	62	408	95



Larger United States - Monthly Atmospheric																				
OI	Moist Flux	Re	Storage	Precipitation	Absolute Error	Bias	Slope	Intercept	R <sup>2</sup>	R <sup>2</sup> 1.1	Corr Coef	RMSE	AER	BR	SR	CCR	RMR	TR	BR	
1	Method 1	1	Yes	reanalysis	43.78	-24.15	-0.8554	9.35	0.1560	-0.9471	-0.3949	55.45	7	2	17	18	7	51	9	
2	Method 1	1	Yes	gpcp	42.23	-21.03	-0.2820	6.95	0.0795	-0.6504	-0.1398	52.39	1	7	2	2	1	7	1	Plot
3	Method 1	1	Yes	gpcc	42.80	-25.51	-0.3384	2.06	0.0264	-0.7828	-0.1626	54.15	3	3	7	8	2	23	5	
4	Method 1	1	Yes	gpcc trmm	42.63	-31.98	-0.3129	11.96	0.0225	-1.0446	-0.1500	55.83	2	10	6	6	8	32	7	
5	Method 1	1	Yes	ghcn cams grid	43.00	-28.39	-0.2088	13.83	0.0109	-0.8669	-0.1045	54.20	4	7	1	1	3	16	2	
6	Method 1	1	Yes	opi2	45.39	-28.46	-0.5041	1.68	0.0578	-0.9348	-0.2404	57.24	9	8	9	9	9	44	8	
7	Method 1	1	Yes	opi3	44.03	-26.80	-0.3074	5.69	0.0224	-0.8062	-0.1495	55.30	8	6	5	5	6	30	6	
8	Method 1	1	Yes	cmmap	43.42	-26.10	-0.2919	5.42	0.0207	-0.7952	-0.1438	54.56	6	5	3	3	5	22	4	
9	Method 1	1	Yes	cmmap2	43.28	-25.90	-0.2953	5.42	0.0208	-0.7845	-0.1444	54.40	5	4	4	4	4	21	3	
10	Method 1	2	Yes	reanalysis	47.31	-30.74	-0.9532	0.95	0.1541	-0.9961	-0.3926	61.23	10	9	19	17	10	65	10	
11	Method 1	2	Yes	gpcc	51.55	-36.27	-0.9326	-3.89	0.2114	-1.3969	-0.4598	67.09	12	11	18	22	12	75	11	
12	Method 1	2	Yes	gpcc	49.72	-37.52	-1.2112	-12.73	0.2183	-1.2669	-0.4673	64.90	11	12	25	23	11	82	13	
13	Method 1	2	Yes	gpcc trmm	59.15	-53.76	-1.2196	-13.29	0.3043	-2.1452	-0.5517	79.41	18	19	26	30	18	111	26	
14	Method 1	2	Yes	ghcn cams grid	52.01	-38.48	-0.9567	-6.50	0.2058	-1.4432	-0.4536	67.74	13	13	20	20	13	79	12	
15	Method 1	2	Yes	opi2	54.70	-42.79	-1.0443	-13.02	0.2696	-1.7384	-0.5192	71.72	17	17	24	27	17	102	23	
16	Method 1	2	Yes	opi3	53.26	-41.01	-0.9981	-9.98	0.2257	-1.5744	-0.4751	69.53	16	16	22	26	16	96	21	
17	Method 1	2	Yes	cmmap	53.13	-40.89	-0.9985	-9.82	0.2235	-1.5839	-0.4728	69.39	15	15	21	25	15	91	19	
18	Method 1	2	Yes	cmmap2	52.98	-40.68	-0.9986	-9.67	0.2231	-1.5518	-0.4724	69.23	14	14	23	24	14	89	16	
19	Method 2	1	Yes	reanalysis	64.39	-55.16	-1.2973	31.93	0.2060	-1.4832	-0.4539	85.11	21	20	27	21	23	112	27	Plot
20	Method 2	1	Yes	gpcp	61.87	-50.89	-0.8634	28.98	0.0688	-1.2402	-0.2623	81.35	19	18	13	13	19	82	14	
21	Method 2	1	Yes	gpcc	63.25	-55.91	-0.7888	20.83	0.0971	-1.5087	-0.3118	84.24	20	21	15	15	20	91	20	
22	Method 2	1	Yes	gpcc trmm	68.05	-63.46	-0.7706	32.67	0.0799	-1.7093	-0.2826	89.32	27	27	14	14	27	109	25	
23	Method 2	1	Yes	ghcn cams grid	66.82	-60.05	-0.4015	39.33	0.0261	-1.5288	-0.1617	85.88	25	26	8	7	24	90	18	
24	Method 2	1	Yes	opi2	67.12	-59.03	-0.8493	21.15	0.1117	-1.5751	-0.3343	88.06	26	25	16	16	26	109	24	
25	Method 2	1	Yes	opi3	66.00	-57.66	-0.6459	26.20	0.0668	-1.4709	-0.2585	86.26	24	24	12	12	25	97	22	
26	Method 2	1	Yes	cmmap	64.76	-56.21	-0.6125	26.63	0.0607	-1.4252	-0.2464	84.82	23	23	10	11	22	89	17	
27	Method 2	1	Yes	cmmap2	64.56	-55.98	-0.6160	26.70	0.0603	-1.4112	-0.2456	84.58	22	22	11	10	21	86	15	
28	Method 2	2	Yes	reanalysis	71.00	-65.67	-1.4898	25.97	0.1947	-1.6407	-0.4412	93.90	28	28	29	19	28	132	28	Plot
29	Method 2	2	Yes	gpcp	76.90	-71.67	-1.4885	17.44	0.3098	-2.1041	-0.5566	101.80	30	29	28	31	30	148	29	
30	Method 2	2	Yes	gpcc	75.68	-72.28	-1.7404	6.58	0.2945	-2.1019	-0.5427	98.62	29	30	35	29	29	152	31	
31	Method 2	2	Yes	gpcc trmm	93.73	-91.91	-1.9439	-0.58	0.4189	-2.7238	-0.6472	120.07	36	36	36	36	36	180	36	
32	Method 2	2	Yes	ghcn cams grid	78.64	-74.51	-1.5045	12.38	0.2886	-2.1737	-0.5372	102.94	31	31	30	28	31	151	30	
33	Method 2	2	Yes	opi2	82.15	-78.69	-1.6282	2.51	0.3780	-2.4614	-0.6148	107.50	35	35	34	35	35	174	35	
34	Method 2	2	Yes	opi3	80.87	-77.18	-1.5727	6.49	0.3277	-2.3333	-0.5724	105.49	34	34	33	34	34	169	34	
35	Method 2	2	Yes	cmmap	80.19	-76.42	-1.5585	8.06	0.3163	-2.2855	-0.5624	104.74	33	33	31	33	33	163	33	
36	Method 2	2	Yes	cmmap2	79.97	-76.19	-1.5646	8.25	0.3155	-2.2710	-0.5617	104.50	32	32	32	32	32	160	32	

Larger United States - Yearly Atmospheric

Ol	Moist Flux	Re	Storage	Precipitation	Absolute Error	Bias	Slope	Oslope	Intercept	R <sup>2</sup>	R <sup>2</sup> 1:1	Corr Coef	RMSE	AER	BR	SR	CCR	RMR	TR	BR	Plot
1	Method 1	1	Yes	gpcp	21.00	-21.00	0.5093	1.9635	15.61	0.0853	-10.6368	0.2920	21.98	1	1	21	19	1	43	1	Plot
2	Method 1	1	Yes	reanalysis	24.15	-24.15	0.2919	3.4258	18.50	0.0146	-4.8931	0.1207	26.70	2	2	33	33	6	76	16	
3	Method 1	1	Yes	cmmap2	25.42	-25.42	0.5687	1.7584	18.78	0.0835	-15.5539	0.2889	26.19	3	3	15	20	2	43	2	
4	Method 1	1	Yes	gpcc	25.49	-25.49	-0.0739		6.67	0.0008	-16.2750	-0.0283	26.40	4	4	36	36	3	83	18	
5	Method 1	1	Yes	cmmap	25.62	-25.62	0.5346	1.8706	18.36	0.0773	-15.8263	0.2780	26.41	5	5	17	21	4	52	4	
6	Method 1	1	Yes	opi3	25.75	-25.75	0.4832	2.0695	17.82	0.0614	-16.0032	0.2478	26.54	6	6	23	25	5	65	9	
7	Method 1	1	Yes	opi2	27.53	-27.53	0.3305	3.0257	15.81	0.0334	-18.3870	0.1827	28.34	7	7	30	31	7	82	17	
8	Method 1	1	Yes	ghcn cams grid	28.05	-28.05	0.7508	1.3319	25.09	0.0398	-6.3513	0.1995	30.09	8	8	7	30	9	62	7	
9	Method 1	1	Yes	gpcc trmm	29.71	-29.71	3.1600	3.1600	58.80	0.7486	-66.1957	0.8652	29.85	9	9	31	3	8	60	6	
10	Method 1	2	Yes	reanalysis	30.72	-30.72	0.0112	89.2857	15.65	0.0000	-13.4189	0.0055	32.13	10	10	35	35	10	100	22	
11	Method 1	2	Yes	gpcp	36.25	-36.25	0.5299	1.8871	26.48	0.0932	-18.3343	0.3052	37.21	11	11	18	15	11	66	10	
12	Method 1	2	Yes	gpcc	37.50	-37.50	1.5160	1.5160	48.22	0.3800	-20.2796	0.6165	38.10	12	12	8	6	12	50	3	
13	Method 1	2	Yes	ghcn cams grid	38.46	-38.46	0.6548	1.5272	30.53	0.1527	-20.5522	0.3908	39.28	13	13	9	11	13	59	5	
14	Method 1	2	Yes	cmmap2	40.67	-40.67	0.6013	1.6631	30.62	0.1141	-23.0345	0.3378	41.48	14	14	12	13	14	67	12	
15	Method 1	2	Yes	cmmap	40.87	-40.87	0.5767	1.7340	30.12	0.1068	-23.2855	0.3268	41.70	15	15	14	14	15	73	13	
16	Method 1	2	Yes	opi3	40.99	-40.99	0.5095	1.9627	28.48	0.0883	-23.4661	0.2972	41.85	16	16	20	17	16	85	19	
17	Method 1	2	Yes	opi2	42.77	-42.77	0.4850	2.0619	28.72	0.1243	-25.5718	0.3526	43.62	17	17	22	12	17	85	20	
18	Method 2	1	Yes	gpcp	50.86	-50.86	0.9291	1.0763	48.92	0.2273	-54.1704	0.4787	51.22	18	18	5	7	18	66	11	
19	Method 1	2	Yes	gpcc trmm	53.90	-53.90	0.9882	1.0119	53.54	0.4973	-401.4675	0.7052	53.93	19	19	2	4	19	63	8	
20	Method 2	1	Yes	reanalysis	55.16	-55.16	0.4710	2.1231	49.81	0.0328	-19.1856	0.1810	56.59	20	20	24	32	23	119	27	Plot
21	Method 2	1	Yes	cmmap	55.38	-55.38	1.0750	1.0750	56.70	0.2059	-64.2911	0.4537	55.72	21	21	4	8	20	74	14	
22	Method 2	1	Yes	cmmap	55.61	-55.61	1.0098	1.0098	55.78	0.1934	-64.8411	0.4398	55.95	22	22	1	9	21	75	15	
23	Method 2	1	Yes	gpcc	55.89	-55.89	0.5808	1.7218	47.65	0.0848	-74.8986	0.2546	56.25	23	23	13	24	22	105	23	
24	Method 2	1	Yes	opi3	56.38	-56.38	0.9370	1.0672	55.19	0.1554	-66.6608	0.3942	56.72	24	24	3	10	24	85	21	
25	Method 2	1	Yes	opi2	57.87	-57.87	0.4381	2.2826	46.53	0.0495	-70.4690	0.2224	58.29	25	25	27	27	25	129	30	
26	Method 2	1	Yes	ghcn cams grid	59.60	-59.60	0.8896	1.1241	57.99	0.0419	-21.9769	0.2047	60.89	26	26	6	29	27	114	26	
27	Method 2	1	Yes	gpcc trmm	59.97	-59.97	2.7894	2.7894	88.02	0.7511	-176.8989	0.8667	60.06	27	27	29	1	26	110	25	
28	Method 2	2	Yes	reanalysis	65.66	-65.66	0.1629	6.1387	52.29	0.0055	-51.7386	0.0744	66.38	28	28	34	34	28	152	34	Plot
29	Method 2	2	Yes	gpcp	71.66	-71.66	0.5549	1.8021	61.88	0.0916	-61.4282	0.3027	72.22	29	29	16	16	29	119	28	
30	Method 2	2	Yes	gpcc	72.26	-72.26	1.5655	1.5655	85.81	0.3874	-83.3336	0.6224	72.54	30	30	10	5	30	105	24	
31	Method 2	2	Yes	ghcn cams grid	74.49	-74.49	0.5177	1.9316	62.53	0.0859	-66.4055	0.2932	75.05	31	31	19	18	31	130	31	
32	Method 2	2	Yes	cmmap2	76.18	-76.18	0.4849	2.1510	62.01	0.0686	-69.4816	0.2619	76.74	32	32	25	22	32	143	32	
33	Method 2	2	Yes	cmmap	76.41	-76.41	0.4582	2.1825	61.93	0.0667	-69.9072	0.2583	76.97	33	33	26	23	33	148	33	
34	Method 2	2	Yes	opi3	77.17	-77.17	0.3949	2.5323	60.54	0.0517	-71.3404	0.2273	77.74	34	34	28	26	34	156	35	
35	Method 2	2	Yes	opi2	78.68	-78.68	0.3120	3.2051	58.73	0.0454	-74.2576	0.2131	79.30	35	35	32	28	35	165	36	
36	Method 2	2	Yes	gpcc trmm	92.06	-92.06	0.6196	1.6139	80.11	0.7492	-999.0000	0.8656	92.07	36	36	11	2	36	121	29	

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OL	Runoff Meth	Re	Storage	Precipitation	Absolute Error	Bias	Slope	Oslope	Intercept	R <sup>2</sup>	R <sup>2</sup> 1.1	Corr Coef	RMSE	AER	BR	SR	CCR	RMR	TR	BR	Plot				
1	Linear	1	Yes	reanalysis	51.27	10.42	3.0313	3.0313	30.96	0.5897	0.3054	0.7679	62.17	299	1	52	24	298	674	137					
2	Linear	1	Yes	gpcp	53.80	13.54	2.6130	2.6130	34.78	0.5254	0.2946	0.7248	65.05	322	19	43	42	307	733	152					
3	Linear	1	Yes	gpcp	53.52	19.90	2.7227	2.7227	54.71	0.5626	0.2673	0.7501	66.66	319	138	49	31	322	859	208					
4	Linear	1	Yes	gpcp trmm	51.26	16.05	2.4205	2.4205	40.75	0.4366	0.2377	0.6607	63.66	298	67	31	52	301	749	160					
5	Linear	1	Yes	ghcn cams grid	51.45	14.86	2.2996	2.2996	34.08	0.4490	0.2659	0.6700	63.94	303	39	28	49	302	721	145					
6	Linear	1	Yes	opi2	51.91	19.96	2.7321	2.7321	55.43	0.5843	0.2826	0.7644	65.38	307	141	46	25	310	829	199					
7	Linear	1	Yes	opi3	52.56	18.58	2.5789	2.5789	48.76	0.5382	0.2785	0.7336	65.57	310	84	37	35	311	777	174					
8	Linear	1	Yes	cmcp	53.01	18.21	2.5574	2.5574	46.78	0.5256	0.2752	0.7250	65.81	313	72	34	40	315	774	171					
9	Linear	1	Yes	cmcp2	53.15	17.99	2.5807	2.5807	46.63	0.5255	0.2742	0.7249	65.86	315	70	39	41	317	782	179					
10	Linear	1	No	reanalysis	16.52	10.67	0.0058	172.4138	0.62	0.0083	-320.9393	0.0912	21.64	15	4	213	196	3	431	33					
11	Linear	1	No	gpcp	18.60	13.88	0.0042	238.0952	0.76	0.0035	-273.7256	0.0592	25.55	31	22	216	210	38	517	59					
12	Linear	1	No	gpcp	21.57	20.13	0.0020	500.0000	0.54	0.0033	-999.0000	0.0570	29.12	94	149	226	211	112	792	182					
13	Linear	1	No	gpcp trmm	21.19	18.54	-0.0028		1.10	0.0005	-120.8500	-0.0226	27.35	74	79	320	244	72	789	181					
14	Linear	1	No	ghcn cams grid	19.79	15.35	-0.0053		0.48	0.0093	-489.1495	-0.0962	26.74	52	43	316	281	58	750	161					
15	Linear	1	No	opi2	22.39	21.18	0.0036	277.7778	0.77	0.0025	-392.0982	0.0504	30.23	145	173	221	212	157	908	220					
16	Linear	1	No	opi3	22.14	19.81	0.0010	1000.0000	0.72	0.0002	-375.7076	0.0144	29.60	135	135	233	228	139	870	210					
17	Linear	1	No	cmcp	21.67	19.05	0.0010	1000.0000	0.72	0.0002	-360.3419	0.0141	29.06	101	109	232	229	110	781	176					
18	Linear	1	No	cmcp2	21.45	18.62	0.0013	769.2308	0.73	0.0003	-352.8202	0.0181	28.76	87	97	230	226	102	742	155					
19	Linear	1	mon	reanalysis	16.32	10.69	-0.0089		0.47	0.0194	-328.5116	-0.1394	21.90	11	5	310	297	5	628	94					
20	Linear	1	mon	gpcp	18.55	13.90	0.0154	64.9351	0.91	0.0459	-269.4490	0.2143	25.39	28	23	189	164	35	439	36					
21	Linear	1	mon	gpcp	21.66	20.23	0.0094	106.3830	0.68	0.0692	-999.0000	0.2631	29.11	100	151	208	139	111	709	140					
22	Linear	1	mon	gpcp trmm	21.49	18.67	0.0235	42.5532	1.58	0.0357	-117.5548	0.1889	27.17	89	88	152	172	64	565	72					
23	Linear	1	mon	ghcn cams grid	19.72	15.39	0.0072	138.8889	0.67	0.0169	-480.9968	0.1301	26.53	50	47	211	186	56	550	68					
24	Linear	1	mon	opi2	22.33	21.21	0.0170	58.8235	1.05	0.0582	-386.6875	0.2413	30.07	143	175	183	151	152	804	190					
25	Linear	1	mon	opi3	22.16	19.85	0.0131	76.3359	0.95	0.0358	-370.7346	0.1893	29.44	138	136	194	171	136	775	172					
26	Linear	1	mon	cmcp	21.54	18.93	0.0127	78.7402	0.93	0.0330	-351.7799	0.1816	28.76	92	103	197	176	101	669	129					
27	Linear	1	mon	cmcp2	21.32	18.71	0.0129	77.5194	0.93	0.0332	-344.3914	0.1822	28.45	79	90	196	175	94	634	103					
28	Linear	1	mon	reanalysis	16.15	10.70	-0.0179		0.38	0.0785	-333.0149	-0.2801	22.06	6	6	297	318	7	634	101					
29	Linear	1	mon	gpcp	18.53	13.95	0.0190	52.6316	0.95	0.0701	-268.9983	0.2647	25.36	27	24	170	136	33	390	19					
30	Linear	1	mon	gpcp	21.70	20.40	0.0135	74.0741	0.75	0.1466	-999.0000	0.3828	29.15	103	158	181	89	117	588	123					
31	Linear	1	mon	gpcp trmm	21.63	18.96	0.0262	38.1679	1.55	0.0470	-126.1865	0.2167	27.30	97	105	137	161	69	589	73					
32	Linear	1	mon	ghcn cams grid	19.63	15.41	0.0126	79.9551	0.75	0.0517	-479.2107	0.2274	26.46	48	50	198	157	52	505	52					
33	Linear	1	mon	opi2	22.15	21.14	0.0225	44.4444	1.15	0.1014	-383.4315	0.3185	29.94	136	171	155	106	149	717	144					
34	Linear	1	mon	opi3	22.01	19.77	0.0178	56.1798	1.03	0.0655	-367.6680	0.2559	29.32	132	132	177	142	134	717	143					
35	Linear	1	mon	cmcp	21.32	18.77	0.0172	58.1395	1.01	0.0596	-345.4282	0.2441	28.49	80	94	182	148	98	602	83					
36	Linear	1	mon	cmcp2	21.10	18.55	0.0173	57.8035	1.01	0.0592	-338.0706	0.2434	28.19	69	80	179	149	86	563	71					
37	Linear	1	mon	reanalysis	16.10	10.73	-0.0281		0.30	0.1678	-337.1653	-0.4096	22.21	3	7	289	329	9	628	95					
38	Linear	1	mon	gpcp	18.39	13.98	0.0224	44.6429	0.99	0.0981	-268.0283	0.3131	25.34	24	25	156	109	32	346	13					
39	Linear	1	mon	gpcp	21.74	20.55	0.0163	61.3497	0.79	0.2196	-999.0000	0.4686	29.21	109	161	187	62	123	642	112					
40	Linear	1	mon	gpcp trmm	21.69	19.30	0.0341	29.3255	1.59	0.0877	-141.0767	0.2961	27.33	102	119	109	120	71	521	60					
41	Linear	1	mon	ghcn cams grid	19.48	15.37	0.0159	82.8931	0.79	0.0821	-479.6795	0.2866	26.37	44	44	188	124	49	449	37					
42	Linear	1	mon	opi2	21.94	21.02	0.0252	39.6825	1.20	0.1268	-378.3377	0.3560	29.77	128	169	142	96	145	680	134					
43	Linear	1	mon	opi3	21.76	19.65	0.0220	45.4545	1.10	0.0997	-362.5495	0.3158	29.14	110	130	158	108	114	620	87					
44	Linear	1	mon	cmcp	21.11	18.66	0.0216	46.2963	1.08	0.0937	-340.6348	0.3062	28.32	70	87	162	115	92	526	63					
45	Linear	1	mon	cmcp2	20.88	18.44	0.0219	45.6621	1.08	0.0942	-333.2392	0.3070	28.02	61	77	159	113	83	493	51					
46	Linear	1	mon	reanalysis	16.22	10.74	-0.0291		0.27	0.2077	-338.0029	-0.4558	22.26	7	8	284	322	11	632	97					
47	Linear	1	mon	gpcp	18.23	14.02	0.0268	37.3134	1.05	0.1397	-266.2978	0.3738	25.31	19	26	134	91	26	296	5					
48	Linear	1	mon	gpcp	21.78	20.82	0.0196	51.0204	0.87	0.3108	-999.0000	0.5575	29.25	114	165	167	58	128	632	98					
49	Linear	1	mon	gpcp trmm	21.37	19.45	0.0495	20.2020	1.83	0.1900	-143.2616	0.4359	27.29	81	122	87	71	68	429	31					
50	Linear	1	mon	ghcn cams grid	19.29	15.32	0.0197	50.7614	0.85	0.1262	-475.3185	0.3552	26.27	41	42	166	97	45	391	21					
51	Linear	1	mon	opi2	21.82	20.91	0.0265	37.7958	1.22	0.1388	-373.8530	0.3726	29.65	119	166	135	92	141	653	119					
52	Linear	1	mon	opi3	21.45	19.50	0.0271	36.9004	1.20	0.1498	-354.7888	0.3870	28.88	86	124	133	83	105	531	65					
53	Linear	1	mon	cmcp	20.88	18.64	0.0272	36.7647	1.18	0.1482	-337.3164	0.3849	28.24	60	85	132	87	87	451	40					
54	Linear	1	mon	cmcp2	20.65	18.41	0.0275	36.3636	1.18	0.1494	-329.9219	0.3865	27.93	56	76	131	85	79	427	20					
55	Linear	2	Yes	reanalysis	21.90	14.29	1.4576	1.4576	21.60	0.5800	0.3324	0.7616	26.76	124	36	24	28	60	272	3	Plot				
56	Linear	2	Yes	gpcp	23.36	20.29	1.2015	1.2015	24.72	0.6455	0.2433	0.8034	28.49	154	154	10	2	96	418	27					
57	Linear	2	Yes	gpcp	26.02	24.16	1.4595	1.4595	35.20	0.8463	0.000														

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OL	Runoff Meth	Re	Storage	Precipitation	Absolute Error	Bias	Slope	Oslope	Intercep	R <sup>2</sup>	R <sup>2</sup> 1.1	Corr Coef	RMSE	AER	BR	SR	CCR	RMR	TR	BR
102	Linear	2	4 mon	gpcp	25.30	24.24	-0.0026		-0.33	0.0000	-20.5632	-0.0068	30.61	177	209	321	239	161	1107	255
103	Linear	2	4 mon	gpcp trmm	32.93	31.95	-0.0352		-1.53	0.0123	-32.9651	-0.1109	39.39	291	314	271	287	288	1451	321
104	Linear	2	4 mon	ghcn cams grid	26.36	24.23	-0.0428		-1.72	0.0161	-21.5606	-0.1269	33.21	188	208	259	293	198	1146	263
105	Linear	2	4 mon	opi2	30.18	28.44	-0.0520		-2.17	0.0285	-27.4172	-0.1631	37.28	258	277	247	310	270	1362	308
106	Linear	2	4 mon	opi3	28.61	26.91	-0.0439		-1.87	0.0176	-24.7576	-0.1327	35.49	238	251	258	296	242	1285	295
107	Linear	2	4 mon	cmap	27.96	26.13	-0.0426		-1.80	0.0163	-23.7327	-0.1278	34.78	228	239	261	294	231	1253	286
108	Linear	2	4 mon	cmmap2	27.75	25.90	-0.0422		-1.78	0.0158	-23.3739	-0.1258	34.52	221	229	263	292	222	1227	274
109	Point	1	Yes	reanalysis	51.39	10.54	3.0378	3.0378	31.14	0.5897	0.3045	<b>0.7679</b>	62.34	300	2	53	23	299	677	133
110	Point	1	Yes	gpcp	53.95	13.70	2.6158	2.6158	34.97	0.5244	0.2931	0.7241	65.25	324	20	45	45	308	742	157
111	Point	1	Yes	gpcp	53.62	20.04	2.7744	2.7744	54.88	0.5619	0.2660	0.7496	66.80	321	144	51	33	323	872	213
112	Point	1	Yes	gpcp trmm	51.40	16.09	2.4316	2.4316	40.98	0.4362	0.2368	0.6604	64.02	301	68	33	53	303	758	167
113	Point	1	Yes	ghcn cams grid	51.59	14.98	2.3028	2.3028	34.24	0.4483	0.2647	0.6695	64.13	305	40	30	50	304	729	149
114	Point	1	Yes	opi2	52.07	20.10	2.7359	2.7359	55.66	0.5836	0.2811	0.7639	65.58	309	147	48	26	312	842	203
115	Point	1	Yes	opi3	52.73	18.73	2.5817	2.5817	48.96	0.5371	0.2769	0.7329	65.78	312	92	41	36	314	795	186
116	Point	1	Yes	cmmap	53.16	18.36	2.5801	2.5801	46.98	0.5246	0.2736	0.7243	66.02	316	74	36	43	318	787	180
117	Point	1	Yes	cmmap2	53.30	18.13	2.5834	2.5834	46.82	0.5244	0.2726	0.7242	66.07	318	71	42	44	319	794	184
118	Point	1	No	reanalysis	16.46	10.79	0.0124	80.6452	0.80	0.0296	-252.0520	0.1721	21.60	13	9	200	178	1	401	23
119	Point	1	No	gpcp	18.68	14.03	0.0070	142.8571	0.95	0.0072	-206.5089	0.0846	25.60	32	27	212	199	39	509	54
120	Point	1	No	gpcp	21.70	20.27	0.0038	263.1579	0.71	0.0086	-999.0000	0.0926	29.19	104	153	219	195	121	792	183
121	Point	1	No	gpcp trmm	21.25	18.57	0.0083	120.4819	1.33	0.0041	-109.8336	0.0637	27.23	77	82	209	206	66	640	109
122	Point	1	No	ghcn cams grid	19.84	15.46	-0.0021		0.65	0.0012	-386.1253	-0.0340	26.75	54	54	322	247	59	736	153
123	Point	1	No	opi2	22.50	21.33	0.0074	135.1351	1.00	0.0082	-296.1020	0.0904	30.29	149	178	210	197	158	892	216
124	Point	1	No	opi3	22.27	19.96	0.0038	263.1579	0.92	0.0022	-283.9133	0.0471	29.68	142	142	218	215	142	859	207
125	Point	1	No	cmmap	21.80	19.20	0.0037	270.2703	0.92	0.0022	-272.3529	0.0485	29.13	117	116	220	216	113	782	177
126	Point	1	No	cmmap2	21.58	18.97	0.0040	250.0000	0.93	0.0024	-266.6874	0.0490	28.82	95	107	217	214	104	737	154
127	Point	1	1 mon	reanalysis	16.22	10.81	-0.0032		0.64	0.0020	-258.2284	-0.0443	21.87	8	10	319	253	4	584	81
128	Point	1	1 mon	gpcp	18.58	14.05	0.0218	45.8716	1.15	0.0696	-202.1927	0.2638	25.37	29	28	161	138	34	390	20
129	Point	1	1 mon	gpcp	21.74	20.37	0.0133	75.1880	0.89	0.1053	-999.0000	0.3245	29.14	108	156	192	104	115	675	132
130	Point	1	1 mon	gpcp trmm	21.39	18.65	0.0388	29.5858	1.75	0.0682	-107.2103	0.2611	27.01	84	86	110	140	62	482	48
131	Point	1	1 mon	ghcn cams grid	19.74	15.50	0.0122	81.9672	0.88	0.0382	-378.8168	0.1955	26.52	51	58	202	169	55	535	66
132	Point	1	1 mon	opi2	22.43	21.36	0.0236	42.3729	1.33	0.0842	-291.1622	0.2902	30.08	148	180	151	123	153	755	164
133	Point	1	1 mon	opi3	22.25	19.99	0.0192	52.0833	1.21	0.0574	-279.1283	0.2395	29.46	141	143	169	152	137	742	156
134	Point	1	1 mon	cmmap	21.64	19.08	0.0187	53.4759	1.19	0.0540	-264.8640	0.2324	28.77	98	112	172	156	103	641	111
135	Point	1	1 mon	cmmap2	21.42	18.86	0.0189	52.9101	1.19	0.0541	-259.3045	0.2326	28.47	85	99	171	155	95	605	84
136	Point	1	2 mon	reanalysis	16.06	10.82	-0.0159		0.51	0.0488	-283.2389	-0.2210	22.09	1	11	300	313	8	633	99
137	Point	1	2 mon	gpcp	18.49	14.10	0.0263	38.0228	1.20	0.1014	-201.8620	0.3184	25.33	26	29	136	107	30	328	10
138	Point	1	2 mon	gpcp	21.76	20.53	0.0180	55.5556	0.97	0.1994	-999.0000	0.4465	29.17	111	160	176	69	118	634	104
139	Point	1	2 mon	gpcp trmm	21.48	18.89	0.0357	28.0112	1.65	0.0900	-128.3038	0.2999	27.11	88	101	106	117	63	475	47
140	Point	1	2 mon	ghcn cams grid	19.61	15.52	0.0184	54.3478	0.95	0.0861	-376.5907	0.2935	26.43	47	60	173	122	50	452	41
141	Point	1	2 mon	opi2	22.23	21.29	0.0295	33.8983	1.44	0.1313	-288.6823	0.3624	29.95	140	176	124	93	150	683	136
142	Point	1	2 mon	opi3	22.06	19.91	0.0248	40.3226	1.31	0.0981	-276.5960	0.3100	29.32	134	139	144	110	133	660	125
143	Point	1	2 mon	cmmap	21.37	18.92	0.0243	41.1523	1.28	0.0988	-259.8522	0.2996	28.49	82	102	149	118	97	548	67
144	Point	1	2 mon	cmmap2	21.14	18.69	0.0245	40.8163	1.28	0.0986	-254.2987	0.2994	28.18	73	89	147	119	85	513	56
145	Point	1	3 mon	reanalysis	16.09	10.84	-0.0265		0.41	0.1354	-267.4238	-0.3680	22.28	2	12	288	319	12	633	100
146	Point	1	3 mon	gpcp	18.35	14.14	0.0291	34.3643	1.23	0.1240	-201.0620	0.3521	25.32	23	30	125	98	29	305	6
147	Point	1	3 mon	gpcp	21.83	20.68	0.0219	45.6621	1.04	0.2992	-999.0000	0.5470	29.22	120	163	160	59	124	628	92
148	Point	1	3 mon	gpcp trmm	21.37	19.16	0.0355	28.1690	1.48	0.1307	-192.8256	0.3615	27.19	83	115	107	94	65	464	43
149	Point	1	3 mon	ghcn cams grid	19.45	15.48	0.0214	46.7290	0.99	0.1180	-376.6239	0.3434	26.35	43	56	163	101	47	410	25
150	Point	1	3 mon	opi2	22.03	21.16	0.0316	31.6456	1.48	0.1497	-284.9352	0.3869	29.79	133	172	119	84	146	654	121
151	Point	1	3 mon	opi3	21.81	19.80	0.0287	34.8432	1.38	0.1279	-272.7459	0.3576	29.14	118	133	128	95	118	590	78
152	Point	1	3 mon	cmmap	21.12	18.81	0.0284	35.2113	1.35	0.1220	-256.2468	0.3492	28.32	72	96	129	100	91	488	50
153	Point	1	3 mon	cmmap2	20.89	18.58	0.0287	34.8432	1.35	0.1225	-250.6734	0.3500	28.01	63	83	127	99	82	454	42
154	Point	1	4 mon	reanalysis	16.30	10.85	-0.0337		0.34	0.2201	-269.7672	-0.4692	22.39	10	13	278	323	13	637	107
155	Point	1	4 mon	gpcp	18.27	14.17	0.0320	31.2500	1.28	0.1503	-200.1764	0.3877	25.31	20	31	118	82	27	278	3
156	Point	1	4 mon	gpcp	21.89	20.96	0.0245	40.8163	1.10	0.2601	-999.0000	0.6051	29.28	123	167	148	56	130	624	91
157	Point	1	4 mon	gpcp trmm	21.19	19.23	0.0399	25.0627	1.43	0.2201	-255.0023	0.4692	27.24	75	117	99	61	67	419	29
158	Point	1	4 mon	ghcn cams grid	19.28	15.43	0.0246	40.6504	1.03	0.1555	-373.3226	0.3944	28.25	40	52	145	81	44	362	15
159	Point	1	4 mon	opi2	21.93	21.05	0.0314	31.84												

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Q1	Runoff Meth	Re	Storage	Precipitation	Absolute Error	Bias	Slope	Oslope	Intercept	R <sup>2</sup>	R <sup>2</sup> 1-1	Corr Coef	RMSE	AER	BR	SR	CCR	RMR	TR	BR							
203	Point	2	3 mon	ghcn cams grid	26.21	24.36	-0.0278		-1.22	0.0052	-16.3393	-0.0719	33.32	186	215	266	265	203	1155	267							
204	Point	2	3 mon	opi2	29.95	28.57	-0.0358		-1.57	0.0096	-20.7760	-0.0979	37.34	257	284	269	283	271	1364	309							
205	Point	2	3 mon	opi3	28.48	27.04	-0.0293		-1.34	0.0060	-18.7938	-0.0774	35.60	235	257	283	271	248	1294	298							
206	Point	2	3 mon	cmcp	27.83	26.27	-0.0293		-1.32	0.0059	-18.0294	-0.0767	34.91	222	247	282	269	235	1255	287							
207	Point	2	3 mon	cmcp2	27.63	26.04	-0.0295		-1.32	0.0059	-17.7657	-0.0767	34.66	217	237	281	270	227	1232	276							
208	Point	2	4 mon	reanalysis	19.61	15.39	-0.0395		-1.27	0.0075	-9.0433	-0.0866	24.86	46	49	264	275	25	659	124							
209	Point	2	4 mon	gpcp	25.34	21.38	-0.0351		-1.41	0.0087	-15.8741	-0.0986	32.22	178	181	272	284	181	1098	253							
210	Point	2	4 mon	gpcp	25.38	23.97	-0.0202		-1.03	0.0025	-17.5805	-0.0495	30.68	179	201	293	258	182	1083	250							
211	Point	2	4 mon	gpcp trmm	32.56	31.19	-0.0489		-2.73	0.0173	-23.5414	-0.1317	39.15	284	308	254	295	284	1425	316							
212	Point	2	4 mon	ghcn cams grid	26.47	24.25	-0.0346		-1.49	0.0084	-17.0412	-0.0914	33.32	189	210	276	277	202	1154	266							
213	Point	2	4 mon	opi2	30.42	28.47	-0.0425		-1.87	0.0141	-21.6655	-0.1188	37.34	259	279	262	289	272	1361	306							
214	Point	2	4 mon	opi3	28.71	26.93	-0.0363		-1.63	0.0096	-19.5888	-0.0978	35.59	242	252	268	282	247	1291	297							
215	Point	2	4 mon	cmcp	28.06	26.16	-0.0350		-1.57	0.0088	-18.7752	-0.0936	34.88	230	242	273	280	234	1259	290							
216	Point	2	4 mon	cmcp2	27.84	25.93	-0.0345		-1.55	0.0084	-18.4913	-0.0919	34.63	223	232	277	279	225	1238	277							
217	Area	1	Yes	reanalysis	51.41	10.66	3.0394	3.0394	31.28	0.5900	0.3040	0.7681	62.38	302	3	54	22	300	681	135							
218	Area	1	Yes	gpcp	53.93	13.85	2.6146	2.6146	35.11	0.5240	0.2923	0.7239	65.28	323	21	44	46	309	743	159							
219	Area	1	Yes	gpcp	53.60	20.08	2.7731	2.7731	54.89	0.5615	0.2656	0.7493	66.81	320	146	50	34	324	874	214							
220	Area	1	Yes	gpcp trmm	51.54	16.58	2.4232	2.4232	41.32	0.4331	0.2325	0.6581	64.20	304	69	32	54	306	765	169							
221	Area	1	Yes	ghcn cams grid	51.61	15.10	2.3015	2.3015	34.34	0.4475	0.2637	0.6690	64.20	306	41	29	51	305	732	151							
222	Area	1	Yes	opi2	52.05	20.26	2.7342	2.7342	55.78	0.5830	0.2798	0.7635	65.63	308	152	47	27	313	847	205							
223	Area	1	Yes	opi3	52.70	18.89	2.5795	2.5795	49.07	0.5364	0.2756	0.7324	65.83	311	100	38	37	316	802	188							
224	Area	1	Yes	cmcp	53.15	18.52	2.5579	2.5579	47.09	0.5238	0.2724	0.7237	66.07	314	78	35	47	320	794	185							
225	Area	1	Yes	cmcp2	53.28	18.29	2.5812	2.5812	46.94	0.5236	0.2714	0.7236	66.11	317	73	40	48	321	799	187							
226	Area	1	No	reanalysis	16.49	10.91	0.0140	71.4286	0.94	0.0363	0.2428	0.1905	21.63	14	14	190	170	2	390	110							
227	Area	1	No	gpcp	18.80	14.18	0.0058	172.4138	1.09	0.0048	-206.1455	0.0693	25.70	35	32	214	202	41	524	62							
228	Area	1	No	gpcp	21.74	20.30	0.0025	400.0000	0.72	0.0039	-999.0000	0.0625	29.23	107	155	225	207	126	820	195							
229	Area	1	No	gpcp trmm	21.73	19.06	-0.0001		1.68	0.0000	-101.1210	-0.0007	27.70	105	110	324	237	77	853	206							
230	Area	1	No	ghcn cams grid	19.95	15.59	-0.0035		0.75	0.0030	-372.8159	-0.0543	26.85	55	63	318	261	61	758	166							
231	Area	1	No	opi2	22.65	21.48	0.0057	175.4366	1.12	0.0048	-295.7219	0.0691	30.43	153	189	215	203	160	920	221							
232	Area	1	No	opi3	22.42	20.11	0.0016	625.0000	1.03	0.0004	-283.6904	0.0196	29.81	147	148	228	225	147	895	217							
233	Area	1	No	cmcp	21.95	19.35	0.0015	666.6667	1.04	0.0003	-272.2596	0.0181	29.27	129	120	229	227	129	834	201							
234	Area	1	No	cmcp2	21.73	19.13	0.0018	555.5556	1.04	0.0005	-266.6060	0.0216	28.96	106	113	227	224	109	779	175							
235	Area	1	mon	reanalysis	16.28	10.93	-0.0055		0.75	0.0056	-249.7511	-0.0745	21.97	9	15	315	267	6	612	85							
236	Area	1	mon	gpcp	18.70	14.20	0.0202	49.5050	1.27	0.0582	-201.9471	0.2433	25.48	39	33	165	150	37	418	28							
237	Area	1	mon	gpcp	21.78	20.40	0.0121	82.6446	0.90	0.0933	-999.0000	0.3055	29.19	113	159	204	116	119	711	141							
238	Area	1	mon	gpcp trmm	21.85	19.14	0.0305	32.7869	2.18	0.0495	-98.2481	0.2225	27.42	121	114	123	159	75	592	79							
239	Area	1	mon	ghcn cams grid	19.83	15.62	0.0116	86.2069	0.97	0.0332	-365.6730	0.1823	26.60	53	65	205	174	57	554	70							
240	Area	1	mon	opi2	22.57	21.52	0.0229	43.6681	1.47	0.0783	-290.4950	0.2798	30.21	152	192	153	130	155	782	178							
241	Area	1	mon	opi3	22.39	20.15	0.0173	57.8035	1.33	0.0462	-278.8115	0.2150	29.59	146	150	180	163	138	777	173							
242	Area	1	mon	cmcp	21.78	19.24	0.0187	59.8802	1.31	0.0426	-264.7079	0.2083	28.91	112	118	168	168	107	691	137							
243	Area	1	mon	cmcp2	21.56	19.01	0.0189	59.1716	1.31	0.0429	-259.1634	0.2070	28.60	93	108	185	167	100	653	118							
244	Area	1	mon	reanalysis	16.11	10.94	-0.0201		0.60	0.0750	-255.2995	-0.2738	22.22	4	16	294	317	10	641	110							
245	Area	1	mon	gpcp	18.58	14.25	0.0276	36.2319	1.37	0.1102	-200.3763	0.3320	25.39	30	34	130	102	36	332	111							
246	Area	1	mon	gpcp	21.60	20.58	0.0181	55.2486	1.01	0.2154	-999.0000	0.4641	29.20	116	162	174	63	122	637	108							
247	Area	1	mon	gpcp trmm	21.97	19.41	0.0435	22.9885	2.31	0.1082	-106.0556	0.3290	27.38	130	121	94	103	74	522	61							
248	Area	1	mon	ghcn cams grid	19.67	15.64	0.0195	51.2821	1.09	0.0938	-363.3391	0.3063	26.48	49	66	168	114	54	451	39							
249	Area	1	mon	opi2	22.37	21.44	0.0101	32.2581	1.63	0.1439	-287.0317	0.3793	30.03	144	186	121	90	151	692	138							
250	Area	1	mon	opi3	22.19	20.07	0.0258	38.7597	1.49	0.1029	-275.1010	0.3207	29.41	139	145	138	105	135	662	126							
251	Area	1	mon	cmcp	21.50	19.08	0.0252	39.6825	1.46	0.0958	-258.5447	0.3094	28.58	90	111	141	112	99	553	69							
252	Area	1	mon	cmcp2	21.28	18.85	0.0255	39.2157	1.4																		

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QI	Runoff Meth	Re	Storage	Precipitation	Absolute Error	Bias	Slope	Oslope	Intercpet	R <sup>2</sup>	R <sup>2</sup> 1-1	Corr Coef	RMSE	AER	BR	SR	CCR	RMR	TR	BR	
304	Area	2	2 mon	opi3	31.42	31.10	0.0133	75.1880	-3.91	0.0020	-36.4198	0.0451	37.92	275	306	193	217	278	1269	291	
305	Area	2	2 mon	cmap	30.76	30.33	0.0126	79.3651	3.88	0.0018	-35.0263	0.0424	37.21	269	297	199	219	268	1252	285	
306	Area	2	2 mon	cmap2	30.54	30.10	0.0130	76.9231	3.89	0.0019	-34.5355	0.0437	36.95	264	292	195	218	259	1228	275	
307	Area	2	3 mon	reanalysis	20.94	19.55	0.0674	14.8368	4.59	0.0350	-18.6587	0.1871	25.99	64	127	80	173	42	486	49	
308	Area	2	3 mon	gpcp	27.06	25.54	0.0475	21.0528	4.56	0.0286	-28.4968	0.1690	33.60	204	225	88	176	206	902	219	
309	Area	2	3 mon	gpcc	27.88	27.65	0.0211	47.3934	3.81	0.0041	-31.9699	0.0641	33.01	225	264	164	205	193	1051	235	
310	Area	2	3 mon	gpcc trmm	35.16	34.55	-0.0110		1.91	0.0019	-59.4556	-0.0440	40.91	296	323	308	252	297	1476	323	
311	Area	2	3 mon	ghcn cams grid	28.94	28.41	0.0423	23.6407	4.56	0.0200	-31.3464	0.1415	35.18	247	273	97	184	238	1039	234	
312	Area	2	3 mon	opi2	32.78	32.61	0.0503	19.8807	4.98	0.0317	-39.0700	0.1781	39.16	288	319	86	177	285	1155	268	
313	Area	2	3 mon	opi3	31.39	31.09	0.0458	21.8341	4.78	0.0245	-35.8400	0.1564	37.55	274	305	90	181	276	1126	259	
314	Area	2	3 mon	cmap	30.72	30.32	0.0449	22.2717	4.72	0.0231	-34.4481	0.1520	36.83	268	296	92	183	258	1097	254	
315	Area	2	3 mon	cmap2	30.50	30.08	0.0452	22.1239	4.71	0.0231	-33.9656	0.1521	36.58	263	291	91	182	257	1084	245	
316	Area	2	4 mon	reanalysis	20.75	19.46	0.0954	10.4822	4.97	0.0731	-16.7809	0.2703	25.60	58	123	60	132	40	413	26	
317	Area	2	4 mon	gpcp	26.72	25.44	0.0787	12.7065	5.17	0.0817	-28.6724	0.2859	33.07	195	222	76	125	194	812	194	
318	Area	2	4 mon	gpcc	27.93	27.69	0.0801	12.4844	5.15	0.0618	-32.2251	0.2481	32.46	227	266	74	147	187	901	218	
319	Area	2	4 mon	gpcc trmm	35.15	34.55	0.0254	39.3701	3.02	0.0104	-57.7013	0.1018	40.64	295	324	140	194	295	1248	283	
320	Area	2	4 mon	ghcn cams grid	28.92	28.32	0.0795	12.5786	5.42	0.0739	-31.5917	0.2718	34.66	246	270	75	131	226	948	225	
321	Area	2	4 mon	opi2	32.86	32.54	0.0779	12.8370	5.70	0.0793	-39.7619	0.2816	38.76	290	316	77	129	282	1094	251	
322	Area	2	4 mon	opi3	31.39	31.00	0.0815	12.2699	5.68	0.0807	-36.2421	0.2841	37.05	273	300	72	126	264	1035	233	
323	Area	2	4 mon	cmap	30.67	30.23	0.0819	12.2100	5.63	0.0801	-34.7814	0.2830	36.31	266	293	71	128	255	1013	231	
324	Area	2	4 mon	cmap2	30.44	29.99	0.0825	12.1212	5.63	0.0804	-34.2815	0.2836	36.06	261	288	70	127	253	999	230	

Larger United States - Yearly Surface																				
OI	Runoff Meth	Re	Storage	Precipitation	Absolute Error	Bias	Slope	Oslope	Intercept	R <sup>2</sup>	R <sup>2</sup> 1:1	Corr Coef	RMSE	AER	BR	SR	CCR	RMR	TR	BR
1	Linear	1	Yes	reanalysis	12.74	11.14	0.0096	104.1667	1.73	0.0000	-2.3944	0.0059	14.44	5	11	55	58	5	134	10
2	Linear	1	Yes	gpcp	13.48	12.09	-0.1769		-3.41	0.0056	-2.3411	-0.0750	15.26	10	15	92	73	10	200	21
3	Linear	1	Yes	gpcp	17.78	17.78	-0.3453		-8.63	0.0184	-6.3458	-0.1356	19.54	36	42	100	78	41	297	66
4	Linear	1	Yes	gpcp trmm	13.11	9.85	-4.4754		-76.95	0.2494	-0.7322	-0.4994	16.62	9	3	105	102	19	238	41
5	Linear	1	Yes	ghcn cams grid	16.05	15.67	-0.2940		-3.25	0.0118	-4.1397	-0.1086	17.94	23	27	97	75	25	247	48
6	Linear	1	Yes	opi2	19.18	19.11	0.0573	17.4520	0.08	0.0006	-5.3937	0.0240	21.11	48	53	34	54	55	244	47
7	Linear	1	Yes	opi3	17.67	17.60	0.2093	4.7778	2.83	0.0053	-4.5143	0.0727	19.60	35	40	28	47	43	193	19
8	Linear	1	Yes	cmap	17.20	16.84	-0.0807		-2.52	0.0008	-4.2207	-0.0290	19.07	30	33	69	63	36	231	36
9	Linear	1	Yes	cmapp2	17.02	16.61	-0.1198		-3.19	0.0017	-4.1103	-0.0418	18.87	27	30	77	67	32	233	39
10	Linear	1	No	reanalysis	10.67	10.67	0.0114	87.7193	0.68	0.0286	-999.0000	0.1690	11.70	1	6	53	32	1	93	3
11	Linear	1	No	gpcp	13.88	13.88	0.0208	48.0769	0.98	0.0303	-999.0000	0.1740	14.31	13	18	43	29	4	107	6
12	Linear	1	No	gpcp	20.13	20.13	0.0171	58.4795	0.83	0.4548	-999.0000	0.6744	20.32	55	62	46	1	52	216	29
13	Linear	1	No	gpcp trmm	17.21	17.21	-0.0816		0.06	0.0015	-999.0000	-0.7755	17.27	31	38	70	108	21	268	55
14	Linear	1	No	ghcn cams grid	15.19	15.19	-0.0062		0.48	0.0031	-999.0000	-0.0556	15.48	18	22	61	69	12	182	15
15	Linear	1	No	opi2	20.89	20.89	0.0154	64.9351	1.02	0.0162	-999.0000	0.1273	21.18	58	65	50	41	56	270	57
16	Linear	1	No	opi3	19.38	19.38	0.0083	120.4819	0.86	0.0322	-999.0000	0.0567	19.60	50	55	58	50	42	255	52
17	Linear	1	No	cmapp	18.63	18.63	0.0135	74.0741	0.95	0.0091	-999.0000	0.0956	18.87	44	49	52	44	31	220	31
18	Linear	1	No	cmapp2	18.40	18.40	0.0167	59.8802	1.00	0.0131	-999.0000	0.1146	18.62	42	47	48	42	29	208	25
19	Linear	2	Yes	reanalysis	24.52	10.87	2.4800	2.4800	34.45	0.1260	-0.0565	0.3550	29.96	68	8	21	10	78	185	16
20	Linear	2	Yes	gpcp	26.42	16.82	1.3197	1.3197	23.84	0.0510	-0.2848	0.2257	33.04	75	32	18	19	88	232	37
21	Linear	2	Yes	gpcp	27.47	23.45	2.7415	2.7415	65.21	0.1593	-1.0867	0.3991	31.17	81	78	25	6	85	275	60
22	Linear	2	Yes	gpcp trmm	14.38	8.46	-6.6001		-230.33	0.4165	-0.4548	-0.6453	18.06	16	2	106	104	27	255	51
23	Linear	2	Yes	ghcn cams grid	27.52	19.65	0.8465	1.1813	15.84	0.0226	-0.4325	0.1503	34.89	82	59	13	38	93	285	63
24	Linear	2	Yes	opi2	29.79	23.83	1.0590	1.0590	25.54	0.0514	-0.6171	0.2268	37.07	94	82	9	18	102	305	73
25	Linear	2	Yes	opi3	28.86	22.32	1.0349	1.0349	23.28	0.0349	-0.5516	0.1868	36.31	91	75	5	26	100	297	68
26	Linear	2	Yes	cmapp	28.40	21.57	0.9496	1.0531	20.22	0.0282	-0.5193	0.1679	35.93	86	73	8	33	97	297	67
27	Linear	2	Yes	cmapp2	28.25	21.34	0.9044	1.0158	20.82	0.0302	-0.5055	0.1739	35.77	84	69	3	31	96	283	67
28	Linear	2	No	reanalysis	15.39	15.39	0.1274	7.8493	1.46	0.0798	-70.4840	0.2824	15.92	20	24	32	15	16	107	7
29	Linear	2	No	gpcp	21.39	21.39	-0.0862		-2.47	0.0521	-137.3065	-0.2282	22.14	61	70	72	82	61	346	87
30	Linear	2	No	gpcp	23.83	23.83	-0.2167		-5.34	0.2225	-279.4668	-0.4717	24.17	64	81	94	99	64	402	94
31	Linear	2	No	gpcp trmm	31.02	31.02	0.6468	1.5461	19.92	0.2643	-283.5390	0.5141	31.06	102	106	19	3	83	313	77
32	Linear	2	No	ghcn cams grid	24.23	24.23	-0.0573		-2.00	0.0248	-173.9626	-0.1574	24.91	65	84	65	80	65	359	88
33	Linear	2	No	opi2	28.41	28.41	-0.0799		-2.89	0.0701	-240.3610	-0.2647	29.25	87	100	68	83	76	414	98
34	Linear	2	No	opi3	26.90	26.90	-0.1102		-3.81	0.0948	-213.6272	-0.3078	27.59	77	94	75	86	72	404	96
35	Linear	2	No	cmapp	26.14	26.14	-0.1052		-3.39	0.0828	-201.8532	-0.2878	26.82	72	91	74	85	70	392	93
36	Linear	2	No	cmapp2	25.91	25.91	-0.1012		-3.26	0.0766	-198.3911	-0.2768	26.59	70	89	73	84	68	384	90
37	Point	1	Yes	reanalysis	12.79	11.26	0.0246	40.6504	1.39	0.0002	-2.4464	0.0153	14.49	6	13	41	56	7	123	9
38	Point	1	Yes	gpcp	13.53	12.24	-0.1653		-3.10	0.0050	-2.4125	-0.0704	15.35	11	16	90	71	11	199	20
39	Point	1	Yes	gpcp	17.92	17.92	-0.3412		-8.41	0.0180	-6.4447	-0.1340	19.66	39	45	99	77	44	304	71
40	Point	1	Yes	gpcp trmm	12.65	9.90	-4.2233		-72.91	0.2378	-0.7837	-0.4876	16.30	4	4	103	100	18	229	35
41	Point	1	Yes	ghcn cams grid	16.13	15.79	-0.2986		-3.20	0.0123	-4.2375	-0.1108	18.02	25	28	98	76	26	253	49
42	Point	1	Yes	opi2	19.32	19.26	0.0091	109.8901	-0.74	0.0000	-5.5458	0.0038	21.26	49	54	56	59	57	275	59
43	Point	1	Yes	opi3	17.86	17.75	0.1930	5.1813	2.68	0.0045	-4.6385	0.0674	19.73	38	41	29	48	46	202	53
44	Point	1	Yes	cmapp	17.26	17.00	-0.0842		-2.43	0.0009	-4.3357	-0.0304	19.19	33	36	71	64	39	243	46
45	Point	1	Yes	cmapp2	17.08	16.76	-0.1224		-3.09	0.0018	-4.2230	-0.0429	18.99	28	31	78	68	34	239	43
46	Point	1	No	reanalysis	10.79	10.79	0.0263	38.0228	0.94	0.0829	-706.4602	0.2879	11.78	2	7	39	14	2	64	1
47	Point	1	No	gpcp	14.03	14.03	0.0324	30.8642	1.29	0.0357	-565.1518	0.1890	14.45	14	19	37	25	6	101	4
48	Point	1	No	gpcp	20.27	20.27	0.0213	46.9484	1.05	0.3842	-999.0000	0.6198	20.46	56	63	42	2	53	216	30
49	Point	1	No	gpcp trmm	17.25	17.25	0.1705	5.8651	4.10	0.1009	-522.3576	0.3177	17.31	32	39	30	12	22	135	12
50	Point	1	No	ghcn cams grid	15.30	15.30	-0.0108		0.53	0.0050	-999.0000	-0.0710	15.59	19	23	63	72	14	191	18
51	Point	1	No	opi2	21.04	21.04	-0.0328		0.20	0.0357	-999.0000	-0.1889	21.36	59	67	64	81	58	329	80
52	Point	1	No	opi3	19.54	19.54	-0.0080		0.71	0.0015	-999.0000	-0.0383	19.76	53	58	62	65	47	285	62
53	Point	1	No	cmapp	18.78	18.78	0.0101	99.0099	1.04	0.0025	-979.8507	0.0497	19.02	46	51	54	51	35	237	40
54	Point	1	No	cmapp2	18.55	18.55	0.0140	71.4288	1.11	0.0045	-954.8687	0.0673	18.78	43	48	51	49	30	221	32
55	Point	2	Yes	reanalysis	24.43	10.90	2.5197	2.5197	35.17	0.1294	-0.0568	0.3597	30.05	67	9	22	9	79	186	17
56	Point	2	Yes	gpcp	26.31	16.90	1.2632	1.2632	22.69	0.0464	-0.2900	0.2154	33.20	74	34	14	21	89	232	38
57	Point	2	Yes	gpcp	27.36	23.25	2.6879	2.6879	63.71	0.1532	-1.0687	0.3914	31.02	79	77	24	7	82	269	56
58	Point	2	Yes	gpcp trmm	14.78	8.36	-7.3133		-252.83	0.5700	-0.5145	-0.7550	17.45	17	1	108	107	23	256	53
59	Point	2	Yes	ghcn cams grid	27.78	19.74	0.7663	1.3050	13.94	0.0184	-0.4392	0.1357	35.07	83	61	17	39	94	294	65
60	Point	2	Yes	opi2	30.24	23.92	0.9705	1.0304	23.06	0.0249	-0.6266	0.2072	37.28	96	83	4	23	103	309	75
61	Point	2	Yes	opi3	29.31	22.41	0.9662	1.0350	21.48	0.0302	-0.5576	0.1739	36.48	93	76	6	30	101	306	74
62	Point	2	Yes	cmapp	28.86	21.66	0.8906	1.1226	18.74	0.0247	-0.5245	0.1570	36.09	92	74	12	37	99	314	78
63	Point	2	Yes	cmapp2	28.71	21.43	0.9235	1.0828	19.40	0.0265	-0.5109	0.1626	35.93	90	71	10	35	98	304	72
64	Point	2	No	reanalysis	15.48	15.48	0.1671	5.9844	2.18	0.0958	-49.4908	0.3095	16.01	22	26	31	13	17	109	8
65	Point	2	No	gpcp	21.48	21.48	-0.1427		-3.62	0.0997	-97.1899	-0.3158	22.33	62	72	86	88	62	370	89
66	Point	2	No	gpcp	23.62	23.62	-0.2704		-6.84	0.3849	-305.9907	-0.6204	23.98	63	79	95	103	63	403	95
67	Point	2	No	gpcp trmm	30.92	30.92	-0.0694		-2.58	0.0008	-83.6914	-0.0287	31.15	100	104	66	62	84	418	99
68	Point	2	No	ghcn cams grid	24.32	24.32	-0.1374		-3.90	0.0996	-123.1785	-0.3156								

Larger United States - Yearly Surface																					
Ol	Runoff Meth	Re	Storage	Precipitation	Absolute Error	Bias	Slope	Oslope	Intercept	R <sup>2</sup>	R <sup>2</sup> 1:1	Corr Coef	RMSE	AER	BR	SR	CCR	RMR	TR	BR	
102	Area	2	No	gpcc	27.43	27.43	-0.4142		-6.48	0.4481	-203.9594	-0.6694	27.83	80	97	102	106	74	459	106	
103	Area	2	No	gpcc trmm	33.78	33.78	0.2641	3.7864	10.66	0.1659	-999.0000	0.4073	33.81	108	108	26	5	91	338	84	
104	Area	2	No	ghcn cams grid	28.34	28.34	-0.1339		0.21	0.1277	-222.5245	-0.3573	29.00	85	99	82	90	75	431	101	
105	Area	2	No	ppi2	32.52	32.52	-0.1242		-0.06	0.1597	-294.0751	-0.3998	33.32	106	107	79	96	90	478	108	
106	Area	2	No	ppi3	31.01	31.01	-0.1330		-0.12	0.1302	-265.0007	-0.3609	31.63	101	105	81	93	86	466	107	
107	Area	2	No	cmap	30.26	30.26	-0.1354		-0.08	0.1294	-252.3407	-0.3598	30.87	97	103	83	91	81	455	105	
108	Area	2	No	cmap2	30.03	30.03	-0.1356		-0.06	0.1295	-248.6302	-0.3599	30.64	95	102	84	92	80	453	103	



Northwest Brazil - Monthly Atmospheric																							
OI	Moist Flux	Re	Storage	Precipitation	Absolute Error	Bias	Abias	Slope	Oslope	Intercept	R <sup>2</sup>	R <sup>2</sup> 1:1	Corr Coef	RMSE	AER	BR	SR	CCR	RMR	TR	BR		
1	Method 1	1	Yes	reanalysis	31.53	-4.36	4.36	0.9214	1.0853	10.29	0.5495	0.5398	0.7413	39.13	1	4	10	4	1	20	2	Plot	
2	Method 1	1	Yes	gpcp	58.89	-30.64	30.64	0.4038	2.4765	69.31	0.1727	-0.4647	0.4156	72.61	9	20	37	37	11	114	23		
3	Method 1	1	Yes	gpcp	62.95	-46.09	46.09	0.3913	2.5556	79.80	0.1557	-0.9224	0.3945	76.30	14	37	41	38	13	143	31		
4	Method 1	1	Yes	gpcp trmm	52.10	-0.40	0.40	0.6124	1.6329	36.14	0.3799	0.2277	0.6163	63.78	5	1	27	10	3	46	3		
5	Method 1	1	Yes	ghcn cams grid	50.09	4.77	4.77	0.4018	2.4888	45.83	0.1827	-0.2289	0.4274	64.00	3	5	39	35	4	86	10		
6	Method 1	1	Yes	gpi	81.89	23.82	23.82	0.1294	7.7280	78.84	0.0306	-1.4937	0.1751	101.85	32	12	52	51	32	179	42		
7	Method 1	1	Yes	trmm 3a46	62.04	35.48	35.48	0.5359	1.8660	25.40	0.2521	-0.1528	0.5020	82.00	12	26	31	26	15	110	16		
8	Method 1	1	Yes	trmm 3b43	54.17	-3.95	3.95	0.5683	1.7658	42.78	0.3207	0.1296	0.5663	67.41	6	3	29	18	5	61	4		
9	Method 1	1	Yes	ssmi	69.38	25.65	25.65	0.2817	3.5499	54.80	0.0870	-0.6267	0.2949	85.04	24	13	43	41	23	144	33		
10	Method 1	1	Yes	opi2	62.12	-37.85	37.85	0.4019	2.4882	72.50	0.1453	-0.5760	0.3812	75.18	13	32	38	39	12	134	30		
11	Method 1	1	Yes	opi3	58.72	-31.43	31.43	0.4293	2.3294	68.16	0.1918	-0.4226	0.4380	71.42	8	24	35	33	9	109	15		
12	Method 1	1	Yes	cmmap	58.46	-31.35	31.35	0.4326	2.3116	68.22	0.1924	-0.4139	0.4386	71.07	7	23	34	32	8	104	14		
13	Method 1	1	Yes	cmmap2	58.94	-30.71	30.71	0.4168	2.3992	68.97	0.1825	-0.4389	0.4272	71.69	10	21	36	36	10	113	20		
14	Method 1	2	Yes	reanalysis	40.91	13.83	13.83	0.0582	1.0582	-17.64	0.7439	0.7204	0.8625	50.13	2	8	6	1	2	19	1		
15	Method 1	2	Yes	gpcp	69.47	38.32	38.32	0.8658	1.1550	-26.26	0.3454	0.1737	0.5877	86.17	25	33	18	14	26	116	24		
16	Method 1	2	Yes	gpcp	63.74	30.19	30.19	0.8659	1.1549	-18.95	0.3285	0.2063	0.5731	79.58	15	18	17	16	14	80	6		
17	Method 1	2	Yes	gpcp trmm	80.51	48.72	48.72	0.9489	1.0539	-43.30	0.4590	0.2589	0.6776	93.71	30	40	5	8	29	110	18		
18	Method 1	2	Yes	ghcn cams grid	73.82	51.59	51.59	0.9141	1.0940	-42.74	0.3841	0.0845	0.6198	90.70	27	42	13	8	27	117	25		
19	Method 1	2	Yes	gpi	115.59	95.94	95.94	0.4600	2.1739	-18.20	0.1830	-1.0322	0.4278	139.37	52	49	33	34	52	220	49		
20	Method 1	2	Yes	trmm 3a46	98.02	83.96	83.96	0.1063	1.0163	-86.39	0.4796	-0.0911	0.6825	116.11	46	48	1	5	43	143	32		
21	Method 1	2	Yes	trmm 3b43	81.68	48.00	48.00	0.9537	1.0485	-43.13	0.4521	0.2561	0.6724	93.78	31	39	4	7	30	111	19		
22	Method 1	2	Yes	ssmi	111.76	96.94	96.94	0.9911	1.4470	-55.12	0.2528	-0.7423	0.5028	131.65	51	51	25	25	51	203	45		
23	Method 1	2	Yes	opi2	67.69	30.63	30.63	0.9219	1.0847	-24.22	0.3361	0.2293	0.5797	83.22	21	19	9	15	18	82	7		
24	Method 1	2	Yes	opi3	67.69	36.61	36.61	0.9185	1.0887	-29.43	0.3768	0.2247	0.6139	83.47	20	27	11	12	20	90	11		
25	Method 1	2	Yes	cmmap	67.67	37.05	37.05	0.9334	1.0714	-31.15	0.3809	0.2262	0.6172	83.39	19	28	7	9	19	82	8		
26	Method 1	2	Yes	cmmap2	68.30	37.63	37.63	0.9177	1.0897	-30.29	0.3733	0.2127	0.6110	84.11	23	30	12	13	21	99	13		
27	Method 2	1	Yes	reanalysis	58.96	-56.63	56.63	1.2221	1.2221	40.54	0.6811	-0.0362	0.8253	69.15	17	43	21	3	7	85	9	Plot	
28	Method 2	1	Yes	gpcp	91.85	-81.41	81.41	0.2255	4.4346	138.70	0.0482	-2.2701	0.2196	111.30	42	46	47	46	40	221	50		
29	Method 2	1	Yes	gpcp	86.58	-73.84	73.84	0.1978	5.0556	131.61	0.0367	-2.0851	0.1915	105.25	37	44	49	49	33	212	46		
30	Method 2	1	Yes	gpcp trmm	66.61	-37.51	37.51	0.5190	1.9268	88.13	0.2359	-0.2176	0.4857	82.64	17	29	32	30	16	124	28		
31	Method 2	1	Yes	ghcn cams grid	66.69	-37.80	37.80	0.2952	3.3875	102.01	0.0495	-0.5417	0.2225	84.45	18	31	42	44	22	157	37		
32	Method 2	1	Yes	gpi	79.28	-29.98	29.98	0.1560	6.4103	125.26	0.0339	-1.1703	0.1842	96.24	29	17	51	50	31	178	41		
33	Method 2	1	Yes	trmm 3a46	51.68	-7.23	7.23	0.6493	1.5401	56.12	0.2942	0.1994	0.5424	68.39	4	6	26	20	6	62	5		
34	Method 2	1	Yes	trmm 3b43	68.06	-46.54	46.54	0.5403	1.8508	91.25	0.2445	-0.3203	0.4945	85.87	22	38	30	29	25	144	34		
35	Method 2	1	Yes	ssmi	69.99	-17.42	17.42	0.1803	5.5483	114.70	0.0301	-0.8614	0.1736	85.70	26	10	50	52	24	162	38		
36	Method 2	1	Yes	opi2	107.45	-102.79	102.79	0.2651	3.7722	141.16	0.0421	-0.3462	0.2052	124.35	50	52	44	48	46	240	52		
37	Method 2	1	Yes	opi3	101.43	-96.12	96.12	0.2631	3.8008	139.51	0.0489	-2.7527	0.2212	119.78	47	50	45	45	44	231	51		
38	Method 2	1	Yes	cmmap	91.17	-82.65	82.65	0.2580	3.8760	136.91	0.0531	-2.2024	0.2305	109.74	41	47	46	43	39	216	47		
39	Method 2	1	Yes	cmmap2	89.57	-78.93	78.93	0.2253	4.4385	138.46	0.0435	-2.1276	0.2087	108.45	39	45	48	47	38	217	48		
40	Method 2	2	Yes	reanalysis	65.20	-45.93	45.93	1.2939	1.2939	21.72	0.7109	0.5288	0.8431	82.68	16	36	23	2	17	94	12	Plot	
41	Method 2	2	Yes	gpcp	84.82	-27.70	27.70	0.9646	1.0367	31.26	0.2480	0.1947	0.4979	108.08	36	15	3	27	37	118	26		
42	Method 2	2	Yes	gpcp	74.49	-9.79	9.79	0.8479	1.1794	25.32	0.2458	0.2290	0.4958	91.15	28	7	19	28	28	110	17		
43	Method 2	2	Yes	gpcp trmm	95.67	-33.70	33.70	1.1227	1.1227	19.02	0.3143	0.2577	0.5608	126.29	43	25	14	19	49	150	35		
44	Method 2	2	Yes	ghcn cams grid	83.97	-16.15	16.15	1.0344	1.0344	12.29	0.2161	0.1979	0.4649	107.87	35	9	2	31	36	113	22		
45	Method 2	2	Yes	gpi	103.91	22.72	22.72	0.4001	2.4994	61.55	0.0713	-0.1256	0.2671	126.01	49	11	40	42	48	190	44		
46	Method 2	2	Yes	trmm 3a46	95.83	-0.72	0.72	1.3777	1.3777	-59.50	0.3769	0.3486	0.6139	122.09	44	2	24	11	45	126	29		
47	Method 2	2	Yes	trmm 3b43	96.24	-38.79	38.79	1.1548	1.1548	21.05	0.3221	0.2463	0.5676	127.25	45	34	18	17	50	162	39		
48	Method 2	2	Yes	ssmi	103.05	31.27	31.27	0.5806	1.7224	29.05	0.0923	-0.0190	0.3039	125.64	48	22	28	40	47	185	43		
49	Method 2	2	Yes	opi2	89.89	-50.21	50.21	1.2492	1.2492	30.74	0.2651	0.0808	0.5149	115.47	40	41	22	23	42	168	40		
50	Method 2	2	Yes	opi3	87.76	-43.72	43.72	1.1848	1.1848	28.10	0.2755	0.1370	0.5249	111.88	38	35	20	21	41	155	36		
51	Method 2	2	Yes	cmmap	83.91	-29.52	29.52	1.1268	1.1268	17.00	0.2755	0.2119	0.5248	106.92	34	16	15	22	35	122	27		
52	Method 2	2	Yes	cmmap2	83.78	-25.90	25.90	1.0762	1.0762	18.09	0.2652	0.2176	0.5149	106.53	33	14	8	24	34	113	21		

Northwest Brazil - Yearly Atmospheric

DI	Moist Flux	Re	Storage	Precipitation	Absolute Error	Bias	Abias	Slope	Oslope	Intercept	R <sup>2</sup>	R <sup>2</sup> 1.1	Corr Coef	RMSE	AER	BR	SR	CCR	RMR	TR	BR	Plot
1	Method 1	1	Yes	reanalysis	18.97	-4.33	4.33	1.1379	1.1379	-6.07	0.6224	0.5997	0.7889	23.59	1	3	4	15	1	24	1	Plot
2	Method 1	1	Yes	gpcp	45.88	-30.68	30.68	0.1359	7.3584	86.72	0.0050	-0.7814	0.0706	53.53	21	20	37	37	12	127	29	
3	Method 1	1	Yes	gpcc	51.25	-46.05	46.05	-0.0449		103.93	0.0003	-1.9135	-0.0185	59.79	34	36	42	42	29	183	44	
4	Method 1	1	Yes	gpcc trmm	32.84	-6.60	6.60	2.3474	2.3474	-112.81	0.7786	0.5066	0.8824	37.24	5	5	22	10	5	47	5	
5	Method 1	1	Yes	ghcn cams grid	32.94	4.86	4.86	0.1175	8.5106	69.75	0.0029	-0.1742	0.0536	40.76	6	4	38	38	6	92	11	
6	Method 1	1	Yes	gpi	43.48	23.86	23.86	-0.2817		127.28	0.0173	-0.6483	-0.1317	55.33	13	12	47	47	17	136	31	
7	Method 1	1	Yes	trmm 3a46	40.46	38.29	38.29	2.6925	2.6925	-254.41	0.9065	0.1425	0.9521	55.65	7	31	27	2	19	86	8	
8	Method 1	1	Yes	trmm 3b43	28.15	-4.04	4.04	3.0500	3.0500	-179.47	0.8095	0.4366	0.8997	35.68	3	2	30	8	4	47	4	
9	Method 1	1	Yes	ssmi	46.82	28.33	28.33	0.1943	5.1467	62.07	0.0069	-0.5347	0.0833	54.02	24	15	36	35	13	123	27	
10	Method 1	1	Yes	opi2	51.22	-38.36	38.36	0.0490	20.4082	92.74	0.0008	-1.1979	0.0274	59.48	33	32	39	39	28	171	40	
11	Method 1	1	Yes	opi3	43.76	-32.39	32.39	0.3992	2.5050	70.33	0.0372	-0.6992	0.1928	52.28	15	23	24	30	9	101	15	
12	Method 1	1	Yes	cmmap	43.95	-31.94	31.94	0.3350	2.9851	74.24	0.0262	-0.7116	0.1620	52.47	16	22	28	32	10	108	20	
13	Method 1	1	Yes	cmmap2	44.22	-31.37	31.37	0.2730	3.6630	78.02	0.0188	-0.7252	0.1365	52.68	18	21	33	34	11	117	24	
14	Method 1	2	Yes	reanalysis	21.48	13.77	13.77	1.2365	1.2365	-29.24	0.7834	0.6698	0.8851	27.15	2	8	7	9	2	28	2	
15	Method 1	2	Yes	gpcc	44.91	38.26	38.26	0.9157	1.0921	-30.89	0.0970	-0.5595	0.3115	59.01	19	30	3	26	26	104	18	
16	Method 1	2	Yes	gpcc	42.55	30.11	30.11	1.0377	1.0377	-33.27	0.0799	-0.3340	0.2827	54.06	8	18	2	27	14	69	7	
17	Method 1	2	Yes	gpcc trmm	50.18	48.81	48.81	2.5720	2.5720	-215.27	0.8845	-0.2500	0.9405	60.85	30	40	25	3	33	131	30	
18	Method 1	2	Yes	ghcn cams grid	52.75	51.54	51.54	1.3414	1.3414	-86.75	0.1292	-1.0689	0.3594	67.96	38	42	8	23	40	151	35	
19	Method 1	2	Yes	gpi	95.95	95.95	95.95	0.0405	24.6914	42.19	0.0003	-3.3115	0.0159	111.91	49	49	40	40	50	228	48	
20	Method 1	2	Yes	trmm 3a46	89.39	89.39	89.39	2.5974	2.5974	-320.94	0.8311	-1.5106	0.9116	99.47	48	48	26	6	48	176	41	
21	Method 1	2	Yes	trmm 3b43	49.90	48.09	48.09	3.4833	3.4833	-309.27	0.9719	-0.3027	0.9858	62.13	29	39	32	1	36	137	32	
22	Method 1	2	Yes	ssmi	101.14	101.14	101.14	0.4239	2.3590	-23.46	0.0203	-3.8577	0.1424	114.40	51	51	23	33	52	210	45	
23	Method 1	2	Yes	opi2	42.87	30.58	30.58	0.4795	2.0855	12.20	0.0430	-0.4265	0.2073	56.43	10	19	19	29	22	99	13	
24	Method 1	2	Yes	opi3	43.11	36.55	36.55	1.5848	1.5848	-88.10	0.2562	-0.3771	0.5062	55.45	12	26	14	17	18	87	9	
25	Method 1	2	Yes	cmmap	43.72	37.00	37.00	1.5583	1.5583	-86.46	0.2395	-0.4043	0.4894	55.99	14	27	13	19	21	94	12	
26	Method 1	2	Yes	cmmap2	44.01	37.57	37.57	1.3850	1.3850	-71.91	0.2024	-0.4456	0.4498	56.81	17	28	10	22	23	100	14	
27	Method 2	1	Yes	reanalysis	56.61	-56.61	56.61	1.5140	1.5140	19.38	0.6825	-0.9891	0.8262	63.26	40	43	12	13	37	145	33	Plot
28	Method 2	1	Yes	gpcp	81.80	-81.44	81.44	-0.3824		183.69	0.0441	-5.1401	-0.2099	94.00	46	46	49	50	47	238	51	
29	Method 2	1	Yes	gpcc	73.79	-73.79	73.79	-0.4014		174.72	0.0246	-4.0230	-0.1567	85.42	44	44	51	48	44	231	49	
30	Method 2	1	Yes	gpcc trmm	46.27	-46.27	46.27	1.8637	1.8637	-39.97	0.8511	-0.8084	0.8069	54.17	22	37	16	14	15	104	17	
31	Method 2	1	Yes	ghcn cams grid	47.40	-38.08	38.08	-0.0214		130.87	0.0000	-0.8118	-0.0068	60.92	25	29	41	41	34	170	39	
32	Method 2	1	Yes	gpi	49.36	-29.94	29.94	-0.6254		213.44	0.0972	-1.1910	-0.3118	55.77	28	17	52	52	20	169	38	
33	Method 2	1	Yes	trmm 3a46	29.30	-7.17	7.17	2.2037	2.2037	-156.32	0.8183	0.5486	0.9035	30.97	4	8	21	7	3	41	3	
34	Method 2	1	Yes	trmm 3b43	46.66	-46.66	46.66	2.1369	2.1369	-63.93	0.5843	-1.2346	0.7644	54.25	23	38	20	16	16	113	22	
35	Method 2	1	Yes	ssmi	42.72	-16.43	16.43	-0.3679		177.79	0.0402	-0.7101	-0.2005	48.71	9	10	48	49	8	124	28	
36	Method 2	1	Yes	opi2	103.94	-103.94	103.94	-0.3952		175.75	0.0443	-8.0145	-0.2104	113.90	52	52	50	51	51	256	52	
37	Method 2	1	Yes	opi3	97.46	-97.46	97.46	-0.0489		158.24	0.0005	-6.8468	-0.0232	106.27	50	50	43	43	49	235	50	
38	Method 2	1	Yes	cmmap	83.25	-83.25	83.25	-0.1352		165.16	0.0044	-5.1256	-0.0667	93.89	47	47	44	44	46	228	47	
39	Method 2	1	Yes	cmmap2	79.63	-79.63	79.63	-0.1916		169.93	0.0096	-4.7897	-0.0982	91.12	45	45	45	46	45	226	46	
40	Method 2	2	Yes	reanalysis	46.53	-45.97	45.97	1.3706	1.3706	15.44	0.7117	0.0462	0.8437	57.32	27	35	9	11	25	107	19	Plot
41	Method 2	2	Yes	gpcp	51.82	-27.75	27.75	1.1709	1.1709	10.55	0.1205	-0.1056	0.3471	61.71	37	14	5	24	35	115	23	
42	Method 2	2	Yes	gpcc	43.10	-9.83	9.83	1.2071	1.2071	-11.32	0.1107	0.0689	0.3326	48.31	11	7	6	25	7	56	6	
43	Method 2	2	Yes	gpcc trmm	45.49	-33.68	33.68	2.9955	2.9955	-205.09	0.7055	0.1222	0.8399	60.71	20	25	29	12	31	117	25	
44	Method 2	2	Yes	ghcn cams grid	50.39	-16.19	16.19	1.0112	1.0112	14.94	0.0326	-0.0434	0.1807	59.95	31	9	1	31	30	102	16	
45	Method 2	2	Yes	gpi	53.28	22.75	22.75	-0.2743		156.27	0.0084	-0.3198	-0.0914	67.94	39	11	46	45	39	180	43	
46	Method 2	2	Yes	trmm 3a46	48.00	1.20	1.20	3.4430	3.4430	-381.09	0.8346	0.4141	0.9135	57.25	26	1	31	5	24	87	10	
47	Method 2	2	Yes	trmm 3b43	51.69	-38.77	38.77	4.2233	4.2233	-330.53	0.8535	-0.0017	0.9239	64.85	36	33	35	4	38	146	34	
48	Method 2	2	Yes	ssmi	59.21	33.46	33.46	0.2482	4.0290	73.50	0.0059	-0.3237	0.0770	73.37	42	24	34	36	42	178	42	
49	Method 2	2	Yes	opi2	63.74	-50.25	50.25	0.7023	1.4239	73.50	0.0473	-0.6942	0.2176	76.40	43	41	11	28	43	166	37	
50	Method 2	2	Yes	opi3	57.70	-43.77	43.77	2.0151	2.0151	-42.10	0.2534	-0.3670	0.5033	68.62	41	34	18	18	41	152	36	
51	Method 2	2	Yes	cmmap	51.31	-28.57	28.57	1.8957	1.8957	-58.92	0.2362	-0.0703	0.4860	60.72	35	16	17	20	32	120	26	
52	Method 2	2	Yes	cmmap2	50.41	-25.94	25.94	1.7370	1.7370	-49.54	0.2125	-0.0211	0.4610	59.31	32	13	15	21	27	108	21	

Northwest Brazil - Monthly Surface

OI	Runoff Meth	Re	Storage	Precipitation	Absolute Error	Bias	Slope	Intercept	R <sup>2</sup>	R <sup>2</sup> 1.1	Corr Coef	RMSE	AER	BR	SR	CCR	RMR	TR	BR	BR	Pilot
1	Linear	1	Yes	reanalysis	100.09	-99.59	-0.0578	-22.98	0.0078	-13.5744	-0.0882	114.76	102	108	310	304	89	913	162	162	Pilot
2	Linear	1	Yes	gpcp	110.68	-107.58	-0.1911	-19.47	0.1170	-14.7338	-0.3420	132.90	142	134	409	442	174	1301	269	269	
3	Linear	1	Yes	gpcp	96.09	-93.05	-0.0887	-14.64	0.0882	-14.9900	-0.2969	113.68	76	83	335	417	86	997	183	183	
4	Linear	1	Yes	gpcp trmm	174.84	-174.84	-0.1190	-57.10	0.0444	-23.4075	-0.2108	195.49	370	370	366	374	383	1863	416	416	
5	Linear	1	Yes	ghcn cams grid	118.66	-118.40	-0.1470	-13.91	0.0629	-19.2998	-0.2508	135.36	195	195	388	391	187	1356	290	290	
6	Linear	1	Yes	gpi	151.86	-151.09	0.0465	-43.45	0.0095	-20.7885	0.0974	172.01	269	274	137	143	270	1093	207	207	
7	Linear	1	Yes	trmm 3a46	219.05	-219.05	-0.2316	-47.34	0.1892	-46.7049	-0.4350	234.74	446	446	438	458	447	2335	481	481	
8	Linear	1	Yes	trmm 3b43	166.04	-166.04	-0.1457	-54.59	0.0627	-21.2354	-0.2504	187.59	326	331	386	390	360	1793	397	397	
9	Linear	1	Yes	ssmi	160.70	-160.63	-0.0086	-40.94	0.0002	-19.6704	-0.0142	177.47	301	303	227	224	289	1344	287	287	
10	Linear	1	Yes	opi2	88.44	-85.84	-0.2352	-21.36	0.1158	-9.8186	-0.3403	108.75	48	50	443	441	60	1042	195	195	
11	Linear	1	Yes	opi3	94.95	-92.51	-0.1938	-22.23	0.0929	-11.2596	-0.3047	115.77	73	76	410	421	93	1073	203	203	
12	Linear	1	Yes	cmmap	108.00	-106.67	-0.2001	-18.90	0.1093	-14.1707	-0.3306	129.15	127	131	413	434	145	1250	255	255	
13	Linear	1	Yes	cmmap2	111.50	-110.39	-0.1881	-19.08	0.1037	-15.1181	-0.3221	133.12	147	156	407	431	177	1318	274	274	
14	Linear	1	No	reanalysis	100.77	-99.57	-0.2674	-7.77	0.2493	-21.7687	-0.4993	117.26	108	107	451	463	99	1228	251	251	
15	Linear	1	No	gpcp	109.84	-107.64	-0.1809	-20.29	0.1108	-15.5356	-0.3329	132.44	139	136	405	436	172	1288	266	266	
16	Linear	1	No	gpcp	94.56	-92.93	-0.0524	-17.13	0.0918	-12.7102	-0.3030	111.61	71	81	299	420	71	942	171	171	
17	Linear	1	No	gpcp trmm	174.78	-174.78	-0.0965	-59.41	0.0229	-18.2454	-0.1513	195.97	369	369	347	341	386	1812	399	399	
18	Linear	1	No	ghcn cams grid	118.36	-118.27	-0.1093	-17.21	0.0517	-28.3483	-0.2275	133.41	191	192	356	380	178	1297	268	268	
19	Linear	1	No	gpi	153.50	-150.91	-0.1190	-24.58	0.0635	-22.7058	-0.2520	177.40	282	271	365	392	288	1598	350	350	
20	Linear	1	No	trmm 3a46	218.60	-218.60	-0.3027	-36.96	0.2373	-34.5041	-0.4871	236.44	445	445	458	461	449	2258	463	463	
21	Linear	1	No	trmm 3b43	165.98	-165.98	-0.1279	-56.27	0.0381	-16.6685	-0.1953	188.22	325	327	374	368	363	1757	391	391	
22	Linear	1	No	ssmi	160.58	-160.37	-0.1171	-27.80	0.0369	-20.1937	-0.1920	179.72	299	300	363	364	298	1624	363	363	
23	Linear	1	No	opi2	88.16	-85.53	-0.2274	-21.46	0.1141	-10.2647	-0.3378	108.09	47	45	435	438	54	1019	190	190	
24	Linear	1	No	opi3	93.62	-92.20	-0.1595	-23.94	0.0663	-11.6355	-0.2576	114.47	63	70	395	394	88	1100	188	188	
25	Linear	1	No	cmmap	107.24	-106.64	-0.1782	-20.47	0.0916	-14.8467	-0.3027	128.38	124	130	403	419	142	1218	247	247	
26	Linear	1	No	cmmap2	110.92	-110.35	-0.1685	-20.56	0.0880	-15.8529	-0.2966	132.40	144	155	399	416	171	1285	263	263	
27	Linear	1	mon	reanalysis	100.61	-99.49	-0.2274	-10.68	0.1802	-21.4799	-0.4245	116.59	106	106	436	457	96	1201	244	244	
28	Linear	1	mon	gpcp	109.25	-107.42	-0.0926	-26.90	0.0289	-14.8416	-0.1701	129.82	133	133	338	351	150	1105	213	213	
29	Linear	1	mon	gpcp	94.06	-92.79	0.0025	-21.08	0.0002	-11.81322	0.0147	109.93	66	80	210	202	68	626	89	89	
30	Linear	1	mon	gpcp trmm	174.96	-174.96	0.0741	-77.92	0.0136	-17.3501	0.1167	192.02	371	371	395	126	376	1339	286	286	
31	Linear	1	mon	ghcn cams grid	118.41	-118.32	-0.0258	-24.79	0.0029	-27.6239	-0.0537	131.80	193	193	259	266	166	1077	206	206	
32	Linear	1	mon	gpi	153.63	-150.95	-0.0957	-27.29	0.0411	-22.4436	-0.2028	176.77	285	272	345	372	283	1557	337	337	
33	Linear	1	mon	trmm 3a46	221.05	-221.05	-0.1754	-55.51	0.0804	-34.7488	-0.2835	236.54	449	449	402	409	450	2159	458	458	
34	Linear	1	mon	trmm 3b43	165.39	-165.39	0.0084	-70.09	0.0002	-15.8178	0.0128	184.31	320	322	202	204	337	1385	301	301	
35	Linear	1	mon	ssmi	160.97	-160.59	-0.0487	-36.10	0.0064	-19.8380	-0.0800	178.44	303	302	293	294	1481	328	328	328	
36	Linear	1	mon	opi2	87.84	-85.71	-0.1010	-28.09	0.0225	-9.7133	-0.1500	105.56	46	47	350	340	49	832	145	145	
37	Linear	1	mon	opi3	93.28	-92.41	-0.0562	-30.06	0.0082	-11.1054	-0.0807	112.20	58	75	308	306	78	825	143	143	
38	Linear	1	mon	cmmap	106.97	-106.59	-0.0757	-28.03	0.0165	-14.2140	-0.1286	125.97	122	129	330	332	128	1041	194	194	
39	Linear	1	mon	cmmap2	110.62	-110.28	-0.0685	-28.30	0.0146	-15.1830	-0.1206	129.92	141	154	322	326	151	1094	208	208	
40	Linear	1	mon	reanalysis	100.40	-99.45	-0.1516	-16.18	0.0801	-20.9140	-0.2830	115.20	105	105	390	407	91	1098	209	209	
41	Linear	1	mon	gpcp	109.00	-107.59	-0.0023	-33.59	0.0000	-14.2278	-0.0042	127.49	130	135	218	218	138	857	147	147	
42	Linear	1	mon	gpcp	93.64	-92.63	0.0476	-24.22	0.0764	-11.5377	0.2765	108.50	64	79	134	40	56	373	29	29	
43	Linear	1	mon	gpcp trmm	173.97	-173.97	0.1721	-87.87	0.0171	-16.4355	0.2704	188.71	367	367	26	43	364	1167	237	237	
44	Linear	1	mon	ghcn cams grid	118.41	-118.35	0.0548	-32.11	0.0130	-26.9232	0.1141	130.21	192	194	121	131	155	793	136	136	
45	Linear	1	mon	gpi	153.60	-151.04	-0.0324	-34.45	0.0047	-21.8127	-0.0687	174.80	284	273	269	280	279	1385	300	300	
46	Linear	1	mon	trmm 3a46	222.85	-222.85	0.0183	-82.81	0.0008	-33.4098	0.0290	234.88	450	450	189	194	448	1731	383	383	
47	Linear	1	mon	trmm 3b43	165.09	-165.09	0.1285	-81.42	0.0383	-15.0014	0.1956	181.22	318	319	48	77	311	1073	204	204	
48	Linear	1	mon	ssmi	161.28	-160.86	0.0140	-43.56	0.0005	-19.4276	0.0229	177.31	305	305	193	196	286	1285	264	264	
49	Linear	1	mon	opi2	87.32	-85.70	0.0039	-33.61	0.0000	-9.2265	0.0057	103.30	43	46	207	208	46	550	68	68	
50	Linear	1	mon	opi3	92.87	-92.24	0.0496	-36.33	0.0064	-10.4896	0.0799	109.49	56	71	131	156	66	480	53	53	
51	Linear	1	mon	cmmap	106.38	-106.11	0.0244	-35.36	0.0017	-13.4356	0.0409	122.91	120	121	178	180	111	710	109	109	
52	Linear	1	mon	cmmap2	110.00	-109.75	0.0267	-35.62	0.0021	-14.3632	0.0463	126.79	140	148	172	171	135	766	129	129	
53	Linear	1	mon	reanalysis	100.16	-99.39	-0.0504	-23.52	0.0089	-20.1747	-0.0941	113.32	104	103	295	310	82	894	158	158	
54	Linear	1	mon	gpcp	109.22	-107.79	0.0571	-38.01	0.0109	-13.8230	0.1046	126.01	132	138	118	136	129	653	95	95	
55	Linear	1	mon	gpcp	94.21	-93.00	0.0543	-24.56	0.1022	-11.09232	0.3197	108.55	67	82	125	31	57	362	26	26	
56	Linear	1	mon	gpcp trmm	174.69	-174.69	0.2811	-98.90	0.1931	-15.9468	0.4394	186.45	368	368	9	11	350	1106	214	214	
57	Linear	1	mon	ghcn cams grid	118.28	-118.23	0.0997	-36.15	0.0431	-26.5129	0.2077	129.19	190	191	72	75	146	674	101	101	
58	Linear	1	mon	gpi	153.13	-151.21	0.0366	-42.24	0.0060	-21.1437	0.0775	172.64	280	276	151	157	272	1136	228	228	
59	Linear	1	mon	trmm 3a46	220.49	-220.49	0.2691	-117.43	0.1804	-31.1212	0.4248	228.26	448	448	11	17	443	1367	291	291	
60	Linear	1	mon	trmm 3b43	163.88	-163.88	0.2366	-90.98	0.1311	-14.2706	0.3621	177.36	308	309	12	25	287	941	170	170	
61	Linear	1	mon	ssmi	161.08	-160.82	0.1016	-54.02	0.0278	-18.8514	0.1667	175.30	304	304	69	92	280	1049	197	197	
62	Linear	1	mon	opi2	87.46	-85.71	0.0628	-36.72	0.0067	-8.9450	0.0832	102.04	44	48	111	145	44	392	34	34	
63	Linear	1	mon	opi3	92.86	-92.14	0.1017	-39.41	0.0268	-10.1708	0.1637	108.14	55	66	68	94</					

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Of	Runoff Meth	Re	Storage	Precipitation	Absolute Error	Bias	Slope	Intercept	R <sup>2</sup>	R <sup>2</sup> 1:1	Corr Coef	RMSE	AER	BR	SR	CCR	RMR	TR	BR	
102	Linear	2	No	opi3	150.17	-149.72	-0.2659	-42.65	0.1045	-13.9016	-0.3322	169.42	258	263	450	432	259	1662	371	
103	Linear	2	No	cmmap	164.08	-163.92	-0.2598	-39.47	0.1103	-16.4327	-0.3321	183.25	310	310	449	435	324	1828	406	
104	Linear	2	No	cmmap2	167.65	-167.54	-0.2574	-38.78	0.1142	-17.1845	-0.3379	187.16	345	345	447	439	357	1933	431	
105	Linear	2	1 mon	reanalysis	149.93	-147.57	-0.2137	-47.73	0.1466	-15.8944	-0.3829	180.33	254	250	424	454	303	1685	376	
106	Linear	2	1 mon	gpcp	166.89	-165.67	-0.2075	-44.48	0.0864	-17.0980	-0.2940	186.65	340	323	421	414	351	1849	412	
107	Linear	2	1 mon	gpcp	165.04	-164.72	-0.2203	-40.27	0.1289	-23.2358	-0.3590	184.06	317	315	432	448	332	1844	410	
108	Linear	2	1 mon	gpcp lrmm	197.35	-197.35	-0.0946	-67.62	0.0216	-20.2538	-0.1470	218.22	408	408	342	338	414	1910	428	
109	Linear	2	1 mon	ghcn cams grid	177.53	-177.49	-0.1555	-47.86	0.0370	-18.3679	-0.1923	193.08	375	375	392	366	379	1887	422	
110	Linear	2	1 mon	gpi	205.91	-205.56	-0.0851	-53.63	0.0271	-29.6356	-0.1648	226.43	422	424	334	349	437	1966	438	
111	Linear	2	1 mon	lrmm 3a46	235.32	-235.32	-0.2853	-28.39	0.1933	-32.6160	-0.4396	253.85	456	456	455	459	463	2289	465	
112	Linear	2	1 mon	lrmm 3b43	192.24	-192.24	-0.1146	-65.84	0.0307	-19.3662	-0.1751	213.62	400	400	361	355	401	1917	429	
113	Linear	2	1 mon	ssmi	208.62	-208.40	-0.1867	-38.16	0.0794	-26.4121	-0.2818	226.14	435	435	406	405	435	2116	454	
114	Linear	2	1 mon	opi2	143.67	-143.25	-0.2059	-49.26	0.0543	-12.4688	-0.2331	161.02	231	231	417	383	241	1503	328	
115	Linear	2	1 mon	opi3	150.03	-149.70	-0.2202	-46.73	0.0717	-13.7718	-0.2678	168.63	255	261	431	403	255	1605	355	
116	Linear	2	1 mon	cmmap	164.01	-163.80	-0.2198	-43.66	0.0787	-16.2823	-0.2805	182.39	309	308	430	404	316	1767	394	
117	Linear	2	1 mon	cmmap2	167.62	-167.42	-0.2161	-43.24	0.0803	-17.0217	-0.2833	186.25	344	344	429	408	347	1872	418	
118	Linear	2	2 mon	reanalysis	149.77	-147.44	-0.1437	-53.68	0.0663	-15.4179	-0.2574	177.81	251	248	384	393	291	1567	340	
119	Linear	2	2 mon	gpcp	167.15	-166.00	-0.1116	-54.24	0.0250	-16.7651	-0.1581	184.96	343	328	358	344	340	1713	380	
120	Linear	2	2 mon	gpcp	164.84	-164.56	-0.1054	-51.65	0.0298	-22.8594	-0.1725	181.66	314	314	353	353	313	1647	369	
121	Linear	2	2 mon	gpcp lrmm	200.49	-200.49	-0.0437	-74.29	0.0044	-20.2862	-0.0666	219.49	415	415	283	276	416	1805	398	
122	Linear	2	2 mon	ghcn cams grid	177.94	-177.94	-0.0595	-58.76	0.0054	-18.1411	-0.0734	191.99	378	378	313	283	375	1727	382	
123	Linear	2	2 mon	gpi	206.19	-205.43	-0.0958	-51.92	0.0346	-29.6970	-0.1859	226.69	424	423	346	362	439	1994	442	
124	Linear	2	2 mon	lrmm 3a46	237.64	-237.64	-0.1710	-47.26	0.0868	-32.2328	-0.2619	254.02	463	463	400	464	2190	460		
125	Linear	2	2 mon	lrmm 3b43	195.13	-195.13	-0.0576	-72.92	0.0075	-19.3524	-0.0888	214.63	403	403	309	301	405	1821	403	
126	Linear	2	2 mon	ssmi	209.22	-208.82	-0.1467	-44.15	0.0492	-26.2683	-0.2218	225.85	437	437	387	378	433	2072	452	
127	Linear	2	2 mon	opi2	143.57	-143.57	-0.1229	-55.86	0.0193	-12.2990	-0.1391	160.03	230	233	370	335	238	1406	305	
128	Linear	2	2 mon	opi3	150.12	-150.07	-0.0952	-57.40	0.0134	-13.4518	-0.1157	166.83	256	267	343	323	252	1441	316	
129	Linear	2	2 mon	cmmap	164.13	-164.13	-0.0997	-55.62	0.0162	-15.9419	-0.1273	180.63	312	313	349	331	308	1613	360	
130	Linear	2	2 mon	cmmap2	167.75	-167.75	-0.0969	-55.55	0.0161	-16.6646	-0.1270	184.44	347	349	348	330	338	1712	379	
131	Linear	2	3 mon	reanalysis	149.40	-147.19	-0.0911	-57.79	0.0268	-15.0883	-0.1636	175.83	248	245	337	347	281	1458	321	
132	Linear	2	3 mon	gpcp	166.33	-166.00	-0.0706	-58.15	0.0100	-16.6375	-0.1001	184.11	328	329	324	316	333	1630	365	
133	Linear	2	3 mon	gpcp	165.23	-165.17	-0.0773	-54.36	0.0158	-22.4699	-0.1255	181.54	319	320	331	329	312	1611	358	
134	Linear	2	3 mon	gpcp lrmm	202.88	-202.86	0.1258	-95.55	0.0359	-19.5891	0.1894	217.62	417	417	52	79	413	1378	296	
135	Linear	2	3 mon	ghcn cams grid	177.94	-177.94	-0.0147	-63.60	0.0003	-18.0432	-0.0181	191.30	379	379	239	230	371	1598	351	
136	Linear	2	3 mon	gpi	206.19	-205.40	-0.0441	-59.10	0.0073	-29.1762	-0.0856	225.32	425	422	285	299	431	1862	415	
137	Linear	2	3 mon	lrmm 3a46	236.42	-236.42	-0.0648	-65.13	0.0096	-30.8201	-0.0982	251.02	459	459	320	315	458	2011	445	
138	Linear	2	3 mon	lrmm 3b43	197.54	-197.36	0.0670	-87.97	0.0100	-18.8599	0.0999	213.74	409	409	106	139	402	1465	322	
139	Linear	2	3 mon	ssmi	210.08	-209.77	-0.0639	-56.88	0.0095	-26.3547	-0.0973	225.19	441	440	319	313	430	1943	434	
140	Linear	2	3 mon	opi2	143.95	-143.59	-0.0545	-60.99	0.0038	-12.1493	-0.0617	158.96	232	234	307	272	233	1278	262	
141	Linear	2	3 mon	opi3	150.33	-150.10	-0.0528	-60.78	0.0041	-13.3570	-0.0642	166.10	260	269	301	273	249	1352	289	
142	Linear	2	3 mon	cmmap	164.09	-164.06	-0.0596	-59.37	0.0058	-15.8368	-0.0762	179.88	311	311	314	286	300	1522	332	
143	Linear	2	3 mon	cmmap2	167.71	-167.69	-0.0528	-59.85	0.0048	-16.5403	-0.0694	183.60	346	347	302	281	329	1605	357	
144	Linear	2	4 mon	reanalysis	149.57	-146.71	-0.0521	-60.91	0.0087	-14.7219	-0.0933	174.01	250	243	298	309	277	1377	295	
145	Linear	2	4 mon	gpcp	166.89	-166.03	-0.0050	-64.65	0.0001	-16.3485	-0.0071	182.79	341	330	223	218	320	1432	312	
146	Linear	2	4 mon	gpcp	165.60	-165.31	0.0111	-63.01	0.0003	-22.2553	0.0180	179.78	321	321	197	199	299	1337	283	
147	Linear	2	4 mon	gpcp lrmm	204.03	-203.28	0.0296	-83.22	0.0020	-19.7993	0.0446	220.54	418	418	166	173	419	1594	349	
148	Linear	2	4 mon	ghcn cams grid	177.93	-177.93	0.0465	-70.39	0.0033	-17.8216	0.0574	190.39	377	377	138	165	368	1425	310	
149	Linear	2	4 mon	gpi	205.87	-205.30	0.0154	-67.15	0.0009	-28.8329	0.0299	223.60	421	421	191	191	425	1649	370	
150	Linear	2	4 mon	lrmm 3a46	234.57	-234.57	0.0200	-77.99	0.0009	-29.8995	0.0303	247.92	455	455	186	190	454	1740	384	
151	Linear	2	4 mon	lrmm 3b43	198.72	-197.84	0.0376	-84.00	0.0031	-18.7881	0.0559	215.11	413	410	148	167	407	1545	336	
152	Linear	2	4 mon	ssmi	209.82	-209.79	0.0289	-70.13	0.0019	-25.8503	0.0441	223.50	440	441	167	174	424	1646	368	
153	Linear	2	4 mon	opi2	144.24	-143.50	0.0088	-65.84	0.0001	-11.9529	0.0100	157.94	234	232	201	206	231	1104	212	
154	Linear	2	4 mon	opi3	150.51	-150.16	0.0331	-67.96	0.0018	-13.0859	0.0402	164.70	261	270	157	181	246	1115	219	
155	Linear	2	4 mon	cmmap	164.34	-164.10	0.0247	-67.60	0.0010	-15.5369	0.0316	178.46	313	312	177	188	295	1285	265	
156	Linear	2	4 mon	cmmap2	167.87	-167.72	0.0341	-68.65	0.0020	-16.2148	0.0448	182.08	348	348	155	172	314	1337	285	
157	Point	1	Yes	reanalysis	78.34	-77.65	0.2055	-20.10	0.1861	-13.5946	0.4076	88.34	13	27	15	19	13	87	7	Plot
158	Point	1	Yes	gpcp	84.72	-79.04	0.0182	-8.42	0.0030	-24.1071	0.0545	100.56	41	33	190	168	41	473	52	
159	Point	1	Yes	gpcp	81.23	-75.50	-0.0250	-1.67	0.0069	-30.2409	-0.0829	97.67	29	16	254	295	30	624	88	
160	Point	1	Yes	gpcp lrmm	115.66	-114.28	0.1403	-23.82	0.1408	-24.1687	0.3752	131.46	174	175	37	22	164	572	117	
161	Point	1	Yes	ghcn cams grid	97.19	-96.43	-0.0162	-3.85	0.0013	-22.4903	-0.0359	112.01	83	90	240	244	76	733	117	
162	Point	1	Yes	gpi	120.56	-119.00	0.1206	-19.73	0.1855	-40.0267	0.4307	138.37	197	200	56	16	200	669	99	
163	Point	1	Yes	lrmm 3a46	146.28	-146.28	0.1903	-33.38	0.2084	-33.7735	0.4566	156.94	240	242	22	9	229	742	120	
164	Point	1	Yes	lrmm 3b43	107.37	-106.16	0.1559	-24.05	0.1679	-21.4011	0.4098	123.15	125	123	30	18	113	409	40	
165	Point	1	Yes	ssmi	126.48	-125.69	0.1063	-19.63	0.0910	-37.4276	0.3017	139.76	221	225	63	37	203	749	124	
166	Point	1	Yes	opi2	63.34	-57.56	0.0256	-6.70	0.0037	-13.6495	0.0610	76.73	6	6	175	163	6	356	23	
167	Point	1	Yes	opi3	69.89	-64.24	0.0112	-6.02	0.0008	-16.8427	0.0291	84.68	12	12	196	193	12	425		

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OI	Runoff Meth	Re	Storage	Precipitation	Absolute Error	Bias	Slope	Intercept	R <sup>2</sup>	R <sup>2</sup> 1:1	Corr Coef	RMSE	AER	BR	SR	CCR	RMR	TR	BR		
203	Point	1	2 mon	trmm 3b43	105.78	-105.45	0.1418	-23.05	0.1389	-21.1043	<b>0.3727</b>	123.35	117	116	36	23	116	408	39		
204	Point	1	2 mon	ssmi	126.61	-125.82	0.0630	-14.35	0.0544	-64.4029	0.2333	140.53	222	227	110	60	208	827	144		
205	Point	1	2 mon	opi2	<b>61.02</b>	<b>-57.37</b>	0.0743	-8.97	0.0649	-26.8669	0.2548	<b>73.79</b>	1	4	92	48	1	146	2		
206	Point	1	2 mon	opi3	<b>67.04</b>	<b>-63.91</b>	0.0736	-9.41	0.0749	-32.7090	0.2737	<b>81.16</b>	7	9	96	42	7	161	3		
207	Point	1	2 mon	cmap	<b>79.32</b>	<b>-77.63</b>	0.0638	-9.74	0.0610	-43.8897	0.2469	<b>93.76</b>	21	25	109	51	19	225	6		
208	Point	1	2 mon	cmap2	82.87	-81.27	0.0612	-9.77	0.0600	-47.6426	0.2450	97.82	33	39	113	53	31	269	12		
209	Point	1	3 mon	reanalysis	<b>79.04</b>	<b>-77.52</b>	-0.0454	-1.92	0.0373	-71.4750	-0.1932	<b>91.86</b>	19	21	267	367	17	711	110		
210	Point	1	3 mon	gpcp	83.73	-79.17	0.0316	-7.51	0.0179	-47.8487	0.1337	99.20	38	35	162	113	36	384	31		
211	Point	1	3 mon	gpcp	<b>79.56</b>	<b>-75.73</b>	-0.0004	-3.33	0.0000	-268.3331	-0.0044	95.79	23	17	213	217	27	497	56		
212	Point	1	3 mon	gpcp trmm	115.56	-114.59	0.0906	-18.72	0.0589	-24.6615	0.2426	133.92	173	179	80	55	180	667	97		
213	Point	1	3 mon	ghcn cams grid	96.59	-96.32	0.0200	-6.98	0.0090	-101.2247	0.0950	109.18	79	87	185	144	64	559	69		
214	Point	1	3 mon	gpi	121.96	-119.09	0.0640	-13.23	0.0963	-75.4141	0.3104	140.36	202	201	108	32	206	749	125		
215	Point	1	3 mon	trmm 3a46	147.43	-147.43	0.2012	-34.79	0.2333	-34.4524	<b>0.4830</b>	157.65	244	247	17	4	230	742	121		
216	Point	1	3 mon	trmm 3b43	105.12	-104.48	0.1062	-19.12	0.0790	-21.7205	0.2810	124.11	112	112	64	39	120	447	50		
217	Point	1	3 mon	ssmi	126.32	-125.69	0.0741	-15.64	0.0753	-63.7298	0.2745	140.19	220	224	94	41	205	784	134		
218	Point	1	3 mon	opi2	<b>62.09</b>	<b>-57.36</b>	0.0388	-7.12	0.0177	-27.6356	0.1332	<b>74.94</b>	3	3	144	114	4	268	10		
219	Point	1	3 mon	opi3	<b>68.00</b>	<b>-63.79</b>	0.0330	-7.02	0.0150	-33.8587	0.1226	<b>82.44</b>	11	8	158	120	10	307	17		
220	Point	1	3 mon	cmap	<b>79.65</b>	<b>-77.26</b>	0.0298	-7.26	0.0131	-44.1299	0.1145	<b>94.39</b>	24	20	164	129	20	357	24		
221	Point	1	3 mon	cmap2	83.14	-80.86	0.0279	-7.22	0.0123	-48.0387	0.1108	98.39	34	38	170	134	34	410	41		
222	Point	1	4 mon	reanalysis	<b>79.42</b>	<b>-77.60</b>	-0.0202	-3.74	0.0074	-70.4241	-0.0861	<b>91.25</b>	22	24	247	300	15	608	83		
223	Point	1	4 mon	gpcp	84.04	-79.26	0.0048	-5.52	0.0004	-48.7548	0.0201	100.30	39	36	205	198	40	518	60		
224	Point	1	4 mon	gpcp	<b>79.97</b>	<b>-76.25</b>	-0.0142	-2.37	0.0202	-273.5651	-0.1421	96.68	27	18	237	336	28	646	94		
225	Point	1	4 mon	gpcp trmm	117.07	-116.43	0.0366	-13.71	0.0098	-26.4560	0.0988	137.48	184	184	150	141	199	858	153		
226	Point	1	4 mon	ghcn cams grid	96.56	-96.24	-0.0140	-3.87	0.0044	-102.5069	-0.0666	109.92	78	85	236	275	67	741	119		
227	Point	1	4 mon	gpi	122.19	-119.35	0.0377	-10.28	0.0332	-76.6967	0.1822	141.78	204	203	147	83	216	853	151		
228	Point	1	4 mon	trmm 3a46	147.21	-147.21	0.1272	-25.07	0.0960	-35.7867	0.3099	159.43	243	246	50	33	237	809	140		
229	Point	1	4 mon	trmm 3b43	107.04	-106.73	0.0618	-15.61	0.0268	-22.9417	0.1637	127.44	123	132	112	95	137	599	80		
230	Point	1	4 mon	ssmi	127.13	-126.31	0.0511	-12.91	0.0354	-64.2949	0.1881	141.28	225	228	129	80	215	877	154		
231	Point	1	4 mon	opi2	<b>62.93</b>	<b>-57.04</b>	-0.0204	-4.02	0.0049	-28.6329	-0.0697	<b>76.36</b>	5	1	248	282	5	541	64		
232	Point	1	4 mon	opi3	<b>67.87</b>	<b>-63.30</b>	-0.0022	-4.96	0.0001	-33.8804	-0.0080	<b>82.85</b>	9	7	216	219	11	462	51		
233	Point	1	4 mon	cmap	<b>79.85</b>	<b>-77.25</b>	-0.0069	-4.61	0.0007	-45.0991	-0.0266	95.57	25	19	224	235	26	529	62		
234	Point	1	4 mon	cmap2	83.15	-80.84	-0.0099	-4.36	0.0015	-49.0928	-0.0394	99.62	35	37	228	251	38	589	78		
235	Point	2	Yes	reanalysis	93.51	-89.52	0.1269	-17.59	0.0608	-7.7286	0.2465	119.34	60	56	51	52	102	321	20		
236	Point	2	Yes	gpcp	109.79	-107.74	0.1355	-20.77	0.0435	-8.8416	0.2086	126.72	137	137	41	73	134	522	61		
237	Point	2	Yes	gpcp	112.59	-110.90	0.1310	-22.19	0.0487	-11.5527	0.2206	127.74	156	157	43	69	139	564	71		
238	Point	2	Yes	gpcp trmm	122.93	-121.09	0.1957	-24.85	0.0920	-8.0414	0.3034	141.96	208	211	19	34	218	690	104		
239	Point	2	Yes	ghcn cams grid	120.24	-119.30	0.1310	-21.83	0.0308	-10.0481	0.1756	134.26	196	202	44	86	184	712	111		
240	Point	2	Yes	gpi	149.41	-148.30	0.1308	-26.19	0.0714	-17.6849	0.2672	167.74	249	253	45	45	254	846	150		
241	Point	2	Yes	trmm 3a46	155.62	-155.62	0.2211	-31.43	0.1119	-13.4688	0.3345	169.52	289	289	13	28	261	860	155		
242	Point	2	Yes	trmm 3b43	117.78	-116.00	0.2114	-25.65	0.1041	-7.3813	0.3226	136.68	187	182	14	30	190	803	81		
243	Point	2	Yes	ssmi	152.01	-151.14	0.1439	-28.01	0.0567	-16.6900	0.2380	165.57	270	275	33	57	247	882	156		
244	Point	2	Yes	opi2	88.63	-85.24	0.1743	-20.75	0.0459	-5.4370	0.2142	102.48	49	43	24	71	45	232	7		
245	Point	2	Yes	opi3	94.48	-91.72	0.1302	-18.15	0.0296	-6.4468	0.1721	110.23	69	62	47	88	69	335	21		
246	Point	2	Yes	cmap	107.52	-105.92	0.1303	-20.01	0.0328	-8.3029	0.1810	123.20	126	120	46	84	114	490	55		
247	Point	2	Yes	cmap2	111.20	-109.54	0.1231	-19.74	0.0308	-8.8892	0.1758	127.02	146	147	55	87	136	571	72		
248	Point	2	No	reanalysis	97.61	-90.17	-0.0071	-7.20	0.0002	-11.5961	-0.0158	124.96	88	80	225	226	123	720	114		
249	Point	2	No	gpcp	110.83	-108.39	-0.0119	-6.59	0.0004	-12.6884	-0.0210	130.17	143	142	232	233	153	903	161		
250	Point	2	No	gpcp trmm	112.52	-111.02	-0.0188	-7.01	0.0014	-17.1561	-0.0371	130.73	154	156	242	247	159	960	174		
251	Point	2	No	gpcp trmm	123.58	-122.87	0.1365	-19.54	0.0546	-10.3828	0.2336	144.29	216	213	39	58	221	747	123		
252	Point	2	No	ghcn cams grid	120.94	-119.95	-0.0269	-4.77	0.0017	-14.0954	-0.0413	136.80	198	205	260	252	192	1107	215		
253	Point	2	No	gpi	150.63	-148.42	-0.0285	-3.93	0.0045	-25.2368	-0.0671	172.65	262	257	262	277	273	1331	278		
254	Point	2	No	trmm 3a46	157.93	-157.93	0.0906	-12.92	0.0252	-19.3316	0.1587	173.62	291	291	81	98	275	1036	193		
255	Point	2	No	trmm 3b43	118.42	-117.78	0.1436	-19.66	0.0585	-9.6065	0.2419	139.29	194	189	35	56	202	678	102		
256	Point	2	No	ssmi	152.94	-151.87	0.0222	-4.66	0.0018	-23.2264	-0.0421	169.18	279	279	249	253	258	1318	275		
257	Point	2	No	opi2	89.59	-85.89	-0.0124	-6.82	0.0003	-7.9868	-0.0174	105.55	50	52	233	227	48	610	84		
258	Point	2	No	opi3	95.28	-92.37	-0.0323	-5.05	0.0024	-9.3274	-0.0490	113.15	74	73	267	260	79	753	126		
259	Point	2	No	cmap	108.22	-106.57	-0.0350	-4.33	0.0031	-11.8774	-0.0557	126.35	128	128	275	268	132	931	167		
260	Point	2	No	cmap2	111.86	-110.19	-0.0398	-3.71	0.0042	-12.6865	-0.0652	130.26	148	153	279	274	156	1010	189		
261	Point	2	1 mon	reanalysis	97.74	-90.07	-0.0114	-6.87	0.0006	-11.5993	-0.0253	125.20	87	59	230	234	126	736	118		
262	Point	2	1 mon	gpcp	112.35	-108.17	-0.0468	-3.13	0.0067	-12.7989	-0.0821	131.02	151	140	289	294	161	1035	191		
263	Point	2	1 mon	gpcp	113.29	-111.11	-0.0525	-3.77	0.0109	-17.4424	-0.1042	131.87	160	159	300	317	167	1103	211		
264	Point	2	1 mon	gpcp trmm	123.31	-121.59	0.0928	-14.07	0.0249	-10.2776	0.1579	145.11	211	212	79	99	223	824	142		
265	Point	2	1 mon	ghcn cams grid	121.69	-119.99	-0.0255	-4.95	0.0015	-14.0559	-0.0392	136.86	201	206	258	250	193	1108	217		
266	Point	2	1 mon	gpi	150.16	-148.09	-0.0324	-3.54	0.0058	-25.1294	-0.0781	172.48	257	252	268	285	271	1333	280		
267	Point	2	1 mon	trmm 3a46	159.20	-159.20	-0.0190	4.84	0.0011	-19.8678	-0.0329	177.71	294	299	244	241	290	1368	292		
268	Point	2	1 mon	trmm 3b43	118.10	-118.48	0.08														

Northwest Brazil - Monthly Surface

ID	Flumoff Meth	Re	Storage	Precipitation	Absolute Error	Bias	Slope	Intercept	R <sup>2</sup>	R <sup>2</sup> 1.1	Corr Coef	RMSE	AER	BR	SR	CCR	RMR	TR	BR
304	Point	2	4 mon	ghcn cams grid	122.85	-120.37	-0.0349	-3.65	0.0029	-14.0677	-0.0535	137.41	207	210	274	265	197	1153	234
305	Point	2	4 mon	gpi	150.84	-148.37	0.0238	-11.41	0.0031	-24.4296	0.0561	170.96	263	258	179	166	265	1129	224
306	Point	2	4 mon	trmm 3a46	156.48	-156.48	0.0484	-4.45	0.0068	-17.9681	0.0824	173.92	290	290	133	153	276	1142	232
307	Point	2	4 mon	trmm 3b43	129.43	-120.36	-0.0188	0.14	0.0009	-10.0538	-0.0304	147.79	227	208	243	238	226	1142	231
308	Point	2	4 mon	ssmi	153.24	-152.96	0.0476	-15.99	0.0081	-22.6422	0.0902	168.88	281	288	135	149	257	1110	218
309	Point	2	4 mon	opi2	93.14	-85.93	-0.0598	-2.91	0.0070	-8.1016	-0.0840	106.80	57	53	316	296	52	774	131
310	Point	2	4 mon	opi3	98.59	-92.59	-0.0252	5.45	0.0014	-9.2246	-0.0379	113.19	94	78	257	248	80	757	127
311	Point	2	4 mon	cmmap	111.03	-106.53	-0.0276	-4.86	0.0019	-11.7095	-0.0438	126.20	145	127	261	255	130	918	164
312	Point	2	4 mon	cmmap2	114.25	-110.16	-0.0196	-5.58	0.0010	-12.4403	-0.0319	129.78	165	152	246	240	149	952	173
313	Area	1	Yes	reanalysis	105.05	-104.62	-0.1243	-23.19	0.0301	-12.5610	-0.1735	121.05	110	115	371	354	104	1054	198
314	Area	1	Yes	gpcp	116.80	-114.10	-0.2326	-22.93	0.1296	-12.1900	-0.3600	140.66	182	171	439	449	210	1451	318
315	Area	1	Yes	gpcp	99.62	-97.14	-0.0956	-18.23	0.0994	-43.4317	-0.3152	117.29	100	95	344	428	100	1067	199
316	Area	1	Yes	gpcp trmm	188.82	-188.82	-0.1681	-65.91	0.0612	-18.6309	-0.2475	210.91	394	394	398	388	396	1970	440
317	Area	1	Yes	ghcn cams grid	123.65	-123.44	-0.1651	-17.29	0.0664	-17.3517	-0.2577	140.76	217	219	397	395	213	1441	315
318	Area	1	Yes	gpi	159.08	-158.38	0.0296	-48.83	0.0028	-16.7108	0.0533	180.34	293	296	165	169	304	1227	250
319	Area	1	Yes	trmm 3a46	236.23	-236.23	-0.3251	-51.48	0.0550	-37.0212	-0.5050	253.42	458	458	464	464	462	2306	466
320	Area	1	Yes	trmm 3b43	179.87	-179.87	-0.2062	-62.53	0.0870	-17.0322	-0.2949	203.09	386	386	418	415	391	1996	443
321	Area	1	Yes	ssmi	168.55	-168.55	-0.0333	-45.93	0.0022	-15.7822	-0.0468	186.71	351	357	270	257	352	1587	347
322	Area	1	Yes	opi2	94.58	-92.28	-0.2865	-25.12	0.1288	-8.3030	-0.3588	116.51	72	72	456	447	95	1142	230
323	Area	1	Yes	opi3	100.76	-98.95	-0.2307	-26.50	0.0986	-9.4172	-0.3140	123.29	107	102	437	427	115	1188	241
324	Area	1	Yes	cmmap	114.24	-113.15	-0.2377	-22.63	0.1154	-11.7296	-0.3398	136.75	164	168	444	440	191	1407	306
325	Area	1	Yes	cmmap2	117.72	-116.86	-0.2228	-22.89	0.1090	-12.4702	-0.3301	140.67	186	188	434	433	212	1453	319
326	Area	1	No	reanalysis	105.75	-104.60	-0.3340	-7.98	0.2758	-16.9791	-0.5252	123.71	116	114	465	465	118	1278	261
327	Area	1	No	gpcp	116.10	-114.16	-0.2223	-23.75	0.1191	-12.2180	-0.3451	140.42	178	173	433	443	207	1434	314
328	Area	1	No	gpcp	98.44	-97.02	-0.0583	-20.73	0.0853	-95.0491	-0.2820	115.40	92	94	312	411	92	1001	184
329	Area	1	No	gpcp trmm	188.76	-188.76	-0.1455	-68.23	0.0366	-14.7869	-0.1814	211.69	393	385	363	398	1932	430	
330	Area	1	No	ghcn cams grid	123.39	-123.31	-0.1274	-20.59	0.0499	-21.6474	-0.2235	139.13	213	216	373	379	201	1382	298
331	Area	1	No	gpi	160.56	-158.20	-0.1380	-29.95	0.0589	-17.4393	-0.2426	185.65	298	292	379	387	344	1700	377
332	Area	1	No	trmm 3a46	235.78	-235.78	-0.3963	-41.10	0.2815	-27.6626	-0.5306	255.29	457	457	468	466	466	2314	467
333	Area	1	No	trmm 3b43	179.82	-179.82	-0.1884	-64.21	0.0582	-13.6027	-0.2413	204.02	384	384	408	386	393	1955	435
334	Area	1	No	ssmi	168.40	-168.29	-0.1418	-32.79	0.0383	-15.6212	-0.1958	189.03	350	352	382	369	366	1819	402
335	Area	1	No	opi2	94.30	-91.97	-0.2787	-25.22	0.1219	-8.2404	-0.3491	116.11	68	64	454	444	94	1124	223
336	Area	1	No	opi3	99.71	-98.64	-0.1964	-28.20	0.0715	-9.2494	-0.2674	122.29	101	99	411	402	109	1122	221
337	Area	1	No	cmmap	113.52	-113.12	-0.2158	-24.20	0.0955	-11.6838	-0.3091	136.21	162	167	428	424	189	1370	293
338	Area	1	No	cmmap2	117.18	-116.83	-0.2032	-24.37	0.1010	-12.4308	-0.3017	140.17	185	187	414	418	204	1408	307
339	Area	1	1 mon	reanalysis	105.55	-104.51	-0.2902	-11.16	0.2082	-16.7536	-0.4563	123.01	115	113	457	460	112	1257	256
340	Area	1	1 mon	gpcp	115.48	-113.94	-0.1147	-31.79	0.0315	-11.6229	-0.1776	137.43	172	170	362	357	198	1259	258
341	Area	1	1 mon	gpcp	97.95	-96.85	0.0036	-25.21	0.0003	-91.3707	0.0177	113.52	89	92	208	200	85	674	100
342	Area	1	1 mon	gpcp trmm	188.89	-188.89	0.0690	-91.32	0.0083	-13.9278	0.0910	207.00	395	395	101	148	394	1433	313
343	Area	1	1 mon	ghcn cams grid	123.43	-123.35	-0.0241	-29.97	0.0018	-21.0103	-0.0423	137.21	214	217	253	254	194	1132	226
344	Area	1	1 mon	gpi	160.69	-158.24	-0.1134	-32.58	0.0409	-17.2476	-0.2024	185.08	300	293	360	371	341	1665	372
345	Area	1	1 mon	trmm 3a46	238.23	-238.23	-0.2326	-64.64	0.0970	-27.5108	-0.3115	255.00	465	465	440	425	465	2260	464
346	Area	1	1 mon	trmm 3b43	179.17	-179.17	-0.0100	-82.09	0.0002	-12.7886	-0.0127	199.38	381	381	229	223	388	1602	354
347	Area	1	1 mon	ssmi	168.84	-168.51	-0.0610	-42.55	0.0071	-15.3077	-0.0844	187.56	353	355	317	297	359	1681	373
348	Area	1	1 mon	opi2	93.95	-92.15	-0.1267	-33.18	0.0252	-7.7654	-0.1586	113.26	65	67	372	345	81	930	166
349	Area	1	1 mon	opi3	99.47	-98.85	-0.0722	-35.56	0.0096	-8.7920	-0.0981	119.71	99	101	325	314	103	942	172
350	Area	1	1 mon	cmmap	113.27	-113.07	-0.0934	-33.21	0.0179	-11.1463	-0.1338	133.50	159	166	339	334	179	1177	239
351	Area	1	1 mon	cmmap2	116.90	-116.76	-0.0841	-33.58	0.0156	-11.8634	-0.1248	137.39	183	186	332	327	196	1224	248
352	Area	1	2 mon	reanalysis	105.33	-104.47	-0.1991	-17.77	0.0980	-16.2753	-0.3131	121.44	113	111	412	426	107	1169	238
353	Area	1	2 mon	gpcp	115.26	-114.12	-0.0079	-39.70	0.0001	-11.1057	-0.0122	134.81	170	172	226	222	185	975	178
354	Area	1	2 mon	gpcp	97.43	-96.67	0.0545	-28.75	0.0279	-89.2202	0.2701	111.93	85	91	223	44	75	418	43
355	Area	1	2 mon	gpcp trmm	187.87	-187.87	0.1970	-104.36	0.0670	-13.1416	0.2589	203.14	391	391	18	47	392	1239	253
356	Area	1	2 mon	ghcn cams grid	123.43	-123.37	0.0735	-38.84	0.0166	-20.4045	0.1288	135.35	215	218	97	117	186	833	146
357	Area	1	2 mon	gpi	160.72	-158.33	-0.0487	-39.91	0.0076	-16.7858	-0.0869	183.17	302	295	292	302	322	1513	330
358	Area	1	2 mon	trmm 3a46	240.08	-240.08	0.0040	-98.00	0.0000	-28.3888	0.0053	252.96	468	468	206	209	461	1812	400
359	Area	1	2 mon	trmm 3b43	178.84	-178.84	0.1439	-96.64	0.0336	-12.0736	0.1833	195.71	380	380	34	81	385	1260	260
360	Area	1	2 mon	ssmi	169.20	-168.81	0.0089	-50.91	0.0002	-14.9987	0.0123	186.36	358	359	200	205	349	1471	324
361	Area	1	2 mon	opi2	93.44	-92.14	-0.0028	-39.71	0.0000	-7.3572	-0.0035	110.77	59	65	219	215	70	628	91
362	Area	1	2 mon	opi3	99.11	-98.68	0.0532	-42.98	0.0052	-8.2782	0.0722	116.72	97	100	126	158	97	578	75
363	Area	1	2 mon	cmmap	112.72	-112.59	0.0235	-41.77	0.0011	-10.5163	0.0333	130.21	157	165	180	186	154	842	149
364	Area	1	2 mon	cmmap2	116.34	-116.23	0.0264	-42.08	0.0015	-11.2054	0.0386	134.05	180	183	173	182	182	900	160
365	Area	1	3 mon	reanalysis	105.05	-104.40	-0.0732	-26.91	0.0132	-15.6283	-0.1150	119.22	111	109	328	321	101	970	177
366	Area	1	3 mon	gpcp	115.45	-114.31	0.0689	-45.41	0.0113	-10.7327	0.1065	132.96	171	176	102	135	176	760	128
367	Area	1	3 mon	gpcp	97.91	-97.02	0.0681	-29.58	0.1170	-91.5094	0.3420	111.75	88	93	104	27	74	386	32
368	Area	1	3 mon	gpcp trmm	188.53	-188.53	0.3360	-118.53	0.1926	-12.6582	0.4389	200.32	392	392	5	12	389	1190	242
369	Area	1	3 mon	ghcn cams grid	123.29	-123.25	0.1284	-43.78	0.0507	-20.0453	0.2251	134.17	210	215	49	66	183	723	115
370	Area	1	3 mon	gpi	160.28	-1													

Northwest Brazil - Monthly Surface

Of	Runoff Meth	Re	Storage	Precipitation	Absolute Error	Bias	Slope	Intercept	R <sup>2</sup>	R <sup>2</sup> 1.1	Corr Coef	RMSE	AER	BR	SR	CCR	RMR	TR	BR
405	Area	2	No	gpcc	168.92	-168.27	-0.2597	-41.55	0.1224	-16.0573	-0.3499	190.57	354	351	448	445	369	1967	439
406	Area	2	No	gpcc	166.69	-166.69	-0.2032	-43.85	0.0995	-21.5753	-0.3155	185.73	337	340	415	430	345	1867	417
407	Area	2	No	gpcc trmm	198.33	-198.33	-0.1171	-64.66	0.0312	-19.5123	-0.1767	219.78	410	411	364	356	418	1959	437
408	Area	2	No	ghcn cams grid	179.83	-179.83	-0.2687	-37.53	0.0993	-17.3040	-0.3152	197.41	385	385	452	429	387	2038	447
409	Area	2	No	gpi	208.26	-207.55	-0.1222	-49.90	0.0518	-28.0986	-0.2275	229.58	434	432	368	381	446	2061	450
410	Area	2	No	trmm 3a46	234.40	-234.40	-0.2343	-37.58	0.1254	-30.9810	-0.3541	252.20	454	454	442	446	459	2255	482
411	Area	2	No	trmm 3b43	193.24	-193.24	-0.1074	-66.37	0.0254	-18.5357	-0.1594	214.48	401	401	354	346	404	1906	426
412	Area	2	No	ssmi	209.31	-209.26	-0.1790	-39.70	0.0677	-24.6769	-0.2602	227.06	438	438	404	398	441	2119	455
413	Area	2	No	opi2	146.45	-145.77	-0.3400	-41.12	0.1338	-11.9242	-0.3658	165.88	242	236	466	450	248	1642	367
414	Area	2	No	opi3	152.54	-152.25	-0.3203	-40.57	0.1372	-13.0814	-0.3705	173.15	273	283	462	451	274	1743	386
415	Area	2	No	cmmap	166.51	-166.45	-0.3123	-36.82	0.1441	-15.4145	-0.3797	186.95	332	335	461	452	356	1936	432
416	Area	2	No	cmmap2	170.08	-170.08	-0.3053	-36.40	0.1454	-16.0977	-0.3813	190.80	363	363	459	453	370	2008	444
417	Area	2	1	reanalysis	152.60	-150.10	-0.2457	-47.63	0.1752	-14.9101	-0.4185	184.04	274	268	446	456	331	1775	395
418	Area	2	1	mon gpcc	169.08	-168.20	-0.2098	-46.78	0.0799	-15.8640	-0.2826	189.48	357	350	422	406	367	1902	424
419	Area	2	1	mon gpcc	166.45	-166.43	-0.2151	-42.51	0.1118	-21.4779	-0.3343	185.86	330	334	427	437	346	1874	420
420	Area	2	1	mon gpcc trmm	198.45	-198.45	-0.1078	-67.16	0.0270	-19.7316	-0.1643	219.74	411	412	355	348	417	1943	433
421	Area	2	1	mon ghcn cams grid	180.01	-180.01	-0.1340	-52.80	0.0248	-16.9726	-0.1575	195.61	387	387	377	343	384	1878	421
422	Area	2	1	mon gpi	207.57	-206.99	-0.1404	-47.32	0.0680	-27.9675	-0.2607	229.56	430	430	381	399	445	2085	453
423	Area	2	1	mon trmm 3a46	237.25	-237.25	-0.3245	-24.02	0.2443	-32.5217	-0.4942	256.42	461	461	463	462	468	2315	468
424	Area	2	1	mon trmm 3b43	193.34	-193.34	-0.1325	-64.90	0.0395	-18.8925	-0.1987	215.25	402	402	376	370	409	1959	436
425	Area	2	1	mon smi	209.78	-209.53	-0.2115	-35.72	0.0951	-24.9621	-0.3063	227.94	439	439	423	423	442	2166	459
426	Area	2	1	mon opi2	146.09	-145.78	-0.2033	-51.99	0.0479	-11.8109	-0.2188	163.85	237	237	416	376	245	1511	329
427	Area	2	1	mon opi3	152.34	-152.22	-0.2145	-49.73	0.0616	-12.7953	-0.2481	171.38	272	282	426	389	267	1836	386
428	Area	2	1	mon cmmap	168.39	-168.33	-0.2138	-46.78	0.0673	-15.0980	-0.2594	185.12	329	339	425	397	342	1826	405
429	Area	2	1	mon cmmap2	170.00	-169.94	-0.2069	-46.71	0.0665	-15.7622	-0.2579	188.91	362	362	419	396	365	1904	425
430	Area	2	2	mon reanalysis	152.19	-149.97	-0.1399	-56.52	0.0568	-14.2684	-0.2382	180.35	271	265	380	384	305	1605	356
431	Area	2	2	mon gpcc	169.46	-168.53	-0.0698	-60.97	0.0088	-15.4009	-0.0941	186.92	361	356	323	311	355	1706	378
432	Area	2	2	mon gpcc	166.15	-166.15	-0.0676	-57.10	0.0112	-20.9089	-0.1059	182.68	327	332	321	319	318	1617	361
433	Area	2	2	mon gpcc trmm	201.63	-201.63	0.0234	-83.55	0.0012	-19.4266	0.0350	219.13	416	416	181	184	415	1612	359
434	Area	2	2	mon ghcn cams grid	180.47	-180.47	0.0117	-69.30	0.0002	-16.8291	0.0137	193.79	390	390	194	203	380	1557	338
435	Area	2	2	mon gpi	207.61	-206.78	-0.1224	-49.54	0.0521	-27.9140	-0.2282	228.96	432	429	369	382	444	2056	449
436	Area	2	2	mon trmm 3a46	239.64	-239.64	-0.1356	-55.01	0.0422	-31.8792	-0.2055	255.34	467	467	378	373	467	2152	457
437	Area	2	2	mon trmm 3b43	196.27	-196.27	0.0050	-81.30	0.0001	-18.5489	0.0074	214.37	405	405	204	207	403	1624	364
438	Area	2	2	mon smi	210.33	-209.87	-0.1316	-47.36	0.0369	-24.6042	-0.1920	226.84	442	442	375	365	440	2064	451
439	Area	2	2	mon opi2	146.10	-146.10	-0.0633	-63.04	0.0046	-11.3242	-0.0681	162.03	238	241	318	279	243	1319	276
440	Area	2	2	mon opi3	152.60	-152.80	-0.0389	-64.70	0.0020	-12.3692	-0.0449	168.76	275	284	277	256	256	1348	288
441	Area	2	2	mon cmmap	166.66	-166.66	-0.0438	-63.67	0.0028	-14.6393	-0.0532	182.53	336	339	284	264	317	1540	335
442	Area	2	2	mon cmmap2	170.28	-170.28	-0.0398	-63.92	0.0025	-15.2885	-0.0496	186.28	366	366	280	263	348	1623	362
443	Area	2	3	mon reanalysis	151.60	-149.70	-0.0731	-61.78	0.0156	-13.8939	-0.1249	177.88	267	262	327	328	292	1476	325
444	Area	2	3	mon gpcc	168.65	-168.51	0.0026	-68.03	0.0000	-15.1783	0.0035	185.39	352	354	209	211	343	1469	323
445	Area	2	3	mon gpcc	166.70	-166.70	-0.0223	-61.55	0.0012	-20.6673	-0.0346	182.17	338	341	250	243	315	1487	327
446	Area	2	3	mon gpcc trmm	204.08	-204.08	0.2041	-106.39	0.0912	-18.7582	0.3020	217.08	419	419	16	36	411	1301	270
447	Area	2	3	mon ghcn cams grid	180.45	-180.45	0.0831	-77.13	0.0095	-16.4753	0.0975	192.68	389	389	85	142	378	1383	299
448	Area	2	3	mon gpi	207.53	-206.70	-0.0360	-61.55	0.0045	-27.1980	-0.0671	226.59	428	428	276	278	438	1848	411
449	Area	2	3	mon trmm 3a46	238.53	-238.53	-0.0239	-73.81	0.0013	-30.5671	-0.0360	252.35	466	466	252	245	460	1889	423
450	Area	2	3	mon trmm 3b43	198.58	-198.58	0.1465	-98.52	0.0460	-18.0511	0.2146	213.16	412	413	32	70	400	1327	277
451	Area	2	3	mon smi	211.10	-210.79	-0.0251	-63.47	0.0014	-24.5587	-0.0369	225.72	444	444	256	246	432	1822	404
452	Area	2	3	mon opi2	146.18	-146.09	0.0384	-70.78	0.0017	-11.1166	0.0414	160.44	239	240	146	179	240	1044	196
453	Area	2	3	mon opi3	152.68	-152.61	0.0365	-70.87	0.0018	-12.2010	0.0422	167.47	277	285	152	178	253	1145	233
454	Area	2	3	mon cmmap	168.57	-168.57	0.0279	-70.53	0.0012	-14.4570	0.0340	181.21	334	337	171	185	310	1337	284
455	Area	2	3	mon cmmap2	170.20	-170.20	0.0343	-71.28	0.0018	-15.0889	0.0429	184.88	365	365	154	177	339	1400	304
456	Area	2	4	mon reanalysis	151.60	-149.17	-0.0291	-65.24	0.0025	-13.5446	-0.0496	175.85	286	259	283	281	282	1331	279
457	Area	2	4	mon gpcc	169.06	-168.49	0.0582	-73.58	0.0064	-14.9622	0.0799	184.22	356	353	116	155	336	1316	273
458	Area	2	4	mon gpcc	166.75	-166.75	0.0561	-69.12	0.0077	-20.6160	0.0878	180.52	339	342	119	150	307	1257	257
459	Area	2	4	mon gpcc trmm	204.88	-204.88	0.0965	-93.11	0.0202	-18.9711	0.1422	220.66	420	420	76	107	420	1443	317
460	Area	2	4	mon ghcn cams grid	180.39	-180.39	0.1236	-81.55	0.0211	-16.3456	0.1452	192.03	388	388	53	105	377	1311	272
461	Area	2	4	mon gpi	207.08	-206.55	0.0444	-72.48	0.0069	-26.8061	0.0834	224.21	427	427	141	152	427	1574	342
462	Area	2	4	mon trmm 3a46	237.15	-237.15	0.1117	-95.23	0.0279	-28.9580	0.1669	248.78	480	480	59	91	456	1526	333
463	Area	2	4	mon trmm 3b43	199.46	-199.46	0.1139	-94.65	0.0276	-17.9598	0.1661	215.00	414	414	58	93	406	1385	302
464	Area	2	4	mon smi	210.81	-210.77	0.0779	-78.15	0.0131	-24.0744	0.1147	223.78	443	443	88	127	426	1527	334
465	Area	2	4	mon opi2	146.32	-145.96	0.1043	-75.78	0.0126	-10.9406	0.1124	159.33	241	239	86	133	238	915	163
466	Area	2	4	mon opi3	152.62	-152.62	0.1040	-76.46	0.0145	-12.0146	0.1203	166.34	276	286	67	123	250	1002	185
467	Area	2	4	mon cmmap	166.56	-166.56	0.0938	-76.90	0.0130	-14.2519	0.1142	180.07	333	336	78	130	301	1178	240
468	Area	2	4	mon cmmap2	170.18	-170.18	0.1007	-77.95	0.0159	-14.8708	0.1260	183.69	364	364	70	118	330	1246	254

Northwest Brazil - Yearly Surface

DI	Runoff Meth	Re	Storage	Precipitation	Absolute Error	Bias	Slope	Intercept	R <sup>2</sup>	R <sup>2</sup> 1:1	Corr Coef	RMSE	AER	BR	SR	CCR	RMR	TR	BR	Plot
1	Linear	1	Yes	reanalysis	99.08	-99.08	-0.2249	-10.36	0.0823	-29.0088	-0.2868	105.14	34	34	73	89	34	264	41	Plot
2	Linear	1	Yes	gpcp	107.36	-107.36	-0.4908	2.91	0.1765	-20.9288	-0.4201	113.92	48	48	107	116	55	374	77	
3	Linear	1	Yes	gpcp	95.65	-95.65	0.0849	-29.74	0.0421	-246.4599	0.2052	96.80	29	29	38	23	22	141	12	
4	Linear	1	Yes	gpcp trmm	176.95	-176.95	-0.9969	22.42	0.8217	-87.6108	-0.9065	180.53	125	125	150	153	125	678	142	
5	Linear	1	Yes	ghcn cams grid	117.61	-117.61	-0.2912	-0.31	0.0461	-37.8001	-0.2147	120.54	65	65	80	77	63	350	72	
6	Linear	1	Yes	gpi	151.57	-151.57	0.4707	-91.81	0.1153	-33.9191	0.3396	153.86	96	96	4	10	91	297	53	
7	Linear	1	Yes	trmm 3a46	208.96	-208.96	-0.5175	-2.86	0.2237	-103.8516	-0.4730	211.71	150	150	110	126	149	685	143	
8	Linear	1	Yes	trmm 3b43	163.05	-163.05	-0.3727	-29.53	0.0290	-33.2766	-0.1702	166.40	104	104	93	63	104	468	100	
9	Linear	1	Yes	ssmi	160.34	-160.34	0.2314	-69.68	0.0270	-31.7848	0.1644	163.54	101	101	14	32	101	349	71	
10	Linear	1	Yes	opi2	84.86	-84.86	-0.5687	-4.12	0.2230	-13.6412	-0.4722	93.09	16	16	123	125	17	297	52	
11	Linear	1	Yes	opi3	91.34	-91.34	-0.4533	7.12	0.1125	-15.1415	-0.3354	97.74	22	22	103	99	27	273	45	
12	Linear	1	Yes	cmmap	105.54	-105.54	-0.5883	-7.62	0.1912	-20.0878	-0.4373	111.71	42	42	122	118	49	373	76	
13	Linear	1	Yes	cmmap2	109.17	-109.17	-0.5599	9.04	0.2003	-21.4911	-0.4475	115.37	53	53	119	121	57	403	89	
14	Linear	1	No	reanalysis	99.57	-99.57	-0.2214	-11.11	0.0868	-31.8546	-0.2946	105.44	35	35	71	91	36	268	43	
15	Linear	1	No	gpcp	107.64	-107.64	-0.5279	5.38	0.1992	-20.5609	-0.4463	114.39	49	49	111	120	56	385	81	
16	Linear	1	No	gpcp	92.93	-92.93	-0.0390	-18.10	0.0393	-999.0000	-0.1983	94.25	27	27	60	74	20	208	28	
17	Linear	1	No	gpcp trmm	180.61	-180.61	-0.5379	-27.06	0.6321	-238.8509	-0.7950	182.74	132	132	114	149	130	657	140	
18	Linear	1	No	ghcn cams grid	118.11	-118.11	-0.2959	-0.39	0.0518	-41.5185	-0.2277	120.91	67	67	82	79	65	360	75	
19	Linear	1	No	gpi	150.91	-150.91	0.5077	-95.32	0.1218	-30.4904	0.3490	153.34	95	95	3	8	90	291	50	
20	Linear	1	No	trmm 3a46	212.86	-212.86	-0.2675	-40.71	0.1967	-353.2092	-0.4436	214.44	152	152	76	119	151	650	138	
21	Linear	1	No	trmm 3b43	165.98	-165.98	0.1673	-84.98	0.0068	-39.8911	0.0825	168.39	111	111	24	41	108	395	85	
22	Linear	1	No	ssmi	160.79	-160.79	0.3336	-82.18	0.0524	-29.6978	0.2289	163.91	102	102	8	19	102	333	65	
23	Linear	1	No	opi2	85.14	-85.14	-0.6040	-2.58	0.2453	-13.4280	-0.4952	93.57	17	17	126	132	18	310	58	
24	Linear	1	No	opi3	91.62	-91.62	-0.4843	-5.60	0.1252	-14.8826	-0.3538	98.18	24	24	106	103	28	285	48	
25	Linear	1	No	cmmap	105.82	-105.82	-0.6255	11.46	0.2259	-19.7520	-0.4753	112.22	43	43	130	127	51	394	84	
26	Linear	1	No	cmmap2	109.44	-109.44	-0.6197	13.29	0.2393	-21.1323	-0.4891	115.89	54	54	129	131	58	426	91	
27	Linear	2	Yes	reanalysis	145.71	-145.71	-0.2846	-39.89	0.1100	-24.2201	-0.3316	155.70	84	84	79	98	94	439	94	Plot
28	Linear	2	Yes	gpcp	163.94	-163.94	-0.5094	-12.09	0.0817	-28.5960	-0.2859	168.66	106	106	108	87	110	517	115	
29	Linear	2	Yes	gpcp	166.07	-166.07	-0.2459	-38.87	0.0354	-85.7825	-0.1882	167.89	112	112	74	68	107	473	102	
30	Linear	2	Yes	gpcp trmm	182.14	-182.14	-1.0006	57.25	0.5187	-53.6166	-0.7202	186.56	133	133	152	145	133	696	146	
31	Linear	2	Yes	ghcn cams grid	175.49	-175.49	-0.5536	-1.24	0.0351	-32.2825	-0.1873	178.86	123	123	117	67	124	554	129	
32	Linear	2	Yes	gpi	203.97	-203.97	0.1890	-90.05	0.0362	-109.2188	0.1903	205.50	143	143	22	27	143	478	106	
33	Linear	2	Yes	trmm 3a46	208.49	-208.49	-0.8821	53.09	0.3750	-90.8658	-0.6124	211.87	149	149	138	141	150	727	152	
34	Linear	2	Yes	trmm 3b43	177.06	-177.06	-1.2271	78.11	0.4747	-50.2838	-0.6890	180.78	126	126	154	143	126	675	141	
35	Linear	2	Yes	ssmi	200.40	-200.40	0.0879	-70.63	0.0085	-113.5349	0.0920	202.09	141	141	37	38	141	498	109	
36	Linear	2	Yes	opi2	141.43	-141.43	-0.2396	-39.93	0.0309	-21.3608	-0.1757	146.60	79	79	83	65	82	368	82	
37	Linear	2	Yes	opi3	147.92	-147.92	-0.6502	-8.34	0.0945	-23.2769	-0.3075	152.76	88	88	134	95	89	494	108	
38	Linear	2	Yes	cmmap	162.12	-162.12	-0.5637	-7.64	0.0748	-27.8445	-0.2736	166.51	103	103	120	83	105	514	113	
39	Linear	2	Yes	cmmap2	165.74	-165.74	-0.5321	-8.84	0.0715	-29.1001	-0.2873	170.09	109	109	112	82	116	528	120	
40	Linear	2	No	reanalysis	147.52	-147.52	-0.3897	-33.03	0.5431	-65.9858	-0.7369	156.36	87	87	95	146	96	511	111	
41	Linear	2	No	gpcp	165.74	-165.74	-0.4282	-22.05	0.1521	-76.8036	-0.3900	168.51	110	110	99	114	109	542	123	
42	Linear	2	No	gpcp	164.93	-164.93	-0.6841	4.97	0.2570	-84.6282	-0.5069	167.25	108	108	137	135	106	594	135	
43	Linear	2	No	gpcp trmm	197.27	-197.27	-0.3334	-37.72	0.1670	-179.6623	-0.4087	199.21	139	139	87	115	138	618	137	
44	Linear	2	No	ghcn cams grid	177.30	-177.30	-0.3193	-29.32	0.0307	-86.8197	-0.1753	178.83	127	127	84	64	123	525	117	
45	Linear	2	No	gpi	206.08	-206.08	0.2699	-103.51	0.0830	-125.2622	0.2882	207.33	145	145	10	13	145	458	97	
46	Linear	2	No	trmm 3a46	228.66	-228.66	-0.7355	41.22	0.9983	-249.5580	-0.9992	231.24	154	154	144	156	155	763	155	
47	Linear	2	No	trmm 3b43	192.18	-192.18	-0.5366	-11.13	0.2633	-170.0357	-0.5132	193.83	135	135	113	136	133	654	139	
48	Linear	2	No	ssmi	206.75	-206.75	0.0313	-68.93	0.0010	-108.2371	0.0310	208.59	146	146	46	50	146	534	122	
49	Linear	2	No	opi2	143.24	-143.24	-0.1997	-49.54	0.0361	-57.4807	-0.1901	146.10	80	80	70	70	80	380	78	
50	Linear	2	No	opi3	149.72	-149.72	-0.4487	-27.19	0.1185	-62.5345	-0.3443	152.28	91	91	100	101	88	471	101	
51	Linear	2	No	cmmap	163.92	-163.92	-0.4712	-18.58	0.1377	-74.8265	-0.3711	166.36	105	105	105	107	103	525	116	
52	Linear	2	No	cmmap2	167.54	-167.54	-0.4518	-18.87	0.1357	-78.1769	-0.3684	169.99	116	116	101	106	115	554	128	
53	Point	1	Yes	reanalysis	77.15	-77.15	0.0151	-5.81	0.0024	-114.7806	0.0490	81.18	7	7	51	48	9	122	9	
54	Point	1	Yes	gpcp	78.82	-78.82	-0.0289	-2.86	0.0064	-136.1613	-0.0800	81.97	11	11	59	59	11	151	15	
55	Point	1	Yes	gpcp	78.09	-78.09	0.0793	-11.78	0.0434	-195.9986	0.2083	79.48	10	10	40	22	6	88	2	
56	Point	1	Yes	gpcp trmm	106.35	-106.35	-0.4092	34.35	0.6617	-154.1656	-0.8135	109.27	45	45	98	151	42	381	79	
57	Point	1	Yes	ghcn cams grid	95.58	-95.58	-0.0881	3.08	0.0261	-161.8060	-0.1614	97.12	28	28	68	61	24	209	29	
58	Point	1	Yes	gpi	119.48	-119.48	0.0199	-8.83	0.0031	-328.4046	0.0559	121.08	70	70	49	46	66	301	55	
59	Point	1	Yes	trmm 3a46	140.93	-140.93	-0.0041	-4.56	0.0001	-213.2685	-0.0079	142.53	78	78	54	54	78	342	68	
60	Point	1	Yes																	



Northwest Brazil - Yearly Surface

OI	Runoff Meth	Re	Storage	Precipitation	Absolute Error	Bias	Slope	Intercept	R <sup>2</sup>	R <sup>2</sup> 1:1	Corr Coef	RMSE	AER	BR	SR	CCR	RMR	TR	BR
102	Point	2	No	opi3	92.37	-92.37	0.2240	-26.73	0.0824	-66.0747	0.2870	93.71	26	26	16	15	19	102	3
103	Point	2	No	cmmap	106.57	-106.57	0.2185	-29.37	0.0826	-87.7221	0.2874	107.78	46	46	17	14	40	163	19
104	Point	2	No	cmmap2	110.19	-110.19	0.2105	-29.34	0.0821	-93.8192	0.2865	111.42	56	56	18	16	48	194	24
105	Area	1	Yes	reanalysis	104.11	-104.11	-0.2819	-11.26	0.0946	-23.3825	-0.3075	110.81	39	39	78	94	44	294	51
106	Area	1	Yes	gpcp	113.89	-113.89	-0.8067	4.96	0.1909	-16.8561	-0.4369	121.52	61	61	127	117	68	434	92
107	Area	1	Yes	gpcp	99.74	-99.74	0.0797	-33.46	0.0351	-252.4498	0.1873	100.87	36	36	39	30	31	172	20
108	Area	1	Yes	gpcp trmm	193.44	-193.44	-1.1186	18.08	0.8424	-85.0314	-0.9178	197.12	137	137	153	154	137	718	151
109	Area	1	Yes	ghcn cams grid	122.68	-122.68	-0.3396	-0.98	0.0459	-30.0756	-0.2142	126.11	72	72	88	78	75	383	80
110	Area	1	Yes	gpi	158.86	-158.86	0.5625	-109.47	0.1147	-25.8833	0.3387	161.76	100	100	2	11	100	313	59
111	Area	1	Yes	trmm 3a46	225.35	-225.35	-0.6584	-0.10	0.2889	-98.2992	-0.5375	228.35	153	153	135	137	153	731	153
112	Area	1	Yes	trmm 3b43	176.89	-176.89	-0.3619	-44.41	0.0193	-27.6096	-0.1389	180.90	124	124	92	60	127	527	119
113	Area	1	Yes	ssmi	168.55	-168.55	0.2500	-80.09	0.0221	-24.5160	0.1485	172.58	120	120	11	33	119	403	90
114	Area	1	Yes	opi2	91.38	-91.38	-0.6966	-4.06	0.2367	-11.1527	-0.4866	100.81	23	23	140	129	30	345	69
115	Area	1	Yes	opi3	97.87	-97.87	-0.5647	-7.19	0.1235	-12.2770	-0.3515	105.37	32	32	121	102	35	322	83
116	Area	1	Yes	cmmap	112.07	-112.07	-0.7043	10.90	0.2078	-16.0263	-0.4558	119.33	58	58	141	122	60	439	93
117	Area	1	Yes	cmmap2	115.69	-115.69	-0.6939	12.67	0.2177	-17.0831	-0.4665	122.98	63	63	139	124	70	459	98
118	Area	1	No	reanalysis	104.60	-104.60	-0.2785	-12.00	0.0974	-24.8999	-0.3121	111.14	40	40	77	96	46	299	54
119	Area	1	No	gpcp	114.16	-114.16	-0.6438	7.43	0.2087	-16.2834	-0.4568	122.01	62	62	133	123	69	449	95
120	Area	1	No	gpcp	97.02	-97.02	-0.0441	-21.82	0.0365	-817.5961	-0.1910	98.31	31	31	62	71	29	224	32
121	Area	1	No	gpcp trmm	197.10	-197.10	-0.6596	-31.40	0.7024	-209.9839	-0.8381	199.36	138	138	136	152	139	703	149
122	Area	1	No	ghcn cams grid	123.17	-123.17	-0.3442	-1.06	0.0498	-32.0342	-0.2232	126.49	74	74	89	78	76	391	83
123	Area	1	No	gpi	158.20	-158.20	0.5994	-112.98	0.1193	-23.4802	0.3454	161.30	99	99	1	9	99	307	57
124	Area	1	No	trmm 3a46	229.24	-229.24	-0.4084	-37.95	0.3098	-276.7720	-0.5566	231.07	155	155	97	138	154	699	147
125	Area	1	No	trmm 3b43	179.82	-179.82	0.1780	-99.86	0.0053	-31.8926	0.0725	182.92	130	130	23	44	132	459	99
126	Area	1	No	ssmi	169.00	-169.00	0.3523	-92.59	0.0410	-23.0115	0.2025	173.00	121	121	7	26	122	397	86
127	Area	1	No	opi2	91.66	-91.66	-0.7319	-2.52	0.2538	-10.9214	-0.5038	101.33	25	25	143	134	32	359	74
128	Area	1	No	opi3	98.14	-98.14	-0.5957	-5.67	0.1335	-12.0070	-0.3653	105.85	33	33	125	105	37	333	64
129	Area	1	No	cmmap	112.35	-112.35	-0.7614	14.74	0.2359	-15.6799	-0.4856	119.86	60	60	147	128	61	456	96
130	Area	1	No	cmmap2	115.97	-115.97	-0.7537	16.92	0.2493	-16.7143	-0.4993	123.52	64	64	145	133	71	477	105
131	Area	2	Yes	reanalysis	148.24	-148.24	-0.3479	-37.20	0.1473	-22.5478	-0.3838	158.94	89	89	90	111	97	476	104
132	Area	2	Yes	gpcp	166.47	-166.47	-0.5390	-11.63	0.0820	-26.4170	-0.2864	171.50	114	114	115	88	116	549	126
133	Area	2	Yes	gpcp	167.83	-167.83	-0.2943	-35.68	0.0400	-89.0065	-0.1999	169.95	117	117	81	75	114	504	110
134	Area	2	Yes	gpcp trmm	183.20	-183.20	-1.2969	91.63	0.6476	-40.6366	-0.8048	188.70	134	134	155	150	134	707	150
135	Area	2	Yes	ghcn cams grid	178.02	-178.02	-0.8139	2.99	0.0386	-29.7701	-0.1966	181.69	128	128	73	129	586	134	
136	Area	2	Yes	gpi	205.44	-205.44	0.1526	-86.40	0.0181	-85.0590	0.1345	207.30	144	144	26	35	144	493	107
137	Area	2	Yes	trmm 3a46	209.95	-209.95	-0.9464	92.72	0.4344	-55.8857	-0.6591	214.55	151	151	149	142	152	745	154
138	Area	2	Yes	trmm 3b43	178.12	-178.12	-1.6263	122.78	0.6198	-38.0055	-0.7873	182.86	129	129	156	148	131	693	145
139	Area	2	Yes	ssmi	201.31	-201.31	0.0692	-68.88	0.0039	-85.7705	0.0627	203.32	142	142	41	45	142	512	112
140	Area	2	Yes	opi2	143.96	-143.96	-0.3242	-40.54	0.0324	-19.8289	-0.1800	149.48	81	81	85	66	85	398	88
141	Area	2	Yes	opi3	150.45	-150.45	-0.7578	-1.77	0.1150	-21.6021	-0.3392	155.72	93	93	146	100	95	527	118
142	Area	2	Yes	cmmap	164.65	-164.65	-0.6343	-3.21	0.0849	-25.7482	-0.2914	169.40	107	107	132	90	113	549	125
143	Area	2	Yes	cmmap2	168.27	-168.27	-0.5942	-5.01	0.0799	-26.8886	-0.2826	172.97	118	118	124	86	121	567	131
144	Area	2	No	reanalysis	150.05	-150.05	-0.4530	-30.35	0.6004	-56.0371	-0.7748	159.53	92	92	102	147	98	531	121
145	Area	2	No	gpcp	168.27	-168.27	-0.4579	-21.60	0.1423	-64.7628	-0.3772	171.29	119	119	104	109	117	568	132
146	Area	2	No	gpcp	166.69	-166.69	-0.7126	8.16	0.2386	-69.7401	-0.4885	189.27	115	115	142	130	112	614	136
147	Area	2	No	gpcp trmm	198.33	-198.33	-0.6296	-3.34	0.3707	-113.5518	-0.6088	201.08	140	140	131	140	140	691	144
148	Area	2	No	ghcn cams grid	179.83	-179.83	-0.3796	-25.09	0.0355	-72.9132	-0.1885	181.60	131	131	94	69	128	553	127
149	Area	2	No	gpi	207.55	-207.55	0.2335	-99.87	0.0495	-101.1752	0.2224	209.07	147	147	13	21	147	475	103
150	Area	2	No	trmm 3a46	230.12	-230.12	-0.9998	80.85	0.9722	-133.6987	-0.9860	233.55	156	156	151	155	156	774	156
151	Area	2	No	trmm 3b43	193.24	-193.24	-0.9358	28.54	0.4984	-107.4307	-0.7060	195.64	136	136	148	144	136	700	148
152	Area	2	No	ssmi	207.66	-207.66	0.0127	-67.19	0.0001	-87.2833	0.0112	209.77	148	148	52	52	148	548	124
153	Area	2	No	opi2	145.77	-145.77	-0.2243	-50.15	0.0373	-48.6970	-0.1931	148.91	85	85	72	72	84	398	87
154	Area	2	No	opi3	152.25	-152.25	-0.5562	-20.62	0.1490	-52.9719	-0.3861	155.18	97	97	118	113	92	517	114
155	Area	2	No	cmmap	166.45	-166.45	-0.5418	-14.15	0.1489	-63.1562	-0.3859	169.19	113	113	116	112	111	565	130
156	Area	2	No	cmmap2	170.08	-170.08	-0.5139	-15.04	0.1436	-65.9340	-0.3790	172.81	122	122	109	110	120	583	133

Rondonia - Monthly Atmospheric																				
OI	Moist Flux	Re	Storage	Precipitation	Absolute Error	Bias	Slope	Intercept	R <sup>2</sup>	R <sup>2</sup> 1:1	Corr Coef	RMSE	AER	BR	SR	CCR	RMR	TR	BR	
1	Method 1	1	Yes	reanalysis	60.95	49.77	0.6329	-45.96	0.1270	-0.5538	0.3564	77.66	2	10	4	3	2	21	2	
2	Method 1	1	Yes	gpcp	101.11	65.92	-0.1258	-25.91	0.0251	-3.0062	-0.1585	130.66	19	17	43	43	21	143	31	
3	Method 1	1	Yes	gpcp	104.81	78.61	-0.2134	-41.01	0.0919	-5.0020	-0.3031	132.17	23	21	48	50	22	164	38	
4	Method 1	1	Yes	gpcp trmm	89.15	78.28	0.1266	-40.19	0.0297	-3.0804	0.1725	121.50	9	20	22	18	13	82	7	
5	Method 1	1	Yes	ghcn cams grid	112.68	82.73	-0.1995	-30.39	0.0781	-4.4450	-0.2758	145.02	25	22	44	49	28	168	40	
6	Method 1	1	Yes	gpi	158.42	143.95	-0.0425	-39.12	0.0064	-9.3782	-0.0799	196.89	45	42	36	36	48	207	49	
7	Method 1	1	Yes	trmm 3a46	124.83	114.26	0.0236	-38.51	0.0017	-6.4520	0.0406	163.73	34	31	31	30	35	161	37	
8	Method 1	1	Yes	trmm 3b43	90.98	77.45	0.1122	-39.42	0.0253	-3.2451	0.1591	122.89	10	19	23	22	14	88	8	
9	Method 1	1	Yes	ssmi	130.77	115.31	-0.0373	-39.92	0.0034	-6.0829	-0.0584	165.32	36	32	34	34	36	172	43	
10	Method 1	1	Yes	opi2	94.92	57.18	-0.1035	-27.04	0.0151	-2.4819	-0.1229	120.67	12	11	40	38	12	113	18	
11	Method 1	1	Yes	opi3	100.03	64.47	-0.1017	-26.34	0.0169	-2.9569	-0.1299	128.63	17	15	39	39	19	129	24	
12	Method 1	1	Yes	cmap	99.29	64.73	-0.1061	-28.52	0.0176	-2.8919	-0.1326	127.74	16	16	41	40	17	130	27	
13	Method 1	1	Yes	cmap2	98.18	62.76	-0.1096	-28.62	0.0186	-2.8282	-0.1364	128.69	15	14	42	41	15	127	23	
14	Method 1	2	Yes	reanalysis	119.09	116.90	0.6341	-111.59	0.2772	-1.4835	0.5720	141.48	29	35	3	2	25	94	10	
15	Method 1	2	Yes	gpcp	147.84	140.77	0.1548	-108.32	0.0229	-3.0988	0.1513	182.50	42	41	21	23	41	188	41	
16	Method 1	2	Yes	gpcp	163.80	163.35	0.0243	-128.81	0.0008	-6.3792	0.0276	195.52	48	48	30	32	47	205	48	
17	Method 1	2	Yes	gpcp trmm	162.74	161.77	0.0707	-120.89	0.0110	-8.7065	0.1050	193.05	47	47	26	26	45	191	46	
18	Method 1	2	Yes	ghcn cams grid	150.34	144.04	0.1588	-108.99	0.0254	-3.2401	0.1593	185.82	43	43	19	21	44	170	42	
19	Method 1	2	Yes	gpi	231.69	230.90	0.0582	-132.37	0.0094	-11.8996	0.0967	268.46	52	52	29	28	52	213	50	
20	Method 1	2	Yes	trmm 3a46	198.13	195.40	-0.0381	-117.47	0.0046	-13.0592	-0.0679	235.91	50	50	35	35	50	220	52	
21	Method 1	2	Yes	trmm 3b43	162.40	161.03	0.0650	-120.59	0.0096	-8.7324	0.0980	193.31	46	46	28	27	46	193	47	
22	Method 1	2	Yes	ssmi	209.76	208.89	0.0180	-135.03	0.0007	-10.9047	0.0267	243.35	51	51	33	33	51	219	51	
23	Method 1	2	Yes	opi2	140.12	133.08	0.2004	-108.53	0.0338	-2.6840	0.1838	173.02	37	37	11	16	37	138	30	
24	Method 1	2	Yes	opi3	146.66	139.77	0.1755	-108.94	0.0293	-3.0223	0.1713	180.79	40	40	18	19	40	157	35	
25	Method 1	2	Yes	cmap	145.64	139.06	0.1827	-109.08	0.0306	-2.9616	0.1750	179.42	39	39	15	17	39	149	32	
26	Method 1	2	Yes	cmap2	144.30	137.12	0.1783	-108.57	0.0289	-2.8998	0.1701	178.01	38	38	17	20	38	151	33	
27	Method 2	1	Yes	reanalysis	54.38	18.55	0.6653	-17.84	0.1120	0.0110	0.3346	68.47	7	2	2	4	7	10	1	Plot
28	Method 2	1	Yes	gpcp	93.71	33.99	-0.2113	6.84	0.0701	-2.5180	-0.2648	119.85	11	9	47	48	11	126	22	
29	Method 2	1	Yes	gpcp	96.27	32.64	-0.2988	8.82	0.1525	-3.0367	-0.3905	118.38	14	8	51	52	10	135	29	
30	Method 2	1	Yes	gpcp trmm	76.23	32.37	0.0870	0.15	0.0135	-1.7892	0.1161	95.83	3	6	25	25	3	62	4	Plot
31	Method 2	1	Yes	ghcn cams grid	102.39	57.55	-0.3459	-2.03	0.1478	-2.7875	-0.3844	134.05	21	12	52	51	23	159	36	
32	Method 2	1	Yes	gpi	128.30	97.98	-0.0831	8.37	0.0206	-6.0232	-0.1436	162.90	35	26	38	42	34	175	44	
33	Method 2	1	Yes	trmm 3a46	100.13	66.11	0.0188	-3.48	0.0009	-3.7816	0.0302	127.57	18	18	32	31	16	115	19	
34	Method 2	1	Yes	trmm 3b43	77.33	32.49	0.0691	1.08	0.0092	-1.9905	0.0961	98.59	4	7	27	29	4	71	5	
35	Method 2	1	Yes	ssmi	104.24	62.17	-0.0584	4.26	0.0073	-3.3924	-0.0854	130.00	22	13	37	37	20	129	25	
36	Method 2	1	Yes	opi2	80.61	13.93	-0.2280	4.36	0.0487	-1.4131	-0.2207	98.89	5	1	50	44	5	105	13	
37	Method 2	1	Yes	opi3	82.61	19.98	-0.2258	5.69	0.0569	-1.7188	-0.2386	104.97	6	3	49	47	6	111	16	
38	Method 2	1	Yes	cmap	85.81	29.20	-0.2083	6.60	0.0547	-1.9946	-0.2338	110.00	7	5	45	45	7	109	15	
39	Method 2	1	Yes	cmap2	85.86	28.52	-0.2104	6.52	0.0582	-2.0056	-0.2371	110.20	8	4	46	46	8	112	17	
40	Method 2	2	Yes	reanalysis	96.03	93.50	0.7111	-89.44	0.4457	-0.8331	0.6676	115.31	13	24	7	1	9	48	3	Plot
41	Method 2	2	Yes	gpcp	118.83	109.84	0.2758	-87.81	0.0793	-2.1313	0.2815	150.71	28	29	10	11	30	108	14	
42	Method 2	2	Yes	gpcp	120.69	117.48	0.1791	-92.88	0.0464	-3.8131	0.2154	151.75	31	36	16	15	31	129	26	
43	Method 2	2	Yes	gpcp trmm	120.32	116.60	0.1840	-89.86	0.0654	-4.5874	0.2557	150.21	30	34	14	13	29	120	21	
44	Method 2	2	Yes	ghcn cams grid	121.31	114.23	0.2918	-89.58	0.0813	-2.1963	0.2851	152.27	33	30	9	10	33	115	20	
45	Method 2	2	Yes	gpi	185.20	184.17	0.1982	-106.75	0.0998	-8.0445	0.3159	217.04	49	49	13	6	49	166	39	
46	Method 2	2	Yes	trmm 3a46	147.70	144.82	0.0901	-93.33	0.0206	-7.0757	0.1435	184.14	41	44	24	24	42	175	45	
47	Method 2	2	Yes	trmm 3b43	120.85	116.53	0.1574	-88.98	0.0492	-4.7227	0.2217	152.02	32	33	20	14	32	131	28	
48	Method 2	2	Yes	ssmi	153.36	152.40	0.1994	-104.49	0.0773	-5.7580	0.2780	184.92	44	45	12	12	43	156	34	
49	Method 2	2	Yes	opi2	102.31	91.64	0.4151	-84.50	0.1118	-1.2879	0.3344	128.26	20	23	5	5	18	71	8	
50	Method 2	2	Yes	opi3	107.12	97.21	0.3567	-85.77	0.0960	-1.5189	0.3099	135.17	24	25	6	9	24	88	9	
51	Method 2	2	Yes	cmap	113.61	105.36	0.3447	-88.36	0.0997	-1.7910	0.3157	142.28	27	28	7	8	27	97	12	
52	Method 2	2	Yes	cmap2	113.16	104.72	0.3434	-88.11	0.0997	-1.7768	0.3158	141.92	26	27	8	7	26	94	11	

Rondonia - Yearly Atmospheric																					
Ol	Moist Flux	Re	Storage	Precipitation	Absolute Error	Bias	Slope	Oslope	Intercept	R <sup>2</sup>	R <sup>2</sup> 1.1	Corr Coef	RMSE	AER	BR	SR	CCR	RMR	TR	BR	
1	Method 1	1	Yes	reanalysis	49.88	49.88	1.2411	1.2411	-52.39	0.5639	-2.0344	0.7509	54.13	10	10	3	4	10	37	3	
2	Method 1	1	Yes	gpcp	67.21	65.98	-0.1089		-26.56	0.0024	-4.6903	-0.0486	74.63	17	17	38	41	17	130	28	
3	Method 1	1	Yes	gpc	78.72	78.72	0.7808	1.2807	-71.92	0.1843	-15.5971	0.4293	80.76	20	20	4	12	20	78	8	
4	Method 1	1	Yes	gpc trmm	81.58	81.58	1.4889	1.4889	-104.59	0.5064	-6.0842	0.7116	84.94	21	21	9	7	21	79	9	
5	Method 1	1	Yes	ghcn cams grid	83.55	83.25	0.3535	2.8288	-55.01	0.0173	-7.0881	0.1313	89.18	22	22	19	20	22	105	20	
6	Method 1	1	Yes	gpi	144.08	144.08	0.2288	4.3706	-66.53	0.0244	-39.5724	0.1562	146.36	43	43	23	18	38	185	37	
7	Method 1	1	Yes	trmm 3a46	112.04	112.04	4.9325	4.9325	-409.52	0.3440	-9.2158	0.5865	117.17	30	30	24	10	30	124	24	
8	Method 1	1	Yes	trmm 3b43	77.72	77.72	1.3791	1.3791	-93.96	0.5105	-6.9373	0.7145	80.44	19	19	6	6	19	69	6	
9	Method 1	1	Yes	ssmi	116.37	116.37	0.5891	1.8975	-86.39	0.0652	-23.4986	0.2553	118.74	32	32	13	16	31	124	25	
10	Method 1	1	Yes	opi2	60.26	58.29	0.1420	7.0423	-34.39	0.0047	-3.6369	0.0683	67.37	12	11	28	26	11	88	10	
11	Method 1	1	Yes	opi3	66.48	64.98	0.1506	6.6401	-35.64	0.0055	-4.4852	0.0744	73.27	16	16	27	24	16	99	18	
12	Method 1	1	Yes	cmap	65.70	64.26	0.1531	6.5317	-35.61	0.0051	-4.3708	0.0714	72.50	15	15	26	25	15	96	16	
13	Method 1	1	Yes	cmap2	64.17	62.33	0.1048	9.5420	-33.78	0.0024	-4.1436	0.0491	70.95	14	13	31	29	14	101	19	
14	Method 1	2	Yes	reanalysis	116.89	116.89	1.7120	1.7120	-127.23	0.7470	-4.1004	0.8643	121.53	35	35	14	1	34	119	23	
15	Method 1	2	Yes	gpcp	140.76	140.76	-0.7418		-73.89	0.0273	-6.9649	-0.1651	151.87	41	41	49	48	42	221	48	
16	Method 1	2	Yes	gpc	163.29	163.29	0.5999	1.6669	-149.13	0.0150	-14.2673	0.1225	168.87	48	48	12	21	48	177	39	
17	Method 1	2	Yes	gpc trmm	162.13	162.13	1.7122	1.7122	-183.47	0.4905	-21.2962	0.7004	164.34	47	47	15	8	47	164	38	
18	Method 1	2	Yes	ghcn cams grid	144.03	144.03	0.3533	2.8305	-117.09	0.0035	-7.1714	0.0588	153.83	42	42	20	28	43	175	36	
19	Method 1	2	Yes	gpi	230.92	230.92	-0.0686		-119.01	0.0007	-33.1501	-0.0262	234.96	52	52	34	35	52	225	52	
20	Method 1	2	Yes	trmm 3a46	196.65	196.65	-1.7488		2.63	0.0828	-26.4286	-0.2878	200.80	50	50	50	50	50	250	49	
21	Method 1	2	Yes	trmm 3b43	161.39	161.39	1.3955	1.3955	-178.50	0.3300	-21.2003	0.5745	163.98	46	46	7	11	46	156	35	
22	Method 1	2	Yes	ssmi	209.23	209.23	-0.1179		-125.18	0.0010	-37.8513	-0.0322	212.24	51	51	40	36	51	229	50	
23	Method 1	2	Yes	opi2	133.07	133.07	-0.1339		-98.26	0.0012	-6.1968	-0.0340	144.37	37	37	41	37	37	189	42	
24	Method 1	2	Yes	opi3	139.77	139.77	-0.1901		-95.26	0.0021	-6.8254	-0.0457	150.54	40	40	44	39	41	204	47	
25	Method 1	2	Yes	cmap	139.05	139.05	-0.0841		-99.29	0.0004	-6.7389	-0.0194	149.70	39	39	36	34	40	188	41	
26	Method 1	2	Yes	cmap2	137.25	137.12	-0.1566		-96.93	0.0013	-6.5632	-0.0364	147.99	38	38	43	38	39	196	45	
27	Method 2	1	Yes	reanalysis	21.42	18.67	1.3662	1.3662	-19.39	0.5186	0.1477	0.7202	20.65	2	2	5	5	2	16	1	Plot
28	Method 2	1	Yes	gpcp	37.79	34.07	-0.2200		7.06	0.0176	-3.2265	-0.1328	42.81	7	8	45	45	8	113	21	
29	Method 2	1	Yes	gpc	38.15	32.80	-0.0829		1.77	0.0023	-2.9358	-0.0484	40.83	8	7	35	40	7	97	17	
30	Method 2	1	Yes	gpc trmm	41.31	41.31	1.7286	1.7286	-70.70	0.6642	-1.9098	0.8150	44.96	9	9	16	3	9	48	5	
31	Method 2	1	Yes	ghcn cams grid	59.48	58.57	0.1191	8.3963	-22.15	0.0015	-3.4283	0.0393	67.39	11	12	30	31	12	96	15	
32	Method 2	1	Yes	gpi	98.14	98.14	0.0212	47.1698	-2.04	0.0002	-21.1965	0.0142	101.48	26	26	32	32	25	141	32	
33	Method 2	1	Yes	trmm 3a46	70.24	70.24	-3.5400		218.86	0.2880	-6.3758	-0.5366	76.67	18	18	51	51	18	156	34	
34	Method 2	1	Yes	trmm 3b43	33.23	32.78	0.6147	1.6284	-18.85	0.1261	-1.6413	0.3557	40.83	5	6	10	14	6	41	4	Plot
35	Method 2	1	Yes	ssmi	63.84	63.84	-0.2349		13.11	0.0177	-12.7146	-0.1331	67.57	13	14	46	46	13	132	30	
36	Method 2	1	Yes	opi2	20.86	15.87	0.5508	1.8155	-8.90	0.1011	-0.5523	0.3180	25.82	1	1	17	15	1	35	2	
37	Method 2	1	Yes	opi3	27.34	21.44	-0.1525		2.86	0.0074	-1.4826	-0.0858	32.66	3	3	42	44	3	95	12	
38	Method 2	1	Yes	cmap	33.45	29.59	-0.0923		2.34	0.0028	-2.4250	-0.0528	38.38	6	5	37	42	5	95	13	
39	Method 2	1	Yes	cmap2	32.83	28.95	-0.1125		2.88	0.0042	-2.3546	-0.0645	37.96	4	4	39	43	4	94	11	
40	Method 2	2	Yes	reanalysis	93.50	93.50	1.2293	1.2293	-98.73	0.6740	-5.8607	0.8210	95.98	24	24	2	2	23	75	7	Plot
41	Method 2	2	Yes	gpcp	109.83	109.83	-0.4150		-66.81	0.0187	-9.1832	-0.1366	116.93	29	29	47	47	29	181	40	
42	Method 2	2	Yes	gpc	117.48	117.48	-0.5731		-70.34	0.0510	-20.3989	-0.2258	121.32	36	36	48	49	33	202	48	
43	Method 2	2	Yes	gpc trmm	116.87	116.87	2.0084	2.0084	-149.91	0.4027	-7.6263	0.6346	121.98	34	34	18	9	35	130	29	
44	Method 2	2	Yes	ghcn cams grid	114.22	114.22	0.6012	1.6633	-100.35	0.0190	-9.7072	0.1379	119.90	31	31	11	19	32	124	26	
45	Method 2	2	Yes	gpi	184.20	184.20	0.2440	4.0984	-111.20	0.0130	-36.7129	0.1140	186.98	49	49	22	22	49	191	44	
46	Method 2	2	Yes	trmm 3a46	152.96	152.96	-4.7880		167.60	0.7551	-14.9680	-0.8690	159.86	45	45	52	52	45	239	51	
47	Method 2	2	Yes	trmm 3b43	116.80	116.80	1.0786	1.0786	-119.37	0.1339	-7.6854	0.3660	123.11	33	33	1	13	36	116	22	
48	Method 2	2	Yes	ssmi	152.67	152.67	0.1414	7.0721	-101.91	0.0023	-28.5187	0.0480	155.55	44	44	29	30	44	191	43	
49	Method 2	2	Yes	opi2	91.64	91.64	0.6965	1.4358	-87.93	0.0489	-6.2164	0.2166	98.43	23	23	8	17	24	95	14	
50	Method 2	2	Yes	opi3	97.21	97.21	0.0032	312.5000	-79.48	0.0000	-7.1223	0.0009	104.43	25	25	33	33	26	142	33	
51	Method 2	2	Yes	cmap	105.35	105.35	0.1948	5.1335	-84.48	0.0036	-8.3250	0.0600	111.89	28	28	25	27	28	136	31	
52	Method 2	2	Yes	cmap2	104.72	104.72	0.2602	3.8432	-86.01	0.0065	-8.2143	0.0807	111.23	27	27	21	23	27	125	27	

Rondonia - Monthly Surface																										
OI	Runoff Meth	Re	Storage	Precipitation	Absolute Error	Bias	Abias	Slope	Oslope	Intercept	R <sup>2</sup>	R <sup>2</sup> 1.1	Corr Coef	RMSE	AER	BR	SR	CCR	RMR	TR	BR	BR	BR	BR	BR	Plot
1	Linear	1	Yes	reanalysis	60.22	7.01	7.01	1.0155	1.0155	6.98	0.2095	0.2007	0.4566	68.88	152	55	2	151	95	455	57	455	57	455	57	Plot
2	Linear	1	Yes	gpcp	62.07	-20.36	20.36	0.5554	1.8005	-5.37	0.3209	0.0481	0.5665	76.62	171	203	36	72	164	646	103	646	103	646	103	
3	Linear	1	Yes	gpcp	63.13	-23.22	23.22	0.5688	1.7581	-9.46	0.3358	0.0484	0.5795	73.75	187	228	33	65	137	650	105	650	105	650	105	
4	Linear	1	Yes	gpcp trmm	49.37	-7.82	7.82	0.8596	1.1833	-2.82	0.5646	0.5416	0.7514	59.38	48	70	8	12	42	180	3	180	3	180	3	
5	Linear	1	Yes	ghcn cams grid	65.74	-32.41	32.41	0.5882	1.7001	-15.42	0.3419	-0.0028	0.5847	77.11	214	275	30	62	170	751	123	751	123	751	123	
6	Linear	1	Yes	gpi	90.86	-83.98	83.98	0.5219	1.9161	-37.04	0.4798	-1.0239	0.6926	113.86	389	443	41	30	391	1294	279	391	279	1294	279	
7	Linear	1	Yes	trmm 3a46	40.84	-34.03	34.03	0.8633	1.1583	-25.30	0.4156	0.6514	0.8031	53.00	17	289	7	1	20	334	7	334	7	334	7	
8	Linear	1	Yes	trmm 3b43	55.22	-8.50	8.50	0.7572	1.3207	0.16	0.4760	0.4173	0.6899	66.42	102	82	12	31	76	303	26	303	26	303	26	
9	Linear	1	Yes	ssmi	60.47	-47.82	47.82	0.6751	1.4813	-27.43	0.5675	0.0899	0.7534	77.54	153	380	19	11	176	739	121	739	121	739	121	
10	Linear	1	Yes	opi2	54.25	-2.89	2.89	0.7557	1.3233	0.75	0.3515	0.3134	0.5929	65.11	86	24	13	60	68	251	14	251	14	251	14	
11	Linear	1	Yes	opi3	57.93	-8.95	8.95	0.8517	1.5344	-1.65	0.3111	0.2093	0.5578	69.87	131	87	22	76	107	423	50	423	50	423	50	
12	Linear	1	Yes	cmcp	60.99	-17.09	17.09	0.6039	1.6559	-5.36	0.3007	0.1241	0.5484	73.51	159	164	27	79	136	565	85	565	85	565	85	
13	Linear	1	Yes	cmcp2	61.01	-16.41	16.41	0.5962	1.6773	-4.72	0.2957	0.1164	0.5438	73.83	160	157	28	82	140	567	87	567	87	567	87	
14	Linear	1	No	reanalysis	25.39	7.14	7.14	1.1952	5.1230	8.86	0.1928	-3.2976	0.4391	31.92	5	57	147	162	2	373	38	373	38	373	38	
15	Linear	1	No	gpcp	62.77	-20.51	20.51	0.0998	10.0200	9.84	0.1464	-12.7389	0.3827	77.41	181	204	270	200	174	1029	204	1029	204	1029	204	
16	Linear	1	No	gpcp	64.30	-23.36	23.36	0.0611	16.3666	6.61	0.2836	-73.6704	0.5326	76.37	199	229	319	96	162	1005	198	1005	198	1005	198	
17	Linear	1	No	gpcp trmm	54.60	-7.34	7.34	0.2496	4.0064	19.39	0.2608	-2.1360	0.5107	66.34	94	63	101	110	75	443	56	443	56	443	56	
18	Linear	1	No	ghcn cams grid	64.03	-31.94	31.94	0.0798	12.5313	6.02	0.1571	-25.0015	0.3963	78.63	196	270	297	195	189	1147	235	1147	235	1147	235	
19	Linear	1	No	gpi	107.67	-83.71	83.71	0.0625	16.0000	8.34	0.0794	-30.3864	0.2817	132.04	423	441	318	250	429	1861	391	1861	391	1861	391	
20	Linear	1	No	trmm 3a46	78.56	-27.67	27.67	0.1656	6.0386	25.60	0.1484	-4.0889	0.3853	91.06	316	241	174	199	277	1207	256	1207	256	1207	256	
21	Linear	1	No	trmm 3b43	56.87	-8.12	8.12	0.2406	4.1563	19.26	0.2623	-2.3978	0.5122	68.65	115	74	104	108	93	494	67	494	67	494	67	
22	Linear	1	No	ssmi	84.92	-47.44	47.44	0.0648	15.4321	11.25	0.0525	-14.3237	0.2291	100.34	366	375	314	292	349	1696	356	1696	356	1696	356	
23	Linear	1	No	opi2	47.34	-1.74	1.74	0.1573	6.3753	10.80	0.2266	-6.1177	0.4697	55.08	34	14	180	147	27	402	46	402	46	402	46	
24	Linear	1	No	opi3	50.96	-7.80	7.80	0.1298	7.7042	10.43	0.1787	-8.0008	0.4227	61.94	61	69	220	179	54	583	89	583	89	583	89	
25	Linear	1	No	cmcp	54.52	-16.39	16.39	0.1155	8.6580	9.82	0.1584	-9.7676	0.3981	67.89	89	156	240	193	87	765	129	765	129	765	129	
26	Linear	1	No	cmcp2	54.63	-15.71	15.71	0.1138	8.7873	9.94	0.1553	-9.8358	0.3941	68.10	95	146	245	197	89	772	132	772	132	772	132	
27	Linear	1	mon	reanalysis	24.55	7.22	7.22	1.065	9.3897	9.04	0.0573	-4.1895	0.2393	35.10	1	60	258	283	5	607	92	607	92	607	92	
28	Linear	1	mon	gpcp	59.01	-20.01	20.01	0.1346	7.4294	8.74	0.2640	-11.5565	0.5138	74.13	142	201	211	107	141	802	141	802	141	802	141	
29	Linear	1	mon	gpcp	62.47	-22.95	22.95	0.0805	12.4224	5.93	0.4980	-71.2023	0.7057	74.77	176	225	296	27	150	874	160	874	160	874	160	
30	Linear	1	mon	gpcp trmm	47.38	-7.73	7.73	0.3464	2.8888	15.84	0.5018	-1.3261	0.7082	57.62	35	67	64	25	32	223	11	223	11	223	11	
31	Linear	1	mon	ghcn cams grid	61.69	-32.03	32.03	0.1068	9.3809	4.89	0.2798	-23.7051	0.5290	76.68	170	271	257	101	165	964	184	964	184	964	184	
32	Linear	1	mon	gpi	105.22	-83.27	83.27	0.0976	10.2459	4.90	0.1927	-28.7093	0.4390	128.76	414	438	276	164	420	1712	361	1712	361	1712	361	
33	Linear	1	mon	trmm 3a46	64.75	-29.78	29.78	0.3059	3.2890	16.19	0.4938	-2.5800	0.7027	77.21	208	258	78	28	171	743	122	743	122	743	122	
34	Linear	1	mon	trmm 3b43	48.10	-6.81	6.81	0.3444	2.9036	16.07	0.5267	-1.4124	0.7258	58.33	29	52	65	20	36	202	7	202	7	202	7	
35	Linear	1	mon	ssmi	79.37	-46.87	46.87	0.1198	8.3472	7.82	0.1784	-12.7730	0.4224	95.39	324	366	231	180	314	1415	310	1415	310	1415	310	
36	Linear	1	mon	opi2	44.47	-1.91	1.91	0.1947	5.1361	10.23	0.3370	-5.4395	0.5806	52.48	23	17	149	63	19	271	19	271	19	271	19	
37	Linear	1	mon	opi3	47.66	-7.99	7.99	0.1668	5.9952	9.63	0.2944	-7.2034	0.5426	59.23	37	72	172	83	41	405	47	405	47	405	47	
38	Linear	1	mon	cmcp	51.85	-16.70	16.70	0.1474	6.7843	8.84	0.2569	-8.9814	0.5068	65.48	67	160	189	115	70	601	91	601	91	601	91	
39	Linear	1	mon	cmcp2	51.91	-16.02	16.02	0.1480	6.8493	8.98	0.2543	-9.0362	0.5042	65.66	69	149	192	118	71	599	90	599	90	599	90	
40	Linear	1	mon	reanalysis	25.36	7.23	7.23	0.0142	70.4225	9.26	0.0010	-5.1288	0.0318	38.16	4	61	363	370	7	805	143	805	143	805	143	
41	Linear	1	mon	gpcp	59.93	-19.79	19.79	0.1284	7.7882	8.99	0.2397	-11.6949	0.4896	74.67	148	198	223	131	148	848	151	848	151	848	151	
42	Linear	1	mon	gpcp	62.93	-22.33	22.33	0.0750	13.3333	6.04	0.4366	-72.5886	0.6607	74.87	183	218	299	43	151	894	168	894	168	894	168	
43	Linear	1	mon	gpcp trmm	48.18	-8.40	8.40	0.3366	2.9709	15.90	0.4725	-1.4108	0.6874	59.16	40	78	68	32	39	257	15	257	15	257	15	
44	Linear	1	mon	ghcn cams grid	62.50	-32.14	32.14	0.0980	10.2041	5.22	0.2365	-24.1393	0.4863	77.39	177	272	274	133	173	1029	303	1029	303	1029	303	
45	Linear	1	mon	gpi	105.49	-83.37	83.37	0.0980	10.2041	4.81	0.1946	-28.6831	0.4412	128.97	416	440	275	181	421	1713	362	1713	362	1713	362	
46	Linear	1	mon	trmm 3a46	63.28	-32.91	32.91	0.3565	2.8050	11.76	0.6388	-2.0777	0.7992	72.45	189	278	60	6	129	662	108	662	108	662	108	
47	Linear	1	mon	trmm 3b43	49.87	-7.19	7.19	0.3237	3.0893	16.52	0.4645	-1.5														

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OL	Runoff Meth	Re	Storage	Precipitation	Absolute Error	Bias	Abias	Slope	Oslope	Intercept	R <sup>2</sup>	R <sup>2</sup> 1.1	Corr Coef	RMSE	AER	BR	SR	CCR	RMR	TR	BR	BR	BR	BR	BR	BR								
102	Linear	2	No	opi3	77.71	-21.61	21.61	-0.1293		-1.53	0.0467	-3.7488	-0.2160	96.53	309	212	438	421	323	1703	357													
103	Linear	2	No	cmcp2	81.80	-29.76	29.76	-0.1218		-0.67	0.0460	-4.3097	-0.2145	102.07	342	257	432	420	358	1809	383													
104	Linear	2	No	cmcp2	81.67	-29.12	29.12	-0.1182		-0.84	0.0437	-4.2980	-0.2091	101.95	341	247	428	416	356	1788	378													
105	Linear	2	1 mon	reanalysis	76.24	-17.58	17.58	-0.0189		-3.58	0.0012	-3.5225	-0.0341	94.37	297	170	392	391	303	1553	335													
106	Linear	2	1 mon	gpcp	84.87	-33.71	33.71	-0.0236		-3.14	0.0021	-4.5732	-0.0460	104.76	365	286	394	395	363	1803	382													
107	Linear	2	1 mon	gpcp	81.96	-32.37	32.37	-0.0666		-0.91	0.0177	-5.1269	-0.1331	103.23	343	274	413	406	361	1797	380													
108	Linear	2	1 mon	gpcp trmm	87.98	-41.87	41.87	-0.0817		-10.24	0.0678	-14.2361	-0.2603	105.13	378	335	416	432	365	1926	410													
109	Linear	2	1 mon	ghcn cams grid	83.04	-38.37	38.37	-0.0119		-3.43	0.0005	-4.3516	-0.0223	102.65	353	311	388	386	360	1798	381													
110	Linear	2	1 mon	gpi	135.84	-103.01	103.01	-0.0631		-1.23	0.0330	-16.0211	-0.1816	164.37	463	459	408	413	463	2206	459													
111	Linear	2	1 mon	trmm 3a46	117.12	-70.66	70.66	-0.1150		-4.53	0.1841	-23.6755	-0.4290	137.05	445	411	424	451	441	2172	452													
112	Linear	2	1 mon	trmm 3b43	88.99	-42.19	42.19	-0.0805		-10.25	0.0689	-14.8000	-0.2624	107.05	382	339	415	433	373	1942	416													
113	Linear	2	1 mon	ssmi	115.00	-72.20	72.20	-0.1155		-6.84	0.1039	-13.8049	-0.3224	135.41	439	415	425	440	437	2156	451													
114	Linear	2	1 mon	opi2	69.99	-15.71	15.71	-0.0386		-3.39	0.0035	-2.6924	-0.0595	85.27	252	147	399	398	242	1438	316													
115	Linear	2	1 mon	opi3	72.42	-21.16	21.16	-0.0092		-3.68	0.0002	-3.0366	-0.0153	89.15	272	205	384	383	264	1508	329													
116	Linear	2	1 mon	cmcp2	76.34	-29.29	29.29	-0.0107		-3.57	0.0004	-3.5717	-0.0188	94.88	298	248	387	384	308	1625	348													
117	Linear	2	1 mon	cmcp2	76.13	-28.66	28.66	-0.0058		-3.70	0.0001	-3.5482	-0.0103	94.63	296	242	381	381	306	1606	345													
118	Linear	2	2 mon	reanalysis	72.02	-17.69	17.69	0.0322	31.0559	-4.55	0.0034	-3.2225	0.0582	90.88	269	174	344	359	276	1422	313													
119	Linear	2	2 mon	gpcp	79.61	-33.77	33.77	0.0655	15.2672	-6.06	0.0165	-3.9285	0.1285	98.18	328	287	312	333	332	1592	341													
120	Linear	2	2 mon	gpcp	75.27	-32.48	32.48	0.0635	15.7480	-5.47	0.0167	-4.2441	0.1292	93.77	289	276	316	332	301	1514	332													
121	Linear	2	2 mon	gpcp trmm	83.39	-39.55	39.55	-0.0065		-11.65	0.0004	-12.7706	-0.0210	98.96	358	328	382	385	389	1792	379													
122	Linear	2	2 mon	ghcn cams grid	78.51	-38.41	38.41	0.0726	13.7741	-6.61	0.0187	-3.7879	0.1367	96.77	315	312	303	327	325	1582	339													
123	Linear	2	2 mon	gpi	132.00	-103.21	103.21	-0.0084		-7.01	0.0008	-15.8467	-0.0245	160.06	460	460	383	387	461	2151	449													
124	Linear	2	2 mon	trmm 3a46	115.09	-72.98	72.98	-0.0683		-6.28	0.0615	-23.0154	-0.2481	133.64	440	417	411	429	434	2131	448													
125	Linear	2	2 mon	trmm 3b43	85.04	-39.75	39.75	-0.0143		-11.43	0.0222	-13.4390	-0.0473	101.33	367	329	390	396	353	1835	387													
126	Linear	2	2 mon	ssmi	111.22	-72.90	72.90	-0.0618		-10.10	0.0299	-13.0202	-0.1728	131.60	431	414	407	412	426	2090	440													
127	Linear	2	2 mon	opi2	65.58	-15.91	15.91	0.0548	18.2482	-4.76	0.0072	-2.2728	0.0850	80.01	213	148	330	344	202	1237	289													
128	Linear	2	2 mon	opi3	67.10	-21.24	21.24	0.0879	11.3766	-5.62	0.0215	-2.5235	0.1465	83.02	223	208	285	325	217	1258	274													
129	Linear	2	2 mon	cmcp2	71.41	-29.36	29.36	0.0806	12.4069	-8.15	0.0201	-3.0379	0.1419	88.87	262	252	295	326	261	1396	303													
130	Linear	2	2 mon	cmcp2	71.17	-28.74	28.74	0.0888	11.5207	-6.25	0.0235	-3.0017	0.1533	88.47	260	243	288	323	258	1372	293													
131	Linear	2	3 mon	reanalysis	67.87	-17.80	17.80	0.1074	9.3110	-5.74	0.0379	-2.7414	0.1947	85.53	233	175	254	280	243	1213	280													
132	Linear	2	3 mon	gpcp	76.89	-33.99	33.99	0.1248	8.0128	-7.99	0.0602	-3.4908	0.2453	93.71	306	288	226	279	300	1399	304													
133	Linear	2	3 mon	gpcp	73.08	-33.10	33.10	0.1563	6.3980	-8.84	0.1052	-3.6366	0.3243	86.77	273	281	182	224	250	1210	257													
134	Linear	2	3 mon	gpcp trmm	79.81	-39.02	39.02	0.0473	21.1416	-12.31	0.0245	-12.0878	0.1564	95.29	330	320	316	326	320	1616	347													
135	Linear	2	3 mon	ghcn cams grid	76.53	-38.68	38.68	0.1360	7.3529	-8.96	0.0659	-3.3577	0.2567	92.31	301	315	208	269	290	1383	300													
136	Linear	2	3 mon	gpi	128.70	-103.67	103.67	0.0548	18.3150	-13.52	0.0284	-15.0695	0.1624	155.15	459	461	311	319	458	2025	423													
137	Linear	2	3 mon	trmm 3a46	112.71	-74.35	74.35	0.0015	666.6667	-9.53	0.0000	-22.0727	0.0056	129.07	435	421	375	374	422	2027	426													
138	Linear	2	3 mon	trmm 3b43	79.78	-39.38	39.38	0.0578	17.3010	-12.62	0.0381	-12.3265	0.1953	96.15	329	326	324	307	320	1606	346													
139	Linear	2	3 mon	ssmi	106.68	-71.13	71.13	0.0259	38.6100	-14.52	0.0053	-11.6405	0.0730	124.68	418	412	348	350	412	1940	415													
140	Linear	2	3 mon	opi2	63.10	-16.22	16.22	0.1381	7.2411	-5.93	0.0460	-1.8818	0.2146	75.07	185	154	202	208	153	992	192													
141	Linear	2	3 mon	opi3	65.48	-21.52	21.52	0.1448	6.9061	-6.78	0.0585	-2.2193	0.2419	79.34	212	211	193	282	194	1092	221													
142	Linear	2	3 mon	cmcp2	69.80	-29.62	29.62	0.1385	7.2202	-7.79	0.0596	-2.6951	0.2442	85.00	250	255	201	281	240	1227	266													
143	Linear	2	3 mon	cmcp2	69.60	-28.99	28.99	0.1397	7.1582	-7.74	0.0612	-2.6889	0.2473	84.91	249	244	197	276	239	1205	254													
144	Linear	2	4 mon	reanalysis	64.45	-18.22	18.22	0.1399	7.1480	-6.55	0.0656	-2.5827	0.2560	83.07	202	185	196	270	220	1073	214													
145	Linear	2	4 mon	gpcp	74.85	-34.44	34.44	0.1415	7.0671	-8.86	0.0788	-3.4385	0.2808	92.46	283	294	195	252	291	1315	284													
146	Linear	2	4 mon	gpcp trmm	68.69	-34.43	34.43	0.2029	4.9285	-11.14	0.1876	-3.4788	0.4332	83.06	244	293	138	171	219	1065	212													
147	Linear	2	4 mon	gpcp trmm	73.43	-39.77	39.77	0.1003	9.9701	-13.90	0.1096	-10.9406	0.3310	92.04	274	330	268	282	286	1380	295													
148	Linear	2	4 mon	ghcn cams grid	75.00	-39.25	39.25	0.1591	6.2854	-10.15	0.0917	-3.2704	0.3027	90.69	286	323	178	236	274	1297	280													
149	Linear	2	4 mon	gpi	128.40	-104.44	104.44	0.1028	9.9278	-18.75	0.0983	-14.9770	0.3135	151.47	457	462	263	231	455	1888	396													
150	Linear	2	4 mon	trmm 3a46	109.48	-74.96	74.96	0.0628	15.9238	-13.56	0.0570	-20.2128	0.2387	125.34	428	422	317	284	413	1864	394													
151	Linear	2	4 mon	trmm 3b43	74.16	-40.11	40.11	0.1033	9.8805	-14.																								

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OL	Runoff Meth	Re	Storage	Precipitation	Absolute Error	Bias	Abias	Slope	Oslope	Intercept	R <sup>2</sup>	R <sup>2</sup> 1.1	Corr Coef	RMSE	AER	BR	SR	CCR	RMR	TR	BR
203	Point	1	2 mon	lrmm 3b43	46.42	7.95	7.95	0.4912	2.0358	25.79	0.4479	-0.0512	0.6692	59.99	31	71	46	40	45	233	13
204	Point	1	2 mon	ssmi	73.94	-38.74	38.74	0.2076	4.8170	10.62	0.2265	-4.0297	0.4759	88.88	277	316	133	142	262	1130	231
205	Point	1	2 mon	opi2	44.26	5.31	5.31	0.2771	3.8088	16.32	0.2850	-1.6820	0.5339	52.47	22	36	89	94	18	259	16
206	Point	1	2 mon	opi3	46.89	-0.87	0.87	0.2396	4.1736	15.41	0.2531	-2.2960	0.5031	58.16	33	94	107	121	34	303	24
207	Point	1	2 mon	cmnp	50.80	-9.40	9.40	0.2085	4.7962	14.42	0.2145	-2.9643	0.4631	63.92	59	94	132	148	61	494	66
208	Point	1	2 mon	cmnp2	50.91	-8.73	8.73	0.2067	4.8379	14.61	0.2128	-2.9946	0.4613	64.16	60	85	135	149	63	492	65
209	Point	1	3 mon	reanalysis	29.28	12.57	12.57	-0.1124	14.93	0.0270	-2.8938	-0.1643	0.4684	46.84	13	122	422	410	14	981	188
210	Point	1	3 mon	gpcc	62.75	-12.46	12.46	0.1365	7.3260	16.08	0.1133	-4.5712	0.3366	76.60	180	118	207	199	163	887	165
211	Point	1	3 mon	gpcc	63.92	-17.84	17.84	0.0741	13.4953	10.36	0.1823	-30.0238	0.4270	74.65	195	177	301	177	146	996	194
212	Point	1	3 mon	gpcc lrmm	57.08	6.14	6.14	0.3984	2.5100	28.68	0.2753	-0.3633	0.5247	69.35	116	46	56	102	103	423	49
213	Point	1	3 mon	ghcn cams grid	62.78	-26.97	26.97	0.0995	10.0503	10.47	0.1028	-9.6082	0.3207	77.38	182	238	271	229	172	1092	220
214	Point	1	3 mon	gpi	101.72	-76.00	76.00	0.1140	8.7719	10.99	0.1107	-10.9073	0.3327	126.04	409	426	244	220	415	1714	363
215	Point	1	3 mon	lrmm 3a46	58.52	-15.46	15.46	0.4583	2.1820	23.43	0.4266	-0.2259	0.6531	71.86	137	145	52	45	124	503	70
216	Point	1	3 mon	lrmm 3b43	61.63	7.50	7.50	0.3527	2.8353	30.51	0.2305	-0.5618	0.4801	73.76	167	64	61	138	139	569	88
217	Point	1	3 mon	ssmi	76.10	-38.97	38.97	0.1904	5.2521	11.59	0.1907	-4.2150	0.4367	90.75	295	319	151	166	275	1206	255
218	Point	1	3 mon	opi2	50.50	4.92	4.92	0.1809	5.5279	17.70	0.1202	-2.3690	0.3468	58.90	55	34	161	212	37	499	68
219	Point	1	3 mon	opi3	53.51	-1.19	1.19	0.1534	6.5189	17.19	0.1032	-3.0393	0.3212	64.50	80	9	184	228	65	566	86
220	Point	1	3 mon	cmnp	57.36	-9.44	9.44	0.1321	7.5700	16.70	0.0861	-3.7190	0.2935	69.86	120	96	216	240	106	778	133
221	Point	1	3 mon	cmnp2	57.47	-8.78	8.78	0.1302	7.6805	16.84	0.0844	-3.7570	0.2906	70.14	126	86	219	244	108	783	137
222	Point	1	4 mon	reanalysis	32.60	12.52	12.52	-0.1817	15.11	0.0705	-3.1886	-0.2655	0.4660	46.60	15	120	456	436	16	1041	209
223	Point	1	4 mon	gpcc	70.22	-12.52	12.52	0.0300	33.3333	19.56	0.0055	-5.8706	0.0739	85.20	255	121	346	349	241	1312	283
224	Point	1	4 mon	gpcc	68.66	-18.52	18.52	0.0121	82.6446	12.00	0.0051	-35.6480	0.0712	79.69	243	193	365	352	198	1531	290
225	Point	1	4 mon	gpcc lrmm	75.89	4.02	4.02	0.1329	7.5245	37.36	0.0310	-1.2937	0.1761	89.84	294	29	214	312	269	1118	227
226	Point	1	4 mon	ghcn cams grid	67.82	-27.20	27.20	0.0203	49.2611	13.69	0.0043	-11.2929	0.0655	83.18	231	240	354	354	221	1400	305
227	Point	1	4 mon	gpi	108.85	-76.40	76.40	0.0462	21.6450	17.53	0.0182	-12.0967	0.1347	132.40	426	426	337	328	430	1949	417
228	Point	1	4 mon	lrmm 3a46	82.28	-17.39	17.39	0.2356	4.2445	37.99	0.1143	-1.1593	0.3380	95.84	346	167	111	217	317	1158	239
229	Point	1	4 mon	lrmm 3b43	77.73	5.55	5.55	0.1173	8.5251	37.65	0.0259	-1.4484	0.1608	92.21	310	39	236	317	288	1190	247
230	Point	1	4 mon	ssmi	84.73	-39.26	39.26	0.1163	8.5985	16.10	0.0711	-5.0019	0.2667	97.64	364	324	237	260	328	1513	331
231	Point	1	4 mon	opi2	57.42	4.73	4.73	0.0448	22.4215	19.79	0.0073	-3.3645	0.0854	67.15	123	33	339	343	84	922	175
232	Point	1	4 mon	opi3	60.63	-1.38	1.38	0.0347	28.8184	19.73	0.0053	-4.0764	0.0727	72.42	155	11	343	351	128	988	191
233	Point	1	4 mon	cmnp	64.40	-9.33	9.33	0.0212	47.1698	20.00	0.0022	-4.8081	0.0471	77.63	200	93	352	366	178	1169	246
234	Point	1	4 mon	cmnp2	64.52	-8.69	8.69	0.0212	47.1698	20.01	0.0022	-4.8387	0.0473	77.83	203	83	353	365	180	1184	243
235	Point	2	Yes	reanalysis	51.88	-2.10	2.10	0.5081	1.9681	4.82	0.3130	0.0188	0.5594	71.94	68	19	45	75	125	332	29
236	Point	2	Yes	gpcc	58.38	-18.43	18.43	0.3988	2.5075	-0.15	0.2282	-0.3549	0.4777	84.53	136	191	55	141	233	756	125
237	Point	2	Yes	gpcc	57.78	-13.97	13.97	0.4166	2.4004	3.51	0.2248	-0.2527	0.4741	81.81	129	139	53	144	212	677	112
238	Point	2	Yes	gpcc lrmm	56.78	-40.14	40.14	0.3074	3.2531	-17.44	0.3524	-2.2058	0.5938	81.90	113	333	77	57	213	793	139
239	Point	2	Yes	ghcn cams grid	54.69	-22.83	22.83	0.4680	2.1368	-4.32	0.2877	-0.1827	0.5364	78.98	97	223	49	89	190	648	104
240	Point	2	Yes	gpi	103.01	-90.76	90.76	0.3006	3.3267	-23.22	0.2538	-2.8707	0.5037	134.98	413	451	81	119	436	1500	327
241	Point	2	Yes	lrmm 3a46	75.78	-62.67	62.67	0.3078	3.2489	-23.49	0.4601	-3.6576	0.6783	101.09	293	397	76	38	352	1156	238
242	Point	2	Yes	lrmm 3b43	57.34	-40.07	40.07	0.3030	3.3003	-17.28	0.3518	-2.2765	0.5931	82.80	119	331	80	88	215	803	142
243	Point	2	Yes	ssmi	79.47	-66.15	66.15	0.2826	3.5866	-23.22	0.2909	-3.2042	0.5394	106.54	326	403	87	87	372	1275	277
244	Point	2	Yes	opi2	50.52	-0.24	0.24	0.5081	1.9681	5.77	0.2305	0.0145	0.4801	72.09	56	3	44	139	126	368	36
245	Point	2	Yes	opi3	51.37	-5.81	5.81	0.4812	2.0781	3.42	0.2402	-0.0455	0.4901	74.25	64	42	47	130	142	425	51
246	Point	2	Yes	cmnp	52.96	-13.96	13.96	0.4609	2.1697	0.02	0.2452	-0.1272	0.4952	77.10	77	138	51	128	169	563	83
247	Point	2	Yes	cmnp2	52.65	-13.32	13.32	0.4621	2.1640	0.28	0.2483	-0.1219	0.4983	76.92	73	132	50	125	167	547	80
248	Point	2	No	reanalysis	80.93	-1.90	1.90	0.0483	20.7039	11.49	0.0037	-1.4184	0.0606	99.27	335	16	334	356	341	1382	299
249	Point	2	No	gpcc	88.73	-18.23	18.23	-0.0247	12.93	0.0011	-2.0309	-0.0337	111.13	381	186	395	390	384	1736	372	
250	Point	2	No	gpcc	91.78	-14.28	14.28	-0.0644	17.62	0.0064	-1.7889	-0.0801	111.80	392	142	409	399	385	1727	369	
251	Point	2	No	gpcc lrmm	95.40	-42.56	42.56	-0.1544	14.73	-0.1341	-8.6659	-0.3662	115.79	400	344	446	444	398	2032	427	
252	Point	2	No	ghcn cams grid	86.20	-22.62	22.62	0.0036	277.7778	12.05	0.0000	-1.8132	0.0047	107.06	371	221	374	375	374	1715	364
253	Point	2	No	gpi	132.55	-90.79	90.79	-0.0544	11.13	0.0103	-5.8278	-0.1016	163.91	461	452	405	402	462	2182	455	
254	Point	2	No	lrmm 3a46	115.53	-84.55	84.55	-0.1489	0.46	0.1598	-12.1721	-0.3998	139.56	442	399	444	446	444	2175	453	
255	Point	2	No	lrmm 3b43	94.90	-42.49	42.49	-0.1558	-4.69	0.1403	-8.8805	-0.3746	117.07	399	341	448	445	400	2033	428	
256	Point	2	No	ssmi	117.04	-66.62	66.62	-0.1505	2.23	0.1004	-7.7728	-0.3169	139.45	444	405	445	439	443	2176	454	
257	Point	2	No	opi2	78.27	-0.04	0.04	-0.0380	12.64	0.0017	-1.2433	-0.0409	95.61	313	1	398	393	315	1420	312	
258	Point	2	No	opi3	79.15	-5.61	5.61	-0.0098	12.35	0.0001	-1.3773	-0.0114	98.42	321	40	385	382	335	1463	321	
259	Point	2	No	cmnp	81.49	-13.75	13.75	-0.0049	12.30	0.0000	-1.5552	-0.0060	102.03	340	136	380	380	357	1593	343	
260	Point	2	No	cmnp2	81.30	-13.12	13.12	0.0003	3333.3333	12.17	0.0000	-1.5469	0.0003	101.87	338	129	376	376	355	1574	337
261	Point	2	1 mon	reanalysis	75.38	-1.85	1.85	0.1386	7.2150	10.28	0.0300	-1.1292	0.1732	93.31	291	12	200	313	294	1110	224
262	Point	2	1 mon	gpcc	81.20	-17.88	17.88	0.0973	10.2775	9.28	0.0174	-1.5565	0.1319	102.25	337	173	277	329	359	1475	323
263	Point	2	1 mon	gpcc	79.20	-18.46	18.46	0.0567	17.6367	11.36	0.0062	-1.7929	0.0789	100.07	322	158	326	347	346	1499	326
264	Point	2	1 mon	gpcc lrmm	87.00	-38.90	38.90														

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QI	Runoff Meth	Re	Storage	Precipitation	Absolute Error	Bias	Abias	Slope	Oslope	Intercept	R <sup>2</sup>	R <sup>2</sup> 1:1	Corr Coef	RMSE	AER	BR	SR	CCR	RMR	TR	BR
304	Point	2	4 mon	ghcn cams grid	77.10	-23.84	23.84	0.1731	5.7770	4.77	0.0529	-1.2992	0.2301	95.29	307	232	166	290	311	1306	282
305	Point	2	4 mon	gpi	121.27	-94.44	94.44	0.1347	7.4239	-11.78	0.0857	-6.6088	0.2927	148.60	451	456	210	241	452	1810	384
306	Point	2	4 mon	trmm 3a46	105.34	-70.15	70.15	0.1016	9.8425	-11.29	0.0793	-9.6530	0.2816	121.80	415	410	265	251	409	1750	376
307	Point	2	4 mon	trmm 3b43	72.19	-36.09	36.09	0.1371	7.2939	-10.99	0.1154	-5.4470	0.3397	92.08	271	305	206	215	287	1284	278
308	Point	2	4 mon	ssmi	97.97	-64.81	64.81	0.1340	7.4627	-14.67	0.0810	-5.2305	0.2847	116.48	405	401	213	248	399	1666	354
309	Point	2	4 mon	opi2	65.28	-1.25	1.25	0.1969	5.0787	8.40	0.0465	-0.7274	0.2156	82.60	209	10	146	296	214	875	161
310	Point	2	4 mon	opi3	68.42	-6.62	6.62	0.1772	5.8433	7.68	0.0435	-0.9056	0.2086	86.75	240	53	162	304	249	1008	199
311	Point	2	4 mon	cmnp	71.85	-14.71	14.71	0.1723	5.8038	6.37	0.0458	-1.0661	0.2140	90.33	268	143	168	299	271	1149	236
312	Point	2	4 mon	cmnp2	71.63	-14.08	14.08	0.1728	5.7870	6.47	0.0465	-1.0680	0.2155	90.37	263	141	167	297	272	1140	232
313	Area	1	Yes	reanalysis	60.61	6.00	6.00	0.9852	1.0150	6.03	0.1963	0.1902	0.4431	69.28	154	44	1	159	100	458	60
314	Area	1	Yes	gpcp	63.38	-21.77	21.77	0.5430	1.8416	-6.36	0.3077	0.0126	0.5547	77.91	190	213	39	78	183	703	115
315	Area	1	Yes	gpcp	63.88	-23.94	23.94	0.5644	1.7718	-10.04	0.3292	0.0332	0.5738	74.49	194	233	35	68	143	673	110
316	Area	1	Yes	gpcp trmm	52.37	-10.69	10.69	0.8289	1.2064	-4.59	0.5297	0.4921	0.7278	62.23	70	109	11	18	55	263	17
317	Area	1	Yes	ghcn cams grid	66.86	-33.43	33.43	0.5773	1.7322	-15.99	0.3297	-0.0356	0.5742	78.33	221	284	32	67	187	791	138
318	Area	1	Yes	gpi	92.34	-85.52	85.52	0.5159	1.9384	-37.99	0.4708	-1.0908	0.6862	115.47	394	449	43	34	396	1316	285
319	Area	1	Yes	trmm 3a46	44.18	-38.15	38.15	0.8475	1.1799	-28.41	0.8041	0.5932	0.8967	56.60	21	310	10	2	30	373	39
320	Area	1	Yes	trmm 3b43	58.01	-11.41	11.41	0.7287	1.3723	-1.62	0.4447	0.3657	0.6669	68.98	132	113	16	41	97	399	45
321	Area	1	Yes	ssmi	61.62	-49.54	49.54	0.6892	1.4943	-28.78	0.5607	0.0502	0.7488	79.00	166	388	21	13	191	779	136
322	Area	1	Yes	opi2	55.30	-4.31	4.31	0.7360	1.3587	-0.38	0.3347	0.2886	0.5785	66.15	104	31	15	66	73	289	23
323	Area	1	Yes	opi3	59.08	-10.37	10.37	0.6353	1.5741	-2.73	0.2968	0.1815	0.5448	70.95	144	106	24	81	116	471	61
324	Area	1	Yes	cmnp	62.31	-18.52	18.52	0.5891	1.6975	-6.35	0.2872	0.0917	0.5359	74.72	174	194	29	91	149	637	101
325	Area	1	Yes	cmnp2	62.30	-17.85	17.85	0.5816	1.7194	-5.74	0.2824	0.0845	0.5314	75.02	173	178	31	99	152	633	100
326	Area	1	No	reanalysis	25.40	6.13	6.13	0.1650	6.0606	7.91	0.1909	-4.9198	0.4369	31.81	6	45	175	165	1	392	44
327	Area	1	No	gpcp	63.24	-21.92	21.92	0.0873	11.4548	8.85	0.1538	-18.1482	0.3922	78.06	188	215	286	198	185	1072	213
328	Area	1	No	gpcp	64.64	-24.08	24.08	0.0567	17.6367	6.03	0.3154	-98.5099	0.5616	76.81	205	234	325	74	166	1004	197
329	Area	1	No	gpcp trmm	54.97	-10.21	10.21	0.2189	4.5883	17.03	0.2869	-3.4726	0.5356	66.26	99	104	121	92	74	490	64
330	Area	1	No	ghcn cams grid	64.53	-32.97	32.97	0.0690	14.4928	5.44	0.1827	-35.7935	0.4034	79.40	204	279	306	188	195	1172	240
331	Area	1	No	gpi	108.67	-85.26	85.26	0.0565	17.8991	7.39	0.0891	-42.7588	0.2985	132.97	424	447	328	327	431	1867	395
332	Area	1	No	trmm 3a46	79.32	-31.80	31.80	0.1498	6.6756	22.49	0.1753	-6.3702	0.4187	91.19	323	269	187	182	279	1240	270
333	Area	1	No	trmm 3b43	56.79	-10.93	10.93	0.2121	4.7148	17.48	0.2913	-3.8538	0.5397	68.61	114	110	128	85	92	529	76
334	Area	1	No	ssmi	85.27	-49.16	49.16	0.0588	17.0068	9.91	0.0598	-20.3250	0.2446	100.75	368	387	321	380	351	1707	359
335	Area	1	No	opi2	47.70	-3.16	3.16	0.1376	7.2674	9.67	0.2314	-8.8906	0.4811	55.48	38	25	205	137	28	433	55
336	Area	1	No	opi3	51.12	-9.22	9.22	0.1133	8.8261	9.35	0.1869	-11.5234	0.4323	62.40	63	90	246	174	58	629	98
337	Area	1	No	cmnp	54.90	-17.82	17.82	0.1007	9.9305	8.82	0.1651	-14.0330	0.4063	68.50	98	176	266	186	90	816	145
338	Area	1	No	cmnp2	54.98	-17.14	17.14	0.0992	10.0806	8.93	0.1819	-14.1173	0.4024	68.70	100	166	272	190	94	822	148
339	Area	1	1 mon	reanalysis	24.72	6.21	6.21	0.0931	10.7411	8.06	0.0608	-5.9194	0.2465	34.42	2	48	281	277	4	612	94
340	Area	1	1 mon	gpcp	59.87	-21.42	21.42	0.1160	8.6207	7.95	0.2884	-16.7868	0.5181	75.32	147	210	238	105	155	855	154
341	Area	1	1 mon	gpcp	63.04	-23.67	23.67	0.0721	13.8696	5.47	0.5181	-94.6119	0.7198	75.54	184	231	305	22	159	901	170
342	Area	1	1 mon	gpcp trmm	48.83	-10.68	10.68	0.2934	3.4083	14.80	0.5146	-2.5831	0.7173	59.81	45	108	84	23	43	303	25
343	Area	1	1 mon	ghcn cams grid	62.66	-33.05	33.05	0.0910	10.9890	4.51	0.2830	-34.3259	0.5320	77.86	179	280	282	97	182	1020	200
344	Area	1	1 mon	gpi	106.73	-84.81	84.81	0.0855	11.6959	4.54	0.2034	-40.7999	0.4510	130.23	419	445	290	154	423	1731	370
345	Area	1	1 mon	trmm 3a46	68.35	-34.06	34.06	0.2630	3.8023	14.76	0.5261	-4.6115	0.7253	80.49	239	290	95	21	206	851	152
346	Area	1	1 mon	trmm 3b43	48.28	-9.50	9.50	0.2897	3.4518	15.07	0.5329	-2.7612	0.7300	60.91	41	98	85	17	48	289	22
347	Area	1	1 mon	ssmi	80.78	-48.58	48.58	0.1044	9.5785	7.07	0.1871	-18.5078	0.4325	96.61	333	383	261	179	324	1474	322
348	Area	1	1 mon	opi2	45.35	-3.34	3.34	0.1680	6.0241	9.23	0.3358	-8.1807	0.5795	53.52	26	26	173	64	21	310	32
349	Area	1	1 mon	opi3	48.38	-9.42	9.42	0.1432	6.9832	8.70	0.2977	-10.6376	0.5456	60.26	42	95	194	80	46	467	59
350	Area	1	1 mon	cmnp	52.64	-18.13	18.13	0.2593	7.9051	8.03	0.2593	-13.1537	0.5092	66.58	72	183	224	113	77	669	109
351	Area	1	1 mon	cmnp2	52.69	-17.46	17.46	0.1253	7.9808	8.15	0.2565	-13.2227	0.5065	66.75	74	168	225	116	78	661	107
352	Area	1	2 mon	reanalysis	25.28	6.22	6.22	0.0183	54.6448	8.23	0.0024	-6.9744	0.0485	36.96	3	49	357	364	6	779	134
353	Area	1	2 mon	gpcp	60.95	-21.20	21.20	0.1087	9.1996	8.23	0.2354	-16.9949	0.4852	75.94	158	206	253	135	161	913	173
354	Area	1	2 mon	gpcp	63.60	-23.04	23.04	0.0649	15.4083	5.63	0.4261	-97.1856	0.6527	75.76	193	226	313	46	160	938	183
355	Area	1	2 mon	gpcp trmm	49.89	-11.36	11.36	0.2804	3.5663	15.00	0.4688	-2.7451	0.6847	61.86	52	112	86	35	53	340	32
356	Area	1	2 mon	ghcn cams grid	63.43	-33.16	33.16	0.0825	12.1212	4.84	0.2327	-34.9237	0.4824	78.55	192	282	293	336	188	1091	219
357	Area	1	2 mon	gpi	107.06	-84.92	84.92	0.0832	12.0192	4.71	0.1927	-40.9293	0.4390	130.88	420	446	292	163	424	1745	375
358	Area	1	2 mon	trmm 3a46	67.68	-37.22	37.22	0.2999	3.3344	11.37	0.6519	-4.0726	0.8074	77.45	228	309	82	5	175	799	140
359	Area	1	2 mon	trmm 3b43	51.75	-10.09	10.09	0.2677	3.7355	15.59	0.4542	-3.0472	0.6739	63.70	66	102	93	39	60	360	34
360	Area	1	2 mon	ssmi	80.90	-48.70	48.70	0.1155	8.6580	6.39	0.2290	-18.1152	0.4785	95.92	334	384	241	40	318	1417	311
361	Area	1	2 mon	opi2	46.80	-3.49	3.49	0.1465	6.8259	9.51	0.2613	-8.6410	0.5112	54.94	32	27	191	109	26	385	43
362	Area	1	2 mon	opi3	49.90	-9.67	9.67	0.1284	7.7882	9.00	0.2382	-11.0381	0.4881	61.39	53	100	222	132	51	558	82
363	Area	1	2 mon	cmnp	54.59	-18.26	18.26	0.1112	8.9928	8.48	0.2003	-13.6471	0.4476	67.85	91	187	250	156	85	769	131
364	Area	1	2 mon	cmnp2	54.59	-17.60	17.60	0.1103	9.0662	8.58	0.1987	-13.7093	0.4458	67.99	93	171	252	157	88	761	127</

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OI	Runoff Meth	Re	Storage	Precipitation	Absolute Error	Bias	Abias	Slope	Oslope	Intercept	R <sup>2</sup>	R <sup>2</sup> 1:1	Corr Coef	RMSE	AER	BR	SR	CCR	RMR	TR	BR
405	Area	2	No	gpcp	98.17	-47.56	47.56	-0.2212		-10.42	0.1991	-7.0833	-0.4462	122.63	406	377	460	454	411	2108	442
406	Area	2	No	gpcp	97.89	-44.14	44.14	-0.2572		-6.46	0.2186	-5.9352	-0.4675	120.51	404	355	467	458	404	2088	439
407	Area	2	No	gpcp trmm	97.18	-50.94	50.94	-0.1940		-11.81	0.3477	-15.8899	-0.5896	119.44	403	390	459	465	402	2119	444
408	Area	2	No	ghcn cams grid	96.53	-51.95	51.95	-0.2319		-9.08	0.2001	-6.8995	-0.4474	121.23	401	394	463	456	407	2121	446
409	Area	2	No	gpi	146.23	-113.34	113.34	-0.1670		-0.65	0.1998	-16.5223	-0.4469	179.80	468	464	452	455	468	2307	468
410	Area	2	No	trmm 3a46	119.69	-74.15	74.15	-0.1716		-7.85	0.3522	-22.2327	-0.5935	143.89	450	420	455	468	449	2242	462
411	Area	2	No	trmm 3b43	96.74	-50.87	50.87	-0.1921		-11.89	0.3501	-16.1983	-0.5917	120.53	402	389	458	467	405	2121	445
412	Area	2	No	ssmi	126.35	-83.34	83.34	-0.2286		-9.82	0.3494	-14.4692	-0.5911	150.81	456	439	462	466	454	2277	466
413	Area	2	No	opi2	82.13	-29.36	29.36	-0.2848		-13.67	0.2052	-4.4351	-0.4530	100.56	345	251	468	457	350	1871	397
414	Area	2	No	opi3	85.36	-34.93	34.93	-0.2509		-12.69	0.1852	-5.0735	-0.4303	106.30	369	299	466	452	371	1957	418
415	Area	2	No	cmap	89.81	-43.08	43.08	-0.2385		-10.97	0.1862	-5.8303	-0.4315	112.73	385	350	465	453	389	2042	431
416	Area	2	No	cmap2	89.68	-42.45	42.45	-0.2355		-11.19	0.1829	-5.8185	-0.4277	112.63	384	340	464	450	388	2026	424
417	Area	2	1 mon	reanalysis	83.26	-30.90	30.90	-0.1319		-15.36	0.0595	-4.8337	-0.2440	104.37	356	264	439	428	362	1849	390
418	Area	2	1 mon	gpcp	92.79	-47.04	47.04	-0.1192		-13.81	0.0572	-6.1894	-0.2392	115.70	397	370	429	426	397	2019	421
419	Area	2	1 mon	gpcp	90.20	-45.09	45.09	-0.1671		-10.67	0.1053	-6.1384	-0.3245	114.63	386	360	453	441	394	2034	429
420	Area	2	1 mon	gpcp trmm	92.07	-47.32	47.32	-0.1426		-13.91	0.1737	-13.5802	-0.4168	112.12	393	373	443	447	387	2043	432
421	Area	2	1 mon	ghcn cams grid	91.72	-51.70	51.70	-0.1221		-12.96	0.0553	-6.0499	-0.2352	114.73	391	392	434	424	395	2036	430
422	Area	2	1 mon	gpi	144.68	-113.51	113.51	-0.1275		-5.56	0.1285	-17.6565	-0.3585	176.25	467	465	436	442	467	2277	467
423	Area	2	1 mon	trmm 3a46	121.96	-76.75	76.75	-0.1658		-7.61	0.3193	-21.9305	-0.5651	144.59	452	429	451	464	450	2246	464
424	Area	2	1 mon	trmm 3b43	93.14	-47.64	47.64	-0.1420		-13.88	0.1804	-14.1188	-0.4248	114.18	398	378	441	449	392	2058	435
425	Area	2	1 mon	ssmi	122.75	-81.35	81.35	-0.1836		-11.99	0.2267	-13.8021	-0.4761	145.78	454	436	457	461	451	2259	465
426	Area	2	1 mon	opi2	76.45	-29.04	29.04	-0.1576		-15.30	0.0624	-3.7568	-0.2498	94.24	300	245	449	430	302	1726	368
427	Area	2	1 mon	opi3	79.56	-34.48	34.48	-0.1252		-15.00	0.0456	-4.2743	-0.2135	99.24	327	296	435	419	340	1817	386
428	Area	2	1 mon	cmap	84.27	-42.62	42.62	-0.1220		-14.07	0.0482	-5.0004	-0.2195	105.85	362	345	433	423	369	1932	412
429	Area	2	1 mon	cmap2	84.02	-41.99	41.99	-0.1176		-14.25	0.0451	-4.9754	-0.2124	105.63	361	337	427	418	368	1911	405
430	Area	2	2 mon	reanalysis	78.47	-30.85	30.85	-0.0532		-16.55	0.0097	-4.2940	-0.0984	99.53	314	262	403	401	343	1723	366
431	Area	2	2 mon	gpcp	87.26	-46.83	46.83	-0.0011		-17.24	0.0000	-5.2098	-0.0021	107.79	373	367	377	377	379	1873	399
432	Area	2	2 mon	gpcp	83.70	-45.16	45.16	-0.0205		-15.73	0.0016	-5.1875	-0.0403	105.13	359	381	393	392	366	1871	398
433	Area	2	2 mon	gpcp trmm	87.47	-44.97	44.97	-0.0484		-15.91	0.0204	-11.9715	-0.1430	104.92	375	359	400	407	364	1905	403
434	Area	2	2 mon	ghcn cams grid	87.29	-51.57	51.57	-0.0031		-17.16	0.0000	-5.1542	-0.0060	107.31	374	391	379	379	377	1900	402
435	Area	2	2 mon	gpi	140.92	-113.68	113.68	-0.0649		-12.09	0.0342	-17.1341	-0.1849	171.66	466	466	410	414	466	2222	460
436	Area	2	2 mon	trmm 3a46	119.65	-79.05	79.05	-0.1073		-9.78	0.1337	-21.0733	-0.3656	140.67	449	430	420	443	446	2188	457
437	Area	2	2 mon	trmm 3b43	89.07	-45.17	45.17	-0.0549		-15.71	0.0276	-12.5447	-0.1661	107.22	383	362	404	411	376	1936	413
438	Area	2	2 mon	ssmi	118.74	-80.98	80.98	-0.1200		-15.68	0.0971	-12.9264	-0.3117	141.52	447	435	430	438	448	2198	458
439	Area	2	2 mon	opi2	71.80	-29.07	29.07	-0.0260		-16.96	0.0017	-3.0991	-0.0413	87.58	266	246	396	394	253	1555	336
440	Area	2	2 mon	opi3	73.97	-34.39	34.39	-0.0093	107.5269	-17.43	0.0003	-3.4859	0.0159	91.62	278	292	369	372	282	1593	342
441	Area	2	2 mon	cmap	79.12	-42.52	42.52	-0.0050	200.0000	-17.40	0.0001	-4.1710	0.0089	98.36	320	343	373	373	334	1743	373
442	Area	2	2 mon	cmap2	78.91	-41.89	41.89	-0.0107	93.4579	-17.53	0.0004	-4.1311	0.0193	97.98	319	336	366	371	330	1722	365
443	Area	2	3 mon	reanalysis	72.11	-30.87	30.87	0.0687	14.5560	-18.29	0.0162	-3.4622	0.1271	91.49	270	263	308	335	281	1457	320
444	Area	2	3 mon	gpcp	82.83	-47.06	47.06	0.1060	9.4340	-20.51	0.0453	-4.3546	0.2128	100.22	350	371	259	301	347	1628	349
445	Area	2	3 mon	gpcp	79.84	-45.89	45.89	0.1148	8.7108	-20.25	0.0527	-4.2746	0.2298	96.05	331	365	243	291	319	1549	334
446	Area	2	3 mon	gpcp trmm	83.27	-43.96	43.96	0.0395	25.3165	-17.03	0.0150	-11.3380	0.1226	98.49	357	354	342	336	336	1725	367
447	Area	2	3 mon	ghcn cams grid	82.89	-51.75	51.75	0.1125	8.8889	-21.23	0.0470	-4.3050	0.2167	99.76	351	393	248	294	344	1630	350
448	Area	2	3 mon	gpi	136.64	-114.06	114.06	0.0193	51.8135	-20.55	0.0031	-16.1380	0.0555	165.44	464	467	358	363	464	2114	443
449	Area	2	3 mon	trmm 3a46	116.36	-79.89	79.89	-0.0154		-13.98	0.0030	-20.7873	-0.0549	134.05	443	431	391	397	435	2097	441
450	Area	2	3 mon	trmm 3b43	83.23	-44.32	44.32	0.0482	20.7469	-17.29	0.0234	-11.5956	0.1529	99.52	354	356	335	324	342	1711	360
451	Area	2	3 mon	ssmi	113.08	-80.08	80.08	-0.0105		-21.36	0.0008	-11.7341	-0.0279	133.33	437	432	386	389	433	2077	437
452	Area	2	3 mon	opi2	67.22	-29.29	29.29	0.1173	8.5251	-18.78	0.0346	-2.3836	0.1860	79.67	225	249	235	309	197	1215	262
453	Area	2	3 mon	opi3	70.25	-34.58	34.58	0.1244	8.0386	-19.50	0.0450	-2.8235	0.2121	84.69	256	297	227	303	237	1320	286
454	Area	2	3 mon	cmap	75.37	-42.69	42.69	0.1180	8.4746	-20.35	0.0451	-3.4462	0.2124	91.33	290	347	233	302	280	1452	318
455	Area	2	3 mon	cmap2	75.12	-42.07	42.07	0.1193	8.3622	-20.31	0.0465	-3.4291	0.2157	91.15	287	338	232	295	278	1430	315
456	Area	2	4 mon	reanalysis	66.27	-31.23	31.23	0.1494	6.6934	-19.69	0.0775	-2.9589	0.2784	85.75	218	267	188	255	244	1172	241
457	Area	2	4 mon	gpcp	77.94	-47.45	47.45	0.1712	5.8411	-22.78	0.1197	-3.8970	0.3459	95.37	311	376	189	213	313	1382	297
458	Area	2	4 mon	gpcp	74.08	-47.01	47.01	0.2112	4.7348	-23.96	0.1875	-3.7509	0.4330	89.09	279	369	130	172	263	1213	261
459	Area	2	4 mon	gpcp trmm	75.27	-44.44	44.44	0.1128	8.8652	-18.94	0.1228	-9.9283	0.3504	93.61	288	357	247	210	298	1400	306
460	Area	2	4 mon	ghcn cams grid	78.01	-52.26	52.26	0.1887	5.2994	-24.19	0.1338	-3.8096	0.3658	94.51	312	396	155	203	305	1371	292
461	Area	2	4 mon	gpi	133.04	-114.81	114.81	0.0971	10.2987	-28.57	0.0816	-15.5333	0.2856	159.61	462	468	278	246	460	1914	407
462	Area	2	4 mon	trmm 3a46	112.57	-80.22	80.22	0.0667	14.9925	-19.07	0.0565	-18.6288	0.2377	128.63	434	433	310	285	419	1881	400
463	Area	2	4 mon	trmm 3b43	75.41	-44.78	44.78	0.1175	8.5106	-19.11	0.1392	-10.2034	0.3730	94.78	292	358	234	201	307	1392	301
464	Area	2	4 mon	ssmi	107.60	-80.84	80.84	0.0905	11.0497	-28.17	0.0559	-10.1361	0.2365	126.56	422	434	283	287	417	1843	389
465	Area	2	4 mon	opi2	61.56	-29.67	29.67	0.2206	4.5331	-20.31	0.1241	-1.8990	0.3523	73.38	163	258	11				



Rondonia - Yearly Surface

OI	Runoff Meth	Re	Storage	Precipitation	Absolute Error	Bias	Abias	Slope	Oslope	Intercept	R <sup>2</sup>	R <sup>2</sup> 1:1	Corr Coef	RMSE	AER	BR	SR	CCR	RMR	TR	BR	
1	Linear	1	Yes	reanalysis	16.11	8.33	8.33	0.2365	4.2283	9.96	0.0956	-1.2971	0.3092	20.06	151	32	98	67	27	375	64	Plot
2	Linear	1	Yes	gpcp	21.73	-19.27	19.27	0.6516	1.5347	-7.53	0.2281	-1.1111	0.4776	24.81	90	66	47	36	45	284	27	
3	Linear	1	Yes	gpcp	22.74	-22.74	22.74	0.2268	4.4092	1.94	0.1580	-12.7054	0.3975	25.36	81	76	102	54	48	361	60	
4	Linear	1	Yes	gpcp trmm	16.07	1.70	1.70	0.6931	1.4428	14.12	0.2073	0.1585	0.4553	17.38	144	9	39	42	21	255	19	
5	Linear	1	Yes	ghcn cams grid	32.16	-30.84	30.84	0.1188	8.4175	5.60	0.0088	-5.8107	0.0940	34.85	62	95	123	118	69	467	112	
6	Linear	1	Yes	gpi	82.38	-82.38	82.38	-0.0540		21.12	0.0018	-20.3290	-0.0419	85.80	11	146	155	133	147	592	153	
7	Linear	1	Yes	trmm 3a46	24.21	-19.85	19.85	-1.9806		169.95	0.1538	-1.0380	-0.3922	30.85	89	68	133	152	57	499	127	
8	Linear	1	Yes	trmm 3b43	14.67	-0.78	0.78	1.1089	1.1089	-4.71	0.5397	0.5332	0.7346	14.91	138	5	13	10	8	174	2	
9	Linear	1	Yes	ssmi	45.01	-43.28	43.28	0.2339	4.2753	4.46	0.0145	-4.7741	0.1203	48.32	38	119	100	114	119	490	124	
10	Linear	1	Yes	opi2	12.65	-1.08	1.08	0.7442	1.3437	2.89	0.2720	0.2359	0.5215	14.93	137	6	29	25	9	206	8	
11	Linear	1	Yes	opi3	11.46	-6.65	6.65	0.8750	1.1429	-4.01	0.3564	0.1975	0.5970	15.30	127	21	15	15	15	193	4	
12	Linear	1	Yes	cmcp	18.76	-14.80	14.80	0.6430	1.5552	-4.36	0.1991	-0.6130	0.4462	21.69	105	51	48	46	32	282	25	
13	Linear	1	Yes	cmcp2	18.31	-14.16	14.16	0.6216	1.6088	-3.34	0.1873	-0.5700	0.4328	21.40	106	47	54	48	31	288	29	
14	Linear	1	Yes	reanalysis	13.41	7.14	7.14	0.2263	4.4189	8.79	0.2365	-3.3121	0.4863	16.73	149	24	103	34	20	330	46	
15	Linear	1	No	gpcp	20.86	-20.51	20.51	0.2201	4.5434	5.78	0.0668	-4.4711	0.2584	24.94	87	70	107	83	46	393	81	
16	Linear	1	No	gpcp	23.36	-23.36	23.36	-0.1976		14.86	0.6316	-83.7939	-0.7947	27.49	80	77	147	154	52	510	137	
17	Linear	1	No	gpcp trmm	10.20	-7.47	7.47	0.4955	2.0182	12.93	0.2611	-0.3933	0.5110	14.24	124	27	68	28	6	253	17	
18	Linear	1	No	ghcn cams grid	32.37	-32.05	32.05	0.0589	16.9779	6.86	0.0059	-17.0382	0.0766	34.52	57	109	127	122	67	473	115	
19	Linear	1	No	gpi	83.71	-83.71	83.71	-0.1827		32.41	0.0474	-49.9031	-0.2177	86.24	9	148	148	143	148	596	155	
20	Linear	1	No	trmm 3a46	30.05	-30.05	30.05	-1.1646		107.79	0.1291	-5.0103	-0.3594	30.00	83	94	137	146	65	505	136	
21	Linear	1	No	trmm 3b43	11.05	-8.12	8.12	0.7936	1.2601	-0.68	0.6035	0.2600	0.7768	12.70	120	31	23	7	3	184	3	
22	Linear	1	No	ssmi	46.61	-46.61	46.61	-0.1147		22.85	0.0076	-12.5025	-0.0874	49.88	30	127	153	137	123	570	145	
23	Linear	1	No	opi2	9.29	-2.31	2.31	0.4808	2.0799	5.74	0.2911	-0.0955	0.5395	11.16	134	11	71	24	2	242	15	
24	Linear	1	No	opi3	11.60	-7.89	7.89	0.3165	3.1596	6.53	0.1196	-0.9850	0.3459	15.02	123	28	92	64	10	317	40	
25	Linear	1	No	cmcp	17.38	-16.03	16.03	0.1381	7.2411	9.17	0.0235	-3.1542	0.1534	21.73	101	55	116	107	35	414	94	
26	Linear	1	No	cmcp2	16.93	-15.40	15.40	0.1325	7.5472	9.41	0.0218	-2.9991	0.1478	21.32	103	53	120	110	30	416	96	
27	Linear	2	Yes	reanalysis	28.76	-16.42	16.42	0.3348	2.9869	-7.06	0.0745	-0.5189	0.2730	36.99	98	58	89	76	77	398	86	Plot
28	Linear	2	Yes	gpcp	36.24	-32.76	32.76	0.2894	3.4554	-11.15	0.0135	-1.2589	0.1163	45.12	56	101	95	115	113	480	118	
29	Linear	2	Yes	gpcp	32.49	-31.82	31.82	0.1197	8.3542	-5.44	0.0025	-1.8222	0.0505	41.15	59	98	122	127	96	502	131	
30	Linear	2	Yes	gpcp trmm	35.83	-29.23	29.23	0.5030	1.9802	-13.01	0.0553	-1.0620	0.2352	40.69	67	90	66	87	95	405	89	
31	Linear	2	Yes	ghcn cams grid	39.30	-37.15	37.15	0.7030	1.4225	-26.81	0.0387	-1.4997	0.1968	47.46	47	110	35	93	117	402	87	
32	Linear	2	Yes	gpi	101.77	-101.77	101.77	0.7882	1.2687	-81.31	0.1308	-10.6493	0.3616	105.84	4	153	24	60	154	395	82	
33	Linear	2	Yes	trmm 3a46	46.71	-46.71	46.71	-0.7265		48.91	0.0288	-2.3951	-0.1698	57.24	29	128	141	141	132	571	147	
34	Linear	2	Yes	trmm 3b43	32.03	-29.16	29.16	0.9172	1.0903	-26.45	0.2105	-0.8505	0.4588	38.54	68	89	9	40	84	290	32	
35	Linear	2	Yes	ssmi	65.70	-65.68	65.68	0.9165	1.0911	-60.74	0.0792	-4.2221	0.2814	72.38	20	137	10	71	140	378	68	
36	Linear	2	Yes	opi2	28.15	-14.56	14.56	0.4712	2.1222	-8.10	0.0320	-0.2436	0.1788	33.48	106	50	74	98	63	391	80	
37	Linear	2	Yes	opi3	27.46	-20.13	20.13	0.9436	1.0598	-19.13	0.1123	-0.3379	0.3352	34.72	88	69	8	66	68	299	34	
38	Linear	2	Yes	cmcp	33.03	-28.28	28.28	0.7002	1.4282	-20.50	0.0693	-0.8309	0.2832	40.62	70	87	36	79	94	366	62	
39	Linear	2	Yes	cmcp2	32.65	-27.65	27.65	0.7520	1.3298	-21.37	0.0812	-0.7758	0.2849	40.00	72	85	28	70	89	344	51	
40	Linear	2	No	reanalysis	29.10	-17.90	17.90	0.2348	4.2589	-7.13	0.0629	-1.2153	0.2507	34.11	94	62	99	85	66	406	90	
41	Linear	2	No	gpcp	36.22	-34.24	34.24	0.1358	7.3638	-7.96	0.0051	-2.4340	0.0715	42.46	52	105	119	123	100	499	128	
42	Linear	2	No	gpcp	37.34	-31.06	31.06	-0.9588		27.64	0.1549	-2.0134	-0.3935	43.70	61	96	138	153	105	553	143	
43	Linear	2	No	gpcp trmm	45.46	-45.46	45.46	0.3218	3.1075	-23.24	0.3149	-37.1627	0.5612	46.76	33	124	91	18	115	381	70	
44	Linear	2	No	ghcn cams grid	39.39	-38.63	38.63	1.1934	1.1934	-45.36	0.1915	-2.8552	0.4376	43.81	45	112	19	47	106	329	45	
45	Linear	2	No	gpi	102.58	-102.58	102.58	0.5574	1.7940	-59.84	0.1198	-20.0062	0.3461	105.00	3	154	64	63	153	437	102	
46	Linear	2	No	trmm 3a46	68.36	-68.36	68.36	0.6091	1.6418	-46.71	0.2571	-61.2817	0.5071	68.83	17	140	58	29	138	382	71	
47	Linear	2	No	trmm 3b43	45.40	-45.40	45.40	0.4564	2.1911	-27.62	0.7305	-36.2786	0.8547	46.21	34	123	76	3	114	505	55	
48	Linear	2	No	ssmi	73.60	-73.60	73.60	0.6517	1.5344	-53.01	0.2371	-31.8118	0.4870	74.55	16	141	46	33	141	377	66	
49	Linear	2	No	opi2	27.51	-16.04	16.04	0.0145	88.9655	-4.01	0.0001	-0.7300	0.0072	30.14	100	56	129	129	56	470	113	
50	Linear	2	No	opi3	26.53	-21.61	21.61	0.3823	2.8157	-10.63	0.0316	-0.9405	0.1779	31.92	84	73	80	89	60	398	83	
51	Linear	2	No	cmcp	32.86	-29.76	29.76	0.3794	2.6357	-13.87	0.0349	-1.7449	0.1868	37.96	64	93	82	96	81	416	97	
52	Linear	2	No	cmcp2	32.94	-29.12	29.12	0.4399	2.2732	-14.96	0.0477	-1.6450	0.2183	37.27	69	88	77	90	79	403	98	
53	Point	1	Yes	reanalysis	18.56	13.69	13.69	0.3565	2.8050	15.06	0.1395	-1.0023	0.3735	23.37	154	45	84	58	43	384	74	
54	Point	1	Yes	gpcp	18.53	-11.89	11.89	0.7857	1.2728	-4.67	0.1997	-0.1072	0.4469	23.16	114	39	25	45	42	265	20	
55	Point	1	Yes	gpcp	18.13	-18.13	18.13	0.1380	7.3529	9.45	0.0562	-9.1402	0.2371	21.94	93	83	118	86	38	398	85	
56	Point	1	Yes	gpcp trmm	20.21	19.32	19.32	1.1674	1.1674	12.55	0.3083	-0.2436	0.5552	29.17	156	87	17	19	55	314	38	
57	Point	1	Yes	ghcn cams grid	28.84	-25.48	25.48	0.1424	7.0225	9.98	0.0081	-2.6246	0.0902	31.73	76	81	115	120	59	451	104	
58	Point	1	Yes	gpi	74.45	-74.45	74.45	-0.1575		39.20	0.0086	-9.6617	-0.0925	80.11	15	142	152	138	143	590	151	
59	Point	1	Yes	trmm 3a46	26.28	-3.33	3.33	-2.7196		233.52	0.1486	-0.1416	-0.3855	32.25	132	13	132	151	62	490	123	
60	Point	1	Yes	trmm 3b43	19.01	14.38	14.38	1.6137	1.6137	-7.75	0.6071	0.2887	0.7792	25.25	155	49	56	6	47	313	37	
61	Point	1	Yes	ssmi	41.53	-34.73	34.73	0.1857	5.3850	16.01	0.0051	-1.7631	0.0714	44.67	49	108	112	124	110	503	133	
62	Point	1	Yes	opi2	14.64	6.30	6.30	1.0069	1.0069</													

Rondonia - Yearly Surface

Stn	Runoff Meth	Re	Storage	Precipitation	Absolute Error	Bias	Abias	Slope	Oslope	Intercept	R <sup>2</sup>	R <sup>2</sup> 1:1	Corr Coef	RMSE	AER	BR	SR	CCR	RMR	TR	BR
102	Point	2	No	opi3	31.01	-5.61	5.61	0.4049	2.4697	4.97	0.0149	-0.0423	0.1219	36.15	128	18	79	113	72	410	92
103	Point	2	No	cmmap	31.65	-13.75	13.75	0.4784	2.0903	-0.23	0.0232	-0.1553	0.1524	38.06	109	46	72	109	82	418	99
104	Point	2	No	cmmap2	31.09	-13.12	13.12	0.5946	1.8818	-2.87	0.0365	-0.1178	0.1910	37.44	112	42	59	94	80	387	76
105	Area	1	Yes	reanalysis	15.85	7.32	7.32	0.2023	4.9432	9.02	0.0774	-1.4648	0.2782	19.76	150	25	109	72	25	381	69
106	Area	1	Yes	gpcc	22.23	-20.68	20.68	0.6379	1.5676	-8.48	0.2426	-1.4633	0.4925	25.44	86	71	51	30	49	287	28
107	Area	1	Yes	gpcc	23.46	-23.46	23.46	0.2575	3.8835	0.24	0.2089	-13.5585	0.4571	25.81	79	78	97	41	51	346	52
108	Area	1	Yes	gpcc trmm	14.11	-1.56	1.56	0.6705	1.4914	11.77	0.2421	0.1752	0.4921	15.40	136	8	43	31	16	234	14
109	Area	1	Yes	ghcn cams grid	32.79	-31.86	31.86	0.1081	9.4251	5.11	0.0078	-6.8419	0.0883	35.55	58	99	124	121	71	473	116
110	Area	1	Yes	gpi	83.92	-83.92	83.92	-0.0306		17.27	0.0006	-23.5804	-0.0251	87.01	8	149	156	132	149	594	154
111	Area	1	Yes	trmm 3a46	23.79	-23.68	23.68	-1.7329		150.34	0.1454	-1.6988	-0.3813	31.95	78	79	135	150	61	503	132
112	Area	1	Yes	trmm 3b43	13.30	-3.59	3.59	1.0408	1.0408	-5.06	0.5734	0.5398	0.7572	13.48	131	14	5	8	4	162	1 Plot
113	Area	1	Yes	ssmi	45.75	-45.01	45.01	0.2876	3.4771	-0.61	0.0245	-5.7437	0.1566	49.31	35	122	96	106	121	480	119
114	Area	1	Yes	opi2	12.25	-2.49	2.49	0.6989	1.4308	2.18	0.2681	0.1932	0.5159	14.58	133	12	37	27	7	216	10
115	Area	1	Yes	opi3	11.72	-8.06	8.06	0.8490	1.1779	-4.88	0.3723	0.1134	0.6102	15.26	121	30	18	14	14	197	7
116	Area	1	Yes	cmmap	19.27	-16.20	16.20	0.6419	1.5579	-5.74	0.2201	-0.8475	0.4692	22.04	99	57	50	39	39	284	26
117	Area	1	Yes	cmmap2	18.81	-15.57	15.57	0.6204	1.6119	-4.72	0.2071	-0.7930	0.4550	21.71	102	54	55	43	34	288	30
118	Area	1	No	reanalysis	13.32	6.13	6.13	0.1921	5.2056	7.85	0.2218	-4.4526	0.4709	16.49	148	19	110	38	18	331	47
119	Area	1	No	gpcc	21.92	-21.92	21.92	0.2085	4.8426	4.83	0.0748	-6.4084	0.2734	25.72	83	74	108	75	50	390	78
120	Area	1	No	gpcc	24.08	-24.08	24.08	-0.1669		13.16	0.6889	-128.7033	-0.8179	27.90	77	80	149	156	54	516	140
121	Area	1	No	gpcc trmm	10.74	-10.74	10.74	0.4729	2.1146	10.58	0.3360	-1.2008	0.5797	15.06	117	38	73	16	12	254	18
122	Area	1	No	ghcn cams grid	33.15	-33.07	33.07	0.0463	21.5983	6.37	0.0047	-23.5369	0.0686	35.29	55	102	128	126	70	481	120
123	Area	1	No	gpi	85.26	-85.26	85.26	-0.1592		28.57	0.0459	-65.7131	-0.2141	87.50	7	150	151	142	150	600	156
124	Area	1	No	trmm 3a46	33.88	-33.88	33.88	-0.9169		88.18	0.1121	-8.7361	-0.3348	36.57	53	104	139	145	74	515	139
125	Area	1	No	trmm 3b43	12.24	-10.93	10.93	0.7255	1.3784	-1.03	0.6606	-0.1520	0.8128	13.85	116	37	33	5	5	196	6
126	Area	1	No	ssmi	48.34	-48.34	48.34	-0.0610		17.78	0.0027	-18.9110	-0.0523	51.01	27	130	154	134	125	570	148
127	Area	1	No	opi2	9.34	-3.72	3.72	0.4355	2.2982	5.04	0.3039	-0.3620	0.5513	11.03	130	15	78	21	1	245	16
128	Area	1	No	opi3	11.55	-9.30	9.30	0.2906	3.4412	5.66	0.1283	-1.6038	0.3582	15.25	118	34	94	61	13	320	42
129	Area	1	No	cmmap	18.13	-17.44	17.44	0.1370	7.2993	7.79	0.0295	-4.5463	0.1718	22.26	95	61	117	100	41	414	95
130	Area	1	No	cmmap2	17.55	-16.81	16.81	0.1314	7.6104	8.03	0.0273	-4.3291	0.1653	21.82	96	60	121	104	37	418	98
131	Area	2	Yes	reanalysis	36.08	-29.75	29.75	0.1915	5.2219	-18.37	0.0295	-1.8843	0.1718	44.71	65	92	111	101	111	480	117
132	Area	2	Yes	gpcc	46.69	-46.08	46.08	0.3798	2.6330	-27.22	0.0282	-2.8983	0.1679	53.88	31	126	81	102	130	470	114
133	Area	2	Yes	gpcc	44.90	-44.90	44.90	0.2262	4.4209	-21.71	0.0109	-4.1224	0.1042	50.77	36	121	105	117	124	503	134
134	Area	2	Yes	gpcc trmm	38.71	-34.70	34.70	0.6692	1.4943	-23.86	0.0958	-1.4033	0.3092	44.29	50	107	44	68	108	377	65
135	Area	2	Yes	ghcn cams grid	51.50	-50.47	50.47	0.4959	2.0165	-32.93	0.0233	-3.4215	0.1527	57.38	26	131	67	108	133	465	110
136	Area	2	Yes	gpi	112.53	-112.53	112.53	0.5583	1.7976	-69.68	0.0709	-14.3072	0.2663	116.29	2	155	85	78	156	456	107
137	Area	2	Yes	trmm 3a46	53.35	-53.35	53.35	-0.6929		40.41	0.0247	-2.8989	-0.1571	63.22	22	135	142	140	135	574	149
138	Area	2	Yes	trmm 3b43	34.91	-34.63	34.63	1.0486	1.0486	-36.22	0.2707	-1.1998	0.5203	42.37	51	106	7	26	99	289	31
139	Area	2	Yes	ssmi	74.97	-74.97	74.97	0.6237	1.6033	-52.73	0.0401	-8.1018	0.2003	80.72	14	143	53	91	145	446	103
140	Area	2	Yes	opi2	31.36	-27.88	27.88	0.8118	1.6345	-23.14	0.0652	-1.0050	0.2554	38.64	71	86	57	84	85	383	72
141	Area	2	Yes	opi3	35.36	-33.46	33.46	0.9070	1.1025	-31.80	0.1256	-1.3788	0.3544	42.09	54	103	11	62	98	328	44
142	Area	2	Yes	cmmap	43.12	-41.60	41.60	0.6325	1.5810	-32.07	0.0684	-2.2787	0.2616	49.41	43	114	52	80	122	411	93
143	Area	2	Yes	cmmap2	42.55	-40.97	40.97	0.6563	1.5237	-32.27	0.0748	-2.1996	0.2735	48.81	44	113	45	74	120	396	84
144	Area	2	No	reanalysis	34.64	-31.22	31.22	0.0915	10.9290	-18.44	0.0131	-3.8295	0.1145	42.97	60	97	125	116	102	500	129
145	Area	2	No	gpcc	47.95	-47.56	47.56	0.2263	4.4189	-24.03	0.0195	-6.1236	0.1396	52.19	28	129	104	111	128	500	130
146	Area	2	No	gpcc	47.11	-44.14	44.14	-0.8523		11.37	0.1410	-4.0663	-0.3755	52.79	37	120	140	149	129	575	150
147	Area	2	No	gpcc trmm	50.94	-50.94	50.94	0.4860	2.0576	-34.10	0.5343	-33.7457	0.7310	51.74	24	133	70	11	127	365	61
148	Area	2	No	ghcn cams grid	51.95	-51.95	51.95	0.9863	1.0139	-51.47	0.1796	-6.8787	0.4238	54.89	23	134	2	51	131	341	50
149	Area	2	No	gpi	113.34	-113.34	113.34	0.3256	3.0713	-48.21	0.0492	-29.6397	0.2219	115.56	1	156	90	89	155	491	125
150	Area	2	No	trmm 3a46	75.00	-75.00	75.00	0.6427	1.5559	-55.21	0.2307	-59.4203	0.4803	75.52	13	144	49	35	142	383	73
151	Area	2	No	trmm 3b43	50.87	-50.87	50.87	0.5878	1.7013	-37.39	0.9012	-33.1362	0.9483	51.28	25	132	61	2	128	348	53
152	Area	2	No	ssmi	82.89	-82.89	82.89	0.3589	2.7863	-45.00	0.0829	-48.9551	0.2880	83.93	10	147	83	69	146	455	106
153	Area	2	No	opi2	33.74	-29.36	29.36	0.1551	6.4475	-19.04	0.0082	-2.4888	0.0903	36.52	66	91	114	119	73	483	108
154	Area	2	No	opi3	37.02	-34.83	34.83	0.3457	2.8827	-23.30	0.0355	-3.2835	0.1885	40.47	48	109	87	95	92	431	101
155	Area	2	No	cmmap	44.16	-43.08	43.08	0.3118	3.2072	-25.23	0.0324	-4.9791	0.1799	47.81	39	118	93	97	118	465	109
156	Area	2	No	cmmap2	43.54	-42.45	42.45	0.3442	2.9053	-25.86	0.0401	-4.8175	0.2002	47.16	42	115	88	92	116	453	105

Southeast United States - Monthly Atmospheric																					
Of	Moist Flux	Re	Storage	Precipitation	Absolute Error	Bias	Abias	Slope	Intercept	R <sup>2</sup>	R <sup>2</sup> 1:1	Corr Coef	RMSE	AER	BR	SR	CCR	RMR	TR	BR	
1	Method 1	1	Yes	reanalysis	43.14	31.18	31.18	0.4861	-17.63	0.2843	-0.5319	0.5332	54.66	6	47	7	2	6	68	7	
2	Method 1	1	Yes	gpcp	54.09	2.91	2.91	0.0453	-2.06	0.0035	-1.5395	0.0589	65.23	33	10	41	41	35	160	37	
3	Method 1	1	Yes	gpc	54.87	-3.03	3.03	0.0234	-2.49	0.0009	-1.5233	0.0295	64.77	37	12	44	44	32	169	39	
4	Method 1	1	Yes	gpc trmm	47.16	-15.40	15.40	0.1861	6.86	0.0387	-1.0877	0.1968	57.49	15	41	30	30	13	129	28	
5	Method 1	1	Yes	ghcn cams grid	52.17	6.67	6.67	0.1104	-5.03	0.0159	-1.0398	0.1262	63.07	28	25	36	34	23	146	30	
6	Method 1	1	Yes	gpi	69.77	16.71	16.71	-0.0143	0.11	0.0006	-3.3647	-0.0252	84.06	47	42	46	46	46	227	48	
7	Method 1	1	Yes	trmm 3b43	47.95	-10.20	10.20	0.1660	6.43	0.0445	-1.1457	0.2110	58.09	17	36	31	28	14	126	27	
8	Method 1	1	Yes	ssmi	38.16	-7.35	7.35	0.2599	1.51	0.0691	-0.5237	0.2629	50.35	2	28	21	20	5	76	11	
9	Method 1	1	Yes	opi2	54.47	-6.77	6.77	-0.0895	-2.19	0.0104	-1.5649	-0.1022	65.14	35	26	48	48	34	191	44	
10	Method 1	1	Yes	opi3	54.26	-3.69	3.69	0.0200	-1.35	0.0006	-1.5817	0.0254	65.10	34	15	45	45	33	172	40	
11	Method 1	1	Yes	cmap	53.86	-2.83	2.83	0.0281	-1.62	0.0013	-1.5329	0.0358	64.69	31	8	43	43	30	155	31	
12	Method 1	1	Yes	cmap2	53.86	-2.83	2.83	0.0281	-1.62	0.0013	-1.5330	0.0358	64.70	32	9	42	42	31	156	34	
13	Method 1	2	Yes	reanalysis	38.46	17.15	17.15	0.3380	-4.30	0.1557	-0.6358	0.3947	49.81	3	43	15	6	3	70	8	
14	Method 1	2	Yes	gpcp	45.82	1.37	1.37	0.1841	1.59	0.0802	-1.1239	0.2454	56.76	13	3	24	21	12	73	9	
15	Method 1	2	Yes	gpc	46.07	-8.15	8.15	0.1755	3.82	0.0518	-1.1374	0.2275	55.23	14	31	26	24	8	103	23	
16	Method 1	2	Yes	gpc trmm	38.68	-11.92	11.92	0.2879	7.59	0.1068	-0.8419	0.3267	49.46	4	39	18	13	2	76	10	
17	Method 1	2	Yes	ghcn cams grid	44.91	1.85	1.85	0.1919	1.47	0.0594	-0.9967	0.2437	55.04	8	5	23	22	7	65	5	
18	Method 1	2	Yes	gpi	62.68	14.09	14.09	0.0910	2.44	0.0296	-3.0566	0.1719	75.94	45	40	38	31	45	199	45	
19	Method 1	2	Yes	trmm 3b43	39.86	-6.45	6.45	0.2723	6.00	0.1078	-0.6899	0.3284	50.17	5	24	19	12	4	64	3	
20	Method 1	2	Yes	ssmi	36.58	-10.17	10.17	0.2886	5.36	0.0953	-0.5554	0.3088	47.52	1	35	17	14	1	68	6	
21	Method 1	2	Yes	opi2	47.23	-7.10	7.10	0.0590	2.55	0.0048	-1.2440	0.0692	58.34	16	27	40	40	15	138	29	
22	Method 1	2	Yes	opi3	45.81	-3.88	3.88	0.1719	2.54	0.0489	-1.0973	0.2212	56.41	12	19	29	27	11	98	20	
23	Method 1	2	Yes	cmap	45.70	-3.66	3.66	0.1746	2.51	0.0504	-1.0853	0.2245	56.24	10	13	27	26	9	85	14	
24	Method 1	2	Yes	cmap2	45.70	-3.66	3.66	0.1745	2.51	0.0504	-1.0854	0.2245	56.25	11	14	28	25	10	88	16	
25	Method 2	1	Yes	reanalysis	48.45	31.91	31.91	0.7598	-26.40	0.3091	-0.0101	0.5559	59.73	18	48	1	1	18	86	15	Plot
26	Method 2	1	Yes	gpcp	58.79	9.88	9.88	0.1660	-7.57	0.0280	-0.8626	0.1813	72.11	42	34	32	32	42	182	42	
27	Method 2	1	Yes	gpc	62.11	-1.53	1.53	0.1116	-2.12	0.0114	-0.7097	0.1066	73.15	44	4	35	35	43	161	38	
28	Method 2	1	Yes	gpc trmm	49.87	-3.84	3.84	0.4766	0.28	0.1382	-0.0318	0.3690	62.96	22	18	8	8	22	78	12	
29	Method 2	1	Yes	ghcn cams grid	55.97	9.71	9.71	0.2601	-9.24	0.0437	-0.3367	0.2091	68.66	36	33	20	29	37	157	35	
30	Method 2	1	Yes	gpi	71.86	23.11	23.11	0.0596	-4.13	0.0053	-1.4833	0.0731	91.11	48	46	39	39	48	220	47	
31	Method 2	1	Yes	trmm 3b43	50.21	0.08	0.08	0.4679	-1.21	0.1497	-0.0439	0.3670	63.08	23	1	9	7	24	64	4	
32	Method 2	1	Yes	ssmi	48.78	-7.73	7.73	0.5190	2.77	0.1315	0.0017	0.3626	59.47	19	30	6	9	16	80	13	
33	Method 2	1	Yes	opi2	60.85	-7.58	7.58	-0.0546	-7.05	0.0020	-0.7496	-0.0443	73.86	43	29	47	47	44	210	46	
34	Method 2	1	Yes	opi3	58.15	-5.23	5.23	0.1028	-5.11	0.0078	-0.5966	0.0884	70.56	41	22	37	38	39	177	41	
35	Method 2	1	Yes	cmap	57.89	1.88	1.88	0.1176	-5.88	0.0106	-0.5864	0.1029	70.56	39	7	33	36	40	155	33	
36	Method 2	1	Yes	cmap2	57.90	1.88	1.88	0.1176	-5.88	0.0106	-0.5864	0.1029	70.56	40	6	34	37	41	158	36	
37	Method 2	2	Yes	reanalysis	48.79	22.00	22.00	0.5807	-14.80	0.1609	-0.0618	0.4011	60.84	20	44	4	5	20	93	17	Plot
38	Method 2	2	Yes	gpcp	52.54	10.58	10.58	0.3805	-7.02	0.1165	-0.2244	0.3413	65.33	29	38	11	11	36	125	26	
39	Method 2	2	Yes	gpc	52.73	-4.13	4.13	0.3360	2.18	0.0933	-0.2764	0.3054	63.73	30	21	16	15	25	107	24	
40	Method 2	2	Yes	gpc trmm	45.12	-1.17	1.17	0.7427	0.25	0.2400	0.2109	0.4899	59.86	9	2	2	4	19	36	1	
41	Method 2	2	Yes	ghcn cams grid	50.87	8.74	8.74	0.3810	-6.32	0.0903	-0.1700	0.3004	63.86	25	32	10	16	26	109	25	
42	Method 2	2	Yes	gpi	64.87	22.09	22.09	0.2012	-4.29	0.0566	-0.9706	0.2378	84.24	46	45	22	23	47	183	43	
43	Method 2	2	Yes	trmm 3b43	44.91	4.00	4.00	0.7223	-3.55	0.2599	0.2179	0.5098	59.60	7	20	3	3	17	50	2	Plot
44	Method 2	2	Yes	ssmi	49.19	-10.36	10.36	0.5444	6.14	0.1306	0.0105	0.3614	60.90	21	37	5	10	21	94	18	
45	Method 2	2	Yes	opi2	54.65	-5.47	5.47	0.1821	-2.96	0.0188	-0.3694	0.1372	69.09	36	23	25	33	38	155	32	
46	Method 2	2	Yes	opi3	50.66	-2.97	2.97	0.3560	-2.05	0.0780	-0.1799	0.2794	64.13	24	11	14	19	29	97	19	
47	Method 2	2	Yes	cmap	50.92	3.70	3.70	0.3850	-4.42	0.0854	-0.1769	0.2922	64.05	26	17	12	17	27	99	21	
48	Method 2	2	Yes	cmap2	50.92	3.70	3.70	0.3850	-4.42	0.0853	-0.1770	0.2921	64.05	27	16	13	18	28	102	22	

Southeast United States - Yearly Atmospheric																						
OI	Moist Flux	Re	Storage	Precipitation	Absolute Error	Bias	Abias	Slope	Oslope	Intercept	R <sup>2</sup>	R <sup>2</sup> 1:1	Corr Coef	RMSE	AER	BR	SR	CCR	RMR	TR	BR	
1	Method 1	1	Yes	reanalysis	31.17	31.17	31.17	0.6356	1.5733	-21.55	0.5440	-4.9137	0.7378	32.99	47	47	6	4	47	151	34	
2	Method 1	1	Yes	gpcp	12.19	2.98	2.98	0.0983	10.1729	-2.18	0.0225	-1.9837	0.1499	15.19	15	11	21	18	14	79	11	
3	Method 1	1	Yes	gpcc	10.98	-3.15	3.15	-0.0993		-3.06	0.0231	-2.9852	-0.1519	14.69	7	14	30	37	13	101	18	
4	Method 1	1	Yes	gpcc trmm	11.45	-11.45	11.45	1.5610	1.5610	17.19	0.8439	-0.2678	0.9186	12.88	10	37	5	1	7	80	6	
5	Method 1	1	Yes	ghcn cams grid	13.17	7.40	7.40	0.2769	3.6114	-5.89	0.0554	-0.6403	0.2353	16.81	19	28	14	15	20	96	17	
6	Method 1	1	Yes	gpi	21.74	16.68	16.68	0.0634	15.7729	-1.15	0.0122	-5.3605	0.1107	25.60	43	42	26	22	42	175	42	
7	Method 1	1	Yes	trmm 3b43	10.20	-10.20	10.20	1.7518	1.7518	13.59	0.6207	-0.0590	0.7878	13.95	6	34	7	3	8	58	5	
8	Method 1	1	Yes	ssmi	14.48	-8.94	8.94	0.0854	11.7096	1.12	0.0073	-1.5076	0.0856	17.23	26	31	22	26	21	126	22	
9	Method 1	1	Yes	opi2	9.15	-5.50	5.50	0.0995	10.0503	-1.33	0.0080	-1.0378	0.0895	12.55	5	24	20	25	6	80	12	
10	Method 1	1	Yes	opi3	11.90	-2.28	2.28	0.0712	14.0449	-1.78	0.0102	-1.7882	0.1008	14.69	14	9	25	23	12	83	13	
11	Method 1	1	Yes	cmmap	11.68	-2.06	2.06	0.0824	12.1359	-1.75	0.0134	-1.7013	0.1157	14.46	11	7	23	21	9	71	7	
12	Method 1	1	Yes	cmmap2	11.68	-2.06	2.06	0.0824	12.1359	-1.75	0.0134	-1.7014	0.1157	14.46	12	8	24	20	10	74	8	
13	Method 1	2	Yes	reanalysis	18.33	17.16	17.16	0.4051	2.4685	-5.61	0.1945	-2.5373	0.4411	21.22	39	43	11	9	32	134	24	
14	Method 1	2	Yes	gpcp	14.60	1.38	1.38	-0.0854		2.57	0.0109	-1.7684	-0.1045	18.78	27	2	29	30	30	118	21	
15	Method 1	2	Yes	gpcc	13.75	-8.27	8.27	-0.1456		2.26	0.0227	-2.2278	-0.1508	16.20	22	29	37	36	19	143	30	
16	Method 1	2	Yes	gpcc trmm	12.32	-12.32	12.32	1.4542	1.4542	15.08	0.3178	-0.7973	0.5638	15.66	16	40	3	6	18	83	14	
17	Method 1	2	Yes	ghcn cams grid	11.72	1.86	1.86	-0.0287		2.37	0.0007	-0.9371	-0.0268	15.71	13	5	28	28	17	91	16	
18	Method 1	2	Yes	gpi	17.81	13.96	13.96	0.1409	7.1023	1.67	0.0498	-3.8643	0.2233	21.45	37	41	17	16	33	144	32	
19	Method 1	2	Yes	trmm 3b43	7.86	-6.84	6.84	1.0852	1.0852	8.90	0.3204	-0.0162	0.5661	11.92	4	26	2	5	4	41	3	
20	Method 1	2	Yes	ssmi	15.00	-11.56	11.56	-0.1192		3.52	0.0128	-2.3941	-0.1132	18.84	29	38	32	31	31	181	37	
21	Method 1	2	Yes	opi2	13.86	-7.09	7.09	-0.3062		0.78	0.0723	-1.8397	-0.2690	18.33	23	27	42	41	26	159	36	
22	Method 1	2	Yes	opi3	15.23	-3.88	3.88	-0.1373		2.03	0.0236	-1.7168	-0.1537	18.60	32	21	35	40	29	157	35	
23	Method 1	2	Yes	cmmap	15.00	-3.66	3.66	-0.1300		2.07	0.0207	-1.6488	-0.1439	18.37	28	16	33	32	27	136	28	
24	Method 1	2	Yes	cmmap2	15.00	-3.66	3.66	-0.1301		2.07	0.0207	-1.6492	-0.1440	18.37	30	17	34	33	28	142	29	
25	Method 2	1	Yes	reanalysis	31.90	31.90	31.90	0.9997	1.0003	-31.89	0.6329	-1.7766	0.7956	34.24	48	48	1	2	48	147	33	Plot
26	Method 2	1	Yes	gpcp	15.30	9.96	9.96	-0.1377		-6.81	0.0220	-2.0985	-0.1484	22.34	33	33	36	34	35	171	40	
27	Method 2	1	Yes	gpcc	11.22	-1.65	1.65	0.4845	2.1529	-0.55	0.1876	-0.0809	0.4337	12.35	8	4	9	10	5	36	2	Plot
28	Method 2	1	Yes	gpcc trmm	5.35	1.96	1.96	0.1777	5.6275	-7.97	0.2211	-5.1465	0.4702	6.10	1	6	16	8	1	32	1	
29	Method 2	1	Yes	ghcn cams grid	18.35	10.46	10.46	0.5063	1.9751	-10.04	0.0628	-0.2659	0.2506	22.70	40	35	8	14	37	134	25	
30	Method 2	1	Yes	gpi	26.11	23.08	23.08	0.1362	7.3421	-5.64	0.0405	-5.0192	0.2013	30.57	46	46	18	17	46	173	41	
31	Method 2	1	Yes	trmm 3b43	11.30	0.10	0.10	0.3448	2.9002	-1.49	0.0161	-0.0419	0.1267	14.57	9	1	13	19	11	53	4	
32	Method 2	1	Yes	ssmi	14.47	-9.29	9.29	0.3538	2.8265	1.88	0.0887	-0.6620	0.2977	17.75	25	32	12	12	22	103	19	
33	Method 2	1	Yes	opi2	12.52	-6.09	6.09	0.1053	9.4967	-5.79	0.0037	-0.4958	0.0611	15.52	17	25	19	27	16	104	20	
34	Method 2	1	Yes	opi3	13.98	-3.59	3.59	-0.1828		-9.16	0.0223	-0.9920	-0.1493	17.91	24	15	41	35	23	138	28	
35	Method 2	1	Yes	cmmap	13.33	3.08	3.08	-0.1792		-7.93	0.0236	-1.0581	-0.1536	18.20	21	13	39	39	25	137	27	
36	Method 2	1	Yes	cmmap2	13.33	3.08	3.08	-0.1792		-7.93	0.0236	-1.0560	-0.1535	18.20	20	12	40	38	24	134	23	
37	Method 2	2	Yes	reanalysis	22.02	22.02	22.02	0.6581	1.5195	-16.15	0.2243	-1.6498	0.4736	26.62	44	45	4	7	43	143	31	Plot
38	Method 2	2	Yes	gpcp	18.73	10.60	10.60	-0.5120		-1.91	0.1959	-1.9331	-0.4428	28.00	41	36	43	45	44	209	48	
39	Method 2	2	Yes	gpcc	12.66	-4.25	4.25	0.4355	2.2982	2.59	0.1053	-0.1596	0.3244	15.42	18	22	10	11	15	76	10	
40	Method 2	2	Yes	gpcc trmm	6.88	-1.50	1.50	-0.1728		-2.68	0.0753	-3.5448	-0.2743	8.18	2	3	98	42	2	87	15	
41	Method 2	2	Yes	ghcn cams grid	15.17	8.76	8.76	-0.5907		-2.55	0.1165	-1.0152	-0.3413	23.21	31	30	44	43	38	186	44	
42	Method 2	2	Yes	gpi	24.07	21.98	21.98	0.1908	5.2411	-3.94	0.0654	-4.4019	0.2558	28.16	45	44	15	13	45	182	38	
43	Method 2	2	Yes	trmm 3b43	7.15	3.67	3.67	0.0517	19.3424	-2.15	0.0080	-3.5973	0.0895	8.23	3	18	27	24	3	75	9	
44	Method 2	2	Yes	ssmi	16.33	-12.13	12.13	-0.1190		0.54	0.0104	-1.7643	-0.1022	21.85	34	39	31	29	34	167	39	
45	Method 2	2	Yes	opi2	18.13	-5.44	5.44	-0.6686		-11.74	0.1503	-0.8964	-0.3876	22.52	38	23	45	44	36	186	43	
46	Method 2	2	Yes	opi3	18.94	-2.95	2.95	-0.7314		-10.56	0.2468	-1.1691	-0.4968	24.08	42	10	48	48	39	187	45	
47	Method 2	2	Yes	cmmap	16.88	3.72	3.72	-0.6982		-5.64	0.2360	-1.2123	-0.4858	24.32	35	20	47	47	40	189	47	
48	Method 2	2	Yes	cmmap2	16.89	3.72	3.72	-0.6980		-5.64	0.2360	-1.2125	-0.4858	24.32	36	19	46	46	41	188	46	

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OI	Runoff Meth	Re	Storage	Precipitation	Absolute Error	Bias	Abias	Slope	Intercept	R <sup>2</sup>	R <sup>2</sup> 1.1	Corr Coef	RMSE	AER	BR	SR	CCR	RMR	TR	BR	
1	Linear	1	Yes	reanalysis	52.52	-15.23	15.23	0.2896	1.07	0.0557	-0.3607	0.2360	62.25	259	135	51	747	238	830	183	Plot
2	Linear	1	Yes	gpcp	34.89	6.44	6.44	0.6438	7.42	0.4475	0.2954	0.6690	43.92	38	20	26	29	27	140	9	
3	Linear	1	Yes	gpcp	32.21	9.50	9.50	0.6256	7.96	0.5085	0.2853	0.7131	39.63	14	67	28	16	10	135	7	
4	Linear	1	Yes	gpcp trmm	38.25	25.66	25.66	0.9705	25.46	0.5986	0.4164	0.7737	45.98	69	256	3	5	51	384	41	
5	Linear	1	Yes	ghcn cams grid	31.14	7.02	7.02	0.8027	7.15	0.5139	0.4656	0.7169	39.08	9	31	13	15	6	74	2	
6	Linear	1	Yes	gpi	47.85	-9.57	9.57	0.3988	2.57	0.2973	-0.4127	0.5453	61.61	222	68	36	49	235	610	98	
7	Linear	1	Yes	trmm 3b43	36.60	20.93	20.93	0.8922	20.70	0.5825	0.4511	0.7632	44.22	51	197	5	8	29	290	29	
8	Linear	1	Yes	ssmi	46.44	20.81	20.81	0.3668	14.28	0.2826	-0.3175	0.2875	60.90	207	196	39	11	228	791	173	
9	Linear	1	Yes	opi2	33.48	22.82	22.82	0.8533	20.79	0.5580	0.3473	0.7470	41.84	23	226	9	11	21	290	27	
10	Linear	1	Yes	opi3	34.87	20.47	20.47	0.7641	17.75	0.5022	0.2981	0.7087	43.39	37	192	18	18	25	290	28	
11	Linear	1	Yes	cmcp	33.30	13.46	13.46	0.7457	12.31	0.4945	0.3699	0.7032	41.25	18	105	20	24	18	185	16	
12	Linear	1	Yes	cmcp2	33.30	13.46	13.46	0.7457	12.31	0.4945	0.3699	0.7032	41.25	19	106	19	23	19	186	17	
13	Linear	1	No	reanalysis	33.34	-15.11	15.11	0.0702	6.23	0.0525	-10.4388	0.2290	45.07	20	130	164	154	40	508	70	
14	Linear	1	No	gpcp	44.17	6.41	6.41	0.0736	8.97	0.0546	-8.7306	0.2337	53.43	176	19	158	151	182	686	127	
15	Linear	1	No	gpcp	44.46	9.36	9.36	0.0320	5.38	0.1127	-106.3622	0.3357	52.79	183	65	222	93	169	732	141	
16	Linear	1	No	gpcp trmm	44.59	26.60	26.60	0.2764	21.68	0.1939	-1.9137	0.4403	51.43	187	263	52	59	146	707	134	
17	Linear	1	No	ghcn cams grid	38.21	7.21	7.21	0.0797	7.79	0.0812	-11.0443	0.2849	46.33	67	33	146	123	54	423	48	
18	Linear	1	No	gpi	55.69	-9.37	9.37	0.0898	9.00	0.1078	-11.2144	0.3284	67.71	295	66	134	98	275	868	188	
19	Linear	1	No	trmm 3b43	43.39	21.85	21.85	0.2702	20.30	0.2131	-1.8764	0.4617	50.68	161	202	53	57	133	606	96	
20	Linear	1	No	ssmi	38.59	22.31	22.31	0.0573	12.59	0.0128	-4.5647	0.1130	49.73	72	214	185	202	117	790	172	
21	Linear	1	No	opi2	42.95	23.01	23.01	0.0903	10.38	0.0587	-7.7446	0.2423	49.99	151	228	131	142	122	774	157	
22	Linear	1	No	opi3	43.19	20.68	20.68	0.0841	10.10	0.0571	-8.2072	0.2390	51.30	158	195	142	146	142	783	163	
23	Linear	1	No	cmcp	41.76	13.67	13.67	0.0857	9.53	0.0613	-7.5631	0.2476	49.65	125	111	139	139	113	627	105	
24	Linear	1	No	cmcp2	41.76	13.68	13.68	0.0857	9.53	0.0613	-7.5637	0.2475	49.65	126	112	138	140	114	630	108	
25	Linear	1	mon	reanalysis	33.05	-15.01	15.01	-0.0012	7.85	0.0000	-11.9662	-0.0038	48.01	16	125	279	279	82	781	160	
26	Linear	1	mon	gpcp	42.70	6.58	6.58	0.1027	8.87	0.1059	-8.1313	0.3254	51.81	145	23	118	101	151	538	79	
27	Linear	1	mon	gpcp	43.85	9.65	9.65	0.0514	5.44	0.2909	-102.2625	0.5394	51.85	168	69	195	50	162	634	111	
28	Linear	1	mon	gpcp trmm	39.45	25.48	25.48	0.3343	20.56	0.3463	-1.8948	0.5885	46.56	81	254	46	39	60	480	60	
29	Linear	1	mon	ghcn cams grid	37.00	7.24	7.24	0.1067	7.76	0.1458	-10.3632	0.3818	44.99	56	34	117	65	38	310	32	
30	Linear	1	mon	gpi	54.04	-8.79	8.79	0.1023	8.81	0.1380	-10.7067	0.3715	66.45	277	59	121	68	264	789	171	
31	Linear	1	mon	trmm 3b43	36.55	21.81	21.81	0.3364	19.31	0.3822	-1.7518	0.6182	45.01	50	201	45	32	39	367	38	
32	Linear	1	mon	ssmi	38.08	22.21	22.21	0.0898	12.84	0.0317	-4.2881	0.1770	48.67	66	211	133	175	101	686	125	
33	Linear	1	mon	opi2	41.36	22.79	22.79	0.1334	10.92	0.1273	-7.0699	0.3573	48.08	120	225	85	79	84	593	93	
34	Linear	1	mon	opi3	41.74	20.41	20.41	0.1186	10.43	0.1131	-7.5873	0.3363	49.59	124	190	98	92	112	616	101	
35	Linear	1	mon	cmcp	40.20	13.35	13.35	0.1181	9.60	0.1154	-6.9426	0.3397	47.86	92	103	102	91	79	467	57	
36	Linear	1	mon	cmcp2	40.20	13.36	13.36	0.1181	9.60	0.1154	-6.9429	0.3397	47.86	93	104	101	90	80	468	58	
37	Linear	1	mon	reanalysis	34.10	-15.09	15.09	-0.0248	8.40	0.0065	-12.4484	-0.0809	48.92	30	129	294	302	107	862	186	
38	Linear	1	mon	gpcp	43.16	6.71	6.71	0.0933	8.88	0.0874	-8.3116	0.2956	52.40	157	25	126	118	161	587	92	
39	Linear	1	mon	gpcp	43.56	10.17	10.17	0.0498	5.41	0.2730	-102.9453	0.5225	51.85	163	78	196	52	153	642	114	
40	Linear	1	mon	gpcp trmm	39.50	23.83	23.83	0.2432	17.62	0.3345	-4.2866	0.5784	46.61	82	233	55	41	61	472	59	
41	Linear	1	mon	ghcn cams grid	37.32	7.06	7.06	0.0949	7.67	0.1178	-10.8929	0.3433	45.48	59	32	123	87	44	345	35	
42	Linear	1	mon	gpi	56.33	-8.80	8.80	0.0678	9.48	0.0606	-11.6170	0.2462	69.16	302	60	169	141	290	962	204	
43	Linear	1	mon	trmm 3b43	38.60	19.96	19.96	0.2293	16.82	0.3270	-4.3543	0.5719	46.49	73	184	56	43	58	414	46	
44	Linear	1	mon	ssmi	39.17	22.19	22.19	0.0607	12.49	0.0143	-4.5017	0.1195	49.81	77	209	180	197	118	781	161	
45	Linear	1	mon	opi2	40.95	22.63	22.63	0.1379	10.93	0.1361	-6.9650	0.3689	47.83	109	223	82	69	78	561	83	
46	Linear	1	mon	opi3	41.83	20.20	20.20	0.1141	10.34	0.1044	-7.6047	0.3230	49.71	128	188	108	103	115	642	113	
47	Linear	1	mon	cmcp	40.50	13.16	13.16	0.1106	9.53	0.1010	-7.0317	0.3177	48.20	97	101	112	109	88	507	69	
48	Linear	1	mon	cmcp2	40.50	13.16	13.16	0.1106	9.53	0.1009	-7.0319	0.3177	48.20	98	102	111	108	89	508	72	
49	Linear	1	mon	reanalysis	36.21	-15.14	15.14	0.0138	7.53	0.0020	-11.5632	0.0449	47.31	46	131	252	237	67	733	142	
50	Linear	1	mon	gpcp	45.30	6.76	6.76	0.0517	8.98	0.0288	-9.1458	0.1638	54.79	196	28	193	183	199	799	176	
51	Linear	1	mon	gpcp	44.86	10.03	10.03	0.0284	5.30	0.0895	-107.4327	0.2974	53.18	189	72	229	116	179	785	168	
52	Linear	1	mon	gpcp trmm	44.07	23.79	23.79	0.1217	15.87	0.0915	-6.1763	0.3025	52.00	172	231	93	112	155	763	154	
53	Linear	1	mon	ghcn cams grid	39.22	6.91	6.91	0.0540	7.62	0.0389	-12.1812	0.1973	47.42	78	30	191	163	69	531	76	
54	Linear	1	mon	gpi	58.55	-8.58	8.58	0.0282	10.33	0.0091	-12.6896	0.0953	72.20	327	55	230	212	315	1139	240	
55	Linear	1	mon	trmm 3b43	42.73	19.77	19.77	0.1340	15.45	0.1228	-6.0597	0.3504	51.12	149	177	84	81	140	631	109	
56	Linear	1	mon	ssmi	39.90	22.16	22.16	0.0199	12.10	0.0015	-4.8022	0.0392	51.30	87	207	237	244	143	918	199	
57	Linear	1	mon	opi2	43.13	22.43	22.43	0.0901	10.27	0.0578	-7.5781	0.2405	49.72	155	220	132	143	116	766	156	
58	Linear	1	mon	opi3	43.38	19.93	19.93	0.0699	9.82	0.0388	-8.2125	0.1970	51.53	159	183	165	164	148	819	180	
59	Linear	1	mon	cmcp	42.13	12.95	12.95	0.0650	9.33	0.0347	-7.7229	0.1864	50.32	136	95	176	170	129	706	131	
60	Linear	1	mon	cmcp2	42.14	12.95	12.95	0.0650	9.33	0.0347	-7.7232	0.1864	50.32	137	96	175	169	130	707	133	
61	Linear	1	mon	reanalysis	36.82	-15.05	15.05	0.0814	5.99	0.0702	-10.1299	0.2649	44.56	52	126	144	134	32	488	66	
62	Linear	1	mon	gpcp	47.33	6.61	6.61	0.0029	9.10	0.0001	-10.0970	0.0091	57.41	218	24	274	274	210	1000	215	
63	Linear	1	mon	gpcp	46.49	10.19	10.19	0.0068	5.23	0.0051	-112.8782	0.0715	54.53	208	79	268	225	194	974	208	
64	Linear	1	mon	gpcp trmm	46.40	22.20	22.20	0.0169	14.53	0.0017	-7.1853	0.0416	55.48	206	210	244	241	200	1101	238	
65	Linear	1	mon	ghcn cams grid	41.01	6.79															

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OI	Runoff Meth	Re	Storage	Precipitation	Absolute Error	Bias	Abias	Slope	Intercept	R <sup>2</sup>	R <sup>2</sup> 1.1	Corr Coef	RMSE	AER	BR	SR	CCR	RM	TR	BR
102	Linear	2	1 mon	gpi	78.14	16.93	16.93	-0.1923	43.49	0.1404	-5.4731	-0.3747	92.86	428	145	361	371	423	1728	357
103	Linear	2	1 mon	trmm 3b43	53.31	34.10	34.10	-0.1180	34.42	0.0408	-5.1415	-0.2019	68.40	267	318	334	342	282	1543	318
104	Linear	2	1 mon	ssmi	57.94	48.27	48.27	-0.0184	38.80	0.0005	-3.2783	-0.0213	72.74	319	407	290	286	319	1621	335
105	Linear	2	1 mon	opi2	66.51	46.34	46.34	-0.4013	31.82	0.2786	-4.9902	-0.5278	82.90	398	397	423	431	385	2034	422
106	Linear	2	1 mon	opi3	63.92	43.91	43.91	-0.3668	33.08	0.2518	-4.9289	-0.5018	82.47	380	374	418	423	380	1975	413
107	Linear	2	1 mon	cmap	61.94	37.23	37.23	-0.3469	35.55	0.2347	-4.5117	-0.4844	79.52	360	330	412	418	362	1882	397
108	Linear	2	1 mon	cmmap2	61.94	37.24	37.24	-0.3469	35.55	0.2347	-4.5119	-0.4844	79.52	361	331	411	417	363	1883	398
109	Linear	2	2 mon	reanalysis	39.72	19.03	19.03	0.1092	34.22	0.0174	-1.4536	0.1317	53.08	85	170	114	193	177	739	144
110	Linear	2	2 mon	gpcp	61.49	30.64	30.64	-0.2177	37.27	0.1161	-4.3326	-0.3407	78.26	357	293	369	361	356	1736	360
111	Linear	2	2 mon	gpcp	68.65	43.63	43.63	-0.3145	38.66	0.1794	-4.2734	-0.4236	87.26	408	371	404	389	405	1977	415
112	Linear	2	2 mon	gpcp trmm	54.87	39.13	39.13	-0.1257	32.54	0.0425	-5.4866	-0.2060	68.34	282	353	338	343	281	1597	329
113	Linear	2	2 mon	ghcn cams grid	58.06	32.44	32.44	-0.2813	37.11	0.1498	-3.8751	-0.3871	74.83	321	303	392	376	330	1725	354
114	Linear	2	2 mon	gpi	74.06	17.24	17.24	-0.1288	42.36	0.0638	-5.0571	-0.2525	89.53	422	149	342	355	414	1682	349
115	Linear	2	2 mon	trmm 3b43	55.48	33.94	33.94	-0.1370	33.18	0.0583	-5.5522	-0.2414	68.68	293	315	345	354	283	1590	327
116	Linear	2	2 mon	ssmi	57.36	48.24	48.24	-0.0399	39.21	0.0023	-3.2997	0.0477	70.97	312	406	210	232	303	1463	299
117	Linear	2	2 mon	opi2	63.92	46.45	46.45	-0.3385	32.58	0.1987	-4.7870	-0.4458	81.52	381	400	409	402	376	1968	411
118	Linear	2	2 mon	opi3	62.40	44.11	44.11	-0.2894	33.76	0.1572	-4.6571	-0.3964	80.60	364	376	395	380	370	1885	399
119	Linear	2	2 mon	cmmap	60.33	37.43	37.43	-0.2765	35.71	0.1494	-4.2534	-0.3865	77.67	349	336	389	374	352	1800	376
120	Linear	2	2 mon	cmmap2	60.33	37.44	37.44	-0.2766	35.71	0.1494	-4.2536	-0.3865	77.68	350	337	390	375	353	1805	377
121	Linear	2	3 mon	reanalysis	42.25	19.13	19.13	0.0114	35.95	0.0002	-1.7415	0.0138	56.19	139	171	257	270	207	1044	223
122	Linear	2	3 mon	gpcp	59.18	30.70	30.70	-0.0715	36.54	0.0125	-3.6196	-0.1119	72.94	336	294	308	314	323	1575	324
123	Linear	2	3 mon	gpcp	66.31	43.74	43.74	-0.1306	39.60	0.0311	-3.6174	-0.1762	81.78	397	373	343	333	377	1823	383
124	Linear	2	3 mon	gpcp trmm	53.35	38.88	38.88	0.0184	32.97	0.0009	-4.6895	0.0302	64.36	268	352	240	260	255	1375	281
125	Linear	2	3 mon	ghcn cams grid	55.20	32.51	32.51	-0.1099	36.55	0.0229	-3.2292	-0.1513	69.79	290	304	330	324	292	1540	317
126	Linear	2	3 mon	gpi	66.14	17.04	17.04	0.0320	38.92	0.0039	-3.8106	0.0627	79.87	396	146	221	229	365	1357	277
127	Linear	2	3 mon	trmm 3b43	53.10	33.47	33.47	0.0027	32.86	0.0000	-4.6579	0.0047	64.18	263	311	275	276	215	1376	282
128	Linear	2	3 mon	ssmi	58.49	48.22	48.22	0.0304	39.16	0.0013	-3.3365	0.0365	71.25	305	405	226	250	306	1492	306
129	Linear	2	3 mon	opi2	60.42	46.38	46.38	-0.1611	34.50	0.0449	-4.1572	-0.2120	77.07	351	398	354	345	349	1797	375
130	Linear	2	3 mon	opi3	59.99	44.14	44.14	-0.1189	35.20	0.0266	-4.0166	-0.1630	76.01	344	377	335	328	342	1726	355
131	Linear	2	3 mon	cmmap	57.50	37.49	37.49	-0.1095	36.00	0.0234	-3.6029	-0.1531	72.81	315	341	328	326	321	1631	336
132	Linear	2	3 mon	cmmap2	57.50	37.49	37.49	-0.1095	36.00	0.0234	-3.6032	-0.1531	72.81	316	342	327	325	322	1632	337
133	Linear	2	4 mon	reanalysis	45.57	19.27	19.27	-0.0851	37.60	0.0105	-2.0220	-0.1025	59.10	199	172	313	312	217	1213	252
134	Linear	2	4 mon	gpcp	54.03	30.53	30.53	0.0858	35.68	0.0180	-2.8300	0.1342	66.53	276	292	137	191	286	1162	245
135	Linear	2	4 mon	gpcp	61.11	44.19	44.19	0.1117	40.80	0.0229	-2.7716	0.1512	73.91	353	378	110	186	327	1354	276
136	Linear	2	4 mon	gpcp trmm	50.12	38.43	38.43	0.1165	33.90	0.0368	-4.0760	0.1913	61.10	243	351	106	166	230	1096	235
137	Linear	2	4 mon	ghcn cams grid	51.27	32.41	32.41	0.0773	35.87	0.0113	-2.5089	0.1063	63.68	249	302	152	207	248	1158	244
138	Linear	2	4 mon	gpi	57.38	16.77	16.77	0.1687	35.95	0.1089	-2.7485	0.3301	70.49	313	141	67	97	297	1151	198
139	Linear	2	4 mon	trmm 3b43	48.63	32.86	32.86	0.1092	33.26	0.0370	-3.8957	0.1924	60.00	236	308	113	165	222	1044	224
140	Linear	2	4 mon	ssmi	59.39	48.16	48.16	-0.0545	38.51	0.0043	-3.5673	-0.0653	73.25	339	404	300	299	325	1667	347
141	Linear	2	4 mon	opi2	58.06	46.30	46.30	0.0517	36.69	0.0046	-3.4053	0.0680	71.36	322	396	192	227	309	1446	295
142	Linear	2	4 mon	opi3	57.08	44.01	44.01	0.0646	36.67	0.0078	-3.3077	0.0884	70.56	310	375	177	217	298	1377	283
143	Linear	2	4 mon	cmmap	54.07	37.37	37.37	0.0659	36.24	0.0085	-2.9021	0.0920	67.16	278	334	173	214	270	1269	264
144	Linear	2	4 mon	cmmap2	54.07	37.37	37.37	0.0658	36.24	0.0084	-2.9026	0.0919	67.16	279	335	174	215	271	1274	265
145	Point	1	Yes	reanalysis	53.02	-13.78	13.78	0.2971	2.35	0.0556	-0.3191	0.2359	62.90	262	114	48	148	243	815	179
146	Point	1	Yes	gpcp	35.27	8.19	8.19	0.6726	9.10	0.4530	0.3230	0.6731	44.70	39	41	25	28	33	166	13
147	Point	1	Yes	gpcp	31.91	10.30	10.30	0.6412	8.82	0.5231	0.3120	0.7233	39.29	10	80	27	14	7	138	8
148	Point	1	Yes	gpcp trmm	40.81	29.55	29.55	1.0542	29.91	0.6049	0.3970	0.7778	50.51	106	287	4	4	132	533	77
149	Point	1	Yes	ghcn cams grid	31.10	8.46	8.46	0.8364	8.57	0.5298	0.4657	0.7278	39.34	7	48	10	13	8	86	4
150	Point	1	Yes	gpi	47.21	-7.52	7.52	0.4306	3.97	0.3150	-0.2551	0.5612	60.92	217	36	33	46	229	581	84
151	Point	1	Yes	trmm 3b43	39.11	24.85	24.85	0.9738	24.80	0.5943	0.4454	0.7709	48.03	76	248	2	6	83	415	47
152	Point	1	Yes	ssmi	48.31	23.13	23.13	0.3845	16.78	0.0815	-0.2977	0.2855	63.81	231	230	37	122	249	869	190
153	Point	1	Yes	opi2	34.29	24.56	24.56	0.8878	23.01	0.5601	0.3426	0.7484	43.61	32	243	6	10	26	317	33
154	Point	1	Yes	opi3	35.66	22.21	22.21	0.7989	19.87	0.5068	0.3032	0.7118	44.89	42	212	14	17	36	321	34
155	Point	1	Yes	cmmap	33.99	15.21	15.21	0.7781	14.21	0.4993	0.3792	0.7066	42.52	27	133	16	20	22	218	19
156	Point	1	Yes	cmmap2	34.00	15.22	15.22	0.7781	14.21	0.4993	0.3792	0.7066	42.52	28	134	15	19	23	219	20
157	Point	1	No	reanalysis	33.48	-13.86	13.86	0.0777	7.50	0.0419	-6.5591	0.2048	45.34	22	109	151	160	41	483	63
158	Point	1	No	gpcp	44.32	8.16	8.16	0.1024	10.64	0.0691	-5.3828	0.2628	53.53	181	40	119	136	183	659	116
159	Point	1	No	gpcp	43.78	10.16	10.16	0.0476	6.24	0.1541	-64.0054	0.3925	52.24	167	77	201	83	158	666	119
160	Point	1	No	gpcp trmm	46.89	30.49	30.49	0.3601	26.14	0.2172	-1.1447	0.4660	54.30	214	291	40	56	191	792	174
161	Point	1	No	ghcn cams grid	37.72	8.65	8.65	0.1134	9.21	0.1074	-6.7338	0.3278	45.93	64	56	109	99	49	377	40
162	Point	1	No	gpi	54.76	-7.33	7.33	0.1216	10.40	0.1292	-6.7123	0.3595	66.56	285	35	94	75	267	756	151
163	Point	1	No	trmm 3b43	45.61	25.78	25.78	0.3518	24.40	0.2386	-1.0631	0.4884	52.81	200	258	42	54	172	726	140
164	Point	1	No	ssmi	40.55	24.62	24.62	0.0750	15.08	0.0143	-3.0539	0.1197	52.50	100	244	154	196	165	859	185
165	Point	1	No	opi2	43.38	24.75	24.75	0.1248	12.60	0.0732	-4.9240	0.2708	50.92	160	247	91	128	134	760	153
166	Point	1	No	opi3	43.57</															

Southeast United States - Monthly Surface

OI	Runoff Meth	Re	Storage	Precipitation	Absolute Error	Bias	Abias	Slope	Intercept	R <sup>2</sup>	R <sup>2</sup> 1.1	Corr Coef	RMSE	AER	BR	SR	CCR	RMR	TR	BR
203	Point	1	3 mon	cmap	42.70	14.67	14.67	0.0853	11.13	0.0391	-4.9370	0.1976	51.35	146	118	141	162	144	711	137
204	Point	1	3 mon	cmmap2	42.70	14.67	14.67	0.0853	11.13	0.0390	-4.9373	0.1976	51.35	147	119	140	161	145	712	138
205	Point	1	4 mon	reanalysis	37.38	-13.64	13.64	0.1011	6.98	0.0708	-6.1926	0.2660	44.33	61	108	122	132	31	454	52
206	Point	1	4 mon	gpcp	48.10	8.34	8.34	0.0046	10.82	0.0001	-6.6355	0.0119	58.89	226	45	272	271	216	1030	219
207	Point	1	4 mon	gpcp	46.65	10.99	10.99	0.0097	8.05	0.0064	-69.5924	0.0802	54.68	210	90	261	220	198	979	210
208	Point	1	4 mon	gpcp trmm	48.14	24.65	24.65	0.0155	16.98	0.0010	-5.0736	0.0313	58.14	227	245	246	258	213	1189	247
209	Point	1	4 mon	ghcn cams grid	41.69	8.19	8.19	0.0146	9.04	0.0019	-8.6795	0.0430	50.35	123	42	250	239	131	785	167
210	Point	1	4 mon	gpi	62.90	-6.48	6.48	-0.0203	13.35	0.0035	-9.0518	-0.0596	76.67	372	21	293	295	345	1326	270
211	Point	1	4 mon	trmm 3b43	48.48	22.60	22.60	0.0449	17.20	0.0094	-5.1781	0.0969	58.13	233	222	202	211	212	1080	232
212	Point	1	4 mon	ssmi	42.64	24.71	24.71	-0.0179	14.12	0.0008	-3.4994	-0.0285	55.99	143	246	289	288	206	1172	246
213	Point	1	4 mon	opi2	46.57	23.99	23.99	0.0208	11.05	0.0020	-5.7436	0.0448	54.63	209	234	234	238	197	1112	239
214	Point	1	4 mon	opi3	46.23	21.51	21.51	0.0170	10.96	0.0015	-6.0480	0.0388	55.85	203	198	243	247	205	1096	234
215	Point	1	4 mon	cmmap	45.08	14.77	14.77	0.0146	10.86	0.0011	-5.6943	0.0339	54.63	191	120	249	253	195	1008	217
216	Point	1	4 mon	cmmap2	45.08	14.77	14.77	0.0146	10.86	0.0011	-5.6947	0.0338	54.63	192	121	248	254	196	1011	218
217	Point	2	Yes	reanalysis	<b>29.96</b>	26.18	26.18	<b>0.9819</b>	26.49	0.6706	0.3838	<b>0.8189</b>	<b>38.39</b>	4	259	1	3	4	271	25
218	Point	2	Yes	gpcp	60.29	37.60	37.60	-0.1073	43.96	0.0135	-2.0157	-0.1162	84.92	348	343	326	315	395	1727	356
219	Point	2	Yes	gpcp	68.44	50.14	50.14	-0.2273	46.53	0.0450	-2.0980	-0.2120	96.75	407	409	374	346	427	1963	410
220	Point	2	Yes	gpcp trmm	52.32	44.74	44.74	0.1023	41.54	0.0122	-2.1082	0.1105	72.58	258	383	120	203	371	1281	286
221	Point	2	Yes	ghcn cams grid	56.42	39.44	39.44	-0.0909	43.70	0.0075	-1.7220	-0.0865	80.68	304	355	318	306	372	1655	343
222	Point	2	Yes	gpi	78.58	24.95	24.95	-0.1582	50.77	0.0465	-2.6737	-0.2156	99.79	429	249	352	347	431	1808	379
223	Point	2	Yes	trmm 3b43	52.27	39.57	39.57	0.0670	41.07	0.0086	-2.0796	0.0774	72.25	257	356	172	222	316	1323	288
224	Point	2	Yes	ssmi	65.10	56.07	56.07	0.1187	47.91	0.0060	-1.6284	0.0928	84.31	385	427	97	213	391	1513	311
225	Point	2	Yes	opi2	65.86	53.64	53.64	-0.2193	41.09	0.0398	-2.3936	-0.1995	90.08	393	421	370	340	417	1941	406
226	Point	2	Yes	opi3	61.32	51.15	51.15	-0.1253	42.37	0.0141	-2.2166	-0.1187	87.70	356	412	337	317	408	1828	385
227	Point	2	Yes	cmmap	58.69	44.47	44.47	-0.1020	43.23	0.0097	-1.9517	-0.0986	84.02	329	379	324	310	389	1731	358
228	Point	2	Yes	cmmap2	58.69	44.48	44.48	-0.1020	43.23	0.0097	-1.9518	-0.0986	84.02	330	380	323	309	390	1732	359
229	Point	2	No	reanalysis	35.37	26.38	26.38	0.6017	33.22	0.3659	-0.2173	0.6049	44.76	40	260	31	33	35	399	44
230	Point	2	No	gpcp	68.43	37.80	37.80	-0.3230	45.41	0.1778	-3.6730	-0.4216	87.70	406	345	406	387	407	1951	409
231	Point	2	No	gpcp	75.79	50.40	50.40	-0.4313	46.18	0.2341	-3.5587	-0.4838	97.61	425	410	429	414	428	2106	429
232	Point	2	No	gpcp trmm	62.11	45.25	45.25	-0.2547	40.78	0.1178	-4.6233	-0.3433	78.22	362	392	381	362	355	1852	394
233	Point	2	No	ghcn cams grid	63.82	39.64	39.64	-0.3529	44.93	0.1641	-3.2019	-0.4051	83.17	379	357	414	383	387	1920	404
234	Point	2	No	gpi	82.96	25.06	25.06	-0.2622	53.19	0.1811	-4.3452	-0.4256	101.04	432	250	387	390	432	1891	401
235	Point	2	No	trmm 3b43	62.70	40.09	40.09	-0.2603	42.10	0.1408	-4.6380	-0.3753	78.33	389	361	386	372	357	1845	391
236	Point	2	No	ssmi	64.70	57.19	57.19	-0.0196	47.74	0.0003	-2.7319	-0.0186	82.54	384	432	292	284	381	1773	368
237	Point	2	No	opi2	72.16	53.85	53.85	-0.4723	38.69	0.2681	-4.0988	-0.5178	91.61	418	422	432	427	420	2119	431
238	Point	2	No	opi3	68.10	51.35	51.35	-0.3818	40.57	0.1901	-3.9021	-0.4360	89.83	405	413	421	397	415	2051	425
239	Point	2	No	cmmap	65.89	44.68	44.68	-0.3553	43.15	0.1713	-3.5339	-0.4139	86.39	394	381	416	385	401	1977	414
240	Point	2	No	cmmap2	65.89	44.68	44.68	-0.3553	43.15	0.1713	-3.5340	-0.4139	86.39	395	382	415	384	402	1978	416
241	Point	2	1 mon	reanalysis	39.82	26.57	26.57	0.3561	37.55	0.1280	-0.7183	0.3578	53.25	86	262	41	78	180	647	115
242	Point	2	1 mon	gpcp	68.71	38.06	38.06	-0.3413	45.53	0.1981	-3.7387	-0.4450	88.44	410	347	410	401	411	1979	417
243	Point	2	1 mon	gpcp	76.32	51.04	51.04	-0.4420	46.38	0.2467	-3.6257	-0.4967	98.31	426	411	430	420	429	2116	430
244	Point	2	1 mon	gpcp trmm	60.61	46.80	46.80	-0.1797	41.05	0.0564	-4.3496	-0.2375	77.01	352	402	360	353	348	1815	381
245	Point	2	1 mon	ghcn cams grid	65.23	39.85	39.85	-0.4058	45.16	0.2167	-3.3458	-0.4655	84.69	386	358	427	409	394	1974	412
246	Point	2	1 mon	gpi	82.73	25.32	25.32	-0.2222	52.54	0.1308	-4.1638	-0.3617	99.28	431	253	372	365	430	1851	393
247	Point	2	1 mon	trmm 3b43	58.78	41.64	41.64	-0.1604	41.97	0.0517	-4.2178	-0.2274	76.06	331	366	353	350	343	1743	362
248	Point	2	1 mon	ssmi	65.57	56.77	56.77	-0.0268	47.22	0.0047	-2.7633	-0.0257	82.31	390	431	296	287	379	1783	371
249	Point	2	1 mon	opi2	72.21	53.99	53.99	-0.4689	38.77	0.2644	-4.0961	-0.5142	91.71	419	424	431	425	421	2120	432
250	Point	2	1 mon	opi3	69.65	51.56	51.56	-0.4256	40.25	0.2360	-4.0221	-0.4857	91.04	413	414	428	419	419	2093	428
251	Point	2	1 mon	cmmap	67.38	44.88	44.88	-0.4021	43.12	0.2192	-3.6663	-0.4682	87.76	402	384	425	411	408	2030	419
252	Point	2	1 mon	cmmap2	67.38	44.88	44.88	-0.4021	43.12	0.2192	-3.6665	-0.4682	87.76	403	385	424	410	409	2031	420
253	Point	2	2 mon	reanalysis	44.06	26.66	26.66	0.1317	41.47	0.0175	-1.1740	0.1324	59.97	171	264	88	192	221	936	202
254	Point	2	2 mon	gpcp	65.83	38.27	38.27	-0.2442	45.04	0.1014	-3.4151	-0.3184	85.47	392	349	379	358	398	1786	396
255	Point	2	2 mon	gpcp	74.08	52.11	52.11	-0.3575	46.97	0.1617	-3.4811	-0.4021	96.32	423	418	417	382	426	2066	427
256	Point	2	2 mon	gpcp trmm	59.84	46.64	46.64	-0.1466	39.93	0.0388	-4.3610	-0.1969	75.81	342	401	348	339	341	1771	366
257	Point	2	2 mon	ghcn cams grid	63.00	40.07	40.07	-0.3200	44.88	0.1346	-3.1258	-0.3668	82.62	373	360	405	369	382	1899	400
258	Point	2	2 mon	gpi	78.01	25.69	25.69	-0.1414	51.09	0.0538	-3.7909	-0.2316	95.28	427	257	348	351	425	1806	378
259	Point	2	2 mon	trmm 3b43	60.24	41.45	41.45	-0.1614	40.88	0.0542	-4.3583	-0.2329	75.79	346	385	355	352	340	1758	363
260	Point	2	2 mon	ssmi	64.69	56.74	56.74	0.0328	47.66	0.0010	-2.7899	0.0324	80.60	383	430	219	256	399	1657	345
261	Point	2	2 mon	opi2	69.31	54.08	54.08	-0.3808	39.76	0.1748	-3.8882	-0.4178	89.93	411	426	420	388	416	2059	426
262	Point	2	2 mon	opi3	68.03	51.73	51.73	-0.3289	41.07	0.1409	-3.7775	-0.3754	88.91	404	416	407	373	412	2012	418
263	Point	2	2 mon	cmmap	65.44	45.06	45.06	-0.3140	43.29	0.1337	-3.4343	-0.3656	85.65	388	388	403	368	399	1946	407
264	Point	2	2 mon	cmmap2	65.44	45.06	45.06	-0.3140	43.29	0.1337	-3.4345	-0.3656	85.66	389	389	402	367	400	1947	408
265	Point	2	3 mon	reanalysis	46.95	26.77	26.77	0.0115	43.59	0.0001	-1.4204	0.0115	63.37	215	267	266	272	245	1255	259
266	Point	2	3 mon	gpcp	63.33	38.34	38.34	-0.0719	44.18	0.0088	-2.8323	-0.0938	79.73	374	3					





Southeast United States - Monthly Surface																				
QI	Runoff Meth	Re	Storage	Precipitation	Absolute Error	Bias	Abias	Slope	Intercept	R <sup>2</sup>	R <sup>2</sup> 1:1	Corr Coef	RMSE	AER	BR	SR	CCR	RMR	TR	BR
405	Area	2	2 mon	opi2	56.96	36.36	36.36	-0.2552	23.35	0.2048	-6.8367	-0.4525	70.45	308	324	382	404	295	1713	353
406	Area	2	2 mon	opi3	55.38	34.01	34.01	-0.2176	24.25	0.1611	-6.7112	-0.4014	69.88	292	316	368	381	293	1650	341
407	Area	2	2 mon	cmap	53.69	27.34	27.34	-0.2079	25.71	0.1531	-6.1945	-0.3913	67.50	273	277	364	378	273	1565	322
408	Area	2	2 mon	cmap2	53.70	27.34	27.34	-0.2079	25.71	0.1531	-6.1948	-0.3913	67.50	274	278	363	377	274	1566	323
409	Area	2	3 mon	reanalysis	37.09	9.04	9.04	0.0106	25.88	0.0003	-2.7177	0.0172	48.57	58	63	259	268	98	746	148
410	Area	2	3 mon	gpcc	53.55	20.61	20.61	-0.0588	26.38	0.0154	-5.6415	-0.1241	64.91	272	194	303	318	257	1344	274
411	Area	2	3 mon	gpcc	59.27	32.81	32.81	-0.1015	28.77	0.0339	-5.2969	-0.1841	71.07	337	307	322	335	305	1606	330
412	Area	2	3 mon	gpcc trmm	47.60	29.22	29.22	0.0067	23.23	0.0002	-6.9111	0.0147	56.63	221	285	289	269	208	1252	258
413	Area	2	3 mon	ghcn cams grid	49.62	22.42	22.42	-0.0881	26.38	0.0267	-4.8378	-0.1633	60.86	241	219	316	329	226	1331	271
414	Area	2	3 mon	gpi	62.42	6.12	6.12	0.0179	28.33	0.0022	-6.6955	0.0471	75.38	365	17	241	233	338	1194	250
415	Area	2	3 mon	trmm 3b43	47.59	23.80	23.80	-0.0052	23.19	0.0002	-7.0867	-0.0123	57.26	220	232	282	282	209	1225	253
416	Area	2	3 mon	ssmi	47.92	37.45	37.45	0.0242	28.34	0.0015	-4.5973	0.0389	60.48	223	338	233	246	225	1265	261
417	Area	2	3 mon	opi2	53.86	36.29	36.29	-0.1263	24.76	0.0501	-6.0123	-0.2239	66.70	275	323	339	348	268	1553	321
418	Area	2	3 mon	opi3	53.17	34.05	34.05	-0.0948	25.30	0.0306	-5.8825	-0.1750	66.08	265	317	320	332	262	1496	308
419	Area	2	3 mon	cmap	51.47	27.40	27.40	-0.0877	25.94	0.0273	-5.3532	-0.1651	63.49	252	279	315	330	246	1422	291
420	Area	2	3 mon	cmap2	51.47	27.40	27.40	-0.0877	25.94	0.0273	-5.3536	-0.1652	63.49	253	280	314	331	247	1425	292
421	Area	2	4 mon	reanalysis	39.45	9.18	9.18	-0.0653	27.18	0.0112	-3.1152	-0.1060	51.19	80	64	306	313	141	904	197
422	Area	2	4 mon	gpcc	49.34	20.44	20.44	0.0550	25.76	0.0134	-4.6030	0.1159	59.73	239	191	190	199	218	1037	221
423	Area	2	4 mon	gpcc	54.55	33.18	33.18	0.0748	29.64	0.0184	-4.1850	0.1356	64.52	280	309	155	190	256	1190	248
424	Area	2	4 mon	gpcc trmm	44.51	28.65	28.65	0.0761	23.92	0.0280	-6.1109	0.1675	53.88	185	282	153	182	186	988	212
425	Area	2	4 mon	ghcn cams grid	45.96	22.32	22.32	0.0476	25.89	0.0078	-3.8906	0.0883	55.80	202	215	200	218	203	1038	222
426	Area	2	4 mon	gpi	55.73	5.80	5.80	0.1214	26.07	0.1012	-5.2491	0.3182	67.93	297	13	95	107	276	788	170
427	Area	2	4 mon	trmm 3b43	43.89	23.08	23.08	0.0724	23.50	0.0293	-6.0716	0.1711	53.81	169	229	180	180	184	922	200
428	Area	2	4 mon	ssmi	50.60	37.36	37.36	-0.0436	27.82	0.0049	-4.9275	-0.0700	62.34	248	333	298	301	240	1420	290
429	Area	2	4 mon	opi2	51.29	36.21	36.21	0.0295	26.37	0.0027	-5.0138	0.0523	61.88	250	321	227	230	237	1265	262
430	Area	2	4 mon	opi3	50.47	33.92	33.92	0.0380	26.37	0.0049	-4.9501	0.0702	61.55	247	314	214	226	234	1235	256
431	Area	2	4 mon	cmap	48.19	27.28	27.28	0.0392	26.12	0.0054	-4.4325	0.0738	58.81	228	274	213	223	214	1152	242
432	Area	2	4 mon	cmap2	48.20	27.29	27.29	0.0392	26.12	0.0054	-4.4332	0.0737	58.82	229	275	212	224	215	1155	243

Southeast United States - Yearly Surface

OI	Runoff Meth	Re	Storage	Precipitation	Absolute Error	Bias	Abias	Slope	Oslope	Intercept	R <sup>2</sup>	R <sup>2</sup> 1.1	Corr Coef	RMSE	AER	BR	SR	CCR	AMR	TR	BR	Plot
1	Linear	1	Yes	reanalysis	19.18	-14.99	14.99	0.1810	5.5249	3.80	0.0528	-2.3821	0.2298	27.69	43	43	62	44	43	235	35	
2	Linear	1	Yes	gpcp	15.80	7.88	7.88	0.0089	112.3596	10.63	0.0001	-1.1716	0.0084	21.36	23	16	79	79	32	229	31	
3	Linear	1	Yes	gpcp	12.91	10.15	10.15	0.2667	3.7495	7.13	0.1835	-3.0031	0.4043	14.61	12	26	47	32	7	124	6	
4	Linear	1	Yes	gpcp trmm	34.47	34.47	34.47	1.9651	1.9651	41.53	0.3144	-2.0465	0.5608	39.80	102	108	35	19	102	366	76	
5	Linear	1	Yes	ghcn cams grid	12.41	7.11	7.11	0.0564	17.7305	7.92	0.0019	-0.8226	0.0433	17.55	9	12	74	77	17	189	19	
6	Linear	1	Yes	gpi	22.61	-8.13	8.13	-0.2005		16.10	0.0563	-2.2342	-0.2373	27.98	61	17	119	119	70	386	84	
7	Linear	1	Yes	trmm 3b43	25.27	25.27	25.27	1.7122	1.7122	26.78	0.1678	-1.1912	0.4096	32.44	78	84	29	31	86	308	54	
8	Linear	1	Yes	ssmi	24.59	24.43	24.43	0.2038	4.9068	15.31	0.0201	-2.4416	0.1419	30.88	74	81	58	65	81	359	75	
9	Linear	1	Yes	opi2	23.94	23.93	23.93	-0.1862		8.18	0.0089	-3.0805	-0.0946	29.27	69	79	118	98	76	440	98	
10	Linear	1	Yes	opi3	22.84	21.43	21.43	-0.1283		9.27	0.0084	-2.8311	-0.0918	28.37	64	69	108	97	71	409	93	
11	Linear	1	Yes	cmapp	18.00	14.76	14.76	-0.0771		10.34	0.0033	-1.6889	-0.0578	23.75	35	41	96	89	45	306	53	
12	Linear	1	Yes	cmapp2	18.00	14.76	14.76	-0.0773		10.33	0.0034	-1.6874	-0.0580	23.76	36	42	97	90	46	311	57	
13	Linear	1	No	reanalysis	16.96	-15.11	15.11	0.1790	5.5866	3.73	0.1992	-9.2975	0.4483	21.05	29	46	64	27	30	196	23	
14	Linear	1	No	gpcp	14.91	6.41	6.41	-0.0947		9.43	0.0224	-3.5224	-0.1498	18.40	19	10	100	112	18	259	42	
15	Linear	1	No	gpcp	12.20	9.36	9.36	0.0125	80.0000	5.30	0.0175	-188.3857	0.1324	14.43	8	23	78	67	6	182	17	
16	Linear	1	No	gpcp trmm	30.20	30.20	30.20	1.1544	1.1544	31.33	0.1717	-10.8927	0.6472	30.58	92	99	7	9	79	286	49	
17	Linear	1	No	ghcn cams grid	10.72	6.98	6.98	-0.0554		7.90	0.0070	-3.6413	-0.0835	14.26	3	11	90	96	5	205	25	
18	Linear	1	No	gpi	21.87	-9.37	9.37	-0.1754		14.35	0.1151	-6.0232	-0.3392	25.22	56	24	115	137	52	384	82	
19	Linear	1	No	trmm 3b43	21.85	21.85	21.85	1.2077	1.2077	22.29	0.3907	-4.2743	0.6250	23.26	55	71	12	17	40	195	22	
20	Linear	1	No	ssmi	24.60	23.91	23.91	0.1669	5.9916	14.36	0.0351	-6.2058	0.1874	27.70	75	78	65	52	69	339	67	
21	Linear	1	No	opi2	22.45	22.45	22.45	-0.4158		3.65	0.1251	-8.0568	-0.3537	26.04	59	73	139	141	58	470	108	
22	Linear	1	No	opi3	20.77	19.96	19.96	-0.1798		7.23	0.0464	-7.2715	-0.2155	24.89	50	63	117	118	49	397	87	
23	Linear	1	No	cmapp	16.31	13.28	13.28	-0.1224		8.67	0.0236	-4.3209	-0.1538	19.96	26	37	106	113	22	304	51	
24	Linear	1	No	cmapp2	16.31	13.28	13.28	-0.1226		8.67	0.0237	-4.3225	-0.1541	19.96	27	38	107	114	23	309	55	
25	Linear	2	Yes	reanalysis	29.43	17.40	17.40	1.9797	1.9797	0.59	0.3406	0.0671	0.5836	38.55	91	53	36	18	99	297	50	Plot
26	Linear	2	Yes	gpcp	39.83	28.82	28.82	0.6174	1.6197	31.02	0.0478	-0.4920	0.2187	48.76	115	96	23	45	120	399	89	
27	Linear	2	Yes	gpcp	40.99	40.99	40.99	0.8455	1.1827	40.53	0.1434	-2.8241	0.3787	46.57	118	124	10	34	118	440	90	
28	Linear	2	Yes	gpcp trmm	23.58	15.69	15.69	4.4020	4.4020	27.82	0.9761	0.0590	0.9880	26.32	67	48	53	2	59	229	32	
29	Linear	2	Yes	ghcn cams grid	40.56	30.66	30.66	0.6878	1.4539	31.88	0.0265	-0.5691	0.1628	50.00	117	101	16	61	125	420	96	
30	Linear	2	Yes	gpi	22.11	12.10	12.10	0.9614	1.0401	12.96	0.2847	0.1133	0.5336	27.56	57	35	2	21	65	180	16	
31	Linear	2	Yes	trmm 3b43	20.50	10.52	10.52	3.5018	3.5018	6.51	0.7335	0.2089	0.8565	24.14	48	30	45	6	47	176	15	
32	Linear	2	Yes	ssmi	46.36	41.82	41.82	0.1830	5.4645	33.36	0.0052	-2.2484	0.0724	51.42	130	125	61	71	127	514	119	
33	Linear	2	Yes	opi2	52.47	44.87	44.87	0.1075	9.3023	35.68	0.0007	-1.3079	0.0255	60.64	139	133	71	78	142	563	128	
34	Linear	2	Yes	opi3	50.11	42.37	42.37	0.6105	1.6380	39.33	0.0289	-1.1099	0.1699	57.98	134	127	25	59	137	482	112	
35	Linear	2	Yes	cmapp	44.63	35.70	35.70	0.6038	1.6562	35.25	0.0296	-0.7831	0.1721	53.30	123	111	27	56	129	446	100	
36	Linear	2	Yes	cmapp2	44.63	35.70	35.70	0.6030	1.6584	35.25	0.0296	-0.7833	0.1713	53.30	124	112	28	57	130	451	103	
37	Linear	2	No	reanalysis	18.75	18.75	18.75	0.8632	1.1585	21.09	0.7211	-1.7533	0.8492	19.85	40	59	8	8	21	136	10	
38	Linear	2	No	gpcp	32.33	30.17	30.17	-0.2475		37.34	0.0855	-8.4484	-0.2925	36.77	96	98	124	126	97	541	123	
39	Linear	2	No	gpcp	42.08	42.08	42.08	-0.1874		38.82	0.0101	-14.5005	-0.1007	45.18	120	126	103	100	114	563	127	
40	Linear	2	No	gpcp trmm	37.81	37.81	37.81	0.3129	3.1959	35.36	0.0531	-21.1024	0.2304	38.89	109	118	41	43	100	411	94	
41	Linear	2	No	ghcn cams grid	32.19	32.01	32.01	-0.3630		37.33	0.0822	-8.2365	-0.2867	36.35	95	104	136	123	95	553	124	
42	Linear	2	No	gpi	17.92	16.72	16.72	0.2434	4.1085	33.58	0.1178	-3.1267	0.3432	23.40	33	51	50	37	42	213	27	
43	Linear	2	No	trmm 3b43	32.64	32.64	32.64	-0.1294		34.45	0.0108	-16.3871	-0.1039	34.49	97	106	109	102	92	506	118	
44	Linear	2	No	ssmi	49.25	49.25	49.25	0.0041	243.9024	38.93	0.0000	-29.2687	0.0050	51.37	132	136	82	81	126	557	125	
45	Linear	2	No	opi2	46.21	46.21	46.21	-0.4698		31.08	0.1386	-18.1435	-0.3723	49.53	129	135	142	144	121	671	142	
46	Linear	2	No	opi3	43.72	43.72	43.72	-0.3604		33.10	0.1120	-14.8413	-0.3347	47.61	122	130	135	136	119	642	140	
47	Linear	2	No	cmapp	37.30	37.04	37.04	-0.3233		35.55	0.0946	-11.0800	-0.3075	41.57	107	115	132	127	104	585	131	
48	Linear	2	No	cmapp2	37.31	37.04	37.04	-0.3235		35.55	0.0947	-11.0817	-0.3078	41.58	108	116	133	128	105	590	133	
49	Point	1	Yes	reanalysis	18.25	-13.54	13.54	0.2156	4.6382	4.48	0.0690	-1.8602	0.2626	22.71	39	39	55	40	36	209	36	
50	Point	1	Yes	gpcp	16.84	9.64	9.64	-0.0014		12.41	0.0000	-1.1848	-0.0013	22.75	28	25	83	83	37	256	40	
51	Point	1	Yes	gpcp	13.34	10.95	10.95	0.2770	3.6101	7.98	0.1781	-3.3058	0.4221	15.08	15	31	46	29	9	130	9	
52	Point	1	Yes	gpcp trmm	38.98	38.98	38.98	2.2368	2.2368	48.03	0.4152	-2.6898	0.6444	43.39	113	121	38	14	111	397	88	
53	Point	1	Yes	ghcn cams grid	13.19	8.55	8.55	0.0604	16.5563	9.36	0.0020	-0.8755	0.0445	18.56	13	21	73	76	19	202	24	
54	Point	1	Yes	gpi	22.91	-6.09	6.09	-0.2308		18.75	0.0655	-1.9308	-0.2558	28.43	65	9	123	120	73	390	85	
55	Point	1	Yes	trmm 3b43	29.20	29.20	29.20	2.0396	2.0396	31.40	0.2322	-1.5594	0.4819	35.50	90	97	37	24	94	342	69	

Southwest United States - Monthly Atmospheric																					
OI	Moist Flux	Re	Storage	Precipitation	Absolute Error	Bias	Slope	Oslope	Interp	R^2	R^2 1:1	Corr Coef	RMSE	AER	BR	SR	CCR	RMR	TR	BR	
1	Method 1	1	Yes	reanalysis	109.91	107.83	-0.2400		-112.64	0.0044	-4.0513	-0.0662	122.11	12	21	25	26	12	96	18	
2	Method 1	1	Yes	gpcp	109.34	107.47	0.4399	2.2732	-107.12	0.0218	-3.8049	0.1475	120.98	11	20	12	12	10	65	12	
3	Method 1	1	Yes	gpcc	108.29	106.66	0.9374	1.0668	-106.95	0.0804	-3.5034	0.2835	119.57	10	19	2	4	9	44	9	
4	Method 1	1	Yes	gpcc trmm	86.76	82.89	0.9855	1.0747	-82.95	0.1082	-2.4091	0.3289	96.48	1	2	1	3	1	8	1	Plot
5	Method 1	1	Yes	ghcn cams grid	112.99	111.74	0.5490	1.8215	-111.81	0.0310	-4.2247	0.1761	124.12	16	28	11	10	13	78	13	
6	Method 1	1	Yes	ssmi	100.90	99.36	0.8481	1.1791	-100.09	0.1676	-3.3447	0.4094	110.59	4	10	3	2	2	21	2	
7	Method 1	1	Yes	opi2	104.66	101.85	0.5500	1.8182	-103.57	0.0261	-3.3311	0.1615	115.98	5	13	10	11	4	43	7	
8	Method 1	1	Yes	opi3	105.34	102.94	0.6120	1.6340	-104.00	0.0393	-3.3880	0.1981	116.74	6	14	8	9	5	42	5	
9	Method 1	1	Yes	cmap	105.69	103.79	0.6148	1.6271	-104.82	0.0413	-3.5579	0.2032	117.06	7	16	7	8	6	44	8	
10	Method 1	1	Yes	cmap2	106.15	104.32	0.6374	1.5889	-105.10	0.0421	-3.5912	0.2051	117.48	9	17	6	7	8	47	10	
11	Method 1	2	Yes	reanalysis	115.69	113.27	-0.3346		-127.04	0.0133	-3.7657	-0.1154	130.93	21	31	29	30	20	131	28	
12	Method 1	2	Yes	gpcp	116.12	114.19	-0.2850		-126.26	0.0096	-3.8105	-0.0979	131.54	22	32	28	28	21	131	29	
13	Method 1	2	Yes	gpcc	111.86	110.14	-0.2849		-122.46	0.0055	-3.7366	-0.0744	125.82	14	25	27	27	14	107	24	
14	Method 1	2	Yes	gpcc trmm	95.96	93.23	0.1887	5.3562	-107.63	0.0030	-2.6118	0.0546	110.77	2	4	14	14	3	37	3	
15	Method 1	2	Yes	ghcn cams grid	114.18	112.21	-0.1358		-125.13	0.0020	-3.6389	-0.0447	129.17	20	30	20	22	19	111	25	
16	Method 1	2	Yes	ssmi	105.72	104.43	0.7471	1.3385	-107.72	0.0687	-3.5958	0.2582	117.10	8	18	4	5	7	42	6	
17	Method 1	2	Yes	opi2	112.38	110.02	-0.1446		-125.55	0.0023	-3.5066	-0.0479	127.32	15	24	22	24	15	100	19	
18	Method 1	2	Yes	opi3	113.27	111.26	-0.1481		-125.41	0.0023	-3.5764	-0.0478	128.30	17	26	23	23	16	105	23	
19	Method 1	2	Yes	cmap	113.33	111.33	-0.1364		-125.26	0.0019	-3.5777	-0.0439	128.32	18	27	21	21	17	104	21	
20	Method 1	2	Yes	cmap2	113.81	111.87	-0.1151		-124.93	0.0013	-3.6045	-0.0367	128.69	19	29	19	19	18	104	22	
21	Method 2	1	Yes	reanalysis	122.29	95.68	-1.7972		-109.81	0.1091	-1.0212	-0.3304	146.18	28	6	38	37	28	137	30	Plot
22	Method 2	1	Yes	gpcp	124.66	103.48	-0.1925		-107.77	0.0016	-1.0653	-0.0403	148.38	30	15	24	20	29	118	27	
23	Method 2	1	Yes	gpcc	130.30	108.81	-0.2789		-119.49	0.0030	-1.1284	-0.0548	153.66	32	23	26	25	32	138	31	
24	Method 2	1	Yes	gpcc trmm	99.73	73.67	0.4242	2.3574	-78.23	0.0080	-0.5913	0.0894	121.54	3	1	13	13	11	41	4	
25	Method 2	1	Yes	ghcn cams grid	117.98	96.47	0.1246	8.0257	-100.03	0.0006	-0.9124	0.0254	142.13	23	8	15	15	23	84	14	
26	Method 2	1	Yes	ssmi	110.25	94.81	1.7477	1.7477	-86.22	0.1867	-0.7398	0.4321	132.10	13	5	9	1	22	50	11	
27	Method 2	1	Yes	opi2	120.69	96.09	-0.5362		-111.23	0.0104	-0.9373	-0.1018	144.00	25	7	30	29	25	116	26	
28	Method 2	1	Yes	opi3	120.03	97.03	-0.0253		-106.17	0.0000	-0.9224	-0.0051	143.45	24	9	18	18	24	93	15	
29	Method 2	1	Yes	cmap	121.10	99.86	0.0473	21.1416	-106.72	0.0001	-0.9841	0.0098	144.70	26	11	16	16	26	95	17	
30	Method 2	1	Yes	cmap2	121.27	100.06	0.0448	22.3214	-106.75	0.0001	-0.9873	0.0092	144.81	27	12	17	17	27	100	20	
31	Method 2	2	Yes	reanalysis	141.88	121.65	-1.6077		-152.85	0.1147	-1.5301	-0.3386	166.97	40	40	37	40	40	197	40	Plot
32	Method 2	2	Yes	gpcp	140.67	121.35	-1.3854		-150.61	0.0918	-1.5167	-0.3029	166.52	39	39	34	35	39	186	38	
33	Method 2	2	Yes	gpcc	137.50	120.77	-1.9738		-154.28	0.1040	-1.4277	-0.3224	165.31	35	38	40	36	38	187	39	
34	Method 2	2	Yes	gpcc trmm	124.42	90.81	-1.8823		-148.81	0.1133	-0.8282	-0.3366	149.36	29	3	39	39	31	141	32	
35	Method 2	2	Yes	ghcn cams grid	138.93	119.94	-1.2597		-150.85	0.0716	-1.4645	-0.2677	164.79	38	37	31	31	37	174	37	
36	Method 2	2	Yes	ssmi	125.05	108.71	1.4128	1.4128	-101.40	0.0591	-1.0219	0.2431	149.02	31	22	5	6	30	94	16	
37	Method 2	2	Yes	opi2	137.02	115.36	-1.5931		-162.70	0.1116	-1.3920	-0.3341	162.35	33	33	36	38	34	174	35	
38	Method 2	2	Yes	opi3	137.04	116.47	-1.4329		-158.19	0.0850	-1.3911	-0.2915	162.32	34	34	35	34	33	170	33	
39	Method 2	2	Yes	cmap	137.91	118.25	-1.3355		-154.14	0.0750	-1.4235	-0.2739	163.41	36	35	32	33	35	171	34	
40	Method 2	2	Yes	cmap2	138.06	118.45	-1.3413		-153.95	0.0749	-1.4268	-0.2737	163.52	37	36	33	32	36	174	36	

Southwest United States - Yearly Atmospheric																					
Ol	Moist Flux	Re	Storage	Precipitation	Absolute Error	Bias	Slope	Oslope	Intercept	R <sup>2</sup>	R <sup>2</sup> 1:1	Corr Coef	RMSE	AER	BR	SR	CCR	RMR	TR	BR	
1	Method 1	1	Yes	reanalysis	107.83	107.83	-0.4656		-113.52	0.0082	-37.3528	-0.0907	109.37	21	21	30	31	21	124	28	
2	Method 1	1	Yes	gpcp	107.49	107.49	-0.8810		-106.30	0.0565	-34.3443	-0.2377	109.36	20	20	34	38	20	132	32	
3	Method 1	1	Yes	gpc	106.69	106.69	-0.3989		-113.06	0.0066	-47.8695	-0.0814	107.88	19	19	27	30	19	114	23	
4	Method 1	1	Yes	gpc trmm	88.23	88.23	-5.1026		-105.67	0.9748	-131.2197	-0.9873	88.71	2	2	40	40	2	86	11	
5	Method 1	1	Yes	ghcn cams grid	112.48	112.48	-1.2732		-112.47	0.0643	-44.3854	-0.2537	113.92	30	30	39	39	28	166	40	
6	Method 1	1	Yes	ssmi	98.58	98.58	1.9696	1.9696	-93.22	0.1320	-34.1942	0.3633	99.88	9	9	15	11	9	53	4	Plot
7	Method 1	1	Yes	opi2	103.32	103.32	-0.1569		-107.41	0.0010	-31.5970	-0.0314	105.02	13	13	26	26	14	92	15	
8	Method 1	1	Yes	opi3	104.56	104.56	-0.9883		-109.15	0.0511	-32.4612	-0.2260	106.41	16	16	38	37	16	123	26	
9	Method 1	1	Yes	cmap	104.62	104.62	-0.8867		-108.84	0.0408	-32.4919	-0.2014	106.46	17	17	36	36	17	123	27	
10	Method 1	1	Yes	cmap2	105.17	105.17	-0.9055		-108.39	0.0375	-32.8128	-0.1936	106.97	18	18	37	35	18	126	30	
11	Method 1	2	Yes	reanalysis	113.30	113.30	0.8677	1.1525	-114.67	0.0337	-24.2445	0.1836	115.54	31	31	4	19	31	116	24	
12	Method 1	2	Yes	gpc	114.22	114.22	0.3433	2.9129	-120.39	0.0058	-24.6886	0.0764	116.55	32	32	18	21	32	135	37	
13	Method 1	2	Yes	gpc	110.16	110.16	2.9608	2.8608	-94.23	0.4265	-34.1639	0.6531	111.33	25	25	17	4	24	95	17	
14	Method 1	2	Yes	gpc trmm	93.12	93.12	8.7952	8.7952	44.96	0.8408	-38.0845	0.9170	94.11	4	4	24	1	4	37	2	
15	Method 1	2	Yes	ghcn cams grid	112.25	112.25	1.6961	1.6961	-104.33	0.1605	-23.6939	0.4006	114.27	29	29	12	8	30	108	21	
16	Method 1	2	Yes	ssmi	103.51	103.51	0.4786	2.0894	-110.90	0.0410	-72.3145	0.2025	104.23	15	15	16	17	13	76	7	
17	Method 1	2	Yes	opi2	110.05	110.05	1.4313	1.4313	-104.20	0.1937	-22.7277	0.4401	112.01	24	24	8	6	25	87	13	
18	Method 1	2	Yes	opi3	111.29	111.29	0.8388	1.1922	-113.28	0.0394	-23.3859	0.1985	113.56	26	26	5	18	26	101	20	
19	Method 1	2	Yes	cmap	111.36	111.36	0.9128	1.0955	-112.43	0.0446	-23.4076	0.2112	113.61	27	27	2	16	27	99	18	
20	Method 1	2	Yes	cmap2	111.90	111.90	1.0473	1.0473	-111.35	0.0576	-23.6228	0.2399	114.11	28	28	1	14	29	100	19	
21	Method 2	1	Yes	reanalysis	95.69	95.69	0.5438	1.8389	-97.99	0.0105	-19.1790	0.1026	98.14	6	6	14	20	6	52	3	Plot
22	Method 2	1	Yes	gpc	103.50	103.50	-0.0009		-107.10	0.0000	-28.3123	-0.0002	105.42	14	14	25	25	15	93	16	
23	Method 2	1	Yes	gpc	108.84	108.84	-0.5453		-121.74	0.0129	-47.4137	-0.1134	110.08	23	23	32	32	22	132	33	
24	Method 2	1	Yes	gpc trmm	75.72	75.72	-0.8002		-86.55	0.0023	-26.7286	-0.0478	77.14	1	1	33	27	1	63	6	
25	Method 2	1	Yes	ghcn cams grid	97.09	97.09	-0.8838		-104.43	0.0158	-19.5782	-0.1256	99.68	7	7	35	33	8	90	14	
26	Method 2	1	Yes	ssmi	94.07	94.07	0.2488	4.0193	-102.87	0.0015	-18.9426	0.0389	96.58	5	5	20	23	5	58	5	
27	Method 2	1	Yes	opi2	97.52	97.52	-0.4430		-111.35	0.0058	-25.1376	-0.0759	99.55	8	8	29	28	7	80	9	
28	Method 2	1	Yes	opi3	98.63	98.63	-0.4139		-110.61	0.0065	-25.7256	-0.0809	100.66	10	10	28	29	10	87	12	
29	Method 2	1	Yes	cmap	100.41	100.41	0.1205	8.2988	-106.29	0.0006	-26.6231	0.0252	102.34	11	11	23	24	11	80	10	
30	Method 2	1	Yes	cmap2	100.61	100.61	0.2287	4.3725	-105.62	0.0022	-26.7209	0.0468	102.52	12	12	21	22	12	79	8	
31	Method 2	2	Yes	reanalysis	121.68	121.68	1.1189	1.1189	-120.25	0.0471	-21.5104	0.2171	124.34	40	40	3	15	40	138	39	Plot
32	Method 2	2	Yes	gpc	121.37	121.37	1.3936	1.3936	-116.54	0.0940	-21.3621	0.3066	123.93	39	39	7	13	39	137	38	
33	Method 2	2	Yes	gpc	120.79	120.79	3.0877	3.0877	-97.26	0.5800	-33.7380	0.7818	121.99	38	38	19	2	37	134	36	
34	Method 2	2	Yes	gpc trmm	90.69	90.69	5.2120	5.2120	-5.93	0.4552	-21.0596	0.6747	92.47	3	3	22	3	3	34	1	
35	Method 2	2	Yes	ghcn cams grid	119.96	119.96	1.7307	1.7307	-109.97	0.1928	-20.7958	0.4391	122.35	37	37	13	7	38	132	35	
36	Method 2	2	Yes	ssmi	108.67	108.67	-0.4707		-136.20	0.0222	-30.6286	-0.1488	110.79	22	22	31	34	23	132	34	
37	Method 2	2	Yes	opi2	115.39	115.39	1.5524	1.5524	-105.30	0.2180	-19.1949	0.4669	117.77	33	33	9	5	33	113	22	
38	Method 2	2	Yes	opi3	116.49	116.49	1.3883	1.3883	-108.84	0.1195	-19.6494	0.3457	119.09	34	34	6	12	34	120	25	
39	Method 2	2	Yes	cmap	118.27	118.27	1.5833	1.5833	-109.31	0.1436	-20.2439	0.3789	120.79	35	35	10	10	35	125	29	
40	Method 2	2	Yes	cmap2	118.48	118.48	1.6716	1.6716	-108.30	0.1572	-20.3065	0.3965	120.97	36	36	11	9	36	128	31	

Southwest United States - Monthly Surface																								
OI	Runoff Meth	Re	Storage	Precipitation	Absolute Error	Bias	Slope	Oslope	Intercept	R <sup>2</sup>	R <sup>2</sup> 1-1	Corr Coef	RMSE	AER	BR	SF	CCR	RMR	TR	BR	Plot			
1	Linear	1	Yes	reanalysis	49.89	4.75	2.9576	2.9576	14.63	0.7649	0.3245	0.7649	60.07	331	45	53	7	334	770	111		Plot		
2	Linear	1	Yes	gpcp	56.93	3.35	2.6489	2.6489	9.28	0.5195	0.3164	0.7208	65.74	346	13	40	24	342	765	106				
3	Linear	1	Yes	gpcp	57.46	8.00	2.9292	2.9292	24.10	0.5630	0.3090	0.7503	67.17	352	114	49	16	356	887	185				
4	Linear	1	Yes	gpcp trmm	49.93	5.24	2.1780	2.1780	14.58	0.4105	0.2847	0.6407	58.39	333	60	31	45	331	800	125				
5	Linear	1	Yes	ghcn cams grid	52.37	3.59	2.3758	2.3758	9.17	0.4632	0.3055	0.6806	60.99	337	16	34	37	338	762	101				
6	Linear	1	Yes	ssmi	64.35	9.17	0.5359	1.8660	3.97	0.0310	-0.0071	0.1760	75.65	359	151	30	114	359	1013	254				
7	Linear	1	Yes	opi2	55.78	8.67	3.1191	3.1191	29.55	0.5990	0.3105	0.7740	65.72	341	136	55	5	341	878	179				
8	Linear	1	Yes	opi3	56.67	7.73	2.7365	2.7365	23.20	0.5217	0.3021	0.7223	66.12	343	110	46	22	347	868	172				
9	Linear	1	Yes	cmcp	57.28	6.22	2.6366	2.6366	18.00	0.5024	0.3027	0.7086	66.29	349	72	37	29	349	836	149				
10	Linear	1	Yes	cmcp2	57.46	6.02	2.6578	2.6578	17.62	0.4994	0.2993	0.7067	66.45	353	70	43	32	352	850	158				
11	Linear	1	No	reanalysis	15.25	5.00	-0.0077		-0.09	0.0222	-411.0841	-0.1492	19.71	9	48	257	283	13	610	43				
12	Linear	1	No	gpcp	17.17	3.62	-0.0129		-0.03	0.0506	-322.4739	-0.2249	22.24	102	18	288	318	71	797	122				
13	Linear	1	No	gpcp	16.89	8.33	-0.0032		-0.05	0.0146	-999.0000	-0.1210	22.38	56	121	229	272	86	764	104				
14	Linear	1	No	gpcp trmm	16.72	8.07	-0.0394		-0.18	0.1290	-102.6784	-0.3591	22.69	44	118	329	346	120	957	219				
15	Linear	1	No	ghcn cams grid	16.87	4.01	-0.0079		-0.09	0.0293	-488.2283	-0.1711	21.53	54	29	259	298	31	671	66				
16	Linear	1	No	ssmi	22.60	11.24	-0.0075		-0.06	0.0143	-306.0153	-0.1194	27.41	261	191	256	271	195	1174	311				
17	Linear	1	No	opi2	16.11	9.86	-0.0173		-0.16	0.0764	-328.5881	-0.2764	22.31	27	165	294	326	79	891	186				
18	Linear	1	No	opi3	17.14	8.92	-0.0129		-0.10	0.0479	-349.0483	-0.2189	23.00	96	144	287	316	147	990	241				
19	Linear	1	No	cmcp	17.32	7.21	-0.0122		-0.08	0.0447	-340.8596	-0.2115	22.81	124	97	284	311	129	945	212				
20	Linear	1	No	cmcp2	17.12	7.01	-0.0120		-0.07	0.0423	-332.1884	-0.2057	22.51	87	92	282	306	101	868	171				
21	Linear	1	1	reanalysis	15.31	5.02	-0.0022		-0.07	0.0019	-407.0055	-0.0436	19.63	12	49	224	226	11	522	19				
22	Linear	1	1	gpcp	17.29	3.65	-0.0091		-0.01	0.0252	-320.2105	-0.1588	22.20	119	20	265	286	65	755	94				
23	Linear	1	1	gpcp	17.10	8.38	-0.0038		-0.05	0.0220	-999.0000	-0.1482	22.48	83	124	238	282	95	822	137				
24	Linear	1	1	gpcp trmm	17.22	8.27	-0.0298		-0.09	0.0737	-101.7288	-0.2715	22.71	109	119	311	324	122	985	238				
25	Linear	1	1	ghcn cams grid	16.98	4.05	-0.0078		-0.08	0.0281	-487.7550	-0.1677	21.52	64	32	258	296	30	680	70				
26	Linear	1	1	ssmi	22.66	11.23	-0.0104		-0.09	0.0272	-307.2559	-0.1649	27.54	263	189	271	292	200	1215	327				
27	Linear	1	1	opi2	16.34	9.90	-0.0136		-0.12	0.0473	-326.8067	-0.2175	22.29	31	167	289	314	73	874	176				
28	Linear	1	1	opi3	17.30	8.95	-0.0097		-0.07	0.0270	-347.2758	-0.1642	22.97	120	146	269	291	144	970	225				
29	Linear	1	1	cmcp	17.38	7.12	-0.0093		-0.05	0.0256	-336.5193	-0.1599	22.70	130	94	268	288	121	901	190				
30	Linear	1	1	cmcp2	17.15	6.92	-0.0089		-0.05	0.0229	-327.7729	-0.1514	22.40	97	87	263	285	90	822	136				
31	Linear	1	2	reanalysis	15.25	5.04	0.0018	555.5556	-0.05	0.0012	-403.7474	0.0345	19.57	8	51	188	175	9	431	13				
32	Linear	1	2	gpcp	17.15	3.69	-0.0011		0.02	0.0004	-315.2060	-0.0190	22.07	99	22	206	209	57	593	38				
33	Linear	1	2	gpcp	17.10	8.51	-0.0020		-0.05	0.0064	-999.0000	-0.0798	22.51	81	129	219	248	99	774	113				
34	Linear	1	2	gpcp trmm	17.22	8.55	-0.0100		0.03	0.0083	-99.8571	-0.0910	22.48	110	132	270	259	93	864	167				
35	Linear	1	2	ghcn cams grid	16.96	4.07	-0.0039		-0.06	0.0072	-490.5439	-0.0847	21.46	67	34	239	252	28	620	50				
36	Linear	1	2	ssmi	22.74	11.24	-0.0081		-0.07	0.0164	-305.8762	-0.1280	27.56	267	190	260	277	201	1135	318				
37	Linear	1	2	opi2	16.28	9.93	-0.0033		-0.02	0.0028	-321.8128	-0.0532	22.15	30	170	231	229	60	720	81				
38	Linear	1	2	opi3	17.28	8.96	-0.0040		-0.02	0.0046	-344.0146	-0.0678	22.90	118	148	240	241	138	885	183				
39	Linear	1	2	cmcp	17.20	6.98	-0.0033		-0.01	0.0031	-328.2521	-0.0561	22.45	106	89	232	232	92	751	91				
40	Linear	1	2	cmcp2	16.99	6.78	-0.0031		-0.01	0.0028	-319.7405	-0.0534	22.16	69	82	226	230	63	670	65				
41	Linear	1	3	reanalysis	15.19	5.06	0.0065	153.8482	-0.02	0.0158	-400.1978	0.1257	19.50	7	54	180	140	7	388	9				
42	Linear	1	3	gpcp	17.11	3.72	0.0013	769.2308	0.03	0.0006	-314.0496	0.0235	22.05	84	24	190	184	53	535	24				
43	Linear	1	3	gpcp	17.07	8.65	-0.0007		-0.04	0.0008	-999.0000	-0.0274	22.55	78	135	197	177	108	735	87				
44	Linear	1	3	gpcp trmm	16.74	9.27	-0.0034		0.13	0.0009	-98.6267	-0.0301	22.24	45	154	234	219	70	722	82				
45	Linear	1	3	ghcn cams grid	16.92	4.08	-0.0019		-0.05	0.0017	-490.9125	-0.0409	21.43	62	35	217	225	25	564	31				
46	Linear	1	3	ssmi	22.99	11.29	-0.0065		-0.04	0.0105	-305.7687	-0.1025	27.61	277	195	250	264	205	1191	317				
47	Linear	1	3	opi2	16.09	9.85	-0.0009		0.00	0.0022	-318.0832	-0.0141	22.05	25	163	198	205	54	645	58				
48	Linear	1	3	opi3	17.27	8.88	-0.0015		0.00	0.0007	-340.4558	-0.0258	22.81	116	142	213	214	131	816	133				
49	Linear	1	3	cmcp	17.10	6.86	-0.0012		0.01	0.0004	-323.1967	-0.0201	22.31	82	85	208	210	76	661	63				
50	Linear	1	3	cmcp2	16.89	6.66	-0.0012		0.01	0.0004	-314.8114	-0.0210	22.02	58	79	209	211	46	603	40				
51	Linear	1	4	reanalysis	15.17	5.08	0.0092	108.6957	-0.01	0.0020	-397.1494	0.1789	19.44	5	56	173	112	5	351	7				
52	Linear	1	4	gpcp	17.22	3.73	-0.0011		0.02	0.0004	-315.7857	-0.0189	22.15	112	25	205	208	59	609	42				
53	Linear	1	4	gpcp	17.12	8.81	-0.0007		-0.04	0.0007	-999.0000	-0.0265	22.61	85	141	196	215	115	752	92				
54	Linear	1	4	gpcp trmm	17.40	9.34	-0.0151		-0.07	0.0202	-110.3023	-0.1420	22.60	132	155	290	280	114	971	226				
55	Linear	1	4	ghcn cams grid	16.90	4.03	-0.0031		-0.06	0.0047	-493.0501	-0.0683	21.44	60	30	227	243	27	587	35				
56	Linear	1	4	ssmi	23.08	11.47	-0.0065		-0.04	0.0104	-305.9593	-0.1021	27.64	264	198	249	263	206	1200	322				
57	Linear	1	4	opi2	16.01	9.73	-0.0010		0.00	0.0003	-314.1493	-0.0164	21.95	23	157									

Southwest United States - Monthly Surface																				
OL	Runoff Meth	Re	Storage	Precipitation	Absolute Error	Bias	Slope	Oslope	Intercept	R <sup>2</sup>	R <sup>2</sup> 1:1	Corr Coef	RMSE	AER	BR	SR	CCR	RMR	TR	BR
102	Linear	2	3 mon	gpcp	21.64	14.65	0.1620	6.1728	4.29	0.0515	-2.1215	0.2270	29.06	228	247	86	82	226	869	173
103	Linear	2	3 mon	gpcp	16.84	12.85	0.0739	13.5318	2.15	0.0297	-7.5960	0.1724	21.89	48	207	135	117	41	548	27
104	Linear	2	3 mon	gpcp trmm	28.78	27.78	0.3129	3.1959	13.26	0.1536	-3.8155	0.3919	34.29	327	360	58	57	319	1121	296
105	Linear	2	3 mon	ghcn cams grid	21.86	16.08	0.1615	6.1920	4.52	0.0483	-2.2097	0.2198	29.47	240	266	87	92	235	920	201
106	Linear	2	3 mon	ssmi	25.25	21.18	-0.0719		2.24	0.0139	-6.7229	-0.1178	30.83	316	343	346	270	265	1540	346
107	Linear	2	3 mon	opi2	24.22	20.68	0.1473	6.7889	5.00	0.0391	-2.8521	0.1978	32.29	303	330	91	106	300	1130	300
108	Linear	2	3 mon	opi3	23.61	19.53	0.1600	6.2500	5.05	0.0436	-2.5650	0.2087	31.06	297	316	89	102	274	1078	277
109	Linear	2	3 mon	cmcp	22.59	17.74	0.1607	6.2228	4.77	0.0446	-2.3364	0.2113	30.05	260	301	88	101	249	999	247
110	Linear	2	3 mon	cmcp2	22.36	17.53	0.1621	6.1690	4.76	0.0450	-2.2926	0.2120	29.85	257	287	85	100	245	974	231
111	Linear	2	4 mon	reanalysis	21.99	14.40	0.0856	11.6822	3.36	0.0133	-2.2713	0.1154	29.79	248	238	125	149	240	1000	249
112	Linear	2	4 mon	gpcp	22.12	14.66	0.0704	14.2045	3.19	0.0097	-2.4824	0.0987	30.74	254	249	137	161	263	1064	271
113	Linear	2	4 mon	gpcp	17.03	12.67	0.0595	16.8067	1.71	0.0231	-9.1795	0.1520	21.81	71	204	141	123	38	577	33
114	Linear	2	4 mon	gpcp trmm	28.17	27.01	0.2490	4.0161	11.11	0.1101	-4.0603	0.3318	34.13	325	357	60	61	318	1121	295
115	Linear	2	4 mon	ghcn cams grid	22.48	16.12	0.0812	12.3153	3.44	0.0122	-2.5113	0.1106	30.86	259	269	127	152	268	1075	274
116	Linear	2	4 mon	ssmi	25.17	21.52	0.0018	555.5556	3.87	0.0000	-7.0922	0.0032	30.11	313	345	189	191	252	1290	337
117	Linear	2	4 mon	opi2	24.75	20.71	0.0889	11.5075	3.92	0.0136	-3.0733	0.1168	33.24	306	331	124	147	313	1221	330
118	Linear	2	4 mon	opi3	24.25	19.55	0.0801	12.4844	3.70	0.0109	-2.8388	0.1045	32.27	304	318	129	156	299	1206	324
119	Linear	2	4 mon	cmcp	23.15	17.76	0.0776	12.8866	3.52	0.0104	-2.6253	0.1021	31.36	285	302	132	159	282	1160	308
120	Linear	2	4 mon	cmcp2	22.91	17.56	0.0785	12.7389	3.52	0.0105	-2.5805	0.1027	31.17	274	288	131	158	278	1128	298
121	Point	1	Yes	reanalysis	49.91	4.78	2.9580	2.9580	14.66	0.5847	0.3243	0.7647	60.11	332	46	54	8	335	775	115
122	Point	1	Yes	gpcp	57.01	3.36	2.6500	2.6500	9.30	0.5192	0.3161	0.7206	65.80	347	14	41	26	344	772	112
123	Point	1	Yes	gpcp	57.51	8.01	2.9298	2.9298	24.12	0.5627	0.3087	0.7501	67.22	355	115	50	17	357	894	188
124	Point	1	Yes	gpcp trmm	50.34	5.06	2.1898	2.1898	14.49	0.4114	0.2846	0.6414	58.65	336	53	32	44	333	798	124
125	Point	1	Yes	ghcn cams grid	52.40	3.81	2.3763	2.3763	9.20	0.4631	0.3053	0.6805	61.02	338	17	35	38	339	767	108
126	Point	1	Yes	ssmi	64.37	9.21	0.5368	1.8629	4.01	0.0311	-0.0070	0.1763	75.68	360	152	29	113	360	1014	256
127	Point	1	Yes	opi2	55.86	8.68	3.1210	3.1210	29.58	0.5989	0.3103	0.7739	65.78	342	137	56	6	343	884	182
128	Point	1	Yes	opi3	56.76	7.74	2.7379	2.7379	23.23	0.5215	0.3018	0.7222	66.18	345	111	47	23	348	874	177
129	Point	1	Yes	cmcp	57.37	6.23	2.6376	2.6376	18.02	0.5021	0.3024	0.7086	66.35	351	73	38	30	351	843	155
130	Point	1	Yes	cmcp2	57.57	6.03	2.6585	2.6585	17.64	0.4990	0.2990	0.7064	66.51	357	71	44	33	354	859	164
131	Point	1	No	reanalysis	15.26	5.02	-0.0072		-0.06	0.0147	-305.5142	-0.1214	19.72	11	50	254	273	14	602	39
132	Point	1	No	gpcp	17.17	3.63	-0.0118		-0.01	0.0304	-230.1909	-0.1745	22.24	103	19	278	301	69	770	110
133	Point	1	No	gpcp	16.90	8.34	-0.0025		-0.03	0.0056	-999.0000	-0.0751	22.37	59	123	225	245	85	737	88
134	Point	1	No	gpcp trmm	16.59	7.88	-0.0276		-0.27	0.0692	-109.8985	-0.2630	22.40	37	112	308	323	91	871	174
135	Point	1	No	ghcn cams grid	16.89	4.04	-0.0074		-0.06	0.0190	-362.6621	-0.1377	21.53	57	31	255	279	32	654	60
136	Point	1	No	ssmi	22.63	11.27	-0.0066		-0.02	0.0082	-226.5685	-0.0906	27.42	262	193	251	258	196	1160	307
137	Point	1	No	opi2	16.10	9.88	-0.0154		-0.13	0.0438	-236.3873	-0.2094	22.30	26	166	291	310	74	867	169
138	Point	1	No	opi3	17.14	8.94	-0.0115		-0.08	0.0275	-251.2406	-0.1660	22.99	93	145	276	294	145	953	216
139	Point	1	No	cmcp	17.33	7.23	-0.0112		-0.06	0.0273	-245.4680	-0.1653	22.80	125	98	274	293	127	917	199
140	Point	1	No	cmcp2	17.13	7.03	-0.0113		-0.05	0.0268	-239.3119	-0.1636	22.52	89	93	275	290	102	849	157
141	Point	1	mon	reanalysis	15.34	5.04	-0.0021		-0.04	0.0133	-302.6579	-0.0360	19.64	13	52	223	223	12	523	20
142	Point	1	mon	gpcp	17.37	3.66	-0.0123		-0.01	0.0330	-230.3740	-0.1816	22.28	131	21	286	303	72	813	130
143	Point	1	mon	gpcp	17.14	8.39	-0.0056		-0.06	0.0280	-999.0000	-0.1672	22.52	92	125	244	295	103	859	163
144	Point	1	mon	gpcp trmm	17.27	8.06	-0.0305		-0.30	0.0845	-110.2167	-0.2906	22.65	117	117	312	332	117	995	244
145	Point	1	mon	ghcn cams grid	17.00	4.08	-0.0091		-0.06	0.0283	-363.4154	-0.1681	21.56	70	37	264	297	33	701	78
146	Point	1	mon	ssmi	22.69	11.26	-0.0108		-0.07	0.0218	-227.8223	-0.1478	27.58	266	192	272	281	202	1213	325
147	Point	1	mon	opi2	16.39	9.91	-0.0161		-0.13	0.0476	-236.6278	-0.2181	22.35	34	168	292	315	83	892	187
148	Point	1	mon	opi3	17.35	8.96	-0.0120		-0.08	0.0301	-251.4530	-0.1736	23.04	128	147	281	300	149	1005	250
149	Point	1	mon	cmcp	17.41	7.13	-0.0122		-0.06	0.0318	-243.8346	-0.1782	22.77	133	95	285	302	125	940	208
150	Point	1	mon	cmcp2	17.21	6.93	-0.0119		-0.06	0.0296	-237.6439	-0.1721	22.48	108	88	279	299	94	868	170
151	Point	1	mon	reanalysis	15.26	5.07	0.0023	434.7826	-0.01	0.0015	-299.9664	0.0393	19.57	10	55	186	172	10	433	14
152	Point	1	mon	gpcp	17.23	3.70	-0.0043		0.02	0.0250	-226.8461	-0.0633	22.15	113	23	241	239	61	677	69
153	Point	1	mon	gpcp	17.12	8.50	-0.0037		-0.07	0.0152	-999.0000	-0.1233	22.54	86	127	237	274	106	830	146
154	Point	1	mon	gpcp trmm	17.34	8.33	-0.0165		-0.25	0.0255	-111.9457	-0.1597	22.50	127	122	293	287	98	927	205
155	Point	1	mon	ghcn cams grid	16.98	4.10	-0.0050		-0.04	0.0089	-367.0796	-0.0945	21.49	68	39	242	260	29	638	55
156	Point	1	mon	ssmi	22.78	11.28	-0.0085		-0.03	0.0135	-227.2325	-0.1162	27.60	269	194	262	269	204	1198	320
157	Point	1	mon	opi2	16.37	9.94	-0.0055		-0.03	0.0056	-232.8297	-0.0748	22.20	33	171	243	244	66	757	95
158	Point	1	mon	opi3	17.35	8.97	-0.0063		-0.03	0.0082	-249.1256	-0.0906	22.97	129	149	248	257	143	926	204
159	Point	1	mon	cmcp	17.26	6.99	-0.0060		-0.02	0.0076	-237.9410	-0.0874	22.53	115	91	245	254	105	810	129
160	Point	1	mon	cmcp2	17.05	6.79	-0.0060		-0.02	0.0074	-231.8148	-0.0860	22.24	72	84	246	253	68	723	84
161	Point	1	mon	reanalysis	15.18	5.09	0.0062	161.2903	0.01	0.0110	-297.7710	0.1047	19.52	6	57	182	155	8	408	12
162	Point	1	mon	gpcp	17.14	3.73	0.0022	454.5455	0.04	0.0111	-224.3103	0.0332	22.05	95	26	187	176	52	536	25
163	Point	1	mon	gpcp	17.09	8.64	-0.0014		-0.05	0.0021	-999.0000	-0.0462	22.57	80	134	211	228	109	762	100
164	Point	1	mon	gpcp trmm	16.67	9.09	-0.0082		-0.10	0.0068	-118.7033	-0.0811	22.23	41	150	261	247	67	766	107
165	Point	1	mon	ghcn cams grid	16.91	4.10	-0.0013		-0.02	0.0008	-364.7504	-0.0248	21.43	61	40	210	213	26	550	28
166	Point	1	mon	ssmi	23.03	11.33	-0.0072		-0.01	0.0098	-227.9335	-0.0991	27.66	280	197	253	262	207	1199	321
167	Point	1	mon	opi2	16.11	9.86	0.0001	10000.0000	0.03	0.0000	-229.1964	0.0019	22.06							



Southwest United States - Monthly Surface																				
OI	Runoff Meth	Re	Storage	Precipitation	Absolute Error	Bias	Slope	Oslope	Intercept	R <sup>2</sup>	R <sup>2</sup> 1:1	Corr Coef	RMSE	AER	BR	SR	CCR	RMR	TR	BR
304	Area	2	Yes	gpcc lrmm	19.74	18.49	0.9035	1.1068	16.55	0.4637	-0.0393	0.6810	26.72	182	309	16	36	185	728	86
305	Area	2	Yes	ghcn cams grid	17.07	11.32	1.0795	1.0795	12.41	0.5689	0.4399	0.7542	23.89	77	196	13	14	156	456	16
306	Area	2	Yes	ssmi	30.26	16.56	-0.1807		-4.35	0.0153	-1.0318	-0.1236	37.58	328	274	352	275	325	1554	347
307	Area	2	Yes	opi2	19.54	15.90	1.0965	1.0965	17.66	0.5719	0.3193	0.7562	26.34	177	259	14	13	179	642	56
308	Area	2	Yes	opi3	18.70	14.79	1.1335	1.1335	17.08	0.5749	0.3522	0.7582	25.69	165	253	18	12	165	613	46
309	Area	2	Yes	cmmap	17.76	13.01	1.1263	1.1263	14.96	0.5771	0.4036	0.7597	24.65	146	212	17	10	159	544	26
310	Area	2	Yes	cmmap2	17.62	12.81	1.1375	1.1375	14.90	0.5825	0.4130	0.7632	24.46	139	206	19	9	157	530	23
311	Area	2	No	reanalysis	18.42	10.65	0.0986	10.1420	-0.13	0.0494	-5.2555	0.2222	24.54	162	173	112	88	158	693	76
312	Area	2	No	gpcc	18.44	10.96	0.1016	9.8425	-0.06	0.0565	-5.6083	0.2378	25.22	163	182	109	72	161	687	72
313	Area	2	No	gpcc	14.88	10.73	0.0584	17.1233	-0.12	0.0765	-28.3939	0.2766	19.85	2	176	143	65	15	401	11
314	Area	2	No	gpcc lrmm	22.83	20.99	-0.1089		-1.33	0.2627	-51.9985	-0.5126	30.55	272	338	349	357	258	1574	349
315	Area	2	No	ghcn cams grid	18.40	12.36	0.1099	9.0992	0.19	0.0825	-5.8203	0.2499	25.24	161	200	102	67	162	692	75
316	Area	2	No	ssmi	21.82	17.67	-0.0338		-0.63	0.0267	-47.3743	-0.1634	25.95	237	294	315	289	175	1310	341
317	Area	2	No	opi2	19.67	16.94	0.1103	9.0662	0.70	0.0612	-6.9075	0.2475	27.59	180	277	101	68	203	829	144
318	Area	2	No	opi3	19.60	15.83	0.1116	8.9606	0.60	0.0590	-6.2852	0.2429	26.48	179	256	99	71	182	787	120
319	Area	2	No	cmmap	18.79	14.05	0.1121	8.9206	0.41	0.0605	-5.7887	0.2460	25.56	166	227	98	69	164	724	85
320	Area	2	No	cmmap2	18.56	13.85	0.1127	8.8731	0.40	0.0605	-5.6853	0.2480	25.37	164	223	96	70	163	716	80
321	Area	2	1 mon	reanalysis	19.23	10.68	0.0689	14.5138	-0.49	0.0241	-5.5572	0.1553	25.17	170	175	138	121	160	764	105
322	Area	2	1 mon	gpcc	19.32	10.98	0.0728	13.7363	-0.42	0.0290	-5.9231	0.1703	25.86	173	184	136	118	173	794	119
323	Area	2	1 mon	gpcc	15.65	10.78	0.0090	111.1111	-0.44	0.0018	-30.6040	0.0424	20.65	16	178	175	169	18	558	30
324	Area	2	1 mon	gpcc lrmm	22.66	21.23	-0.0584		-0.31	0.0752	-49.6782	-0.2742	30.18	264	344	344	325	254	1531	345
325	Area	2	1 mon	ghcn cams grid	18.83	12.38	0.0880	11.3636	-0.11	0.0400	-5.8444	0.2001	25.71	167	201	123	104	166	761	98
326	Area	2	1 mon	ssmi	21.57	17.66	-0.0234		-0.41	0.0129	-47.0278	-0.1134	25.85	223	293	300	266	172	1254	335
327	Area	2	1 mon	opi2	20.21	16.98	0.0811	12.3305	0.17	0.0331	-7.1997	0.1820	28.14	188	278	128	111	212	917	200
328	Area	2	1 mon	opi3	20.00	15.85	0.0885	11.2994	0.21	0.0371	-6.4985	0.1926	26.91	184	257	121	108	187	857	160
329	Area	2	1 mon	cmmap	19.26	14.07	0.0887	11.2740	0.05	0.0379	-6.0095	0.1947	26.02	171	228	119	107	176	801	126
330	Area	2	1 mon	cmmap2	19.02	13.86	0.0907	11.0254	0.06	0.0392	-5.8904	0.1981	25.80	168	224	118	105	169	784	118
331	Area	2	2 mon	reanalysis	19.46	10.78	0.0415	24.0964	-0.80	0.0087	-5.8342	0.0933	25.74	178	177	156	162	167	838	151
332	Area	2	2 mon	gpcc	19.54	11.03	0.0439	22.7790	-0.76	0.0105	-6.2428	0.1027	26.50	178	186	154	157	183	858	161
333	Area	2	2 mon	gpcc	15.83	10.87	-0.0067		-0.67	0.0010	-31.9781	-0.0320	20.92	19	180	252	222	19	692	74
334	Area	2	2 mon	gpcc lrmm	22.96	21.72	0.0338	29.5858	1.65	0.0250	-46.5218	0.1582	29.32	276	346	159	120	230	1131	303
335	Area	2	2 mon	ghcn cams grid	19.26	12.45	0.0631	15.8479	-0.44	0.0205	-6.1065	0.1432	26.25	172	202	139	130	177	820	134
336	Area	2	2 mon	ssmi	21.63	17.63	-0.0214		-0.38	0.0108	-46.8501	-0.1038	25.84	227	292	298	265	171	1253	333
337	Area	2	2 mon	opi2	20.88	17.01	0.0479	20.8768	-0.43	0.0116	-7.5346	0.1075	28.76	196	279	151	154	220	1000	248
338	Area	2	2 mon	opi3	20.50	15.89	0.0568	17.6056	-0.33	0.0153	-6.8026	0.1236	27.50	190	258	145	142	199	934	207
339	Area	2	2 mon	cmmap	19.68	14.11	0.0579	17.2712	-0.41	0.0161	-6.3105	0.1269	26.62	181	229	144	139	184	877	178
340	Area	2	2 mon	cmmap2	19.44	13.91	0.0595	16.8067	-0.40	0.0169	-6.1918	0.1299	26.40	175	225	142	137	181	880	165
341	Area	2	3 mon	reanalysis	19.83	10.82	0.0167	59.8802	-1.07	0.0014	-6.1009	0.0375	26.26	183	179	168	173	178	881	180
342	Area	2	3 mon	gpcc	20.08	11.08	0.0086	116.2791	-1.17	0.0004	-6.6491	0.0200	27.25	186	187	178	185	194	930	206
343	Area	2	3 mon	gpcc	18.16	10.93	-0.0093		-0.74	0.0020	-32.4511	-0.0444	21.01	29	181	266	227	20	723	83
344	Area	2	3 mon	gpcc lrmm	23.50	22.13	0.0543	18.4162	2.18	0.0644	-48.1515	0.2537	29.40	294	348	147	66	232	1087	281
345	Area	2	3 mon	ghcn cams grid	20.08	12.51	0.0196	51.0204	-1.01	0.0020	-6.5753	0.0446	27.12	185	203	167	168	191	914	196
346	Area	2	3 mon	ssmi	21.73	17.69	-0.0183		-0.31	0.0079	-46.8504	-0.0891	25.88	231	297	295	256	174	1253	334
347	Area	2	3 mon	opi2	21.52	17.11	0.0106	94.3398	-1.08	0.0006	-7.9367	0.0238	29.45	222	280	172	183	234	1091	284
348	Area	2	3 mon	opi3	21.35	15.96	0.0128	78.1250	-1.06	0.0008	-7.2419	0.0279	28.29	211	260	171	182	215	1039	263
349	Area	2	3 mon	cmmap	20.54	14.17	0.0135	74.0741	-1.07	0.0009	-6.7583	0.0297	27.44	192	231	170	181	197	971	227
350	Area	2	3 mon	cmmap2	20.30	13.97	0.0146	68.4932	-1.05	0.0010	-6.6408	0.0318	27.24	189	226	169	179	193	958	218
351	Area	2	4 mon	reanalysis	20.17	10.96	-0.0253		-1.42	0.0035	-7.0665	-0.0592	27.00	187	183	301	236	188	1095	286
352	Area	2	4 mon	gpcc	20.57	11.22	-0.0394		-1.60	0.0092	-7.7852	-0.0958	28.14	193	188	328	261	211	1181	314
353	Area	2	4 mon	gpcc	16.41	11.01	-0.0121		-0.78	0.0034	-32.5793	-0.0580	21.13	35	185	283	234	21	758	97
354	Area	2	4 mon	gpcc lrmm	23.56	21.96	0.0446	22.4215	1.74	0.0490	-51.5124	0.2213	29.52	295	347	153	89	236	1120	294
355	Area	2	4 mon	ghcn cams grid	20.52	12.68	-0.0272		-1.49	0.0041	-7.6521	-0.0641	27.96	191	205	306	240	209	1151	305
356	Area	2	4 mon	ssmi	21.86	17.92	-0.0035		0.17	0.0004	-71.2849	-0.0212	25.80	241	306	235	212	170	1164	310
357	Area	2	4 mon	opi2	21.97	17.28	-0.0257		-1.59	0.0036	-9.0037	-0.0599	30.07	247	281	302	237	250	1317	343
358	Area	2	4 mon	opi3	21.77	16.12	-0.0370		-1.75	0.0070	-8.3634	-0.0837	29.09	233	268	323	249	227	1300	339
359	Area	2	4 mon	cmmap	20.95	14.33	-0.0369		-1.69	0.0071	-7.8521	-0.0842	28.28	199	235	322	251	214	1221	329
360	Area	2	4 mon	cmmap2	20.70	14.12	-0.0364		-1.67	0.0068	-7.7243	-0.0825	28.08	194	230	321	248	210	1203	323



Southwest United States - Yearly Surface

OI	Runoff Meth	Re	Storage	Precipitation	Absolute Error	Bias	Abias	Slope	Oslope	Intercept	R <sup>2</sup>	R <sup>2</sup> 1	Corr Coef	RMSE	AER	BR	SR	CCR	RMR	TR	BR	BR	Plot
1	Linear	1	Yes	reanalysis	9.48	5.45	5.45	-0.0087		0.35	0.0000	-0.4992	-0.0037	11.88	38	35	85	75	44	277	56	Plot	
2	Linear	1	Yes	gpcp	8.68	1.39	1.39	-0.4471	2.2366	-0.59	0.0351	-0.0345	0.1874	11.24	30	2	36	38	36	142	6		
3	Linear	1	Yes	gpcp	9.25	4.36	4.36	0.5881	1.7603	0.76	0.0419	-0.2101	0.2047	10.05	35	22	20	29	28	134	4		
4	Linear	1	Yes	gpcp trmm	12.68	-1.57	1.57	6.8387	6.6387	36.58	0.3045	0.0723	0.5518	13.51	62	4	61	2	58	187	22		
5	Linear	1	Yes	ghcn cams grid	7.97	4.31	4.31	0.1782	5.6754	1.10	0.0032	-0.2589	0.0562	10.99	22	20	57	69	33	201	29		
6	Linear	1	Yes	ssmi	12.23	7.24	7.24	1.9307	1.9307	18.14	0.2179	-0.1011	0.4668	14.66	59	51	27	5	63	205	33		
7	Linear	1	Yes	opi2	11.46	7.38	7.38	0.3333	3.0003	0.99	0.0101	-0.4766	0.1007	13.43	55	55	45	60	56	271	53		
8	Linear	1	Yes	opi3	10.18	6.27	6.27	0.7512	1.3312	4.16	0.0670	-0.2626	0.2589	12.41	47	39	14	26	52	178	17		
9	Linear	1	Yes	cmmap	9.68	4.49	4.49	0.5230	1.9120	1.30	0.0371	-0.1590	0.1925	11.89	44	24	26	35	45	174	16		
10	Linear	1	Yes	cmmap2	9.62	4.29	4.29	0.5150	1.9417	1.14	0.0345	-0.1467	0.1857	11.83	42	18	28	40	43	171	14		
11	Linear	1	No	reanalysis	5.65	5.00	5.00	0.0220	45.4545	0.06	0.1244	-623.0347	0.3527	6.42	9	31	66	17	9	132	3		
12	Linear	1	No	gpcp	5.06	3.62	3.62	0.0049	204.0816	0.04	0.0053	-354.8611	0.0731	5.87	6	12	74	64	6	162	13		
13	Linear	1	No	gpcp	8.33	8.33	8.33	0.0099	101.0101	0.06	0.0853	-999.0000	0.2921	8.94	25	59	71	23	22	200	27		
14	Linear	1	No	gpcp trmm	6.99	6.99	6.99	0.1182	8.4602	1.02	0.0409	-107.3393	0.2021	7.09	16	48	62	32	14	172	15		
15	Linear	1	No	ghcn cams grid	4.27	3.85	3.85	-0.0076		-0.08	0.0085	-369.2077	-0.0921	4.98	2	15	82	99	2	200	26		
16	Linear	1	No	ssmi	11.74	11.74	11.74	-0.0017		0.01	0.0002	-886.3537	-0.0140	12.23	56	77	77	80	50	340	81		
17	Linear	1	No	opi2	9.60	9.60	9.60	-0.0423		-0.38	0.2056	-999.0000	-0.4534	10.22	40	64	95	108	29	326	78		
18	Linear	1	No	opi3	8.58	8.50	8.50	-0.0045		-0.02	0.0030	-897.6289	-0.0545	9.32	27	61	80	93	24	285	59		
19	Linear	1	No	cmmap	7.12	6.72	6.72	0.0050	200.0000	0.05	0.0043	-633.6632	0.0657	7.85	17	44	73	67	17	218	37		
20	Linear	1	No	cmmap2	6.91	6.51	6.51	0.0060	186.6667	0.06	0.0058	-600.7277	0.0764	7.63	14	41	72	63	15	205	31		
21	Linear	2	Yes	reanalysis	24.38	10.21	10.21	2.5305	2.5305	28.52	0.1465	0.0007	0.3827	33.60	97	69	42	13	101	322	74	Plot	
22	Linear	2	Yes	gpcp	25.53	10.51	10.51	1.0126	1.0126	10.67	0.0302	-0.0677	0.1737	34.73	111	71	1	47	107	337	79		
23	Linear	2	Yes	gpcp	16.12	7.12	7.12	1.4591	1.4591	12.29	0.1212	-0.0015	0.3481	21.41	77	50	17	87	249	48			
24	Linear	2	Yes	gpcp trmm	22.51	10.57	10.57	-5.8554		-127.58	0.6495	-0.5667	-0.8059	23.18	94	73	120	117	92	496	112		
25	Linear	2	Yes	ghcn cams grid	25.45	11.92	11.92	0.5145	1.9436	5.28	0.0104	-0.1246	0.1018	35.65	108	80	30	59	112	389	91		
26	Linear	2	Yes	ssmi	27.53	22.41	22.41	-0.2222		-0.47	0.0028	-0.8194	-0.0530	35.20	120	117	109	92	109	547	118		
27	Linear	2	Yes	opi2	26.60	16.50	16.50	0.2783	3.5932	3.33	0.0043	-0.2653	0.0653	37.81	117	101	49	68	120	455	105		
28	Linear	2	Yes	opi3	26.29	15.39	15.39	0.7330	1.3643	10.81	0.0202	-0.1921	0.1423	36.70	115	96	16	54	118	399	94		
29	Linear	2	Yes	cmmap	25.88	13.61	13.61	0.7911	1.2641	10.40	0.0218	-0.1437	0.1476	35.95	113	86	9	52	116	378	88		
30	Linear	2	Yes	cmmap2	25.80	13.41	13.41	0.7966	1.2553	10.32	0.0217	-0.1388	0.1473	35.87	112	86	8	53	115	374	86		
31	Linear	2	No	reanalysis	14.27	14.27	14.27	0.4222	2.3685	7.36	0.1407	-6.3450	0.3751	15.51	69	92	40	14	70	285	60		
32	Linear	2	No	gpcp	14.58	14.58	14.58	-0.0454		1.75	0.0021	-7.5994	-0.0457	16.78	71	94	97	89	72	423	97		
33	Linear	2	No	gpcp	12.95	12.95	12.95	-0.2784		-1.46	0.1997	-20.5909	-0.4468	14.78	64	84	111	107	64	430	99		
34	Linear	2	No	gpcp trmm	26.74	26.74	26.74	-1.6318		-26.23	0.7353	-31.5643	-0.8575	27.68	118	120	117	120	96	571	120		
35	Linear	2	No	ghcn cams grid	15.99	15.99	15.99	0.0111	90.0901	2.46	0.0002	-9.1257	0.0129	18.21	74	99	69	74	75	391	93		
36	Linear	2	No	ssmi	22.42	22.42	22.42	-0.5813		-7.18	0.7292	-32.6134	-0.8539	24.58	93	118	115	119	94	539	117		
37	Linear	2	No	opi2	20.56	20.56	20.56	-0.0678		1.07	0.0087	-15.0716	-0.0934	22.94	88	112	103	100	91	494	111		
38	Linear	2	No	opi3	19.46	19.46	19.46	-0.0881		0.80	0.0101	-13.0916	-0.1005	21.48	86	110	105	101	88	490	109		
39	Linear	2	No	cmmap	17.68	17.68	17.68	-0.0526		1.50	0.0033	-10.8711	-0.0578	19.71	83	106	99	94	81	463	106		
40	Linear	2	No	cmmap2	17.47	17.47	17.47	-0.0431		1.66	0.0022	-10.6060	-0.0468	19.49	81	104	96	90	79	450	103		
41	Point	1	Yes	reanalysis	9.52	5.47	5.47	-0.0094		0.38	0.0000	-0.4968	-0.0040	11.93	39	36	86	76	46	283	58		
42	Point	1	Yes	gpcp	8.73	1.41	1.41	0.4411	2.2671	-0.60	0.0334	-0.0360	0.1827	11.38	31	3	37	42	39	152	9		
43	Point	1	Yes	gpcp	9.24	4.37	4.37	0.5579	1.7924	0.68	0.0409	-0.2160	0.2021	10.02	34	23	21	33	27	138	5		
44	Point	1	Yes	gpcp trmm	12.70	-1.81	1.81	6.4625	6.4625	35.15	0.2917	0.0664	0.5401	13.48	63	5	60	3	57	188	24		
45	Point	1	Yes	ghcn cams grid	7.98	4.33	4.33	0.1684	5.9382	1.09	0.0029	-0.2604	0.0534	11.05	23	21	58	70	35	207	34		
46	Point	1	Yes	ssmi	12.30	7.28	7.28	1.9466	1.9466	18.36	0.2157	-0.0994	0.4644	14.85	60	52	31	6	66	215	36		
47	Point	1	Yes	opi2	11.45	7.39	7.39	0.3263	3.0647	0.94	0.0095	-0.4683	0.0974	13.54	54	56	46	61	59	278	55		
48	Point	1	Yes	opi3	10.22	6.28	6.28	0.7484	1.3362	4.15	0.0650	-0.2584	0.2549	12.54	49	40	15	27	55	186	21		
49	Point	1	Yes	cmmap	9.73	4.50	4.50	0.5147	1.9429	1.26	0.0351	-0.1585	0.1872	12.03	48	25	29	39	49	188	23		
50	Point	1	Yes	cmmap2	9.67	4.30	4.30	0.5055	1.9782	1.09	0.0325	-0.1466	0.1802	11.97	43	19	33	46	48	189	25		
51	Point	1	No	reanalysis	5.69	5.02	5.02	0.0213	46.9484	0.08	0.0711	-383.2985	0.2666	6.45	10	32	67	25	10	144	7		
52	Point	1	No	gpcp	5.09	3.63	3.63	-0.0011		0.03	0.0002	-210.6377	-0.0123	5.90	7	13	78	79	7	182	19		
53	Point	1	No	gpcp	8.34	8.34	8.34	-0.0003		-0.01	0.0000	-999.0000	-0.0048	8.96	26	60	75	77	23	261	51		
54	Point	1	No	gpcp trmm	6.75	6.75	6.75	-0.0580		-0.41	0.0102	-104.6804	-0.1008	6.89	13	48	101	102	13	275	54		
55	Point	1	No	ghcn cams grid	4.30	3.87	3.87	-0.0155		-0.08	0.0212	-227.9501	-0.1456	5.02	3	18	90	103	3	215	35		
56	Point	1	No	ssmi	11.78	11.78	11.78	0.0143	69.9301	0.23	0.0085	-551.1304	0.0923	12.25	57	79	68	62	51	317	70		
57	Point	1	No	opi2	9.82	9.82	9.82	-0.0493		-0.44	0.1644	-636.2789	-0.4054	10.24	41	65	98	106	30	340	80		
58	Point	1	No	opi3	8.59	8.51	8.51	-0.0072		-0.03	0.0045	-529.2534	-0.0674	9.34	29	62	81	96	25	293	64		
59	Point	1	No	cmmap	7.13	6.73	6.73	-0.0033		0.01	0.0011	-376.2352	-0.0333	7.88	18	45	78	85	18	244	46		
60	Point	1	No	cmmap2	6.93	6.52	6.52	-0.0036		0.01	0.0012	-355.8320	-0.0348	7.66	15	42	79	86	16	238	44		
61	Point	2	Yes	reanalysis	24.05	10.25	10.25	2.5446	2.5446	28.73	0.1496	0.0006	0.3868	33.43	95	70	43	12	100	320	72		
62	Point	2	Yes	gpcp	25.13	10.56	10.56	1.0477	1.0477	11.14	0.0326	-0.0671	0.1806	34.55	104	72	3	45	106	330	76		
63	Point	2	Yes	gpcp	16.00	7.30	7.30	1.4803	1.4803	12.71	0.1257	-0.0048	0.3545	21.37	75	53	18	16	86	248	47		

Southwest United States - Yearly Surface																									
OI	Runoff Meth	Re	Storage	Precipitation	Absolute Error	Bias	Abias	Slope	Oslope	Intercept	R <sup>2</sup>	R <sup>2</sup> 1:1	Corr Coef	RMSE	AER	BR	SR	CCR	RMR	TR	BR				
102	Area	2	Yes	gpcp	24.70	6.89	6.89	1.3042	1.3042	10.62	0.0514	0.0054	0.2268	33.06	100	47	12	28	99	286	61				
103	Area	2	Yes	gpcp	16.35	4.90	4.90	1.7295	1.7295	13.12	0.1626	0.0836	0.4033	20.96	78	30	19	10	83	220	38				
104	Area	2	Yes	gpcp trmm	17.45	4.82	4.82	-4.5942		-107.75	0.4837	-0.3155	-0.6955	19.31	80	27	118	112	77	414	96				
105	Area	2	Yes	ghcn cams grid	24.64	8.30	8.30	0.7825	1.2780	5.32	0.0246	-0.0400	0.1569	33.80	99	58	10	48	102	317	71				
106	Area	2	Yes	ssmi	24.27	18.70	18.70	0.2041	4.8996	3.81	0.0025	-0.5802	0.0502	31.86	96	108	54	71	97	426	98				
107	Area	2	Yes	opi2	25.46	12.88	12.88	0.5271	1.8972	4.25	0.0157	-0.1478	0.1253	35.51	109	83	25	55	111	383	89				
108	Area	2	Yes	opi3	25.36	11.77	11.77	1.0191	1.0191	12.10	0.0402	-0.0858	0.2006	34.54	106	78	2	34	105	325	75				
109	Area	2	Yes	cmap	25.03	9.99	9.99	1.0715	1.0715	11.09	0.0411	-0.0499	0.2027	33.97	102	67	4	30	104	307	69				
110	Area	2	Yes	cmap2	24.96	9.79	9.79	1.0792	1.0792	10.99	0.0409	-0.0464	0.2024	33.91	101	66	5	31	103	306	68				
111	Area	2	No	reanalysis	10.65	10.65	10.65	0.5010	1.9960	4.68	0.3891	-6.4494	0.6075	11.44	51	74	34	1	40	200	28				
112	Area	2	No	gpcp	10.96	10.96	10.96	0.2462	4.0617	1.71	0.1146	-7.7884	0.3385	12.43	53	76	50	21	53	253	49				
113	Area	2	No	gpcp	10.73	10.73	10.73	-0.0080		-0.63	0.0013	-107.0151	-0.0356	11.96	52	75	84	87	47	345	82				
114	Area	2	No	gpcp trmm	20.99	20.99	20.99	-0.3706		-6.60	0.6602	-334.2063	-0.8125	21.28	90	114	112	118	84	518	114				
115	Area	2	No	ghcn cams grid	12.36	12.36	12.36	0.2791	3.5829	2.50	0.1958	-9.8062	0.4425	13.78	61	82	48	8	61	260	50				
116	Area	2	No	ssmi	18.71	18.71	18.71	-0.1550		-2.90	0.8226	-267.6742	-0.7890	20.06	85	109	108	115	82	499	113				
117	Area	2	No	opi2	16.94	16.94	16.94	0.1810	5.5249	1.99	0.1158	-18.5811	0.3403	18.55	79	103	56	20	76	334	77				
118	Area	2	No	opi3	15.83	15.83	15.83	0.1980	5.0505	2.08	0.0949	-15.7232	0.3081	17.15	73	98	55	22	73	321	73				
119	Area	2	No	cmap	14.05	14.05	14.05	0.2279	4.3879	2.19	0.1162	-12.4523	0.3409	15.38	68	91	52	19	68	298	67				
120	Area	2	No	cmap2	13.85	13.85	13.85	0.2395	4.1754	2.32	0.1261	-12.0557	0.3551	15.15	67	90	51	15	67	290	63				

Amazon Basin - Monthly Atmospheric																				
OI	Runoff	Re	Storage	Precipitation	Absolute Error	Bias	Slope	Intercept	R <sup>2</sup>	R <sup>2</sup> 1.1	Corr Coef	RMSE	AER	BR	SR	CCR	RMR	TR	BR	Plot
1	Surface	1	Yes	reanalysis	111.99	46.36	3.9939	-151.99	0.6583	0.2531	0.8113	109.17	7	14	26	4	2	53	8	Plot
2	Surface	1	Yes	gpcp	105.31	20.71	1.7158	-22.96	0.3042	0.2513	0.5516	111.48	2	6	2	25	6	41	5	
3	Surface	1	Yes	gpcc	95.51	-1.23	2.4461	-40.30	0.4130	0.2686	0.6426	115.90	1	1	11	18	9	40	4	
4	Surface	1	Yes	gpcc trmm	157.61	98.69	2.4789	-57.27	0.2684	0.1692	0.5181	166.67	13	20	12	26	13	84	20	
5	Surface	1	Yes	ghcn cams grid	119.66	62.93	1.8153	-74.18	0.3382	0.2047	0.5815	112.66	9	16	3	21	8	57	10	
6	Surface	1	Yes	gpi	114.81	82.56	1.4802	-98.07	0.6376	0.3964	0.7985	105.28	8	18	1	5	1	33	1	
7	Surface	1	Yes	trmm 3a46	160.43	123.19	2.5675	-244.08	0.5882	-0.1704	0.7669	181.48	17	25	13	7	23	85	21	
8	Surface	1	Yes	trmm 3b43	160.43	105.77	2.2315	-161.54	0.3791	-0.1254	0.6157	179.90	18	22	10	19	21	90	23	
9	Surface	1	Yes	ssmi	127.26	63.78	1.9732	-87.85	0.4561	0.3131	0.6753	117.48	10	17	7	17	10	61	11	
10	Surface	1	Yes	opi2	106.50	16.41	2.1791	-20.17	0.3747	0.2608	0.6121	110.76	3	2	9	20	3	37	2	
11	Surface	1	Yes	opi3	108.22	18.03	1.9602	-19.97	0.3270	0.2465	0.5718	111.83	6	3	6	24	7	46	7	
12	Surface	1	Yes	cmap	106.86	18.56	1.9498	-24.74	0.3371	0.2565	0.5806	111.08	4	4	5	22	4	39	3	Plot
13	Surface	1	Yes	cmap2	107.27	18.66	1.9053	-23.22	0.3276	0.2530	0.5724	111.35	5	5	4	23	5	42	6	
14	Surface	2	Yes	reanalysis	145.51	42.49	3.2774	-136.06	0.7933	0.3993	0.8907	157.70	12	13	16	7	11	53	9	Plot
15	Surface	2	Yes	gpcp	160.08	36.05	3.1746	-100.87	0.5049	0.2646	0.7105	174.49	14	12	15	15	16	72	15	
16	Surface	2	Yes	gpcc	143.88	24.91	3.3965	-75.59	0.4792	0.2367	0.6823	176.54	11	7	17	16	20	71	14	
17	Surface	2	Yes	gpcc trmm	198.12	106.02	3.9897	-220.40	0.5743	0.0594	0.7578	234.41	25	23	25	10	24	107	26	
18	Surface	2	Yes	ghcn cams grid	161.11	49.05	3.5078	-162.08	0.5844	0.2701	0.7644	173.83	21	15	19	8	14	77	17	
19	Surface	2	Yes	gpi	161.59	104.49	2.1525	-193.24	0.6886	0.3207	0.8298	164.93	22	21	8	3	12	86	12	
20	Surface	2	Yes	trmm 3a46	206.57	135.70	3.7977	-357.56	0.7171	0.0092	0.8468	239.25	26	26	24	2	26	104	25	
21	Surface	2	Yes	trmm 3b43	197.95	115.90	3.6997	-245.86	0.6067	0.0537	0.7789	235.12	24	24	22	6	25	101	24	
22	Surface	2	Yes	ssmi	177.81	84.71	2.9241	-193.15	0.5466	0.2462	0.7394	180.33	23	19	14	11	22	89	22	
23	Surface	2	Yes	opi2	160.24	28.61	3.7808	-94.89	0.5779	0.2649	0.7602	174.44	16	8	23	9	15	71	13	
24	Surface	2	Yes	opi3	160.44	30.10	3.5740	-97.38	0.5356	0.2568	0.7318	175.40	19	9	21	13	19	81	19	
25	Surface	2	Yes	cmap	160.23	32.33	3.5313	-105.00	0.5437	0.2624	0.7373	174.74	15	10	20	12	17	74	16	
26	Surface	2	Yes	cmap2	160.58	32.39	3.4770	-102.55	0.5342	0.2612	0.7309	174.88	20	11	18	14	18	81	18	

Amazon Basin - Yearly Atmospheric																				
Of	Runoff	Re	Storage	Precipitation	Absolute Error	Bias	Slope	Oslope	Intercept	R <sup>2</sup>	R <sup>2</sup> 1:1	Corr Coef	RMSE	AER	BR	SR	CCR	RMR	TR	BR
1	Surface	1	Yes	reanalysis	60.15	46.43	2.0482	2.0482	-68.69	0.4205	-0.1661	0.6485	37.20	9	14	12	8	8	51	6
2	Surface	1	Yes	gpcp	40.71	20.74	0.2247	4.4504	25.23	0.0073	-0.0798	0.0855	31.78	5	6	16	21	4	52	8
3	Surface	1	Yes	gpcp	26.93	-1.15	0.8523	1.1733	5.39	0.0445	0.0420	0.2109	32.80	1	1	1	17	5	25	2
4	Surface	1	Yes	gpcp trmm	89.11	89.11	3.7474	3.7474	-189.16	0.7875	-2.5682	0.8874	98.29	19	20	15	6	21	81	17
5	Surface	1	Yes	ghcn cams grid	72.81	61.63	0.5344	1.8713	7.97	0.0341	-0.1833	0.1846	33.27	16	16	11	19	6	68	13
6	Surface	1	Yes	gpi	82.67	82.67	0.6316	1.5833	-24.76	0.0589	-2.8154	0.2428	65.46	18	18	10	14	18	78	16
7	Surface	1	Yes	trmm 3a46	118.36	118.36	4.7411	4.7411	-404.47	0.9408	-4.1024	0.9699	126.63	25	25	18	1	24	93	22
8	Surface	1	Yes	trmm 3b43	105.85	105.85	4.6786	4.6786	-272.46	0.9088	-4.2945	0.9533	113.05	22	22	17	4	22	87	19
9	Surface	1	Yes	ssmi	73.69	67.64	-0.7057		82.17	0.0501	-0.9311	-0.2238	45.80	17	17	25	25	10	94	23
10	Surface	1	Yes	opi2	41.39	13.30	0.1181	8.4674	29.63	0.0027	-0.2199	0.0516	33.78	6	2	22	22	7	59	12
11	Surface	1	Yes	opi3	38.81	14.79	0.6890	1.4514	14.05	0.0609	0.0134	0.2467	30.38	4	3	7	13	3	30	4
12	Surface	1	Yes	cmap	38.38	17.02	0.7358	1.3591	10.89	0.0700	0.0504	0.2646	29.80	2	4	3	11	7	21	1
13	Surface	1	Yes	cmap2	38.71	17.08	0.6882	1.4531	12.37	0.0637	0.0393	0.2523	29.98	3	5	8	12	2	30	3
14	Surface	2	Yes	reanalysis	52.38	42.47	2.2662	2.2662	-85.02	0.7378	0.3346	0.8590	41.60	8	13	13	7	9	50	5
15	Surface	2	Yes	gpcp	64.28	36.03	-0.6133		54.29	0.0112	-0.1194	-0.1058	53.96	14	12	24	24	16	90	20
16	Surface	2	Yes	gpcp	47.84	24.84	3.5079	3.5079	-120.56	0.1744	-0.1717	0.4176	53.03	7	7	14	9	14	51	7
17	Surface	2	Yes	gpcp trmm	106.23	106.23	5.5012	5.5012	-278.44	0.9163	-2.5706	0.9572	118.42	23	23	19	3	23	91	21
18	Surface	2	Yes	ghcn cams grid	67.26	49.03	0.7844	1.2749	-13.56	0.0177	-0.2301	0.1329	56.56	15	15	2	20	17	69	14
19	Surface	2	Yes	gpi	104.75	104.52	1.3662	1.3662	-117.74	0.0916	-2.7580	0.3026	94.97	21	21	4	10	20	76	15
20	Surface	2	Yes	trmm 3a46	139.77	139.77	6.3735	6.3735	-559.98	0.8171	-3.6149	0.9040	153.00	26	26	21	5	26	104	25
21	Surface	2	Yes	trmm 3b43	116.11	116.11	6.2428	6.2428	-368.51	0.9226	-3.1610	0.9605	127.84	24	24	20	2	25	95	24
22	Surface	2	Yes	ssmi	95.17	88.12	-1.8134		150.55	0.0737	-1.1891	-0.2715	76.52	20	19	26	26	19	110	26
23	Surface	2	Yes	opi2	63.67	28.59	-0.1934		35.53	0.0022	-0.0879	-0.0473	53.19	13	8	23	23	15	82	18
24	Surface	2	Yes	opi3	60.40	30.09	1.3797	1.3797	-19.64	0.0429	0.0248	0.2070	50.36	10	9	5	18	11	53	9
25	Surface	2	Yes	cmap	60.83	32.32	1.5731	1.5731	-30.55	0.0560	0.0189	0.2366	50.52	11	10	9	15	12	57	10
26	Surface	2	Yes	cmap2	60.86	32.38	1.3806	1.3806	-23.09	0.0454	0.0130	0.2130	50.67	12	11	6	16	13	58	11

Amazon Basin - Monthly Surface																						
Of	Runoff	Re	Storage	Precipitation	Absolute Error	Bias	Abias	Slope	Oslope	Intercep	R <sup>2</sup>	R <sup>2</sup> 1:1	Corr Coef	RMSE	AER	BR	SR	CCR	RMR	TR	BR	Plot
1	Surface	1	Yes	reanalysis	54.25	54.27	54.21	0.5913	1.6912	-14.54	0.4211	-4.2475	0.6489	58.75	92	136	100	66	48	442	67	
2	Surface	1	Yes	gpcp	63.04	62.08	62.08	0.7346	1.3613	-37.03	0.2553	-2.0250	0.5052	72.03	175	196	65	80	161	677	147	
3	Surface	1	Yes	gpcp	68.25	68.20	68.20	0.6864	1.4569	-37.81	0.2870	-3.4424	0.5358	75.04	210	226	78	77	190	781	176	
4	Surface	1	Yes	ghcn cams grid	48.08	45.65	45.65	0.7839	1.2757	-24.68	0.2976	-0.9974	0.5456	57.20	55	102	44	75	40	316	28	
5	Surface	1	Yes	gpi	45.86	10.43	10.43	2.1766	2.1766	-124.46	0.6846	0.4642	<b>0.8274</b>	53.51	34	7	110	9	27	187	14	
6	Surface	1	Yes	ssmi	<b>37.26</b>	<b>32.81</b>	<b>32.81</b>	1.3623	1.3623	-67.97	0.6732	0.1681	<b>0.8205</b>	<b>44.25</b>	4	25	66	12	4	111	2	
7	Surface	1	Yes	opi2	70.30	70.23	70.23	0.6818	1.4867	-40.20	0.2880	-3.5407	0.5367	77.12	224	242	79	76	202	823	186	
8	Surface	1	Yes	opi3	68.58	67.64	67.64	0.6536	1.5300	-34.94	0.2454	-3.0626	0.4954	75.75	214	222	90	83	194	803	179	
9	Surface	1	Yes	cmep	66.24	65.05	65.05	0.6766	1.4780	-34.53	0.2524	-2.6803	0.5024	73.60	195	212	82	81	176	746	166	
10	Surface	1	Yes	cmep2	66.40	65.16	65.16	0.6809	1.4686	-35.04	0.2511	-2.6385	0.5011	73.82	199	216	80	82	178	755	170	
11	Surface	1	No	reanalysis	54.98	54.14	54.14	-0.0401		46.74	0.0016	-5.5136	-0.0398	65.50	105	134	187	186	105	717	157	
12	Surface	1	No	gpcp	63.36	62.08	62.08	0.6935	1.4420	-33.14	0.1774	-2.1038	0.4212	72.97	176	195	77	108	170	726	158	
13	Surface	1	No	gpcp	68.68	68.21	68.21	0.6057	1.6510	-29.99	0.2002	-3.5552	0.4474	75.99	216	227	99	95	195	832	190	
14	Surface	1	No	ghcn cams grid	49.69	45.58	45.58	0.6490	1.5408	-11.54	0.1668	-1.1505	0.4084	59.35	68	100	92	111	53	424	52	
15	Surface	1	No	gpi	65.97	10.44	10.44	0.0372	26.8817	82.87	0.0002	-0.1402	0.0134	78.06	194	8	165	180	205	752	167	
16	Surface	1	No	ssmi	47.15	32.91	32.91	3.7750	3.7750	38.49	0.0218	-0.6063	0.1476	61.49	48	26	118	149	72	413	50	
17	Surface	1	No	opi2	70.43	70.23	70.23	0.6792	1.4723	-39.94	0.2229	-3.5923	0.4721	77.55	226	241	81	90	204	842	193	
18	Surface	1	No	opi3	68.65	67.63	67.63	0.6449	1.5506	-34.11	0.1863	-3.1088	0.4317	76.18	215	221	94	100	196	826	188	
19	Surface	1	No	cmep	66.26	65.04	65.04	0.6461	1.5477	-31.64	0.1795	-2.7490	0.4237	74.28	196	210	83	106	181	786	177	
20	Surface	1	No	cmep2	66.42	65.15	65.15	0.6691	1.4945	-33.92	0.1891	-2.6911	0.4349	74.36	200	214	84	96	189	779	175	
21	Surface	1	1 mon	reanalysis	54.41	54.20	54.20	0.3261	3.0685	11.19	0.1046	-4.7901	0.3234	61.84	98	135	116	126	79	548	95	
22	Surface	1	1 mon	gpcp	62.52	62.08	62.08	1.1579	1.1579	-76.98	0.4945	-1.7513	0.7032	68.86	171	197	21	52	135	576	104	
23	Surface	1	1 mon	gpcp	68.49	68.26	68.26	0.9754	1.0252	-65.88	0.5188	-3.1299	0.7203	72.63	213	228	3	48	165	657	134	
24	Surface	1	1 mon	ghcn cams grid	47.36	45.57	45.57	1.1615	1.1615	-61.25	0.5339	-0.7413	0.7307	53.47	50	99	22	41	26	238	20	
25	Surface	1	1 mon	gpi	55.21	10.65	10.65	1.3021	1.3021	-39.95	0.2195	0.1866	0.4685	66.16	106	9	52	91	114	372	40	
26	Surface	1	1 mon	ssmi	41.95	33.13	33.13	0.9384	1.0856	-27.14	0.2731	-0.1896	0.5226	53.19	14	27	10	78	24	153	9	
27	Surface	1	1 mon	opi2	70.29	70.29	70.29	1.0850	1.0850	-78.31	0.5691	-3.1918	0.7544	74.24	223	243	13	34	180	693	153	
28	Surface	1	1 mon	opi3	68.22	67.62	67.62	1.0424	1.0424	-71.62	0.4869	-2.7366	0.6978	72.81	208	220	8	55	167	658	136	
29	Surface	1	1 mon	cmep	65.73	65.03	65.03	1.0601	1.0601	-70.70	0.4832	-2.3787	0.6951	70.68	189	209	9	56	149	612	119	
30	Surface	1	1 mon	cmep2	65.86	65.14	65.14	1.0818	1.0818	-72.86	0.4944	-2.3281	0.7031	70.77	191	213	12	53	150	619	122	
31	Surface	1	2 mon	reanalysis	54.39	54.29	54.29	0.5839	1.7126	-13.90	0.3356	-4.2879	0.5793	59.16	97	138	102	71	50	458	70	
32	Surface	1	2 mon	gpcp	62.33	62.16	62.16	1.3023	1.3023	-90.69	0.6258	-1.8406	<b>0.7911</b>	67.60	168	198	53	23	124	566	100	
33	Surface	1	2 mon	gpcp	68.30	68.30	68.30	1.1159	1.1159	-79.54	0.6790	-2.9531	<b>0.8240</b>	71.33	212	229	16	11	156	624	125	
34	Surface	1	2 mon	ghcn cams grid	46.45	45.60	45.60	1.3533	1.3533	-79.90	0.7244	-0.5877	<b>0.8511</b>	51.14	41	101	64	4	17	227	17	
35	Surface	1	2 mon	gpi	48.10	11.29	11.29	2.0984	2.0984	-117.82	0.5752	0.3939	0.7584	57.07	56	11	109	32	39	247	22	
36	Surface	1	2 mon	ssmi	38.66	33.91	33.91	1.3640	1.3640	-69.46	0.5765	0.0525	0.7592	47.50	6	30	67	31	8	142	7	
37	Surface	1	2 mon	opi2	70.36	70.36	70.36	1.2333	1.2333	-92.38	0.7356	-3.0400	<b>0.8577</b>	73.04	225	244	37	3	171	680	148	
38	Surface	1	2 mon	opi3	68.26	67.64	67.64	1.1558	1.1558	-82.34	0.5985	-2.6218	0.7736	71.85	211	223	20	28	159	641	130	
39	Surface	1	2 mon	cmep	65.74	65.04	65.04	1.1831	1.1831	-82.33	0.6018	-2.2608	0.7758	69.60	190	211	28	27	140	596	114	
40	Surface	1	2 mon	cmep2	65.89	65.15	65.15	1.1968	1.1968	-83.73	0.6050	-2.2193	0.7778	69.76	192	215	31	26	141	605	117	
41	Surface	1	3 mon	reanalysis	54.28	54.26	54.26	0.6578	1.5202	-21.06	0.4255	-4.1258	0.6523	58.33	94	137	88	65	47	431	55	
42	Surface	1	3 mon	gpcp	62.48	61.97	61.97	1.0970	1.0970	-71.12	0.4422	-1.7714	0.6650	69.40	170	194	15	64	139	582	107	
43	Surface	1	3 mon	gpcp	67.97	67.97	67.97	0.9618	1.0397	-64.27	0.5016	-3.0674	0.7082	72.57	205	225	5	50	163	648	132	
44	Surface	1	3 mon	ghcn cams grid	46.91	45.53	45.53	1.2144	1.2144	-66.33	0.5826	-0.6905	0.7633	52.85	46	98	34	30	23	231	18	
45	Surface	1	3 mon	gpi	46.36	11.37	11.37	2.2486	2.2486	-132.21	0.6568	0.4304	<b>0.8104</b>	55.48	40	12	111	20	34	217	16	
46	Surface	1	3 mon	ssmi	40.22	33.79	33.79	1.3492	1.3492	-67.83	0.5604	0.0483	0.7486	47.85	9	29	62	38	10	148	8	
47	Surface	1	3 mon	opi2	70.14	70.14	70.14	1.0416	1.0416	-74.07	0.5230	-3.1922	0.7232	74.52	222	240	6	46	187	701	155	
48	Surface	1	3 mon	opi3	68.16	67.43	67.43	0.9595	1.0422	-63.61	0.4108	-2.7684	0.6410	73.42	207	218	7	68	175	675	146	
49	Surface	1	3 mon	cmep	65.72	64.84	64.84	0.9849	1.0153	-63.41	0.4154	-2.4045	0.6445	71.25	188	207	2	67	155	619	123	
50	Surface	1	3 mon	cmep2	65.91	64.96	64.96	0.9864	1.0138	-63.68	0.4093	-2.3711	0.6398	71.53	193	208	1	69	157	628	126	
51	Surface	1	4 mon	reanalysis	54.33	54.12	54.12	0.5701	1.7541	-12.45	0.3189	-4.2882	0.5847	59.18	95	133	105	73	51	457	69	
52	Surface	1	4 mon	gpcp	64.00	61.63	61.63	0.6551	1.5265	-29.18	0.1564	-2.0693	0.3955	73.09	177	189	89	112	172	739	162	
53	Surface	1	4 mon	gpcp	68.25	67.45	67.45	0.6317	1.5830	-31.90	0.2153	-3.3689	0.4640	75.24	209	219	97	93	191	809	181	
54	Surface	1	4 mon	ghcn cams grid	49.44	45.30	45.30	0.8062	1.2404	-26.51	0.2569	-1.0014	0.5068	57.47	67	97	39	79	41	323	30	
55	Surface	1	4 mon	gpi	49.29	10.77	10.77	1.8856	1.8856	-96.27	0.4578	0.3354	0.6766	60.08	63	10	107	62	62	304	26	
56	Surface	1	4 mon	ssmi	44.62	33.55	33.55	1.0299	1.0299	-36.45	0.3262	-0.1368	0.5712	52.58	30	28	4	72	22	156	10	
57	Surface	1	4 mon	opi2	70.05	69.78	69.78	0.6626	1.5092	-38.03	0.2104	-3.5221	0.4587	77.38	221	234	86	94	203	838	192	
58	Surface	1	4 mon	opi3	68.83	67.12	67.12	0.5714	1.7501	-26.78	0.1445	-3.0791	0.3801	76.47	217	217	104	115	199	852	194	
59	Surface	1	4 mon	cmep	66.66	64.51	64.51	0.5897	1.6958	-25.90	0.1477	-2.7105	0.3843	74.44	201	205	101	113	186	806	180	
60	Surface	1	4 mon	cmep2	66.94	64.64	64.64	0.5800	1.7241	-25.12	0.1403	-2.6806	0.3746	74.82	202	206	103	117	188	816	184	
61	Surface	2	Yes	reanalysis	50.76	44																

Amazon Basin - Monthly Surface

OI	Runoff	Re	Storage	Precipitation	Absolute Error	Bias	Abias	Slope	Oslope	Intercept	R <sup>2</sup>	R <sup>2</sup> 1:1	Corr Coef	RMSE	AER	BR	SR	CCR	RMR	TR	BR
102	Surface	2	3 mon	gpcp	54.88	53.19	53.19	1.2844	1.2844	-80.01	0.5018	-0.8709	0.7084	62.67	102	128	46	49	81	406	49
103	Surface	2	3 mon	gpcp	58.66	58.26	58.26	1.1742	1.1742	-76.45	0.5603	-1.5072	0.7485	65.82	128	165	26	39	112	470	75
104	Surface	2	3 mon	ghcn cams grid	43.10	39.60	39.60	1.3274	1.3274	-70.48	0.5650	-0.2567	0.7518	50.03	19	56	58	36	13	182	13
105	Surface	2	3 mon	gpi	49.29	1.66	1.66	2.4295	2.4295	-140.02	0.6839	0.4466	0.8270	57.90	62	5	113	10	46	236	19
106	Surface	2	3 mon	ssmi	35.30	24.54	24.54	1.5360	1.5360	-76.77	0.6161	0.3288	0.7849	43.63	3	17	91	24	3	138	5
107	Surface	2	3 mon	opi2	61.60	61.37	61.37	1.2291	1.2291	-82.96	0.5706	-1.6769	0.7554	67.27	162	187	36	33	120	538	91
108	Surface	2	3 mon	opi3	60.13	58.66	58.66	1.1489	1.1489	-72.51	0.4788	-1.4903	0.6919	66.10	150	168	18	58	113	507	82
109	Surface	2	3 mon	cmapp	57.86	56.06	56.06	1.1723	1.1723	-72.31	0.4811	-1.2524	0.6936	64.10	120	141	24	57	95	437	59
110	Surface	2	3 mon	cmapp2	58.05	56.18	56.18	1.1739	1.1739	-72.58	0.4757	-1.2414	0.6897	64.39	123	142	25	60	97	447	66
111	Surface	2	4 mon	reanalysis	51.72	43.24	43.24	1.2915	1.2915	-70.67	0.3433	-0.2807	0.5860	62.83	80	85	49	70	82	366	38
112	Surface	2	4 mon	gpcp	57.36	52.81	52.81	0.7832	1.2768	-32.41	0.1853	-1.1569	0.4304	67.31	116	127	45	102	121	511	84
113	Surface	2	4 mon	gpcp	59.53	57.66	57.66	0.7763	1.2882	-37.36	0.2446	-1.7933	0.4946	69.37	141	163	47	84	138	573	103
114	Surface	2	4 mon	ghcn cams grid	46.54	39.18	39.18	0.8500	1.1765	-25.06	0.2305	-0.5484	0.4801	55.49	43	51	27	89	35	245	21
115	Surface	2	4 mon	gpi	52.52	0.98	0.98	2.0125	2.0125	-98.72	0.4655	0.3476	0.6823	63.00	87	2	108	61	87	345	33
116	Surface	2	4 mon	ssmi	43.88	24.10	24.10	1.3150	1.3150	-55.00	0.4518	0.2194	0.6722	47.29	23	14	55	63	7	162	11
117	Surface	2	4 mon	opi2	61.80	60.96	60.96	0.7907	1.2647	-41.27	0.2350	-1.9833	0.4848	70.96	166	186	41	87	152	632	127
118	Surface	2	4 mon	opi3	61.76	58.30	58.30	0.6995	1.4296	-30.02	0.1768	-1.7913	0.4204	70.01	165	166	74	109	144	658	135
119	Surface	2	4 mon	cmapp	59.77	55.69	55.69	0.7178	1.3931	-29.13	0.1791	-1.5480	0.4232	68.19	144	139	71	107	127	588	110
120	Surface	2	4 mon	cmapp2	60.04	55.82	55.82	0.7081	1.4122	-28.36	0.1718	-1.5405	0.4145	68.58	148	140	73	110	131	602	116
121	Reanalysis	1	Yes	reanalysis	48.66	-40.58	40.58	0.3812	2.6233	-18.86	0.0814	-1.1219	0.2853	59.44	58	69	114	131	54	426	53
122	Reanalysis	1	Yes	gpcp	62.53	-42.00	42.00	-0.1298		-2.39	0.0190	-2.4711	-0.1379	76.35	172	78	218	199	197	862	197
123	Reanalysis	1	Yes	gpcp	54.85	-32.71	32.71	-0.0456		-2.68	0.0017	-1.6052	-0.0415	63.09	101	23	191	187	89	591	111
124	Reanalysis	1	Yes	gpcp trmm	74.83	-52.29	52.29	-0.4012		-0.72	0.1293	-2.7141	-0.3596	89.56	239	124	273	232	233	1101	242
125	Reanalysis	1	Yes	ghcn cams grid	69.39	-57.56	57.56	-0.0608		-2.52	0.0037	-3.1132	-0.0607	82.62	219	162	201	188	218	988	219
126	Reanalysis	1	Yes	gpi	92.02	-89.72	89.72	0.1979	5.0531	-24.30	0.1132	-6.3027	0.3364	113.58	253	257	124	124	255	1013	227
127	Reanalysis	1	Yes	trmm 3a46	100.87	-99.41	99.41	-0.0712		-16.79	0.0059	-5.9178	-0.0768	122.08	266	273	203	189	269	1200	261
128	Reanalysis	1	Yes	trmm 3b43	80.72	-60.65	60.65	-0.3697		1.38	0.1408	-3.5237	-0.3753	98.04	249	164	272	234	249	1188	258
129	Reanalysis	1	Yes	ssmi	72.30	-70.08	70.08	0.2075	4.8193	-21.74	0.0523	-3.3854	0.2287	89.73	231	238	121	138	234	962	213
130	Reanalysis	1	Yes	opi2	58.44	-35.50	35.50	-0.1814		-3.07	0.0237	-1.9621	-0.1539	70.09	126	36	231	203	145	741	163
131	Reanalysis	1	Yes	opi3	59.39	-37.12	37.12	-0.1579		-2.92	0.0243	-2.1159	-0.1560	71.89	138	42	229	204	160	773	173
132	Reanalysis	1	Yes	cmapp	60.51	-39.41	39.41	-0.1436		-2.71	0.0203	-2.2016	-0.1424	72.89	155	54	225	200	168	802	178
133	Reanalysis	1	Yes	cmapp2	60.67	-39.51	39.51	-0.1495		-2.51	0.0223	-2.2350	-0.1493	73.27	157	55	225	202	173	814	182
134	Reanalysis	1	No	reanalysis	44.63	-40.51	40.51	-0.2372		2.91	0.3614	-20.7739	-0.6012	56.22	31	68	257	264	37	655	133
135	Reanalysis	1	No	gpcp	52.54	-42.08	42.08	-0.2243		0.84	0.3878	-18.3658	-0.6227	69.00	88	78	252	268	137	823	185
136	Reanalysis	1	No	gpcp	41.96	-32.70	32.70	-0.1333		-0.15	0.5321	-63.1398	-0.7294	52.12	15	22	221	275	21	554	99
137	Reanalysis	1	No	gpcp trmm	60.59	-52.09	52.09	-0.4350		0.71	0.4603	-8.3543	-0.6785	81.68	158	122	275	273	214	1040	231
138	Reanalysis	1	No	ghcn cams grid	60.32	-57.31	57.31	-0.1877		4.32	0.3996	-38.1308	-0.6322	75.52	153	157	239	270	192	1011	226
139	Reanalysis	1	No	gpi	100.78	-89.53	89.53	-0.1265		1.86	0.2390	-46.2382	-0.4888	121.09	264	255	216	253	264	1252	269
140	Reanalysis	1	No	trmm 3a46	99.45	-96.69	96.69	-0.0660		16.37	0.6139	-15.9892	-0.7835	122.84	257	263	276	271	1343	276	
141	Reanalysis	1	No	trmm 3b43	67.28	-60.46	60.46	-0.4030		3.08	0.5064	-10.8361	-0.7116	91.17	204	183	274	274	240	1175	255
142	Reanalysis	1	No	ssmi	75.37	-69.80	69.80	-0.1970		3.20	0.2370	-21.8449	-0.4868	91.37	241	235	244	252	241	1213	263
143	Reanalysis	1	No	opi2	46.13	-34.88	34.88	-0.2591		0.28	0.4213	-14.5900	-0.6491	61.20	38	31	259	272	69	669	140
144	Reanalysis	1	No	opi3	47.35	-36.50	36.50	-0.2394		0.12	0.3864	-15.5162	-0.6216	62.99	49	37	256	267	86	695	154
145	Reanalysis	1	No	cmapp	49.11	-39.11	39.11	-0.2340		0.49	0.3702	-16.2607	-0.6084	64.55	61	49	254	266	99	729	159
146	Reanalysis	1	No	cmapp2	49.33	-39.20	39.20	-0.2305		0.40	0.3640	-16.3775	-0.6033	64.77	64	52	253	265	100	734	160
147	Reanalysis	1	1 mon	reanalysis	44.26	-40.43	40.43	-0.1866		1.14	0.2236	-20.1131	-0.4728	55.40	28	67	238	250	33	616	120
148	Reanalysis	1	1 mon	gpcp	51.77	-41.84	41.84	-0.1632		-1.33	0.2036	-17.2362	-0.4512	67.07	81	73	233	247	119	753	168
149	Reanalysis	1	1 mon	gpcp	41.57	-32.60	32.60	-0.1009		-1.00	0.3116	-62.2050	-0.5582	51.38	12	21	212	262	18	525	89
150	Reanalysis	1	1 mon	gpcp trmm	59.62	-52.10	52.10	-0.2829		-4.56	0.1953	-7.5801	-0.4419	78.70	143	123	264	245	208	983	216
151	Reanalysis	1	1 mon	ghcn cams grid	60.12	-57.35	57.35	-0.1368		1.71	0.2126	-37.0858	-0.4611	74.43	149	158	223	248	185	963	214
152	Reanalysis	1	1 mon	gpi	99.97	-89.27	89.27	-0.1110		1.09	0.2026	-45.6458	-0.4501	120.58	262	253	214	246	263	1238	268
153	Reanalysis	1	1 mon	trmm 3a46	99.75	-97.90	97.90	-0.3457		7.83	0.3269	-15.1962	-0.5718	121.19	261	269	271	263	266	1330	275
154	Reanalysis	1	1 mon	trmm 3b43	65.28	-59.48	59.48	-0.2729		-2.78	0.2297	-9.7433	-0.4792	87.37	186	179	283	251	227	1106	243
155	Reanalysis	1	1 mon	ssmi	74.78	-69.54	69.54	-0.1408		-0.19	0.1207	-20.9807	-0.3474	89.85	238	231	224	227	235	1155	253
156	Reanalysis	1	1 mon	opi2	45.55	-35.05	35.05	-0.2001		-1.33	0.2499	-13.8302	-0.4999	59.79	33	32	246	254	56	621	124
157	Reanalysis	1	1 mon	opi3	46.76	-36.68	36.68	-0.1888		-1.93	0.1911	-14.5465	-0.4371	61.22	44	39	235	244	71	633	128
158	Reanalysis	1	1 mon	cmapp	48.62	-39.15	39.15	-0.1675		-1.63	0.1897	-15.3504	-0.4355	62.93	57	50	234	243	85	669	141
159	Reanalysis	1	1 mon	cmapp2	48.81	-39.24	39.24	-0.1624		-1.78	0.1806	-15.4311	-0.4249	63.09	60	53	232	240	88	673	144
160	Reanalysis	1	2 mon	reanalysis	43.97	-40.39	40.39	-0.0958		-2.05	0.0590	-18.9294	-0.2429	53.86	26	66	210	218	29	549	96
161	Reanalysis	1	2 mon	gpcp	51.04	-41.85	41.85	-0.0840		-4.08	0.0540	-16.0048	-0.2323	64.89	78	74	207	215	101	675	145
162	Reanalysis	1	2 mon	gpcp	40.71	-32.48	32.48	-0.0504		-2.35	0.0801	-60.4496	-0.2830	50.13	10	19	195	221	16	461	71
163	Reanalysis	1	2 mon	gpcp trmm	58.80	-51.80	51.80	-0.1458		-9.16	0.0257	-6.9167	-0.2296	75.62	132	121	226	214	193	886	204
164	Reanalysis	1	2 mon	ghcn cams grid	59.86	-57.41</															

Amazon Basin - Monthly Surface

OI	Runoff	Re	Storage	Precipitation	Absolute Error	Bias	Abias	Slope	Oslope	Intercept	R <sup>2</sup>	R <sup>2</sup> 1:1	Corr Coef	RMSE	AER	BR	SR	CCR	RMR	TR	BR
203	Reanalysis	2	Yes	ghcn cams grid	65.13	-63.22	63.22	0.0222	45.0450	-9.01	0.0016	-9.3977	0.0399	81.07	184	200	170	171	212	937	209
204	Reanalysis	2	Yes	gpi	101.94	-97.87	97.87	0.1053	9.4967	-17.30	0.0838	-18.6016	0.2894	121.87	267	268	142	129	268	1074	236
205	Reanalysis	2	Yes	trmm 3a46	95.40	-94.44	94.44	0.1216	8.2237	-24.77	0.0318	-8.3134	0.1784	111.48	254	259	135	143	253	1044	232
206	Reanalysis	2	Yes	trmm 3b43	72.06	-67.76	67.76	0.1038	9.6339	-24.64	0.0195	-4.6416	0.1396	89.88	230	224	143	152	236	985	217
207	Reanalysis	2	Yes	ssmi	76.31	-74.74	74.74	0.1648	6.0680	-17.77	0.0911	-9.4101	0.3019	90.12	244	247	130	126	237	986	218
208	Reanalysis	2	Yes	opi2	49.43	-42.78	42.78	0.0064	156.2500	-8.01	0.0001	-5.7413	0.0109	65.28	65	81	181	181	103	611	118
209	Reanalysis	2	Yes	opi3	50.54	-44.27	44.27	0.0133	75.1880	-8.27	0.0005	-6.0648	0.0232	66.82	73	94	177	178	117	639	129
210	Reanalysis	2	Yes	cmapp	52.06	-46.50	46.50	0.0137	72.9927	-8.31	0.0006	-6.4011	0.0241	68.40	82	109	176	177	128	672	143
211	Reanalysis	2	Yes	cmapp2	52.14	-46.56	46.56	0.0107	93.4579	-8.20	0.0004	-6.4538	0.0189	68.64	84	111	178	179	133	885	150
212	Reanalysis	2	No	reanalysis	75.90	-56.77	56.77	-0.2703		5.32	0.4026	-13.9196	-0.6345	94.08	243	156	262	271	246	1178	256
213	Reanalysis	2	No	gpcc	65.25	-50.33	50.33	-0.2692		3.53	0.2727	-10.0577	-0.5222	80.99	185	120	260	257	211	1033	229
214	Reanalysis	2	No	gpcc	55.86	-41.26	41.26	-0.3114		8.79	0.3884	-10.7283	-0.6232	68.71	109	71	270	269	134	853	195
215	Reanalysis	2	No	gpcc trmm	72.51	-60.01	60.01	-0.2523		-12.10	0.1092	-5.5158	-0.3305	89.45	232	182	258	225	232	1129	245
216	Reanalysis	2	No	ghcn cams grid	71.21	-63.33	63.33	-0.2959		8.51	0.3002	-12.2183	-0.5479	88.55	229	203	268	261	229	1190	259
217	Reanalysis	2	No	gpi	113.49	-97.79	97.79	-0.1588		6.56	0.2155	-25.5584	-0.4642	133.31	276	267	230	249	276	1298	272
218	Reanalysis	2	No	trmm 3a46	106.23	-97.04	97.04	-0.2703		3.70	0.1862	-12.2829	-0.4315	122.35	272	265	261	241	270	1309	273
219	Reanalysis	2	No	trmm 3b43	80.26	-69.90	69.90	-0.1953		-12.35	0.0804	-6.9114	-0.2836	98.56	248	236	242	222	251	1199	260
220	Reanalysis	2	No	ssmi	85.33	-75.10	75.10	-0.1988		6.67	0.1573	-14.1403	-0.3966	99.79	252	250	245	238	252	1237	267
221	Reanalysis	2	No	opi2	59.24	-42.89	42.89	-0.3061		2.82	0.2878	-8.0537	-0.5365	73.29	135	84	269	260	174	922	208
222	Reanalysis	2	No	opi3	59.80	-44.38	44.38	-0.2890		2.65	0.2709	-8.4397	-0.5205	74.83	145	95	265	255	189	949	210
223	Reanalysis	2	No	cmapp	61.35	-46.61	46.61	-0.2935		3.47	0.2812	-8.8438	-0.5302	76.42	159	113	267	259	198	996	221
224	Reanalysis	2	No	cmapp2	61.50	-46.67	46.67	-0.2909		3.38	0.2786	-8.8800	-0.5278	76.56	161	114	266	258	201	1000	222
225	Reanalysis	2	1 mon	reanalysis	73.23	-56.55	56.55	-0.2227		2.95	0.2723	-13.3049	-0.5218	92.29	233	153	251	256	243	1136	249
226	Reanalysis	2	1 mon	gpcc	62.39	-50.10	50.10	-0.1803		-0.28	0.1216	-9.3046	-0.3487	78.33	169	117	236	230	206	958	212
227	Reanalysis	2	1 mon	gpcc	52.89	-40.65	40.65	-0.2028		5.27	0.1884	-11.0960	-0.4340	65.50	90	70	247	242	106	755	169
228	Reanalysis	2	1 mon	gpcc trmm	70.63	-59.29	59.29	-0.1817		-15.34	0.0555	-5.1062	-0.2355	87.31	228	178	237	217	228	1086	238
229	Reanalysis	2	1 mon	ghcn cams grid	68.91	-63.26	63.26	-0.2040		3.41	0.1426	-11.5463	-0.3776	86.43	218	202	248	235	224	1127	244
230	Reanalysis	2	1 mon	gpi	112.07	-97.03	97.03	-0.1330		4.60	0.1579	-25.9723	-0.3974	131.55	275	264	220	239	275	1273	270
231	Reanalysis	2	1 mon	trmm 3a46	107.45	-98.92	98.92	-0.2409		1.47	0.1429	-12.1539	-0.3780	123.01	273	271	257	236	272	1309	274
232	Reanalysis	2	1 mon	trmm 3b43	78.99	-69.36	69.36	-0.1345		-15.74	0.0378	-6.5033	-0.1944	96.79	246	230	222	208	248	1154	252
233	Reanalysis	2	1 mon	ssmi	84.38	-74.85	74.85	-0.1533		3.32	0.0940	-13.8258	-0.3068	98.29	251	248	228	224	250	1201	282
234	Reanalysis	2	1 mon	opi2	56.34	-42.76	42.76	-0.2174		-0.31	0.1449	-7.4697	-0.3807	71.01	112	80	249	237	153	831	189
235	Reanalysis	2	1 mon	opi3	56.85	-44.23	44.23	-0.1913		-0.94	0.1183	-7.7564	-0.3440	72.20	115	92	240	226	162	835	191
236	Reanalysis	2	1 mon	cmapp	58.39	-48.41	48.41	-0.1967		-0.31	0.1257	-8.1428	-0.3545	73.78	125	105	243	231	177	881	199
237	Reanalysis	2	1 mon	cmapp2	58.55	-46.47	46.47	-0.1924		-0.46	0.1213	-8.1623	-0.3483	73.86	127	108	241	229	179	884	203
238	Reanalysis	2	2 mon	reanalysis	69.81	-58.24	58.24	-0.1277		-1.61	0.0895	-12.1997	-0.2992	88.65	220	144	217	223	230	1034	230
239	Reanalysis	2	2 mon	gpcc	59.28	-50.06	50.06	-0.0441		-5.93	0.0073	-8.2928	-0.0854	74.38	136	115	189	193	183	816	183
240	Reanalysis	2	2 mon	gpcc	49.86	-40.27	40.27	-0.0751		0.88	0.0282	-10.6534	-0.1679	61.89	70	63	204	206	75	618	121
241	Reanalysis	2	2 mon	gpcc trmm	67.99	-59.96	59.96	-0.0039		-22.15	0.0000	-4.5099	-0.0050	83.78	206	181	184	184	221	976	215
242	Reanalysis	2	2 mon	ghcn cams grid	66.32	-63.21	63.21	-0.0518		-4.93	0.0092	-10.5076	-0.0961	82.77	197	199	187	197	219	1009	225
243	Reanalysis	2	2 mon	gpi	108.76	-96.84	96.84	-0.0574		-1.89	0.0302	-25.0992	-0.1737	127.99	274	262	198	207	274	1215	265
244	Reanalysis	2	2 mon	trmm 3a46	105.34	-100.51	100.51	-0.1031		-9.65	0.0253	-11.4550	-0.1592	121.12	271	275	213	205	265	1229	286
245	Reanalysis	2	2 mon	trmm 3b43	75.25	-70.05	70.05	0.0240	41.6667	-23.44	0.0012	-5.8343	0.0347	93.30	240	237	169	173	245	1064	234
246	Reanalysis	2	2 mon	ssmi	82.40	-74.95	74.95	-0.0459		-3.86	0.0085	-13.0019	-0.0920	95.55	250	249	192	196	247	1134	247
247	Reanalysis	2	2 mon	opi2	52.88	-42.75	42.75	-0.0587		-5.75	0.0106	-6.5046	-0.1029	66.84	89	79	200	198	118	684	149
248	Reanalysis	2	2 mon	opi3	53.74	-44.20	44.20	-0.0434		-6.22	0.0061	-6.8079	-0.0782	68.18	91	91	188	191	126	687	152
249	Reanalysis	2	2 mon	cmapp	55.39	-46.37	46.37	-0.0507		-5.84	0.0084	-7.1999	-0.0915	69.87	107	103	196	195	142	743	164
250	Reanalysis	2	2 mon	cmapp2	55.41	-46.43	46.43	-0.0451		-6.06	0.0067	-7.2032	-0.0818	69.88	108	106	190	192	143	739	161
251	Reanalysis	2	3 mon	reanalysis	67.01	-56.20	56.20	-0.0390		-5.82	0.0084	-11.2411	-0.0915	85.36	203	143	186	194	223	949	211
252	Reanalysis	2	3 mon	gpcc	56.20	-50.10	50.10	0.0879	11.3766	-11.43	0.0291	-7.3148	0.1705	70.35	111	116	152	145	147	671	142
253	Reanalysis	2	3 mon	gpcc	45.94	-40.08	40.08	0.0717	13.9470	-4.32	0.0277	-9.8747	0.1663	57.59	36	61	156	148	42	443	63
254	Reanalysis	2	3 mon	gpcc trmm	65.36	-60.93	60.93	0.1037	9.6432	-26.37	0.0177	-4.1572	0.1331	81.94	187	185	144	154	216	886	206
255	Reanalysis	2	3 mon	ghcn cams grid	64.67	-63.25	63.25	0.0924	10.8225	-12.84	0.0294	-9.5287	0.1715	79.17	182	201	150	144	209	886	205
256	Reanalysis	2	3 mon	gpi	105.27	-96.40	96.40	0.0192	52.0833	-8.48	0.0035	-24.0848	0.0589	124.39	270	260	172	167	273	1142	251
257	Reanalysis	2	3 mon	trmm 3a46	102.49	-100.52	100.52	0.0380	26.3158	-21.26	0.0034	-10.5743	0.0587	118.21	268	276	164				

Amazon Basin - Yearly Surface																						
OI	Runoff	Re	Storage	Precipitation	Absolute Error	Bias	Abias	Slope	Oslope	Intercept	R <sup>2</sup>	R <sup>2</sup> 1 1	Corr Coef	RMSE	AER	BR	SR	CCR	RMR	TR	BR	Plot
1	Surface	1	Yes	reanalysis	54.96	54.96	54.96	-0.0407		46.82	0.0017	-26.4887	-0.0409	57.18	50	50	65	65	48	278	67	Plot
2	Surface	1	Yes	gpcp	61.85	61.85	61.85	0.5034	1.9865	-15.09	0.2369	-28.2185	0.4867	62.93	67	67	19	23	62	238	44	
3	Surface	1	Yes	gpcp	67.48	67.48	67.48	0.4912	2.0358	-18.53	0.4307	-66.2576	0.6563	68.01	76	76	22	9	75	258	56	
4	Surface	1	Yes	ghcn cams grid	46.41	46.41	46.41	0.6522	1.5333	-15.54	0.4827	-20.5386	0.6948	49.04	36	37	9	2	31	115	11	
5	Surface	1	Yes	gpi	11.46	9.72	9.72	0.7762	1.2883	11.81	0.4458	-0.1604	0.6677	13.87	3	3	7	7	3	23	3	
6	Surface	1	Yes	ssmi	34.43	34.43	34.43	0.4186	2.3889	22.92	0.2710	-11.1009	0.5206	36.36	11	11	26	19	10	77	8	
7	Surface	1	Yes	opi2	69.99	69.99	69.99	0.7145	1.3996	-43.11	0.3624	-27.1473	0.6020	70.88	83	83	8	15	80	269	62	
8	Surface	1	Yes	opi3	67.40	67.40	67.40	0.5128	1.9501	-21.53	0.2778	-37.8521	0.5270	68.26	75	75	17	18	76	261	58	
9	Surface	1	Yes	cmep	64.81	64.81	64.81	0.4986	2.0056	-17.60	0.2604	-34.7413	0.5103	65.74	71	71	20	21	68	251	51	
10	Surface	1	Yes	cmep2	64.92	64.92	64.92	0.5122	1.9524	-18.99	0.2643	-33.5051	0.5141	65.86	72	72	18	20	69	251	52	
11	Surface	1	No	reanalysis	54.14	54.14	54.14	0.0966	10.3520	33.48	0.0061	-25.1626	0.0779	55.79	46	46	52	53	44	241	46	
12	Surface	1	No	gpcp	62.08	62.08	62.08	0.4395	2.2753	-9.18	0.1213	-28.5117	0.3483	63.24	69	69	25	35	63	261	57	
13	Surface	1	No	gpcp	68.21	68.21	68.21	0.5558	1.7992	-25.16	0.3659	-67.5274	0.6049	68.64	80	80	16	13	78	267	59	
14	Surface	1	No	ghcn cams grid	45.58	45.58	45.58	0.6330	1.5798	-13.88	0.3055	-20.8812	0.5527	49.43	33	36	12	16	33	130	13	
15	Surface	1	No	gpi	11.96	10.44	10.44	0.9635	1.0379	-6.91	0.4558	-0.2023	0.6751	14.12	4	4	2	5	4	19	2	
16	Surface	1	No	ssmi	33.39	33.39	33.39	0.5886	1.6989	6.76	0.3733	-10.0131	0.6109	34.69	8	8	14	12	9	51	6	
17	Surface	1	No	opi2	70.23	70.23	70.23	0.9748	1.0259	-67.85	0.4531	-27.1801	0.6731	70.92	84	84	1	6	81	256	55	
18	Surface	1	No	opi3	67.63	67.63	67.63	0.4186	2.4004	-12.56	0.1231	-38.2571	0.3509	68.61	77	77	27	34	77	292	71	
19	Surface	1	No	cmep	65.04	65.04	65.04	0.3911	2.5569	-7.58	0.1077	-35.1391	0.3281	66.11	73	73	30	38	70	284	69	
20	Surface	1	No	cmep2	65.15	65.15	65.15	0.4183	2.4021	-10.06	0.1173	-33.8814	0.3425	66.22	74	74	28	36	71	283	68	
21	Surface	2	Yes	reanalysis	42.32	42.32	42.32	0.3485	2.8694	18.08	0.1015	-5.0462	0.3186	47.53	26	28	35	39	27	155	17	Plot
22	Surface	2	Yes	gpcp	51.77	51.77	51.77	0.0568	17.6056	35.67	0.0130	-38.2222	0.1141	55.09	44	44	56	51	42	237	42	
23	Surface	2	Yes	gpcp	55.23	55.23	55.23	0.2546	3.9277	14.39	0.4672	-93.2144	0.6836	56.61	51	51	42	4	45	193	26	
24	Surface	2	Yes	ghcn cams grid	38.21	38.21	38.21	0.2069	4.8333	35.31	0.1792	-22.0159	0.4233	41.45	17	17	44	28	19	125	12	
25	Surface	2	Yes	gpi	8.70	-2.54	2.54	0.4920	2.0325	49.98	0.5042	-0.0879	0.7100	11.32	2	2	21	1	2	28	4	
26	Surface	2	Yes	ssmi	32.26	32.26	32.26	0.3179	3.1456	40.59	0.2949	-14.3702	0.5430	30.36	6	6	38	17	6	73	7	
27	Surface	2	Yes	opi2	59.92	59.92	59.92	0.3091	4.77	0.1835	-23.9010	0.4284	62.20	63	63	40	27	59	252	53		
28	Surface	2	Yes	opi3	57.32	57.32	57.32	0.1039	9.8246	25.76	0.0575	-60.3438	0.2398	59.93	59	59	51	42	57	268	80	
29	Surface	2	Yes	cmep	54.74	54.74	54.74	0.0945	10.5820	29.22	0.0473	-55.2140	0.2176	57.52	48	48	53	44	50	243	47	
30	Surface	2	Yes	cmep2	54.85	54.85	54.85	0.1047	9.5511	28.16	0.0552	-52.5762	0.2350	57.60	49	49	50	43	52	243	48	
31	Surface	2	No	reanalysis	43.99	43.99	43.99	0.3478	2.8752	17.56	0.0276	-5.2483	0.1660	48.32	30	31	36	47	30	174	20	
32	Surface	2	No	gpcp	53.44	53.44	53.44	0.3985	2.5094	3.33	0.1747	-37.1406	0.4180	54.32	45	45	29	30	41	190	24	
33	Surface	2	No	gpcp	58.75	58.75	58.75	0.3896	2.5667	0.41	0.3634	-102.0142	0.6028	59.20	60	60	31	14	55	220	36	
34	Surface	2	No	ghcn cams grid	39.88	39.88	39.88	0.5921	1.6889	-1.38	0.3997	-21.1023	0.6322	40.61	21	22	13	10	17	83	9	
35	Surface	2	No	gpi	6.82	0.99	0.99	0.7973	1.2542	18.66	0.4397	0.4029	0.6631	8.38	1	1	6	8	1	17	1	Plot
36	Surface	2	No	ssmi	24.93	24.93	24.93	0.4509	2.2178	28.65	0.3993	-10.5629	0.6319	26.33	5	5	24	11	5	50	5	
37	Surface	2	No	opi2	61.59	61.59	61.59	0.9338	1.0709	-55.34	0.4776	-23.9375	0.6911	62.25	66	66	3	3	60	198	29	
38	Surface	2	No	opi3	59.00	59.00	59.00	0.3756	2.6624	-0.06	0.2050	-59.8000	0.4528	59.67	61	61	32	24	56	234	40	
39	Surface	2	No	cmep	56.41	56.41	56.41	0.3502	2.8555	4.92	0.1773	-54.5000	0.4211	57.15	53	54	34	29	47	217	34	
40	Surface	2	No	cmep2	56.52	56.52	56.52	0.3754	2.6638	2.44	0.1937	-51.9369	0.4401	57.25	55	56	33	25	49	218	35	
41	Reanalysis	1	Yes	reanalysis	39.73	-39.73	39.73	-0.1017		-1.06	0.0700	-43.4950	-0.2646	44.57	20	21	69	77	23	210	32	
42	Reanalysis	1	Yes	gpcp	41.39	-41.39	41.39	-0.0149		-5.81	0.0007	-32.9295	-0.0263	44.21	24	25	64	63	22	198	28	
43	Reanalysis	1	Yes	gpcp	33.42	-33.42	33.42	0.2165	4.6189	-10.92	0.1326	-47.5344	0.3641	34.36	9	9	43	31	8	100	10	
44	Reanalysis	1	Yes	gpcp trmm	48.77	-48.77	48.77	-0.5159		6.43	0.6294	-41.8542	-0.7933	52.45	40	40	90	90	39	299	74	
45	Reanalysis	1	Yes	ghcn cams grid	56.44	-56.44	56.44	0.1172	8.5324	-10.72	0.0327	-71.8754	0.1807	57.56	54	55	47	46	51	253	54	
46	Reanalysis	1	Yes	gpi	88.82	-88.82	88.82	0.1425	7.0175	-18.89	0.0819	-163.7118	0.2862	89.89	88	88	46	40	88	350	83	
47	Reanalysis	1	Yes	trmm 3a46	85.93	-85.93	85.93	-0.4760		28.95	0.5739	-147.1636	-0.7576	87.71	86	86	88	89	86	435	91	
48	Reanalysis	1	Yes	trmm 3b43	55.96	-55.96	55.96	-0.4857		11.33	0.3773	-53.1304	-0.6142	58.24	52	53	89	87	53	334	81	
49	Reanalysis	1	Yes	ssmi	67.82	-67.82	67.82	0.1425	7.0175	-15.72	0.0248	-78.2881	0.2069	68.92	78	78	45	45	79	325	80	
50	Reanalysis	1	Yes	opi2	34.07	-33.95	33.95	-0.1206		-3.00	0.0601	-25.1288	-0.2451	38.80	10	10	71	76	12	179	21	
51	Reanalysis	1	Yes	opi3	35.44	-35.44	35.44	0.0284	35.2113	-7.16	0.0022	-24.4140	0.0472	38.26	13	13	59	57	11	153	16	
52	Reanalysis	1	Yes	cmep	37.67	-37.67	37.67	0.0370	27.0270	-7.49	0.0036	-27.0391	0.0597	40.19	15	15	58	55	15	158	18	
53	Reanalysis	1	Yes	cmep2	37.73	-37.73	37.73	0.0262	38.1679	-7.15	0.0019	-27.2770	0.0431	40.36	16	16	60	58	16	166	19	
54	Reanalysis	1	No	reanalysis	40.51	-40.51	40.51	-0.1203		-1.19	0.2452	-112.9467	-0.4952	45.10	22	23	70	85	24	224	39	
55	Reanalysis	1	No	gpcp	42.08	-42.08	42.08	-0.1931		-0.25	0.1887	-57.2183	-0.4344	45.32	25	27	76	83	25	236	41	
56	Reanalysis	1	No	gpcp	32.70	-32.70	32.70	-0.0021		-3.92	0.0010	-999.0000	-0.0308	33.74	7	7	62	64	7	147	15	
57	Reanalysis	1	No	gpcp trmm	54.39	-54.39	54.39	-0.3313		-5.91	0.8068	-155.4871	-0.8982	56.85	47	47	83	91	46	314	78	
58	Reanalysis	1	No	ghcn cams grid	57.24	-57.24	57.24	-0.0838		-1.10	0.0418	-187.5480	-0.2045	58.48	58	58	68	72	54	310	76	
59	Reanalysis	1	No	gpi	89.53	-89.53	89.53	0.1165	8.5837	-17.48	0.0616	-187.3727	0.2482	90.61	89	89	48	41	89	356	86	
60	Reanalysis	1	No	trmm 3a46	93.12	-93.12	93.12	-0.3689		11.58	0.8788	-436.8279	-0.9375	94.44	90	90	85	92	90	447	92	
61	Reanalysis	1	No	trmm 3b43	60.46	-60.46	60.46	-0.4272		4.18	0.3809	-80.0065	-0.6172	62.36	65	65	87	88	61	366	87	
62	Reanalysis	1	No	ssmi	69.92	-69.92	69.92	0.0660	15.1515	-13.18	0.0108	-98.0713	0.1039	71.06	82	82	55	52	82	353	84	
63	Reanalysis	1	No	opi2	34.64	-34.64	3															



Arkansas Basin - Monthly Atmospheric																						
QI	Runoff	Re	Storage	Precipitation	Absolute Error	Bias	Slope	Oslope	Intercept	R <sup>2</sup>	R <sup>2</sup> 1.1	Corr Coef	RMSE	AER	BR	SR	CCR	RMR	TR	BR		
1	Surface	1	Yes	reanalysis	148.13	-144.56	0.0955	10.4712	129.10	0.0006	-2.5537	0.0238	172.41	11	11	19	20	9	70	16	Plot	
2	Surface	1	Yes	gpcc	140.51	-136.57	0.8278	1.2080	137.58	0.0565	-2.0375	0.2378	165.40	4	4	6	4	4	22	1	Plot	
3	Surface	1	Yes	gpcc	145.75	-143.04	0.5672	1.7630	140.20	0.0242	-2.2943	0.1556	171.03	6	7	14	15	8	48	7		
4	Surface	1	Yes	gpcc trmm	147.12	-143.75	-0.8505		111.51	0.0660	-1.5520	-0.2570	168.38	7	8	24	24	5	68	15		
5	Surface	1	Yes	ghcn cams gnd	132.83	-128.57	0.5721	1.7479	133.95	0.0305	-2.2194	0.1748	161.18	3	3	13	14	3	36	3		
6	Surface	1	Yes	gpi	107.46	-86.27	0.0481	20.7900	133.91	0.0008	-0.9920	0.0236	132.99	1	1	21	21	1	45	5		
7	Surface	1	Yes	trmm 3b43	141.27	-137.20	0.6941	1.4407	136.15	0.0346	-1.5232	0.1861	174.99	5	5	10	10	11	41	4		
8	Surface	1	Yes	ssmi	127.97	-114.09	0.3970	2.5189	131.75	0.0586	-1.4687	0.2421	151.97	2	2	17	3	2	26	2		
9	Surface	1	Yes	opi2	152.38	-149.03	-0.1756		137.32	0.0020	-2.5723	-0.0446	179.37	13	13	23	23	13	85	22		
10	Surface	1	Yes	opi3	150.56	-147.36	0.5354	1.8678	143.77	0.0223	-2.4188	0.1494	175.48	12	12	15	16	12	67	12		
11	Surface	1	Yes	cmap	147.54	-144.18	0.8123	1.6332	142.33	0.0307	-2.2949	0.1751	172.27	9	10	12	13	8	52	9		
12	Surface	1	Yes	cmap2	147.51	-144.15	0.8132	1.6308	142.32	0.0308	-2.2941	0.1754	172.25	8	9	11	12	7	47	6		
13	Surface	2	Yes	reanalysis	197.36	-193.47	0.6991	1.4304	188.22	0.0207	-2.9581	0.1438	219.02	18	18	9	17	17	79	20	Plot	
14	Surface	2	Yes	gpcc	190.43	-188.12	1.1044	1.1044	184.24	0.0689	-2.7329	0.2625	212.70	15	16	1	2	15	49	8		
15	Surface	2	Yes	gpcc	191.06	-188.12	0.2339	4.2753	184.19	0.0034	-2.6941	0.0584	223.20	16	15	18	18	21	88	23		
16	Surface	2	Yes	gpcc trmm	213.18	-213.18	1.1371	1.1371	214.50	0.0708	-2.8177	0.2657	245.13	24	24	2	1	24	75	18		
17	Surface	2	Yes	ghcn cams gnd	194.12	-191.76	0.7831	1.2770	186.94	0.0356	-2.8729	0.1888	216.65	17	17	8	9	16	67	13		
18	Surface	2	Yes	gpi	147.86	-139.24	0.0775	12.9032	174.92	0.0012	-1.5479	0.0351	174.27	10	6	20	19	10	65	11		
19	Surface	2	Yes	trmm 3b43	208.49	-208.49	0.8607	1.1618	207.80	0.0437	-2.7192	0.2090	241.95	23	23	3	6	23	78	19		
20	Surface	2	Yes	ssmi	179.18	-169.71	0.4540	2.2028	174.95	0.0525	-2.0232	0.2291	198.11	14	14	16	5	14	63	10		
21	Surface	2	Yes	opi2	202.83	-200.81	-0.0448		183.82	0.0001	-3.2424	-0.0099	226.75	22	22	22	22	22	110	24		
22	Surface	2	Yes	opi3	200.83	-198.77	0.7835	1.2763	192.51	0.0328	-3.0998	0.1810	222.90	21	21	7	11	20	80	21		
23	Surface	2	Yes	cmap	197.51	-195.38	0.8563	1.1678	190.34	0.0413	-2.9808	0.2033	219.64	20	20	5	8	19	72	17		
24	Surface	2	Yes	cmap2	197.48	-195.35	0.8569	1.1670	190.33	0.0414	-2.9801	0.2035	219.62	19	19	4	7	18	67	14		

Arkansas Basin - Yearly Atmospheric																					
OI	Runoff	Re	Storage	Precipitation	Absolute Error	Bias	Slope	Oslope	Intercept	R <sup>2</sup>	R <sup>2</sup> 1:1	Corr Coef	RMSE	AER	BR	SR	CCR	RMR	TR	BR	Plot
1	Surface	1	Yes	reanalysis	144.57	-144.57	0.2966	3.3715	132.58	0.0067	-36.1719	0.0616	146.76	10	10	21	20	17	72	14	Plot
2	Surface	1	Yes	gpcp	136.50	-136.50	1.6393	1.6393	135.95	0.1980	-27.4977	0.4450	139.21	4	4	15	2	5	30	3	
3	Surface	1	Yes	gpcp	143.07	-143.07	2.5599	2.5599	153.29	0.3047	-20.8552	0.5520	145.79	7	7	20	1	7	42	7	
4	Surface	1	Yes	gpcp trmm	145.33	-145.33	0.1485	6.7340	140.98	0.0008	-63.6695	0.0275	146.49	11	11	23	23	10	78	17	
5	Surface	1	Yes	ghcn cams grid	128.41	-128.41	1.9892	1.9892	142.06	0.1755	-28.9239	0.4189	142.65	3	3	19	3	6	34	5	
6	Surface	1	Yes	gpi	86.28	-86.28	1.2049	1.2049	72.00	0.1717	-6.9124	0.4143	87.72	1	1	3	4	1	10	1	Plot
7	Surface	1	Yes	trmm 3b43	137.19	-137.19	0.8225	1.2158	136.58	0.0758	-43.4882	0.2752	136.85	5	5	4	15	4	33	4	
8	Surface	1	Yes	ssmi	113.17	-113.17	0.6347	1.5755	121.87	0.0551	-11.9140	0.2347	116.88	2	2	14	18	2	38	8	
9	Surface	1	Yes	opi2	149.19	-149.19	1.5410	1.5410	155.17	0.1358	-32.7549	0.3682	151.51	13	13	12	9	13	60	11	
10	Surface	1	Yes	opi3	147.15	-147.15	1.6478	1.6478	153.26	0.1585	-31.9357	0.3981	149.66	12	12	16	5	12	57	10	
11	Surface	1	Yes	cmep	143.77	-143.77	1.5388	1.5388	147.05	0.1561	-30.4629	0.3951	146.28	9	9	11	7	9	45	9	
12	Surface	1	Yes	cmep2	143.73	-143.73	1.5456	1.5456	147.06	0.1564	-30.4541	0.3955	146.28	8	8	13	6	8	43	8	
13	Surface	2	Yes	reanalysis	193.42	-193.42	0.5178	1.9312	187.13	0.0045	-34.7521	0.0669	192.49	18	18	18	21	18	83	23	Plot
14	Surface	2	Yes	gpcp	188.07	-188.07	1.8358	1.8358	184.06	0.1543	-32.6000	0.3928	186.61	15	15	17	8	15	70	13	
15	Surface	2	Yes	gpcp	188.15	-188.15	1.3584	1.3584	191.47	0.0696	-25.5812	0.2638	191.55	16	16	9	16	17	74	15	
16	Surface	2	Yes	gpcp trmm	213.33	-213.33	1.4515	1.4515	217.89	0.0996	-95.8695	0.3156	214.34	24	24	10	10	24	92	22	
17	Surface	2	Yes	ghcn cams grid	191.70	-191.70	1.0780	1.0780	187.82	0.0573	-33.8864	0.2393	190.15	17	17	2	17	16	69	12	
18	Surface	2	Yes	gpi	139.30	-139.30	0.9448	1.0584	130.86	0.0804	-11.7998	0.2835	132.92	6	6	1	14	3	30	2	
19	Surface	2	Yes	trmm 3b43	208.64	-208.64	-1.7598		194.92	0.0855	-91.6160	-0.2925	209.92	23	23	24	24	23	117	24	
20	Surface	2	Yes	ssmi	168.04	-168.04	0.1614	8.1968	175.75	0.0621	-14.2258	0.0454	162.76	14	14	22	22	14	86	20	
21	Surface	2	Yes	opi2	200.76	-200.76	0.7666	1.3045	193.66	0.0467	-37.2177	0.2162	199.02	22	22	6	19	22	91	21	
22	Surface	2	Yes	opi3	198.72	-198.72	1.2967	1.2967	197.76	0.0843	-36.4858	0.2904	197.11	21	21	5	13	21	81	19	
23	Surface	2	Yes	cmep	195.33	-195.33	1.3158	1.3158	193.45	0.0908	-35.2051	0.3014	193.71	20	20	8	11	20	79	18	
24	Surface	2	Yes	cmep2	195.30	-195.30	1.3157	1.3157	193.42	0.0905	-35.1977	0.3009	193.69	19	19	7	12	19	76	18	

Arkansas Basin - Monthly Surface

Ol	Runoff	Re	Storage	Precipitation	Absolute Error	Bias	Abias	Slope	Oslope	Intercept	R <sup>2</sup>	R <sup>2</sup> 1:1	Corr Coef	RMSE	AER	BR	SR	CCR	RMR	TR	BR	Plot
1	Surface	1	Yes	reanalysis	28.17	24.14	24.14	0.4639	2.1556	-20.45	0.4400	-1.2662	0.6633	34.36	160	213	97	5	156	631	132	Plot
2	Surface	1	Yes	gpcp	24.53	5.71	5.71	0.4258	2.3485	-1.34	0.3215	-0.3070	0.5670	31.16	136	13	102	21	123	395	66	
3	Surface	1	Yes	gpcp	25.37	14.62	14.62	0.4433	2.2558	-10.13	0.3749	-0.5362	0.6123	32.05	149	115	100	13	144	521	112	
4	Surface	1	Yes	ghcn cams grid	24.46	8.35	8.35	0.4465	2.2396	-4.09	0.3522	-0.2818	0.5935	31.07	132	29	99	18	122	400	67	
5	Surface	1	Yes	gpi	49.43	-45.47	45.47	0.6185	1.6221	48.57	0.2273	-0.8304	0.4768	62.48	237	264	72	31	222	826	167	
6	Surface	1	Yes	ssmi	63.80	-22.92	22.92	-0.7707		34.11	0.2049	-1.0277	-0.4527	83.99	264	210	257	263	264	1258	264	
7	Surface	1	Yes	opi2	24.31	18.07	18.07	0.4936	2.0259	-14.21	0.5523	-0.5905	0.7431	30.41	128	152	87	2	111	480	96	
8	Surface	1	Yes	opi3	25.29	16.23	16.23	0.4879	2.0496	-12.33	0.4473	-0.4210	0.6688	31.57	148	133	88	4	135	508	108	
9	Surface	1	Yes	cmap	24.43	12.81	12.81	0.4838	2.0670	-8.87	0.4187	-0.2806	0.6471	30.71	131	83	92	7	118	431	79	
10	Surface	1	Yes	cmap2	24.42	12.78	12.78	0.4839	2.0665	-8.85	0.4187	-0.2798	0.6471	30.70	130	81	91	8	117	427	77	
11	Surface	1	No	reanalysis	26.82	24.11	24.11	0.4723	2.1173	-20.49	0.0212	-1.1208	0.1458	33.24	154	212	96	140	154	756	153	
12	Surface	1	No	gpcp	21.42	5.83	5.83	0.8048	1.2425	-4.32	0.0504	0.0017	0.2245	27.24	37	15	39	107	34	232	14	
13	Surface	1	No	gpcp	22.46	14.91	14.91	0.6276	1.5934	-11.80	0.0405	-0.3062	0.2012	29.55	72	129	71	119	89	480	95	
14	Surface	1	No	ghcn cams grid	22.94	8.24	8.24	0.3288	3.0414	-3.15	0.0080	-0.1155	0.0894	28.98	82	28	116	173	72	471	94	
15	Surface	1	No	gpi	49.55	-45.19	45.19	1.8960	1.8960	37.70	0.1158	-0.8675	0.3403	63.11	240	262	80	59	231	872	182	
16	Surface	1	No	ssmi	45.33	-20.88	20.88	1.5270	1.5270	16.47	0.0463	-0.0845	0.2152	61.42	216	190	65	110	213	794	161	
17	Surface	1	No	opi2	24.08	18.18	18.18	-0.2614		-8.43	0.0068	-0.7201	-0.0824	31.62	122	155	251	227	140	895	192	
18	Surface	1	No	opi3	24.02	16.34	16.34	0.3846	2.6001	-11.59	0.0122	-0.3999	0.1104	31.33	118	135	108	161	129	651	137	
19	Surface	1	No	cmap	23.10	12.92	12.92	0.4846	2.0636	-8.93	0.0184	-0.2290	0.1358	30.09	89	87	89	147	101	513	110	
20	Surface	1	No	cmap2	23.10	12.90	12.90	0.4845	2.0640	-8.91	0.0184	-0.2283	0.1357	30.08	87	86	90	148	100	511	109	
21	Surface	1	1 mon	reanalysis	26.34	24.10	24.10	0.9713	1.0295	-23.90	0.0899	-1.0232	0.2998	32.49	152	211	3	70	147	583	123	
22	Surface	1	1 mon	gpcp	20.36	5.75	5.75	1.3040	1.3040	-8.11	0.1324	0.0807	0.3638	26.16	13	14	50	49	15	141	4	
23	Surface	1	1 mon	ghcn	21.71	14.82	14.82	1.0906	1.0906	-15.58	0.1224	-0.2054	0.3498	28.46	48	128	9	55	61	301	38	
24	Surface	1	1 mon	ghcn cams grid	21.83	8.21	8.21	1.0655	1.0655	-8.71	0.0838	-0.0059	0.2895	27.55	52	27	6	74	45	204	8	
25	Surface	1	1 mon	gpi	48.48	-44.75	44.75	2.8724	2.8724	29.06	0.2681	-0.7874	0.5177	61.66	230	258	113	25	216	842	171	
26	Surface	1	1 mon	ssmi	47.27	-21.55	21.55	-0.2627		32.04	0.0014	-0.1640	-0.0372	63.64	224	200	252	219	236	131	245	
27	Surface	1	1 mon	opi2	23.89	18.25	18.25	0.3982	2.5113	-13.59	0.0157	-0.5911	0.1255	30.47	118	162	106	155	113	652	138	
28	Surface	1	1 mon	opi3	23.34	16.30	16.30	1.0414	1.0414	-16.62	0.0893	-0.2886	0.2989	30.11	101	134	5	71	103	414	71	
29	Surface	1	1 mon	cmap	22.25	12.87	12.87	1.1413	1.1413	-13.96	0.1022	-0.1235	0.3196	28.81	68	85	20	63	69	305	41	
30	Surface	1	1 mon	cmap2	22.25	12.85	12.85	1.1411	1.1411	-13.94	0.1021	-0.1228	0.3196	28.80	67	84	19	64	68	302	40	
31	Surface	1	2 mon	reanalysis	27.11	24.17	24.17	0.8868	1.4560	-22.02	0.0451	-1.0873	0.2124	32.95	155	214	60	115	150	694	143	
32	Surface	1	2 mon	gpcp	21.43	5.96	5.96	0.8282	1.2074	-4.61	0.0538	0.0037	0.2319	27.18	38	18	34	104	32	226	11	
33	Surface	1	2 mon	gpcp	22.71	15.11	15.11	0.8491	1.1777	-13.85	0.0752	-0.2695	0.2743	29.10	75	130	27	77	73	382	83	
34	Surface	1	2 mon	ghcn cams grid	22.18	8.40	8.40	0.9187	1.0885	-7.78	0.0360	-0.0319	0.2510	27.78	61	30	8	87	51	237	15	
35	Surface	1	2 mon	gpi	48.18	-44.33	44.33	2.1100	2.1100	35.01	0.1460	-0.8204	0.3821	62.15	228	254	94	44	219	839	170	
36	Surface	1	2 mon	ssmi	49.42	-21.20	21.20	-1.8269		43.48	0.0339	-0.2145	-0.2321	65.33	236	193	201	252	242	1184	254	
37	Surface	1	2 mon	opi2	24.14	18.42	18.42	0.6739	1.4839	-15.89	0.0454	-0.5490	0.2131	30.01	124	167	62	114	96	563	121	
38	Surface	1	2 mon	opi3	24.04	16.48	16.48	0.8644	1.1569	-15.42	0.0619	-0.3266	0.2488	30.51	120	138	23	90	114	485	97	
39	Surface	1	2 mon	cmap	22.96	13.05	13.05	0.8812	1.1348	-12.13	0.0613	-0.1710	0.2478	29.37	84	93	17	92	76	362	57	
40	Surface	1	2 mon	cmap2	22.95	13.03	13.03	0.8808	1.1353	-12.10	0.0612	-0.1703	0.2475	29.36	83	91	18	93	75	360	55	
41	Surface	1	3 mon	reanalysis	27.25	24.17	24.17	0.8547	1.5274	-21.80	0.0410	-1.0918	0.2024	33.01	156	215	66	117	152	706	149	
42	Surface	1	3 mon	gpcp	21.75	5.89	5.89	0.7267	1.3761	-3.77	0.0414	-0.0111	0.2035	27.44	49	16	55	116	39	275	28	
43	Surface	1	3 mon	gpcp	23.13	15.16	15.16	0.6856	1.5024	-12.34	0.0481	-0.3074	0.2148	29.65	94	131	64	111	90	490	99	
44	Surface	1	3 mon	ghcn cams grid	22.14	8.40	8.40	0.9926	1.0075	-8.35	0.0735	-0.0207	0.2711	27.67	58	31	1	80	48	218	10	
45	Surface	1	3 mon	gpi	49.13	-44.58	44.58	1.4982	1.4982	49.38	0.0738	-0.8678	0.2717	63.05	234	256	63	79	230	862	181	
46	Surface	1	3 mon	ssmi	49.71	-21.79	21.79	-1.9601		46.67	0.0780	-0.2346	-0.2794	65.99	242	201	262	256	245	1206	258	
47	Surface	1	3 mon	opi2	23.34	18.39	18.39	0.8434	1.1857	-17.17	0.0111	-0.5103	0.2666	29.70	100	165	28	81	93	467	92	
48	Surface	1	3 mon	opi3	24.04	16.46	16.46	0.8541	1.1708	-15.33	0.0604	-0.3257	0.2458	30.57	121	136	24	95	116	492	101	
49	Surface	1	3 mon	cmap	23.07	13.03	13.03	0.8314	1.2028	-11.72	0.0545	-0.1771	0.2335	29.51	85	90	32	98	86	391	64	
50	Surface	1	3 mon	cmap2	23.07	13.01	13.01	0.8307	1.2038	-11.69	0.0544	-0.1765	0.2333	29.51	86	89	33	101	84	393	65	
51	Surface	1	4 mon	reanalysis	27.32	24.22	24.22	0.6932	1.4426	-22.12	0.0461	-1.0901	0.2146	32.99	157	216	59	112	151	695	144	
52	Surface	1	4 mon	gpcp	21.49	5.94	5.94	0.7080	1.4124	-3.68	0.0393	-0.0147	0.1983	27.53	40	17	57	120	44	278	31	
53	Surface	1	4 mon	gpcp	22.84	15.27	15.27	0.7638	1.3092	-13.29	0.0699	-0.2904	0.2467	29.51	77	132	51	94	85	439	82	
54	Surface	1	4 mon	ghcn cams grid	21.97	8.49	8.49	0.9352	1.0693	-8.00	0.0654	-0.0312	0.2558	27.79	54	32	7	84	52	229	13	
55	Surface	1	4 mon	gpi	49.36	-44.78	44.7															

Arkansas Basin - Monthly Surface

QI	Runoff	Re	Storage	Precipitation	Absolute Error	Bias	Abias	Slope	Oslope	Intercept	R <sup>2</sup>	R <sup>2</sup> 1:1	Corr Coef	RMSE	AER	BR	SR	CCR	RMR	TR	BR
102	Surface	2	3 mon	gpcp	21.78	7.66	7.66	4.006	2.4983	-3.00	0.0137	-0.1024	0.1169	27.49	50	23	105	159	41	378	82
103	Surface	2	3 mon	gpcp	23.59	17.84	17.84	0.6369	1.5701	-15.01	0.0355	-0.3836	0.1883	33.29	111	148	70	126	155	810	125
104	Surface	2	3 mon	ghcn cams grid	22.81	11.22	11.22	0.4398	2.2738	-8.87	0.0182	-0.1903	0.1271	28.83	74	67	101	154	70	486	91
105	Surface	2	3 mon	gpi	50.19	-41.90	41.90	1.2595	1.2595	39.71	0.0454	-0.8742	0.2131	63.99	247	248	43	113	239	890	189
106	Surface	2	3 mon	ssmi	47.52	-18.26	18.26	-2.1580	44.79	0.0997	-0.2137	-0.3157	0.1636	63.68	225	163	264	260	237	1149	247
107	Surface	2	3 mon	opi2	24.89	20.15	20.15	0.5173	1.9331	-18.40	0.0284	-0.6824	0.1624	31.56	142	185	82	131	134	674	140
108	Surface	2	3 mon	opi3	24.85	18.22	18.22	0.5280	1.8938	-14.56	0.0250	-0.5058	0.1582	31.29	138	157	79	135	127	836	134
109	Surface	2	3 mon	cmcp	23.49	14.79	14.79	0.5053	1.9790	-10.95	0.0217	-0.3178	0.1474	30.07	105	124	83	138	99	549	117
110	Surface	2	3 mon	cmcp2	23.46	14.77	14.77	0.5046	1.9818	-10.93	0.0216	-0.3170	0.1471	30.07	104	122	84	139	98	547	116
111	Surface	2	4 mon	reanalysis	21.53	13.29	13.29	0.0699	14.3062	-8.09	0.0008	-0.4401	0.0236	27.26	42	99	164	192	35	532	114
112	Surface	2	4 mon	gpcp	22.04	7.71	7.71	0.1979	5.0531	-1.50	0.0033	-0.1378	0.0578	27.97	56	24	126	183	53	442	83
113	Surface	2	4 mon	gpcp	23.54	17.95	17.95	0.7813	1.2799	-16.28	0.0533	-0.3580	0.2309	33.09	107	150	46	105	153	561	119
114	Surface	2	4 mon	ghcn cams grid	22.29	11.36	11.36	0.4189	2.3872	-8.87	0.0148	-0.1992	0.1215	28.89	69	70	103	157	71	470	93
115	Surface	2	4 mon	gpi	51.06	-42.10	42.10	0.6430	1.5552	45.09	0.0118	-0.7118	0.1088	64.91	252	250	67	165	241	975	209
116	Surface	2	4 mon	ssmi	52.16	-17.44	17.44	-2.0799	43.77	0.0631	-0.2053	-0.3052	0.1636	63.63	253	140	263	259	235	1150	248
117	Surface	2	4 mon	opi2	24.84	20.17	20.17	0.3598	2.7809	-15.21	0.0127	-0.7118	0.1129	31.90	140	186	112	160	142	740	151
118	Surface	2	4 mon	opi3	24.72	18.24	18.24	0.4032	2.4902	-13.62	0.0146	-0.5269	0.1208	31.57	139	156	104	158	137	697	145
119	Surface	2	4 mon	cmcp	23.58	14.81	14.81	0.3785	2.6420	-10.00	0.0122	-0.3391	0.1103	30.38	110	127	109	162	109	617	130
120	Surface	2	4 mon	cmcp2	23.56	14.80	14.80	0.3773	2.6504	-9.97	0.0121	-0.3383	0.1100	30.38	108	125	110	163	108	614	127
121	Reanalysis	1	Yes	reanalysis	26.14	18.91	18.91	0.8983	1.1169	17.18	0.3948	0.0661	0.8283	32.14	151	169	12	11	145	468	98
122	Reanalysis	1	Yes	gpcp	24.85	1.42	1.42	0.7284	1.3767	1.92	0.2925	0.2494	0.8408	31.60	141	1	56	24	139	361	56
123	Reanalysis	1	Yes	gpcp	24.47	8.74	8.74	0.7785	1.2845	7.29	0.3152	0.2303	0.8616	31.45	133	35	48	22	132	370	61
124	Reanalysis	1	Yes	gpcp trmm	22.86	11.82	11.82	0.7731	1.2935	9.80	0.4506	0.2867	0.8716	27.72	80	73	49	3	49	254	20
125	Reanalysis	1	Yes	ghcn cams grid	24.16	2.88	2.88	0.8411	1.5598	2.47	0.2649	0.1755	0.5147	30.24	125	10	68	27	106	336	49
126	Reanalysis	1	Yes	gpi	50.80	-44.97	44.97	0.3354	2.9815	-12.88	0.1959	-2.2061	0.4428	63.01	250	261	113	33	229	868	188
127	Reanalysis	1	Yes	trmm 3b43	22.88	8.98	8.98	0.6855	1.4588	5.87	0.3868	0.2586	0.8219	28.01	81	19	61	12	55	228	12
128	Reanalysis	1	Yes	ssmi	63.23	-21.39	21.39	-0.3279	10.40	0.2628	-4.4178	-0.5127	0.1636	61.85	263	198	255	264	263	1241	263
129	Reanalysis	1	Yes	opi2	22.72	14.28	14.28	1.1428	1.588	15.88	0.5598	0.3951	0.7481	28.14	78	109	21	5	264	21	
130	Reanalysis	1	Yes	opi3	23.87	12.81	12.81	0.8888	1.1254	11.54	0.4387	0.3103	0.8623	30.04	114	78	15	6	97	310	42
131	Reanalysis	1	Yes	cmcp	23.19	8.98	8.98	0.8498	1.1767	8.11	0.4108	0.3367	0.8419	29.54	98	45	25	9	88	265	24
132	Reanalysis	1	Yes	cmcp2	23.18	8.95	8.95	0.8493	1.1774	8.08	0.4108	0.3369	0.8409	29.54	97	44	26	10	87	264	22
133	Reanalysis	1	No	reanalysis	24.34	18.93	18.93	0.0098	102.0408	2.37	0.0021	-35.0573	0.0454	30.30	129	170	161	187	107	784	180
134	Reanalysis	1	No	gpcp	22.23	1.81	1.81	-0.0008	3.43	0.0000	-17.7499	-0.0036	27.98	66	2	202	203	54	527	113	
135	Reanalysis	1	No	gpcp	21.64	9.03	9.03	-0.0028	2.45	0.0004	-50.5117	-0.0188	27.73	46	48	204	210	50	558	118	
136	Reanalysis	1	No	gpcp trmm	25.62	13.49	13.49	0.0155	64.5161	5.57	0.0019	-9.3168	0.0435	32.77	150	101	188	198	149	777	157
137	Reanalysis	1	No	ghcn cams grid	21.80	2.75	2.75	-0.0115	2.21	0.0037	-26.9247	-0.0610	27.85	44	12	206	226	47	535	115	
138	Reanalysis	1	No	gpi	50.21	-44.63	44.63	0.0452	12.1239	1.47	0.0875	-78.5855	0.2957	63.28	248	257	179	172	232	988	213
139	Reanalysis	1	No	trmm 3b43	24.99	8.78	8.78	0.0453	22.0751	5.51	0.0174	-8.4632	0.1320	31.19	145	36	178	149	124	632	133
140	Reanalysis	1	No	ssmi	45.27	-20.04	20.04	-0.0407	4.87	0.0871	-63.7693	-0.2952	61.08	215	184	219	258	212	1088	236	
141	Reanalysis	1	No	opi2	22.21	14.63	14.63	0.0141	70.9220	3.53	0.0027	-18.5898	0.0521	28.29	63	117	189	185	59	613	126
142	Reanalysis	1	No	opi3	23.88	12.96	12.96	0.0057	175.4386	3.43	0.0006	-21.7094	0.0241	30.45	115	88	196	191	112	702	147
143	Reanalysis	1	No	cmcp	23.11	9.24	9.24	0.0030	333.3333	3.42	0.0002	-20.2463	0.0127	29.50	90	54	198	195	83	620	131
144	Reanalysis	1	No	cmcp2	23.10	9.21	9.21	0.0032	312.5000	3.42	0.0002	-20.2390	0.0138	29.50	88	53	197	194	82	614	128
145	Reanalysis	1	1 mon	reanalysis	24.23	18.97	18.97	-0.0181	1.90	0.0070	-36.2659	-0.0837	30.82	127	172	209	228	121	857	177	
146	Reanalysis	1	1 mon	gpcp	21.63	1.87	1.87	0.0240	41.8667	3.38	0.0102	-18.8568	0.1010	27.36	45	6	186	188	37	442	84
147	Reanalysis	1	1 mon	gpcp	21.02	8.88	8.88	0.0237	42.1941	2.59	0.0255	-48.3997	0.1596	27.08	31	41	187	133	30	422	78
148	Reanalysis	1	1 mon	gpcp trmm	24.48	13.28	13.28	0.0543	18.4182	5.81	0.0231	-8.8374	0.1519	31.91	134	98	176	136	143	687	142
149	Reanalysis	1	1 mon	ghcn cams grid	20.94	2.72	2.72	0.0134	74.8269	2.22	0.0050	-27.4908	0.0709	27.00	26	11	190	177	26	430	78
150	Reanalysis	1	1 mon	gpi	49.57	-44.30	44.30	0.0555	18.0180	0.97	0.1309	-76.9097	0.3618	62.80	241	253	173	50	224	941	202
151	Reanalysis	1	1 mon	trmm 3b43	23.83	9.18	9.18	0.0879	11.3786	5.83	0.0849	-7.7189	0.2547	30.18	112	52	153	85	104	506	107
152	Reanalysis	1	1 mon	ssmi	48.02	-20.33	20.33	-0.0543	5.17	0.1540	-85.0801	-0.3924	61.90	221	188	225	282	217	1113	242	
153	Reanalysis	1	1 mon	opi2	21.25	14.51	14.51	0.0807	16.4745	4.05	0.0503	-17.1389	0.2244	27.28	35	111	169	108	38	459	89
154	Reanalysis	1	1 mon	opi3	23.11	12.76	12.76	0.0381	26.2467	3.73	0.0254	-20.1920	0.1594	29.47	92	80	181	134	81	568	122
155	Reanalysis	1	1 mon	cmcp	22.15	9.04	9.04	0.0329	30.3851	3.59	0.0195	-18.8222	0.1397	28.55	59	49	185	148	63	502	106
156	Reanalysis	1	1 mon	cmcp2	22.15	9.01	9.01	0.0330	30.3030	3.59	0.0196	-18.8209	0.1400	28.55	60	46	184	143	62	495	

Arkansas Basin - Monthly Surface

Of	Runoff	Re	Storage	Precipitation	Absolute Error	Bias	Ablas	Slope	Oslope	Intercept	R <sup>2</sup>	R <sup>2</sup> 1:1	Corr Coef	RMSE	AER	BR	SR	CCR	RMR	TR	BR
203	Reanalysis	2	Yes	cmcp	33.73	21.32	21.32	0.2180	4.5872	15.01	0.0186	-0.4155	0.1287	54.19	167	195	122	153	168	805	163
204	Reanalysis	2	Yes	cmcp2	33.72	21.26	21.26	0.2185	4.5767	15.00	0.0187	-0.4148	0.1290	54.18	166	194	121	152	167	800	162
205	Reanalysis	2	No	reanalysis	33.97	19.93	19.93	-0.0855		13.25	0.0018	-0.4608	-0.0421	57.29	169	180	229	221	183	982	211
206	Reanalysis	2	No	gpcp	37.47	14.56	14.56	-0.0967		13.70	0.0029	-0.4680	-0.0541	57.44	178	113	231	223	184	929	199
207	Reanalysis	2	No	gpcp	28.03	10.89	10.89	-0.1070		0.85	0.0082	-1.0182	-0.0905	40.17	159	83	234	229	180	845	172
208	Reanalysis	2	No	gpcp trmm	44.99	33.00	33.00	-0.3480		20.00	0.0491	-1.2147	-0.2217	68.09	214	237	256	250	232	1209	259
209	Reanalysis	2	No	ghcn cams grid	38.78	18.22	18.22	-0.2236		12.79	0.0187	-0.6317	-0.1293	60.56	190	158	248	237	208	1037	228
210	Reanalysis	2	No	gpi	50.73	-37.54	37.54	-0.1056		12.85	0.0208	-3.3280	-0.1443	75.62	262	245	233	236	262	1240	262
211	Reanalysis	2	No	trmm 3b43	42.77	28.31	28.31	-0.2953		21.87	0.0387	-1.0881	-0.1987	66.12	208	231	253	246	246	1184	255
212	Reanalysis	2	No	ssmi	44.22	-10.28	10.28	0.0945	10.5820	9.10	0.0196	-1.8526	0.1399	62.38	210	55	151	144	221	781	158
213	Reanalysis	2	No	opl2	38.78	27.27	27.27	-0.3164		9.51	0.0252	-0.7467	-0.1597	62.66	189	226	254	243	225	1137	246
214	Reanalysis	2	No	opl3	39.62	25.23	25.23	-0.2185		11.28	0.0144	-0.7181	-0.1202	62.14	197	219	247	235	218	1118	243
215	Reanalysis	2	No	cmcp	39.15	21.85	21.85	-0.1919		12.23	0.0118	-0.6574	-0.1088	61.03	193	203	245	233	211	1085	235
216	Reanalysis	2	No	cmcp2	39.15	21.81	21.81	-0.1910		12.25	0.0117	-0.6568	-0.1084	61.02	192	202	244	232	210	1080	234
217	Reanalysis	2	1 mon	reanalysis	33.92	19.93	19.93	0.0368	27.1739	14.01	0.0003	-0.4005	0.0181	56.20	168	179	182	193	174	898	194
218	Reanalysis	2	1 mon	gpcp	36.43	14.61	14.61	0.1080	9.2593	13.88	0.0037	-0.3399	0.0604	54.98	178	114	141	182	169	782	159
219	Reanalysis	2	1 mon	gpcp	28.70	10.78	10.78	0.0564	17.7305	2.18	0.0023	-0.7782	0.0478	37.84	153	81	172	196	158	730	150
220	Reanalysis	2	1 mon	gpcp trmm	45.53	33.65	33.65	0.1155	8.6580	24.94	0.0055	-0.8488	0.0741	62.72	217	238	139	178	228	998	216
221	Reanalysis	2	1 mon	ghcn cams grid	38.42	18.25	18.25	-0.0440		13.99	0.0008	-0.5114	-0.0254	58.39	185	181	221	214	187	968	207
222	Reanalysis	2	1 mon	gpi	58.86	-37.47	37.47	-0.0145		8.75	0.0004	-2.9791	-0.0199	72.68	260	243	207	212	258	1180	253
223	Reanalysis	2	1 mon	trmm 3b43	44.83	28.98	28.98	0.1394	7.1736	24.52	0.0086	-0.7149	0.0930	60.40	213	232	134	171	205	955	204
224	Reanalysis	2	1 mon	ssmi	47.83	-10.40	10.40	0.0060	186.8667	11.03	0.0001	-2.2389	0.0089	68.68	228	56	195	198	248	921	197
225	Reanalysis	2	1 mon	opl2	39.78	27.32	27.32	-0.1117		12.28	0.0032	-0.8422	-0.0583	60.86	199	227	235	225	208	1094	237
226	Reanalysis	2	1 mon	opl3	39.23	25.29	25.29	-0.0182		13.58	0.0001	-0.5969	-0.0100	60.02	194	220	210	205	203	1032	226
227	Reanalysis	2	1 mon	cmcp	38.63	21.90	21.90	0.0088	147.0588	13.84	0.0000	-0.5298	0.0038	58.74	187	208	194	198	191	876	210
228	Reanalysis	2	1 mon	cmcp2	38.61	21.86	21.86	0.0091	109.8901	13.86	0.0000	-0.5280	0.0051	58.71	196	204	193	197	190	870	208
229	Reanalysis	2	2 mon	reanalysis	35.42	19.99	19.99	-0.0869		13.30	0.0018	-0.4809	-0.0428	57.50	173	182	230	222	185	992	214
230	Reanalysis	2	2 mon	gpcp	37.16	14.67	14.67	0.0582	18.8919	13.88	0.0011	-0.3710	0.0331	55.70	177	119	171	190	171	828	168
231	Reanalysis	2	2 mon	gpcp	28.79	11.07	11.07	-0.1182		0.58	0.0098	-1.0183	-0.0900	40.49	161	86	238	230	161	854	175
232	Reanalysis	2	2 mon	gpcp trmm	49.49	34.42	34.42	-0.2380		21.96	0.0230	-1.1585	-0.1517	68.26	239	239	249	240	253	1220	280
233	Reanalysis	2	2 mon	ghcn cams grid	39.72	18.36	18.36	-0.0625		13.55	0.0013	-0.5247	-0.0361	58.74	186	184	228	218	192	998	218
234	Reanalysis	2	2 mon	gpi	57.94	-37.39	37.39	-0.0240		9.24	0.0011	-3.0059	-0.0329	73.10	261	242	214	217	259	1193	257
235	Reanalysis	2	2 mon	trmm 3b43	48.18	29.89	29.89	-0.1709		23.42	0.0130	-1.0075	-0.1142	65.86	227	233	239	234	244	1177	250
236	Reanalysis	2	2 mon	ssmi	54.28	-10.99	10.99	-0.1474		14.38	0.0471	-2.8828	-0.2171	73.45	257	80	238	249	261	1085	233
237	Reanalysis	2	2 mon	opl2	39.55	27.42	27.42	0.1080	9.2593	15.30	0.0030	-0.5314	0.0545	58.87	195	229	142	184	194	944	203
238	Reanalysis	2	2 mon	opl3	40.50	25.40	25.40	-0.0065		13.76	0.0000	-0.5909	-0.0038	60.01	208	222	205	204	202	1039	229
239	Reanalysis	2	2 mon	cmcp	39.91	22.00	22.00	-0.0003		13.83	0.0000	-0.5352	-0.0002	58.94	203	209	201	201	198	1010	221
240	Reanalysis	2	2 mon	cmcp2	39.90	21.97	21.97	0.0006	1666.6667	13.84	0.0000	-0.5344	0.0003	58.93	202	208	199	200	195	1004	219
241	Reanalysis	2	3 mon	reanalysis	35.82	19.85	19.85	-0.2109		12.46	0.0108	-0.5182	-0.1038	58.70	174	178	248	231	189	1018	222
242	Reanalysis	2	3 mon	gpcp	37.49	14.55	14.55	-0.0318		13.72	0.0003	-0.4283	-0.0178	58.90	179	112	215	209	179	894	191
243	Reanalysis	2	3 mon	gpcp	29.53	11.31	11.31	-0.1774		0.24	0.0222	-1.1150	-0.1482	41.46	162	89	241	239	162	873	183
244	Reanalysis	2	3 mon	gpcp trmm	50.16	35.15	35.15	-0.2546		22.80	0.0271	-1.1987	-0.1847	69.21	246	240	250	244	254	1234	281
245	Reanalysis	2	3 mon	ghcn cams grid	39.81	18.24	18.24	-0.0856		13.46	0.0014	-0.5247	-0.0379	58.83	198	180	228	220	193	997	217
246	Reanalysis	2	3 mon	gpi	55.83	-37.29	37.29	0.0005	2000.0000	8.30	0.0000	-2.9159	0.0007	72.30	259	241	200	199	257	1156	249
247	Reanalysis	2	3 mon	trmm 3b43	48.24	30.30	30.30	-0.1805		24.22	0.0147	-1.0386	-0.1214	66.81	229	234	242	236	247	1188	256
248	Reanalysis	2	3 mon	ssmi	54.21	-10.81	10.81	-0.1380		14.54	0.0418	-2.8703	-0.2040	73.11	258	59	237	246	260	1080	231
249	Reanalysis	2	3 mon	opl2	38.81	27.32	27.32	0.2078	4.8123	16.57	0.0110	-0.4778	0.1048	57.91	191	228	124	187	186	898	193
250	Reanalysis	2	3 mon	opl3	40.35	25.30	25.30	-0.0025		13.72	0.0000	-0.5856	-0.0014	59.99	205	221	203	202	201	1032	227
251	Reanalysis	2	3 mon	cmcp	39.84	21.90	21.90	-0.0226		13.57	0.0002	-0.5472	-0.0128	59.26	201	205	212	208	196	1023	223
252	Reanalysis	2	3 mon	cmcp2	39.85	21.87	21.87	-0.0235		13.56	0.0002	-0.5478	-0.0133	59.26	201	205	213	207	199	1025	224
253	Reanalysis	2	4 mon	reanalysis	35.18	19.53	19.53	-0.0395		13.06	0.0004	-0.4371	-0.0198	58.59	172	177	218	211	178	958	205
254	Reanalysis	2	4 mon	gpcp	35.98	14.11	14.11	0.1537	6.5082	13.44	0.0078	-0.3111	0.0870	54.05	175	104	132	174	166	751	152
255	Reanalysis	2	4 mon	gpcp	27.98	11.62	11.62	-0.0366		1.71	0.0009	-0.9227	-0.0307	39.81	158	72	218	216	159	821	165
256	Reanalysis	2	4 mon	gpcp trmm	48.00	32.83	32.83	0.2058	4.8591	25.00	0.0196	-0.8062	0.1399	60.35	220	236	125	145	204	930	200
257	Reanalysis	2	4 mon	ghcn cams grid	37.54	17.84	17.84	0.1128	8.8652	13.83	0.0043	-0.4074	0.0859	56.00							

Arkansas Basin - Yearly Surface																						
OI	Runoff	Re	Storage	Precipitation	Absolute Error	Bias	Abias	Slope	Oslope	Intercept	R <sup>2</sup>	R <sup>2</sup> 1:1	Corr Coef	RMSE	AER	BR	SR	CCR	RMR	TR	BR	Plot
1	Surface	1	Yes	reanalysis	24.58	24.58	24.58	0.2652	3.7707	-19.20	0.0942	-14.3923	0.3069	25.99	64	75	29	21	60	249	51	Plot
2	Surface	1	Yes	gpcp	7.54	5.94	5.94	0.6951	1.4386	-3.55	0.2060	-0.5376	0.4539	8.78	4	6	13	16	4	43	3	
3	Surface	1	Yes	gpcp	13.48	13.48	13.48	0.5868	1.7042	-10.62	0.1754	-3.9281	0.4188	14.93	28	35	19	18	25	125	14	
4	Surface	1	Yes	ghcn cams grid	9.37	9.37	9.37	0.5487	1.8225	-5.80	0.2133	-2.8264	0.4619	10.74	13	21	21	15	9	79	5	
5	Surface	1	Yes	gpi	46.62	-46.62	46.62	0.9516	1.0509	46.95	0.1815	-18.7167	0.4280	47.62	88	88	4	17	86	283	66	
6	Surface	1	Yes	ssmi	21.65	-21.65	21.65	0.4091	2.4444	24.11	0.0120	-3.2521	0.1097	24.80	58	69	24	41	57	249	52	
7	Surface	1	Yes	opi2	18.29	18.29	18.29	-0.0157		-10.33	0.0001	-9.1861	-0.0117	19.89	47	59	61	58	40	265	59	
8	Surface	1	Yes	opi3	16.45	16.45	16.45	0.6723	1.4874	-13.88	0.2433	-6.6352	0.4933	17.41	38	49	14	13	36	150	21	
9	Surface	1	Yes	cmmap	13.03	13.03	13.03	0.7270	1.3755	-10.89	0.2519	-3.5709	0.5019	14.32	25	33	11	10	23	102	10	
10	Surface	1	Yes	cmmap2	24.11	24.11	24.11	0.1012	1.0102	-24.16	0.2963	-12.9439	0.5462	24.74	63	74	1	6	56	200	33	
11	Surface	1	No	reanalysis	8.79	5.83	5.83	1.2258	1.2258	-7.57	0.3302	-0.3589	0.5748	8.25	3	5	8	4	3	23	1	Plot
12	Surface	1	No	gpcp	14.91	14.91	14.91	1.3505	1.3505	-17.84	0.4042	-4.5374	0.6358	15.82	35	47	10	3	27	122	13	
13	Surface	1	No	gpcp	8.24	8.24	8.24	0.9321	1.0728	-8.71	0.3173	-2.5114	0.5633	10.29	6	13	7	5	8	39	2	
14	Surface	1	No	ghcn cams grid	45.19	-45.19	45.19	1.8345	1.8345	38.22	0.2933	-17.5231	0.5416	46.15	85	85	22	7	83	282	65	
15	Surface	1	No	gpi	18.08	-18.08	18.08	0.9349	1.0696	18.58	0.0698	-2.1891	0.2642	21.48	44	54	6	23	47	174	30	
16	Surface	1	No	ssmi	18.18	18.18	18.18	0.1004	9.9602	-11.23	0.0029	-8.7409	0.0535	19.45	46	58	42	54	38	236	49	
17	Surface	1	No	opi2	16.34	16.34	16.34	0.9425	1.0610	-15.90	0.2464	-6.4835	0.4964	17.24	36	48	5	12	35	136	15	
18	Surface	1	No	opi3	12.92	12.92	12.92	1.0183	1.0183	-13.06	0.2547	-3.4684	0.5047	14.15	23	31	3	9	21	87	7	
19	Surface	1	No	cmmap	12.90	12.90	12.90	1.0150	1.0150	-13.01	0.2548	-3.4819	0.5048	14.13	22	30	2	8	20	82	6	
20	Surface	1	No	cmmap2	17.85	12.29	12.29	0.1410	7.0922	-6.47	0.4117	-23.5762	0.6416	20.64	42	27	37	2	45	153	24	Plot
21	Surface	2	Yes	reanalysis	17.04	6.66	6.66	0.0417	23.9808	-0.17	0.0131	-7.8633	0.1146	20.51	39	7	52	40	44	182	31	
22	Surface	2	Yes	gpcp	19.34	14.49	14.49	0.1095	9.1324	-9.82	0.0800	-9.2401	0.2828	23.09	50	41	40	22	52	205	36	
23	Surface	2	Yes	ghcn cams grid	18.00	10.07	10.07	0.0440	22.7273	-3.59	0.0136	-8.3854	0.1165	21.90	43	22	51	38	49	203	35	
24	Surface	2	Yes	gpi	45.60	-45.60	45.60	0.1112	8.9928	50.26	0.0346	-18.9094	0.1859	49.74	87	87	39	27	88	328	76	
25	Surface	2	Yes	ssmi	28.21	-23.44	23.44	-0.0604		22.25	0.0097	-6.9800	-0.0966	33.16	72	72	66	65	73	348	81	
26	Surface	2	Yes	opi2	25.90	19.02	19.02	-0.0759		-11.72	0.0251	-9.3975	-0.1584	29.28	67	61	68	69	65	330	77	
27	Surface	2	Yes	opi3	22.59	17.18	17.18	0.0526	19.0114	-10.75	0.0191	-11.8523	0.1381	25.84	62	51	50	34	59	256	55	
28	Surface	2	Yes	cmmap	20.20	13.76	13.76	0.0629	15.8983	-7.40	0.0281	-9.2422	0.1616	23.60	54	37	46	29	54	210	42	
29	Surface	2	Yes	cmmap2	20.19	13.73	13.73	0.0627	15.9490	-7.38	0.0260	-9.2672	0.1612	23.59	53	36	47	30	53	219	40	
30	Surface	2	No	reanalysis	13.24	13.24	13.24	-0.1435		-4.40	0.0131	-10.9378	-0.1144	14.38	26	34	69	67	24	220	41	
31	Surface	2	No	gpcp	8.96	7.61	7.61	-0.2349		1.93	0.0128	-1.5631	-0.1132	11.03	9	11	71	66	13	170	29	
32	Surface	2	No	gpcp	17.60	17.60	17.60	0.0048	208.3333	-9.29	0.0000	-6.1405	0.0021	19.28	40	52	57	57	37	243	50	
33	Surface	2	No	ghcn cams grid	11.55	11.02	11.02	-0.5285		0.80	0.0602	-2.8200	-0.2454	13.97	18	24	78	75	19	214	37	
34	Surface	2	No	gpi	42.49	-42.49	42.49	0.4888	2.0458	46.76	0.0193	-14.5323	0.1388	43.93	83	83	23	33	81	303	72	
35	Surface	2	No	ssmi	14.59	-14.59	14.59	-0.9115		29.37	0.0697	-1.7811	-0.2639	19.58	30	44	81	76	39	270	61	
36	Surface	2	No	opi2	20.56	19.97	19.97	-1.3603		-1.72	0.2472	-5.3339	-0.4971	22.85	56	66	83	84	50	339	79	
37	Surface	2	No	opi3	18.13	18.13	18.13	-0.5182		-6.39	0.0569	-6.7554	-0.2385	20.08	45	55	77	73	42	292	69	
38	Surface	2	No	cmmap	14.73	14.70	14.70	-0.4424		-3.55	0.0366	-4.3578	-0.1991	17.07	33	46	75	71	34	259	57	
39	Surface	2	No	cmmap2	14.70	14.68	14.68	-0.4457		-3.51	0.0404	-4.3629	-0.2009	17.05	32	45	76	72	32	257	56	
40	Surface	2	No	reanalysis	19.42	19.36	19.36	0.1481	8.7522	5.11	0.0170	-7.1655	0.1305	21.50	51	63	36	36	48	234	47	
41	Reanalysis	1	Yes	reanalysis	8.37	1.74	1.74	-0.0270		3.60	0.0006	-0.9741	-0.0250	9.58	7	2	62	62	5	138	16	
42	Reanalysis	1	Yes	gpcp	9.52	7.60	7.60	0.0360	27.7778	1.28	0.0031	-5.2398	0.0555	10.90	14	10	54	52	11	141	18	
43	Reanalysis	1	Yes	gpcp	13.24	9.34	9.34	-2.9075		-10.62	0.7584	-1.3014	-0.8709	17.05	27	20	87	87	33	254	54	
44	Reanalysis	1	Yes	gpcp trmm	7.58	2.86	2.86	0.1006	9.9404	2.69	0.0068	-0.6813	0.0826	9.84	5	4	41	44	6	100	8	
45	Reanalysis	1	Yes	ghcn cams grid	45.50	-45.50	45.50	0.1811	5.5218	-5.97	0.1014	-37.5042	0.3184	47.37	86	86	32	19	85	308	74	
46	Reanalysis	1	Yes	gpi	14.67	6.99	6.99	-0.5335		1.73	0.1339	-1.4467	-0.3659	15.87	31	8	79	81	28	227	44	
47	Reanalysis	1	Yes	trmm 3b43	20.89	-19.61	19.61	-0.2572		7.70	0.1236	-8.2348	-0.3515	25.63	57	64	72	80	58	331	78	
48	Reanalysis	1	Yes	ssmi	14.83	14.43	14.43	0.3353	2.9824	7.20	0.0993	-4.7744	0.3152	16.38	34	40	27	20	30	151	22	
49	Reanalysis	1	Yes	opi2	12.51	12.39	12.39	0.1299	7.6982	4.70	0.0115	-3.8114	0.1072	14.95	21	28	38	43	26	156	27	
50	Reanalysis	1	Yes	cmmap	10.22	9.01	9.01	0.0787	12.7065	3.98	0.0048	-2.3928	0.0689	12.55	16	18	44	48	17	143	20	
51	Reanalysis	1	Yes	cmmap2	10.19	8.97	8.97	0.0775	12.9032	3.97	0.0046	-2.3767	0.0677	12.52	15	17	45	49	16	142	19	
52	Reanalysis	1	No	reanalysis	18.93	18.93	18.93	0.0535	18.6916	3.10	0.0395	-124.7595	0.1987	20.02	49	60	48	26	41	224	43	
53	Reanalysis	1	No	gpcp	5.95	1.61	1.61	0.0178	56.1798	3.40	0.0029	-9.4885	0.0540	6.75	2	1	55	53	1	112	12	
54	Reanalysis	1	No	gpcp	9.18	9.03	9.03	0.0364	27.4725	2.71	0.0123	-445.7478	0.4651	11.12	10	19	53	14	14	110	11	
55	Reanalysis	1	No	gpcp trmm	11.58	11.58	11.58	0.0631	18.8324	6.74	0.0060	-26.9364	0.0773	12.23	19	25	49	47	15	155	25	
56	Reanalysis	1	No	ghcn cams grid	5.66	2.42	2.42	-0.0468		2.22	0.0265	-15.0587	-0.1629	7.18								

Mississippi Basin - Monthly Atmospheric																						
OI	Runoff	Re	Storage	Precipitation	Absolute Error	Bias	Slope	Oslope	Intercept	R^2	R^2 1:1	Corr Coef	RMSE	AER	BR	SR	CCR	RMR	TR	BR		
1	Surface	1	Yes	reanalysis	143.54	-143.43	0.3880	2.5773	139.64	0.0101	-3.5379	0.1004	163.33	3	3	4	4	3	17	2	Plot	
2	Surface	1	Yes	gpcp	141.18	-140.86	-0.3428		134.50	0.0178	-3.5910	-0.1327	166.31	1	1	8	8	4	22	3		
3	Surface	1	Yes	gpcp	150.47	-150.47	-0.2613		132.60	0.0081	-3.6418	-0.0898	174.24	7	9	5	5	10	36	8		
4	Surface	1	Yes	gpcp trmm	154.53	-154.53	-1.2044		128.50	0.1326	-3.8174	-0.3641	182.66	15	15	19	17	18	84	18		
5	Surface	1	Yes	ghcn cams grid	146.25	-146.06	-0.3403		135.67	0.0167	-3.9399	-0.1291	169.94	4	4	7	7	6	28	6		
6	Surface	1	Yes	ssmi	141.39	-141.39	1.2393	1.2393	140.10	0.3484	-2.8111	0.5902	153.09	2	2	2	2	1	9	1	Plot	
7	Surface	1	Yes	opi2	152.94	-152.64	-0.9916		121.10	0.1442	-4.2605	-0.3797	178.02	12	12	16	19	15	74	15		
8	Surface	1	Yes	opi3	151.51	-151.33	-0.5778		128.51	0.0480	-4.0435	-0.2192	174.31	11	11	14	15	11	62	12		
9	Surface	1	Yes	cmap	149.02	-148.83	-0.5162		130.27	0.0378	-3.9328	-0.1943	172.39	6	6	11	13	9	45	10		
10	Surface	1	Yes	cmap2	148.97	-148.78	-0.5158		130.30	0.0376	-3.9293	-0.1939	172.33	5	5	10	12	8	40	9		
11	Surface	2	Yes	reanalysis	150.51	-150.51	1.1389	1.1389	148.83	0.0552	-3.2849	0.2349	168.12	9	10	1	3	5	28	5	Plot	
12	Surface	2	Yes	gpcp	150.49	-150.29	-0.3019		144.92	0.0100	-3.4560	-0.0996	171.44	8	7	6	6	7	34	7		
13	Surface	2	Yes	gpcp	154.03	-154.03	-1.4381		128.17	0.2253	-4.0958	-0.4747	178.95	13	14	20	20	16	83	17		
14	Surface	2	Yes	gpcp trmm	171.12	-171.12	-1.1073		139.28	0.0917	-4.5821	-0.3028	194.03	20	20	17	16	20	93	20		
15	Surface	2	Yes	ghcn cams grid	154.06	-153.94	-0.5013		142.76	0.0248	-3.6430	-0.1575	175.00	14	13	9	9	12	57	11		
16	Surface	2	Yes	ssmi	150.56	-150.35	1.4799	1.4799	142.41	0.4134	-2.6636	0.6429	156.86	10	8	3	1	2	24	4		
17	Surface	2	Yes	opi2	160.34	-160.25	-1.1224		132.38	0.1398	-4.0994	-0.3740	183.40	19	19	18	18	19	93	19		
18	Surface	2	Yes	opi3	158.77	-158.67	-0.6219		139.31	0.0398	-3.8851	-0.1995	179.50	18	18	15	14	17	82	16		
19	Surface	2	Yes	cmap	156.67	-156.56	-0.5406		141.08	0.0296	-3.7812	-0.1720	177.59	17	17	13	11	14	72	14		
20	Surface	2	Yes	cmap2	156.62	-156.51	-0.5398		141.11	0.0294	-3.7780	-0.1715	177.53	16	16	12	10	13	67	13		

Mississippi Basin - Monthly Atmospheric																					
Of	Runoff	Re	Storage	Precipitation	Absolute Error	Bias	Slope	Oslope	Intercpt	R <sup>2</sup>	R <sup>2</sup> 1-1	Corr Coef	RMSE	AER	BR	SR	CCR	RMA	TR	BR	
1	Surface	1	Yes	reanalysis	143.54	-143.43	0.3880	2.5773	139.64	0.0101	-3.5379	0.1004	163.33	3	3	4	4	3	17	2	Plot
2	Surface	1	Yes	gpcp	141.18	-140.86	-0.3428		134.50	0.0176	-3.5910	-0.1327	166.31	1	1	8	8	4	22	3	
3	Surface	1	Yes	gpcp	150.47	-150.47	-0.2613		132.60	0.0081	-3.6418	-0.0898	174.24	7	9	5	5	10	36	8	
4	Surface	1	Yes	gpcp trmm	154.53	-154.53	-1.2044		128.50	0.1326	-3.8174	-0.3641	182.66	15	15	19	17	18	84	18	
5	Surface	1	Yes	ghcn cams grid	146.25	-146.06	-0.3403		135.67	0.0167	-3.9399	-0.1291	169.94	4	4	7	7	6	28	6	
6	Surface	1	Yes	ssmi	141.39	-141.39	1.2393	1.2393	140.10	0.3484	-2.8111	0.5902	153.09	2	2	2	2	1	9	1	Plot
7	Surface	1	Yes	opi2	152.94	-152.64	-0.9916		121.10	0.1442	-4.2605	-0.3797	178.02	12	12	16	19	15	74	15	
8	Surface	1	Yes	opi3	151.51	-151.33	-0.5778		128.51	0.0480	-4.0435	-0.2192	174.31	11	11	14	15	11	62	12	
9	Surface	1	Yes	cmap	149.02	-148.83	-0.5162		130.27	0.0378	-3.9328	-0.1943	172.39	6	6	11	13	9	45	10	
10	Surface	1	Yes	cmap2	148.97	-148.78	-0.5158		130.30	0.0376	-3.9293	-0.1939	172.33	5	5	10	12	8	40	9	
11	Surface	2	Yes	reanalysis	150.51	-150.51	1.1389	1.1389	148.83	0.0552	-3.2849	0.2349	168.12	9	10	1	3	5	28	5	Plot
12	Surface	2	Yes	gpcp	150.49	-150.29	-0.3019		144.92	0.0100	-3.4560	-0.0998	171.44	8	7	6	6	7	34	7	
13	Surface	2	Yes	gpcp	154.03	-154.03	-1.4381		128.17	0.2253	-4.0958	-0.4747	178.95	13	14	20	20	16	83	17	
14	Surface	2	Yes	gpcp trmm	171.12	-171.12	-1.1073		139.28	0.0917	-4.5821	-0.3028	194.03	20	20	17	16	20	93	20	
15	Surface	2	Yes	ghcn cams grid	154.06	-153.94	-0.5013		142.76	0.0248	-3.6430	-0.1575	175.00	14	13	9	9	12	57	11	
16	Surface	2	Yes	ssmi	150.58	-150.35	1.4799	1.4799	142.41	0.4134	-2.8636	0.6429	156.86	10	8	3	1	2	24	4	
17	Surface	2	Yes	opi2	160.34	-160.25	-1.1224		132.38	0.1398	-4.0894	-0.3740	183.40	19	19	18	18	19	93	19	
18	Surface	2	Yes	opi3	158.77	-158.67	-0.6219		139.31	0.0398	-3.8851	-0.1995	179.50	18	18	15	14	17	82	16	
19	Surface	2	Yes	cmap	156.67	-156.56	-0.5406		141.08	0.0296	-3.7812	-0.1720	177.59	17	17	13	11	14	72	14	
20	Surface	2	Yes	cmap2	156.62	-156.51	-0.5398		141.11	0.0294	-3.7780	-0.1715	177.53	16	16	12	10	13	67	13	



Mississippi Basin - Yearly Atmospheric																						
OI	Runoff	Re	Storage	Precipitation	Absolute Error	Bias	Slope	Oslope	Intercept	R <sup>2</sup>	R <sup>2</sup> 1 1	Corr Coef	RMSE	AER	BR	SR	CCR	RMR	TR	BR		
1	Surface	1	Yes	reanalysis	143.43	-143.43	1.0370	1.0370	144.18	0.1300	-50.7855	0.3605	145.15	3	3	1	12	4	23	3	Plot	
2	Surface	1	Yes	gpcp	140.80	-140.80	1.0905	1.0905	142.26	0.1782	-63.2803	0.4198	142.69	2	2	3	7	3	17	2	Plot	
3	Surface	1	Yes	gpcc	150.51	-150.51	3.1172	3.1172	180.51	0.5548	-51.8012	0.7449	151.52	12	12	18	2	12	56	10		
4	Surface	1	Yes	gpcc trmm	145.34	-145.34	5.1827	5.1827	196.61	0.6514	-129.4422	0.8071	145.77	5	5	20	1	5	36	5		
5	Surface	1	Yes	ghcn cams grid	145.23	-145.23	1.8170	1.8170	152.86	0.2186	-66.5916	0.4676	146.32	4	4	16	5	6	35	4		
6	Surface	1	Yes	ssmi	140.73	-140.73	1.2996	1.2996	139.62	0.3095	-54.0739	0.5563	138.53	1	1	11	3	1	17	1		
7	Surface	1	Yes	opi2	150.76	-150.76	1.4180	1.4180	158.15	0.1590	-72.5261	0.3988	152.60	13	13	13	8	14	61	13		
8	Surface	1	Yes	opi3	149.18	-149.18	1.1302	1.1302	151.69	0.1232	-70.8409	0.3510	150.84	8	8	4	13	11	44	8		
9	Surface	1	Yes	cmap	147.08	-147.08	1.1496	1.1496	149.90	0.1357	-69.1495	0.3683	149.06	7	7	7	11	9	41	7		
10	Surface	1	Yes	cmap2	147.03	-147.03	1.1612	1.1612	149.97	0.1373	-69.0957	0.3706	149.00	6	6	8	10	8	38	6		
11	Surface	2	Yes	reanalysis	150.48	-150.48	1.7643	1.7643	148.85	0.0684	-63.7606	0.2616	149.44	11	11	9	19	10	60	12	Plot	
12	Surface	2	Yes	gpcp	150.26	-150.26	1.4112	1.4112	147.74	0.2147	-62.5083	0.4834	147.99	10	10	12	6	7	45	9		
13	Surface	2	Yes	gpcc	154.07	-154.07	1.4901	1.4901	158.99	0.1159	-51.0626	0.3404	155.41	15	15	14	18	17	77	18		
14	Surface	2	Yes	gpcc trmm	171.32	-171.32	3.4597	3.4597	208.47	0.2418	-225.0908	0.4917	171.65	20	20	19	4	20	83	19		
15	Surface	2	Yes	ghcn cams grid	153.91	-153.91	1.5599	1.5599	153.69	0.1443	-65.7318	0.3798	151.70	14	14	15	9	13	65	14		
16	Surface	2	Yes	ssmi	149.88	-149.88	0.4805	2.0812	140.00	0.0270	-48.0726	0.1643	142.45	9	9	17	20	2	57	11		
17	Surface	2	Yes	opi2	160.22	-160.22	0.7837	1.2760	154.48	0.0797	-71.4003	0.2823	158.01	19	19	10	18	19	85	20		
18	Surface	2	Yes	opi3	158.64	-158.64	1.0537	1.0537	155.73	0.1000	-69.7498	0.3182	156.20	18	18	2	17	18	73	17		
19	Surface	2	Yes	cmap	156.54	-156.54	1.1316	1.1316	154.48	0.1174	-68.1363	0.3426	154.41	17	17	5	15	16	70	16		
20	Surface	2	Yes	cmap2	156.49	-156.49	1.1402	1.1402	154.48	0.1183	-68.0858	0.3440	154.35	16	16	6	14	15	67	15		

Mississippi Basin - Monthly Surface																						
OI	Runoff	Re	Storage	Precipitation	Absolute Error	Bias	Abias	Slope	Oslope	Intercept	R2	R2 1.1	Corr Coef	RMSE	AER	BR	SR	CCR	RMR	TR	BR	Plot
1	Surface	1	Yes	reanalysis	40.06	22.41	22.41	0.1955	5.1151	-10.00	0.3159	-6.3083	0.5621	53.63	221	158	100	37	226	742	191	
2	Surface	1	Yes	gpcp	31.88	22.08	22.08	0.4055	2.4661	-12.19	0.6003	-1.2302	0.7748	44.87	180	148	94	13	214	649	141	
3	Surface	1	Yes	ghcn cams grid	35.21	30.39	30.39	0.3950	2.5316	-20.58	0.6395	-2.0561	0.7997	48.59	211	214	95	6	222	748	197	
4	Surface	1	Yes	ghcn cams grid	33.73	25.30	25.30	0.3873	2.5820	-15.00	0.6254	-1.6997	0.7908	47.66	203	177	96	8	220	704	167	
5	Surface	1	Yes	ssmi	64.32	18.74	18.74	-0.1487		-2.02	0.0475	-3.0399	-0.2180	75.07	228	133	214	167	228	970	223	
6	Surface	1	Yes	opi2	35.51	32.02	32.02	0.4283	2.3348	-22.52	0.6843	-1.6961	0.8273	48.80	214	223	88	1	223	749	199	
7	Surface	1	Yes	opi3	34.00	30.23	30.23	0.4207	2.3770	-20.60	0.6726	-1.6568	0.8201	48.00	205	211	90	2	221	729	184	
8	Surface	1	Yes	cmcp	33.29	28.44	28.44	0.4130	2.4213	-18.68	0.6585	-1.6197	0.8115	47.29	196	197	91	4	219	707	170	
9	Surface	1	Yes	cmcp2	33.24	28.39	28.39	0.4128	2.4225	-18.63	0.6595	-1.6211	0.8121	47.24	195	195	92	3	218	703	164	
10	Surface	1	No	reanalysis	24.72	22.41	22.41	0.1941	5.1520	-9.98	0.0064	-1.3792	0.0799	30.60	64	157	101	121	63	506	83	
11	Surface	1	No	gpcp	30.31	22.15	22.15	-0.1194		-3.46	0.0010	-0.6324	-0.0320	38.39	168	149	221	132	170	840	209	
12	Surface	1	No	gpcp	33.81	30.56	30.56	-0.1667		-11.44	0.0021	-1.3074	-0.0454	42.22	204	218	227	133	203	985	226	
13	Surface	1	No	ghcn cams grid	31.52	25.11	25.11	-0.3141		-3.26	0.0079	-0.8803	-0.0891	39.78	176	175	222	146	182	901	212	
14	Surface	1	No	ssmi	38.06	20.60	20.60	-0.6955		7.24	0.0206	-0.4060	-0.1435	44.29	218	143	219	159	211	950	217	
15	Surface	1	No	opi2	35.26	32.09	32.09	-0.4514		-7.86	0.0150	-1.3058	-0.1223	45.13	212	224	228	154	215	1033	228	
16	Surface	1	No	opi3	34.66	30.30	30.30	-0.2921		-8.73	0.0064	-1.1772	-0.0799	43.45	209	212	226	143	210	1000	227	
17	Surface	1	No	cmcp	33.54	28.52	28.52	-0.2807		-7.13	0.0060	-1.0711	-0.0774	42.05	198	202	225	140	202	967	221	
18	Surface	1	No	cmcp2	33.50	28.46	28.46	-0.2805		-7.08	0.0060	-1.0699	-0.0774	41.98	197	198	224	141	200	960	220	
19	Surface	1	1 mon	reanalysis	23.93	22.44	22.44	0.9790	1.0277	-22.02	0.1802	-1.1182	0.4003	28.88	51	159	2	78	41	331	27	
20	Surface	1	1 mon	gpcp	28.35	22.18	22.18	1.4931	1.4931	-30.39	0.1598	-0.4003	0.3998	35.62	145	150	33	79	154	561	107	
21	Surface	1	1 mon	gpcp	32.43	30.70	30.70	1.4455	1.4455	-38.01	0.1546	-1.0735	0.3931	40.13	188	219	29	80	187	703	163	
22	Surface	1	1 mon	ghcn cams grid	29.68	25.17	25.17	1.1708	1.1708	-28.01	0.1105	-0.6435	0.3324	37.22	164	176	13	94	166	613	133	
23	Surface	1	1 mon	ssmi	38.53	20.38	20.38	-0.8500		9.89	0.0305	-0.4092	-0.1747	44.53	219	140	216	166	212	953	218	
24	Surface	1	1 mon	opi2	34.36	32.20	32.20	1.0125	1.0125	-32.41	0.0754	-1.0986	0.2746	43.06	207	225	1	97	208	738	188	
25	Surface	1	1 mon	opi3	33.18	30.33	30.33	1.2683	1.2683	-34.81	0.1201	-0.9425	0.3465	41.12	193	213	19	90	196	711	176	
26	Surface	1	1 mon	cmcp	31.95	28.54	28.54	1.2723	1.2723	-33.08	0.1227	-0.8335	0.3503	39.64	182	203	21	88	180	674	151	
27	Surface	1	1 mon	cmcp2	31.91	28.49	28.49	1.2709	1.2709	-33.00	0.1227	-0.8320	0.3503	39.57	181	200	20	89	178	668	150	
28	Surface	1	2 mon	reanalysis	24.54	22.51	22.51	0.9469	1.0581	-21.69	0.1528	-1.1406	0.3909	28.96	61	162	4	81	43	351	32	
29	Surface	1	2 mon	gpcp	28.23	22.39	22.39	1.7310	1.7310	-34.57	0.2170	-0.3779	0.4659	35.24	143	156	64	71	149	583	118	
30	Surface	1	2 mon	gpcp	32.87	31.04	31.04	1.8145	1.8145	-44.41	0.2487	-1.0594	0.4987	39.72	190	220	71	58	181	720	182	
31	Surface	1	2 mon	ghcn cams grid	29.61	25.33	25.33	1.6561	1.6561	-36.25	0.2230	-0.5777	0.4722	36.36	163	178	56	69	160	626	136	
32	Surface	1	2 mon	ssmi	40.16	20.57	20.57	-1.5415		21.38	0.0992	-0.4679	-0.3149	45.89	222	141	217	179	217	976	225	
33	Surface	1	2 mon	opi2	34.52	32.44	32.44	1.7471	1.7471	-44.89	0.2275	-1.0161	0.4770	42.01	208	228	67	66	201	770	201	
34	Surface	1	2 mon	opi3	33.60	30.54	30.54	1.6974	1.6974	-42.16	0.2172	-0.8970	0.4661	40.52	199	217	62	70	189	737	187	
35	Surface	1	2 mon	cmcp	32.21	28.75	28.75	1.6712	1.6712	-39.94	0.2139	-0.7905	0.4625	39.06	185	210	60	72	176	703	167	
36	Surface	1	2 mon	cmcp2	32.17	28.69	28.69	1.6689	1.6689	-39.84	0.2139	-0.7892	0.4625	38.99	184	207	59	73	175	698	160	
37	Surface	1	3 mon	reanalysis	24.78	22.49	22.49	0.9476	1.0553	-21.68	0.1522	-1.1368	0.3901	28.96	66	160	3	82	42	353	34	
38	Surface	1	3 mon	gpcp	28.23	22.33	22.33	1.9731	1.9731	-38.47	0.2751	-0.3428	0.5245	34.86	142	154	75	47	144	562	108	
39	Surface	1	3 mon	gpcp	33.03	31.13	31.13	2.0961	2.0961	-49.08	0.3322	-1.0159	0.5763	39.41	192	221	85	33	177	708	171	
40	Surface	1	3 mon	ghcn cams grid	29.39	25.36	25.36	2.0041	2.0041	-42.06	0.3266	-0.5208	0.5715	35.75	162	179	81	35	156	613	132	
41	Surface	1	3 mon	ssmi	40.34	20.58	20.58	-1.2630		18.79	0.0860	-0.4402	-0.2568	45.50	223	142	218	168	216	967	222	
42	Surface	1	3 mon	opi2	34.23	32.37	32.37	2.1825	2.1825	-51.98	0.3463	-0.9485	0.5885	41.38	206	227	87	31	197	748	196	
43	Surface	1	3 mon	opi3	33.62	30.50	30.50	2.0844	2.0844	-47.55	0.3028	-0.8454	0.5503	40.05	200	215	82	38	184	719	181	
44	Surface	1	3 mon	cmcp	32.31	28.71	28.71	1.9841	1.9841	-45.03	0.2843	-0.7414	0.5425	38.60	187	208	77	41	172	685	156	
45	Surface	1	3 mon	cmcp2	32.27	28.65	28.65	1.9804	1.9804	-44.91	0.2839	-0.7403	0.5422	38.54	186	205	78	42	171	680	154	
46	Surface	1	4 mon	reanalysis	25.00	22.50	22.50	0.8239	1.2137	-19.79	0.1144	-1.1815	0.3983	29.25	70	161	15	91	51	388	47	
47	Surface	1	4 mon	gpcp	27.81	22.34	22.34	1.9215	1.9215	-37.52	0.2522	-0.3560	0.5022	35.07	133	155	74	57	146	565	109	
48	Surface	1	4 mon	gpcp	32.73	31.35	31.35	2.0844	2.0844	-49.12	0.3312	-1.0349	0.5755	39.59	189	222	84	34	179	708	172	
49	Surface	1	4 mon	ghcn cams grid	29.06	25.45	25.45	1.9975	1.9975	-42.03	0.3256	-0.5277	0.5706	35.81	157	180	79	36	157	609	130	
50	Surface	1	4 mon	ssmi	39.62	20.75	20.75	-0.3833		2.12	0.0060	-0.3688	-0.0775	44.57	220	144	220	142	213	939	214	
51	Surface	1	4 mon	opi2	33.68	32.33	32.33	2.1800	2.1800	-51.78	0.3335	-0.9486	0.5775	41.47	202	226	86	32	198	744	193	
52	Surface	1	4 mon	opi3	33.02	30.51	30.51	2.0492	2.0492	-47.80	0.2988	-0.8488	0.5467	40.12	191	216	83	40	186	716	180	
53	Surface	1	4 mon	cmcp	31.71	28.72	28.72	1.9990	1.9990	-45.18	0.2889	-0.7457	0.5375	38.68	178	209	80	43	174	684	155	
54	Surface	1	4 mon	cmcp2	31.67	28.66	28.66	1.9965	1.9965	-45.08	0.2889	-0.7444	0.5375	38.61	177	206	78	44	173	678	153	
55	Surface	2	Yes	reanalysis	23.05	18.90	18.90	0.3787	2.6406	-9.01	0.2448	-1.6867	0.4948	27.46	40	134	97	59	36	366	41	Plot
56	Surface	2	Yes	gpcp	20.38	17.59	17.59	0.9350	1.0695	-16.55	0.5806	0.1490	0.7620	24.78	22	127	5	19	20	193	10	
57	Surface	2	Yes	gpcp	26.80	26.51	26.51	0.7176	1.3935	-20.34	0.3824	-0.5539	0.6184	32.62	109	183	25	30	109	456	64	
58	Surface	2	Yes	ghcn cams grid	22.93	21.24	21.24	0.8966	1.1153	-19.59	0.5918	-0.1092	0.7693	26.87	38	145	10	16	30	239	12	
59	Surface	2	Yes	ssmi	42.65	15.22	15.22	-0.8459		10.58	0.1280	-0.8861	-0.3578	48.86	225	121	208	196	224	974	224	
60	Surface	2	Yes	opi2	28.47	27.53	27.53	0.9307	1.0745	-26.43	0.5668	-0.4717	0.7529	32.83	147	193	6	20	111	477	71	
61	Surface	2	Yes	opi3	26.80	25.74	25.74	0.8224	1.0841	-24.50	0.6005	-0.3798	0.7749	30.60	110	181	7	12	64	374	43	

Mississippi Basin - Monthly Surface

Ol	Runoff	Re	Storage	Precipitation	Absolute Error	Bias	Abias	Slope	Oslope	Intercept	R <sup>2</sup>	R <sup>2</sup> 1:1	Corr Coef	RMSE	AER	BR	SR	CCR	RMR	TR	BR
102	Surface	2.4 mon	gpcp		28.17	27.07	27.07	1.1283	1.1283	-28.90	0.1086	-0.9359	0.3295	36.50	141	192	11	96	162	602	124
103	Surface	2.4 mon	ghcn cams grid		25.58	22.25	22.25	1.2077	1.2077	-25.68	0.1396	-0.6296	0.3736	32.48	90	152	14	85	104	445	61
104	Surface	2.4 mon	ssmi		42.77	16.23	16.23	-1.4525		24.66	0.0965	-0.3697	-0.3107	42.42	226	126	213	176	206	947	215
105	Surface	2.4 mon	opi2		30.18	28.46	28.46	1.3984	1.3984	-35.03	0.1653	-0.9536	0.4066	37.84	165	199	26	77	168	635	137
106	Surface	2.4 mon	opi3		28.86	26.64	26.64	1.2676	1.2676	-31.05	0.1460	-0.9012	0.3821	36.00	153	187	18	84	158	600	122
107	Surface	2.4 mon	cmmap		27.54	24.85	24.85	1.2174	1.2174	-28.43	0.1370	-0.7885	0.3701	34.62	125	171	17	86	141	540	92
108	Surface	2.4 mon	cmmap2		27.49	24.79	24.79	1.2149	1.2149	-28.33	0.1368	-0.7870	0.3689	34.56	124	169	16	87	140	536	90
109	Reanalysis	1 Yes	reanalysis		41.66	2.84	2.84	1.5314	1.5314	6.32	0.2700	0.2351	0.5197	50.62	224	31	38	52	225	570	111
110	Reanalysis	1 Yes	gpcp		33.67	1.44	1.44	1.5241	1.5241	4.53	0.5643	0.4970	0.7512	41.66	201	20	37	21	199	478	73
111	Reanalysis	1 Yes	gpcp		30.82	11.45	11.45	1.6135	1.6135	20.15	0.6181	0.4884	0.7862	40.80	173	105	52	11	194	535	89
112	Reanalysis	1 Yes	gpcp trmm		33.19	3.61	3.61	1.7484	1.7484	14.09	0.5546	0.4490	0.7447	42.38	194	38	68	22	205	527	86
113	Reanalysis	1 Yes	ghcn cams grid		31.95	5.29	5.29	1.6198	1.6198	10.97	0.5910	0.4961	0.7688	41.11	183	44	54	17	195	493	76
114	Reanalysis	1 Yes	ssmi		56.82	-1.24	1.24	-0.2495		-8.52	0.0214	-0.5160	-0.1463	69.53	227	19	138	161	227	772	202
115	Reanalysis	1 Yes	opi2		30.32	11.32	11.32	1.6129	1.6129	21.30	0.6436	0.5130	0.8022	40.71	169	104	51	5	193	522	86
116	Reanalysis	1 Yes	opi3		30.28	10.02	10.02	1.6106	1.6106	19.16	0.6388	0.5175	0.7992	40.52	167	93	49	7	188	504	81
117	Reanalysis	1 Yes	cmmap		31.02	7.90	7.90	1.6120	1.6120	15.62	0.6243	0.5161	0.7901	40.67	175	70	50	10	192	497	76
118	Reanalysis	1 Yes	cmmap2		31.01	7.85	7.85	1.6145	1.6145	15.58	0.6251	0.5165	0.7906	40.65	174	67	53	9	191	494	77
119	Reanalysis	1 No	reanalysis		16.72	3.01	3.01	-0.0328		-3.76	0.0123	-12.4728	-0.1111	21.30	13	32	144	150	17	356	37
120	Reanalysis	1 No	gpcp		24.92	1.66	1.66	-0.0271		-4.40	0.0117	-16.7769	-0.1080	30.64	67	24	139	148	66	444	58
121	Reanalysis	1 No	gpcp		24.03	11.62	11.62	-0.0179		-2.80	0.0269	-101.7734	-0.1640	30.73	52	107	194	163	67	583	117
122	Reanalysis	1 No	gpcp trmm		24.50	5.41	5.41	-0.0783		-9.69	0.0279	-5.4955	-0.1671	29.03	60	45	149	164	45	463	63
123	Reanalysis	1 No	ghcn cams grid		24.06	5.61	5.61	-0.0227		-3.77	0.0115	-24.3059	-0.1073	29.24	54	48	152	147	50	451	63
124	Reanalysis	1 No	ssmi		23.58	0.50	0.50	0.1147	8.7184	-4.65	0.1850	-10.8445	0.4301	30.39	47	8	105	76	59	295	16
125	Reanalysis	1 No	opi2		26.74	12.06	12.06	-0.0396		-4.86	0.0255	-20.3958	-0.1597	33.25	108	114	201	162	118	703	161
126	Reanalysis	1 No	opi3		26.08	10.76	10.76	-0.0294		-4.68	0.0140	-19.4078	-0.1182	32.47	98	103	193	153	103	650	143
127	Reanalysis	1 No	cmmap		25.33	8.38	8.38	-0.0290		-4.62	0.0133	-18.0989	-0.1153	31.48	82	79	176	152	89	578	115
128	Reanalysis	1 No	cmmap2		25.31	8.33	8.33	-0.0289		-4.61	0.0132	-18.0493	-0.1147	31.44	80	78	175	151	87	571	113
129	Reanalysis	1 1 mon	reanalysis		17.58	3.06	3.06	0.0187	59.8802	-3.44	0.0032	-11.3392	0.0565	20.40	17	33	127	122	15	314	23
130	Reanalysis	1 1 mon	gpcp		26.22	1.75	1.75	-0.0947		-4.71	0.0999	-18.4109	-0.3161	32.07	102	26	140	180	99	547	96
131	Reanalysis	1 1 mon	gpcp		25.14	11.76	11.76	-0.0532		-3.29	0.2389	-108.4129	-0.4887	31.73	72	110	197	218	92	689	157
132	Reanalysis	1 1 mon	gpcp trmm		27.18	5.70	5.70	-0.2010		-11.21	0.1878	-6.7676	-0.4333	31.70	117	51	155	207	91	621	134
133	Reanalysis	1 1 mon	ghcn cams grid		25.37	5.63	5.63	-0.0760		-4.25	0.1292	-26.7160	-0.3594	30.81	83	50	154	197	65	549	100
134	Reanalysis	1 1 mon	ssmi		24.38	0.46	0.46	0.0885	11.2994	-4.80	0.1101	-11.5683	0.3318	31.40	57	7	108	95	85	352	33
135	Reanalysis	1 1 mon	opi2		28.30	12.01	12.01	-0.0992		-5.82	0.1602	-22.2821	-0.4002	34.74	144	113	200	202	142	801	206
136	Reanalysis	1 1 mon	opi3		27.56	10.68	10.68	-0.0849		-5.48	0.1167	-21.1249	-0.3416	33.86	126	101	191	191	133	742	190
137	Reanalysis	1 1 mon	cmmap		26.64	8.15	8.15	-0.0850		-5.30	0.1125	-19.4731	-0.3354	32.64	106	77	174	189	110	656	148
138	Reanalysis	1 1 mon	cmmap2		26.62	8.10	8.10	-0.0851		-5.29	0.1124	-19.4244	-0.3352	32.60	105	76	173	188	108	650	144
139	Reanalysis	1 2 mon	reanalysis		18.13	3.07	3.07	0.0413	24.2131	-3.28	0.0195	-10.7644	0.1396	19.93	19	34	123	113	11	300	17
140	Reanalysis	1 2 mon	gpcp		27.08	1.85	1.85	-0.1110		-4.90	0.1948	-19.3733	-0.4414	32.91	116	27	141	208	113	605	127
141	Reanalysis	1 2 mon	gpcp		25.76	12.16	12.16	-0.0739		-3.56	0.4667	-114.5817	-0.6832	32.19	93	116	203	226	101	739	189
142	Reanalysis	1 2 mon	gpcp trmm		29.20	6.71	6.71	-0.2994		-12.44	0.4224	-7.8957	-0.6500	33.28	158	55	156	225	119	713	178
143	Reanalysis	1 2 mon	ghcn cams grid		26.16	5.62	5.62	-0.1021		-4.48	0.2337	-27.9407	-0.4834	31.27	100	49	153	215	81	598	121
144	Reanalysis	1 2 mon	ssmi		25.50	0.38	0.38	0.0560	17.8571	-4.97	0.0440	-12.4713	0.2099	32.60	86	4	116	102	107	415	53
145	Reanalysis	1 2 mon	opi2		28.97	11.86	11.86	-0.1310		-6.31	0.2773	-23.0989	-0.5266	35.39	154	112	199	222	153	840	208
146	Reanalysis	1 2 mon	opi3		28.50	10.49	10.49	-0.1182		-5.94	0.2230	-21.8592	-0.4722	34.47	148	98	188	212	139	785	205
147	Reanalysis	1 2 mon	cmmap		27.63	8.02	8.02	-0.1190		-5.69	0.2187	-20.3814	-0.4677	33.40	130	75	172	210	122	709	173
148	Reanalysis	1 2 mon	cmmap2		27.61	7.97	7.97	-0.1191		-5.69	0.2189	-20.3143	-0.4679	33.36	127	74	171	211	121	704	165
149	Reanalysis	1 3 mon	reanalysis		17.94	3.12	3.12	0.0518	19.3050	-3.22	0.0307	-10.5190	0.1751	19.73	18	35	118	106	8	285	15
150	Reanalysis	1 3 mon	gpcp		27.20	1.93	1.93	-0.1133		-4.91	0.2028	-19.4385	-0.4501	33.01	118	29	142	209	115	613	131
151	Reanalysis	1 3 mon	gpcp		25.95	12.31	12.31	-0.0755		-3.55	0.5000	-118.1342	-0.7071	32.36	96	117	204	228	102	744	195
152	Reanalysis	1 3 mon	gpcp trmm		29.33	7.24	7.24	-0.3022		-12.02	0.4805	-8.9025	-0.6932	33.53	161	65	165	227	126	744	192
153	Reanalysis	1 3 mon	ghcn cams grid		26.21	5.56	5.56	-0.1062		-4.49	0.2549	-28.2311	-0.5048	31.31	101	47	151	220	83	602	123
154	Reanalysis	1 3 mon	ssmi		26.81	0.33	0.33	0.0325	30.7692	-5.08	0.0149	-13.1340	0.1219	33.48	111	3	124	115	125	478	72
155	Reanalysis	1 3 mon	opi2		28.71	11.69	11.69	-0.1264		-6.20	0.2553	-22.8490	-0.5052	35.11	151	109	196	221	147	824	207
156	Reanalysis	1 3 mon	opi3		28.39	10.28	10.28	-0.1238		-5.96	0.2408	-21.6312	-0.4907	34.35	146	95	186	219	136	782	204
157	Reanalysis	1 3 mon	cmmap		27.63	7.91	7.91	-0.1244		-5.73	0.2380	-20.3931	-0.4879	33.47	131	71					

Mississippi Basin - Monthly Surface																						
OI	Runoff	Re	Storage	Precipitation	Absolute Error	Bias	Abias	Slope	Oslope	Intercept	R <sup>2</sup>	R <sup>2</sup> 1:1	Corr Coef	RMSE	AER	BR	SR	CCR	RMR	TR	BR	
203	Reanalysis	2	2 mon	ghcn cams grid	24.61	4.31	4.31	-0.1081														
204	Reanalysis	2	2 mon	ssmi	23.32	1.63	1.63	0.0548	18.2482	-3.88	0.0839	-8.9367	-0.2897	30.09	63	41	146	169	56	475	69	
205	Reanalysis	2	2 mon	opi2	26.94	10.53	10.53	-0.1081		-2.00	0.0418	-12.4529	0.2045	31.00	44	23	117	103	73	360	38	
206	Reanalysis	2	2 mon	opi3	25.64	9.03	9.03	-0.1124		-4.55	0.0945	-11.0511	-0.3074	33.14	114	99	189	171	117	690	158	
207	Reanalysis	2	2 mon	cmap	25.27	6.94	6.94	-0.1142		-4.44	0.0952	-10.1246	-0.3086	31.84	91	85	179	172	96	623	135	
208	Reanalysis	2	2 mon	cmap2	25.24	6.89	6.89	-0.1141		-4.22	0.0964	-9.8095	-0.3105	31.09	78	60	160	174	76	548	99	
209	Reanalysis	2	3 mon	reanalysis	15.24	0.95	0.95	-0.0372		-3.11	0.0045	-3.4878	-0.0669	19.82	6	16	135	136	10	303	19	
210	Reanalysis	2	3 mon	gpcp	25.13	0.77	0.77	-0.1113		-3.37	0.1002	-9.9030	-0.3165	30.89	71	13	132	181	71	468	66	
211	Reanalysis	2	3 mon	gpcp	21.23	9.39	9.39	-0.0395		-1.51	0.0158	-12.5864	-0.1256	26.77	26	89	183	155	28	481	74	
212	Reanalysis	2	3 mon	gpcp trmm	26.66	12.37	12.37	-0.2121		-7.35	0.2266	-8.6760	-0.4761	31.40	107	118	205	213	84	727	183	
213	Reanalysis	2	3 mon	ghcn cams grid	24.45	4.44	4.44	-0.1115		-3.78	0.0933	-9.3986	-0.3054	30.17	58	42	147	170	57	474	67	
214	Reanalysis	2	3 mon	ssmi	23.18	1.52	1.52	0.0725	13.7931	-1.99	0.0738	-12.0585	0.2717	30.50	41	21	109	98	61	330	26	
215	Reanalysis	2	3 mon	opi2	26.95	10.63	10.63	-0.1146		-4.52	0.1109	-11.6693	-0.3330	33.30	115	100	190	187	120	712	177	
216	Reanalysis	2	3 mon	opi3	25.53	9.14	9.14	-0.1174		-4.38	0.1084	-10.6722	-0.3293	31.96	89	86	180	186	97	638	138	
217	Reanalysis	2	3 mon	cmap	24.97	7.06	7.06	-0.1178		-4.14	0.1071	-10.1112	-0.3272	31.19	69	62	162	185	78	556	104	
218	Reanalysis	2	3 mon	cmap2	24.96	7.01	7.01	-0.1177		-4.13	0.1068	-10.0857	-0.3268	31.15	68	61	161	184	77	551	101	
219	Reanalysis	2	4 mon	reanalysis	15.65	1.11	1.11	-0.0590		-3.07	0.0118	-3.7925	-0.1085	20.06	10	18	137	149	12	326	25	
220	Reanalysis	2	4 mon	gpcp	25.24	0.84	0.84	-0.1115		-3.25	0.1052	-10.3563	-0.3244	30.89	75	15	134	183	70	477	70	
221	Reanalysis	2	4 mon	gpcp	21.78	9.71	9.71	-0.0524		-1.54	0.0284	-13.2803	-0.1686	27.09	31	92	184	165	34	506	82	
222	Reanalysis	2	4 mon	gpcp trmm	25.43	12.69	12.69	-0.1348		-5.67	0.0987	-8.5621	-0.3141	30.39	85	120	207	178	60	650	142	
223	Reanalysis	2	4 mon	ghcn cams grid	24.47	4.54	4.54	-0.1210		-3.73	0.1149	-9.9876	-0.3390	30.38	59	43	148	190	58	498	79	
224	Reanalysis	2	4 mon	ssmi	24.60	1.61	1.61	0.0453	22.0751	-1.95	0.0297	-13.2104	0.1723	31.45	62	22	121	107	88	400	51	
225	Reanalysis	2	4 mon	opi2	27.32	10.72	10.72	-0.1346		-4.67	0.1600	-12.5740	-0.4001	33.77	121	102	192	201	130	746	194	
226	Reanalysis	2	4 mon	opi3	25.80	9.21	9.21	-0.1231		-4.32	0.1246	-11.2616	-0.3530	32.09	94	87	181	194	100	656	147	
227	Reanalysis	2	4 mon	cmap	25.19	7.14	7.14	-0.1211		-4.05	0.1183	-10.6333	-0.3439	31.26	74	64	164	192	80	574	114	
228	Reanalysis	2	4 mon	cmap2	25.17	7.09	7.09	-0.1211		-4.04	0.1182	-10.6087	-0.3439	31.23	73	63	163	193	79	571	112	

Mississippi Basin - Yearly Surface

Ol	Runoff	Re	Storage	Precipitation	Absolute Error	Bias	Abias	Slope	Oslope	Intercept	R <sup>2</sup>	R <sup>2</sup> 1:1	Corr Coef	RMSE	AER	BR	SR	CCR	RMR	TR	BR	Plot
1	Surface	1	Yes	reanalysis	22.91	22.91	22.91	0.3042	3.2873	-11.83	0.0998	-11.0926	0.3159	24.38	50	53	32	30	44	209	51	Plot
2	Surface	1	Yes	gpcp	22.17	22.17	22.17	0.5950	1.8807	-15.40	0.2925	-10.3119	0.5409	23.04	48	51	19	18	40	176	37	
3	Surface	1	Yes	gpcp	29.56	29.56	29.56	0.4122	2.4260	-20.51	0.3802	-35.5859	0.6166	30.14	66	71	26	11	63	237	60	
4	Surface	1	Yes	ghcn cams grid	26.02	26.02	26.02	0.3673	2.7226	-15.24	0.2494	-32.2674	0.4994	26.42	55	61	28	21	49	214	53	
5	Surface	1	Yes	ssmi	20.40	20.40	20.40	1.6210	1.6210	-28.94	0.7114	-5.9080	<b>0.8435</b>	21.00	40	47	18	2	37	144	15	
6	Surface	1	Yes	opi2	32.11	32.11	32.11	0.1031	9.8993	-17.12	0.0164	-42.3928	0.1282	32.97	72	76	44	49	71	312	74	
7	Surface	1	Yes	opi3	30.32	30.32	30.32	0.5094	1.9631	-22.12	0.3295	-30.0676	0.5741	30.80	69	73	24	15	67	248	64	
8	Surface	1	Yes	cmmap	28.53	28.53	28.53	0.5316	1.8811	-20.70	0.3371	-24.9618	0.5806	29.05	65	70	22	14	60	231	59	
9	Surface	1	Yes	cmmap2	28.48	28.48	28.48	0.5305	1.8850	-20.63	0.3383	-25.0609	0.5816	29.00	63	68	23	13	59	226	57	
10	Surface	1	No	reanalysis	22.41	22.41	22.41	1.1403	1.1403	-24.57	0.3562	-9.8592	0.5968	23.11	49	52	7	12	41	161	31	
11	Surface	1	No	gpcp	22.15	22.15	22.15	1.4615	1.4615	-29.85	0.4887	-10.0126	<b>0.6991</b>	22.73	47	50	16	8	38	159	28	
12	Surface	1	No	gpcp	30.56	30.56	30.56	1.3701	1.3701	-36.62	0.7878	-36.8983	<b>0.8762</b>	30.67	70	74	14	1	66	225	56	
13	Surface	1	No	ghcn cams grid	25.11	25.11	25.11	1.1396	1.1396	-28.13	0.6649	-31.0791	<b>0.8154</b>	25.94	54	59	6	3	48	170	35	
14	Surface	1	No	ssmi	23.03	23.03	23.03	1.5218	1.5218	-31.59	0.4689	-7.8952	<b>0.6848</b>	23.83	51	54	17	10	42	174	36	
15	Surface	1	No	opi2	32.09	32.09	32.09	0.6906	1.4480	-26.93	0.2044	-40.9560	0.4521	32.42	71	75	15	22	68	251	65	
16	Surface	1	No	opi3	30.30	30.30	30.30	1.1932	1.1932	-33.52	0.5004	-29.5699	<b>0.7074</b>	30.56	68	72	10	7	64	221	55	
17	Surface	1	No	cmmap	28.52	28.52	28.52	1.2343	1.2343	-32.43	0.5031	-24.5220	<b>0.7093</b>	28.81	64	69	11	6	58	208	50	
18	Surface	1	No	cmmap2	28.46	28.46	28.46	1.2346	1.2346	-32.38	0.5073	-24.6148	<b>0.7123</b>	28.75	62	67	12	5	57	203	48	
19	Surface	2	Yes	reanalysis	27.83	18.79	18.79	0.0748	13.4048	-4.16	0.2650	-60.8418	0.5147	32.81	60	44	47	20	70	241	62	Plot
20	Surface	2	Yes	gpcp	27.11	17.48	17.48	0.0704	14.2045	-2.79	0.1105	-27.3966	0.3324	32.50	58	41	48	28	69	244	63	
21	Surface	2	Yes	gpcp	27.75	26.23	26.23	0.1892	5.2854	-13.11	0.5358	-37.7253	<b>0.7320</b>	30.62	59	62	40	4	65	230	58	
22	Surface	2	Yes	ghcn cams grid	29.96	21.13	21.13	0.0422	23.6967	-5.99	0.0724	-59.0042	0.2691	35.03	67	48	57	33	72	277	68	
23	Surface	2	Yes	ssmi	21.25	20.09	20.09	0.2340	4.2735	-8.16	0.2723	-10.9947	0.5219	24.08	45	46	37	19	43	190	42	
24	Surface	2	Yes	opi2	35.57	27.43	27.43	-0.0159		-11.37	0.0047	-35.9437	-0.0686	40.68	78	65	74	69	76	360	76	
25	Surface	2	Yes	opi3	33.61	25.63	25.63	0.0444	22.5225	-10.53	0.0528	-45.5217	0.2297	38.00	75	60	56	40	75	306	72	
26	Surface	2	Yes	cmmap	32.17	23.85	23.85	0.0494	20.2429	-8.82	0.0641	-41.6480	0.2531	36.72	74	56	55	36	74	295	71	
27	Surface	2	Yes	cmmap2	32.12	23.79	23.79	0.0496	20.1813	-8.77	0.0649	-41.8603	0.2548	36.68	73	55	54	35	73	290	70	
28	Surface	2	No	reanalysis	19.68	19.68	19.68	0.4709	2.1236	-10.85	0.1368	-22.2824	0.3698	20.13	38	45	25	24	36	168	33	
29	Surface	2	No	gpcp	18.37	18.37	18.37	0.2606	3.8373	-6.03	0.0196	-9.2136	0.1400	19.49	35	43	35	47	33	193	43	
30	Surface	2	No	gpcp	26.44	26.44	26.44	0.8703	1.1490	-24.31	0.3178	-28.5547	0.5636	26.75	56	63	8	16	50	193	44	
31	Surface	2	No	ghcn cams grid	22.02	22.02	22.02	-0.0613		-4.30	0.0020	-24.3076	-0.0444	22.75	46	49	66	64	39	264	66	
32	Surface	2	No	ssmi	18.30	18.30	18.30	0.7473	1.3382	-14.16	0.1494	-6.7957	0.3865	19.41	34	42	13	23	32	144	14	
33	Surface	2	No	opi2	28.32	28.32	28.32	-0.5104		-3.10	0.0625	-18.4002	-0.2500	29.47	61	66	64	72	62	325	75	
34	Surface	2	No	opi3	26.52	26.52	26.52	-0.0078		-9.70	0.0000	-23.0081	-0.0046	27.30	57	64	75	62	51	309	73	
35	Surface	2	No	cmmap	24.74	24.74	24.74	0.0334	29.9401	-8.60	0.0004	-19.6724	0.0194	25.56	53	58	60	61	46	278	69	
36	Surface	2	No	cmmap2	24.68	24.68	24.68	0.0337	29.6736	-8.55	0.0004	-19.7281	0.0197	25.51	52	57	59	60	45	273	67	
37	Reanalysis	1	Yes	reanalysis	9.41	3.40	3.40	-0.0550		-3.51	0.0024	-1.0682	-0.0492	11.45	18	15	67	66	19	185	40	
38	Reanalysis	1	Yes	gpcp	8.52	1.56	1.56	0.1315	7.6046	-3.57	0.0084	-0.3880	0.0916	10.52	12	7	42	56	16	133	9	
39	Reanalysis	1	Yes	gpcp	11.24	10.63	10.63	0.3334	2.9994	1.18	0.0777	-3.4086	0.2787	12.52	23	36	30	32	24	145	16	
40	Reanalysis	1	Yes	gpcp trmm	16.65	1.28	1.28	-3.4033		-52.70	0.1798	-0.1276	-0.4241	16.94	32	5	62	73	31	203	67	
41	Reanalysis	1	Yes	ghcn cams grid	8.55	5.78	5.78	0.2878	3.4746	-0.59	0.0316	-0.6796	0.1778	10.42	13	20	33	44	15	125	4	
42	Reanalysis	1	Yes	ssmi	8.81	0.12	0.12	0.0825	12.1212	-6.54	0.0028	-0.3400	0.0526	11.21	16	1	46	57	17	137	10	
43	Reanalysis	1	Yes	opi2	12.46	11.52	11.52	0.5670	1.7637	4.65	0.1044	-1.6194	0.3231	14.45	29	38	20	29	145	17		
44	Reanalysis	1	Yes	opi3	11.80	9.93	9.93	0.3667	2.7270	0.89	0.0437	-1.3235	0.2091	13.61	28	33	29	43	28	181	30	
45	Reanalysis	1	Yes	cmmap	10.85	7.83	7.83	0.2107	4.7461	-1.78	0.0147	-0.9605	0.1214	12.50	22	27	39	50	23	161	29	
46	Reanalysis	1	Yes	cmmap2	10.81	7.78	7.78	0.2109	4.7416	-1.79	0.0147	-0.9493	0.1211	12.47	21	26	38	51	22	158	27	
47	Reanalysis	1	No	reanalysis	6.81	3.01	3.01	-0.0474		-3.88	0.0210	-11.9119	-0.1450	<b>6.86</b>	8	13	69	71	9	170	34	
48	Reanalysis	1	No	gpcp	5.52	1.66	1.66	0.0565	17.6991	-3.91	0.0120	-3.6079	0.1096	<b>6.88</b>	3	8	53	52	3	119	4	
49	Reanalysis	1	No	gpcp	11.62	11.62	11.62	-0.0285		-2.95	0.3264	-999.0000	-0.5713	12.70	27	40	73	74	25	239	61	
50	Reanalysis	1	No	gpcp trmm	3.22	2.29	2.29	-0.2465		-12.99	0.9834	-45.6759	-0.9916	<b>3.38</b>	1	11	65	75	1	153	22	
51	Reanalysis	1	No	ghcn cams grid	6.22	5.38	5.38	0.0617	16.2075	-3.01	0.0171	-9.2050	0.1307	7.49	5	19	52	48	5	129	8	
52	Reanalysis	1	No	ssmi	5.60	1.74	1.74	0.0645	15.5039	-5.05	0.0110	-2.5222	0.1051	7.12	4	9	51	54	4	122	5	
53	Reanalysis	1	No	opi2	11.62	11.62	11.62	0.0910	10.9890	-2.80	0.0209	-15.1928	0.1445	12.90	26	39	45	46	26	182	38	
54	Reanalysis	1	No	opi3	10.06	10.03	10.03	0.1382	7.2359	-2.27	0.0482	-11.6229	0.2195	11.39	19	34	41	41	18	153	23	
55	Reanalysis	1	No	cmmap	8.26	7.93	7.93	0.0649	15.4083	-3.45	0.0108	-8.3600	0.1041	9.81	11	29	50	55	11	156	26	

Poro Velho Basin - Monthly Atmospheric																			
OI	Runoff	Re	Storage	Precipitation	Absolute Error	Bias	Slope	Intercept	R <sup>2</sup>	R <sup>2</sup> 1:1	Corr Coef	RMSE	AER	BR	SR	CCR	RMR	TR	BR
1	Surface	1	Yes	reanalysis	189.41	85.67	3.3707	-161.69	0.3720	-0.1077	0.6099	205.61	8	10	19	14	7	58	10
2	Surface	1	Yes	gpcp	173.39	95.89	1.8630	-103.97	0.2608	0.0130	0.5107	202.63	1	15	6	20	2	44	1
3	Surface	1	Yes	gpcp	173.94	101.77	2.9347	-123.36	0.4110	0.0009	0.6411	211.41	2	16	13	10	9	50	6
4	Surface	1	Yes	gpcp lrmm	205.04	137.09	4.0232	55.13	0.2305	0.0561	0.4801	385.42	12	23	23	21	26	105	26
5	Surface	1	Yes	ghcn cams grid	194.81	87.89	1.4941	-125.39	0.2051	-0.1422	0.4528	208.69	9	11	2	25	8	55	9
6	Surface	1	Yes	gpi	205.10	166.58	1.4684	-202.86	0.3862	-0.2759	0.6215	238.90	13	26	1	12	10	62	11
7	Surface	1	Yes	lrmm 3a46	199.92	144.41	2.7812	-207.15	0.4937	-0.0621	0.7026	250.39	10	25	11	7	12	65	13
8	Surface	1	Yes	lrmm 3b43	204.51	136.76	3.4340	-174.44	0.5399	-0.0357	0.7348	252.26	11	22	20	5	13	71	15
9	Surface	1	Yes	ssmi	185.97	119.78	2.1374	-154.99	0.4751	0.0833	0.6893	202.64	7	21	8	9	3	48	3
10	Surface	1	Yes	opi2	176.39	83.90	2.1139	-77.94	0.2040	0.0121	0.4516	202.73	6	9	7	26	4	52	7
11	Surface	1	Yes	opi3	174.48	88.19	1.7328	-86.18	0.2168	0.0228	0.4656	201.62	3	12	5	24	1	45	2
12	Surface	1	Yes	cmap	175.09	93.73	1.6574	-97.75	0.2283	0.0076	0.4778	203.19	5	13	4	22	5	49	4
13	Surface	1	Yes	cmap2	175.03	93.76	1.6204	-97.48	0.2252	0.0066	0.4746	203.29	4	14	3	23	6	50	5
14	Surface	2	Yes	reanalysis	235.12	109.20	2.4066	-189.23	0.5078	0.2226	0.7126	270.91	16	17	10	6	14	63	12
15	Surface	2	Yes	gpcp	242.70	52.01	3.2474	-49.16	0.3849	0.1804	0.6204	278.17	19	5	18	13	17	72	18
16	Surface	2	Yes	gpcp	217.26	77.44	3.6454	-12.81	0.3136	0.1438	0.5600	303.93	14	8	21	19	22	84	21
17	Surface	2	Yes	gpcp lrmm	291.13	113.48	5.8176	-99.08	0.6612	0.1087	0.8132	340.33	26	18	26	2	25	97	25
18	Surface	2	Yes	ghcn cams grid	241.66	61.61	3.1174	-78.29	0.3972	0.1841	0.6302	277.54	18	6	16	11	16	67	14
19	Surface	2	Yes	gpi	237.75	138.21	2.1692	-215.82	0.4786	0.0940	0.6918	273.94	17	24	9	8	15	73	19
20	Surface	2	Yes	lrmm 3a46	281.49	118.18	4.7828	-181.73	0.6775	0.1438	0.8231	328.44	24	20	24	1	23	92	23
21	Surface	2	Yes	lrmm 3b43	290.27	115.62	5.4896	-111.84	0.6535	0.1135	0.8084	339.41	25	19	25	3	24	96	24
22	Surface	2	Yes	ssmi	230.24	72.66	3.1548	-116.13	0.5750	0.2703	0.7563	250.26	15	7	17	4	11	54	8
23	Surface	2	Yes	opi2	245.66	39.30	3.7757	3.64	0.3241	0.1398	0.5693	284.96	23	1	22	18	21	85	22
24	Surface	2	Yes	opi3	244.41	43.45	3.1020	-20.90	0.3362	0.1891	0.5796	280.07	22	2	15	17	20	76	20
25	Surface	2	Yes	cmap	243.52	49.75	2.9555	-43.55	0.3493	0.1777	0.5911	278.62	20	3	14	15	19	71	17
26	Surface	2	Yes	cmap2	243.53	49.79	2.8937	-44.03	0.3452	0.1785	0.5875	278.49	21	4	12	16	18	71	16

Porto Velho Basin - Yearly Atmospheric																					
Ol	Runoff	Re	Storage	Precipitation	Absolute Error	Bias	Slope	Oslope	Intercept	R <sup>2</sup>	R <sup>2</sup> 1:1	Corr Coef	RMSE	AER	BR	SR	CCR	RMR	TR	BR	Plot
1	Surface	1	Yes	reanalysis	126.02	85.80	-3.7370	3.7370	-170.35	0.5682	-2.8395	<b>0.7538</b>	118.20	20	10	13	3	15	61	11	Plot
2	Surface	1	Yes	gpcp	98.28	96.03	<b>0.8789</b>	1.1378	-87.37	0.0226	-1.9532	0.1503	109.34	12	15	5	12	7	51	8	
3	Surface	1	Yes	gpcp	102.02	102.02	-0.7183		-82.83	0.0055	-3.0228	-0.0741	118.19	13	16	18	18	14	79	18	
4	Surface	1	Yes	gpcp trmm	138.53	138.53	-3.9510		-59.75	0.6850	-20.7386	-0.8277	143.18	24	24	23	24	24	119	26	
5	Surface	1	Yes	ghcn cams grid	136.51	87.38	0.4935	2.0263	-85.50	0.0078	-2.4153	0.0885	117.58	21	11	9	14	13	68	18	
6	Surface	1	Yes	gpi	166.83	166.83	-0.3926		-60.89	0.0014	-8.0632	-0.0378	177.41	26	26	16	16	25	109	24	
7	Surface	1	Yes	trmm 3a46	140.61	140.61	2.8747	2.8747	-205.47	0.9905	-127.8034	<b>0.9952</b>	140.84	25	25	12	1	22	85	20	
8	Surface	1	Yes	trmm 3b43	137.00	137.00	-2.3932		-84.46	0.5445	-25.2690	-0.7379	141.23	22	22	19	23	23	109	23	
9	Surface	1	Yes	ssmi	118.37	118.37	2.1264	2.1264	-148.94	0.1047	-5.3380	<b>0.3236</b>	111.31	19	21	10	5	8	63	12	
10	Surface	1	Yes	opi2	<b>87.88</b>	83.32	-2.9889		-64.62	0.2195	-1.5646	-0.4685	<b>101.89</b>	4	9	21	22	3	59	10	
11	Surface	1	Yes	opi3	90.26	87.47	<b>1.0037</b>	1.0037	-80.46	0.0411	-1.5569	0.2028	<b>101.74</b>	8	12	1	8	2	31	3	
12	Surface	1	Yes	cmep	95.95	93.77	<b>1.0611</b>	1.0611	-88.70	0.0509	-1.8522	0.2256	<b>107.45</b>	10	13	4	6	4	37	6	
13	Surface	1	Yes	cmep2	95.98	93.81	<b>0.9501</b>	1.0525	-87.18	0.0437	-1.8673	0.2092	<b>107.74</b>	11	14	3	7	5	40	7	
14	Surface	2	Yes	reanalysis	115.11	109.24	2.2023	2.2023	-176.66	0.2823	-0.7540	<b>0.5314</b>	139.39	17	17	11	4	21	70	17	Plot
15	Surface	2	Yes	gpcp	86.89	52.05	0.1163	8.5985	-41.47	0.0001	-0.1795	0.0116	114.30	6	5	15	15	12	53	9	
16	Surface	2	Yes	gpcp	102.28	77.50	-2.9808		-87.52	0.0312	-0.5703	-0.1767	131.45	14	8	20	20	19	81	19	
17	Surface	2	Yes	gpcp trmm	113.72	113.72	-8.7258		-142.77	0.7753	-4.1176	-0.8805	129.77	16	19	26	25	18	104	21	
18	Surface	2	Yes	ghcn cams grid	93.19	61.65	-0.6688		-33.23	0.0047	-0.2792	-0.0689	119.04	9	6	17	17	16	65	14	
19	Surface	2	Yes	gpi	138.32	138.32	-3.0573		111.47	0.0250	-1.8685	-0.1581	177.85	23	23	22	19	26	113	25	
20	Surface	2	Yes	trmm 3a46	111.41	111.41	8.2320	8.2320	-221.19	0.9721	-4.3578	<b>0.9859</b>	120.50	15	18	14	2	17	66	15	
21	Surface	2	Yes	trmm 3b43	115.86	115.86	-6.6985		-122.35	0.7801	-4.3297	-0.8832	132.43	18	20	25	26	20	109	22	
22	Surface	2	Yes	ssmi	<b>81.44</b>	<b>66.18</b>	<b>1.0308</b>	<b>1.0308</b>	-43.31	0.0120	-0.4955	0.1095	<b>72.94</b>	1	7	2	13	1	24	1	Plot
23	Surface	2	Yes	opi2	90.19	39.35	-4.4774		-94.36	0.1679	-0.1610	-0.4098	113.40	7	1	24	21	11	64	13	
24	Surface	2	Yes	opi3	<b>66.69</b>	<b>43.49</b>	1.6188	1.6188	-30.80	0.0346	-0.0788	0.1859	109.32	2	2	7	11	6	28	2	
25	Surface	2	Yes	cmep	87.55	49.79	1.6711	1.6711	-42.52	0.0410	-0.1248	0.2024	111.62	3	3	8	9	9	32	4	
26	Surface	2	Yes	cmep2	<b>87.72</b>	<b>49.83</b>	1.6047	1.6047	-42.76	0.0406	-0.1257	0.2014	111.67	5	4	6	10	10	35	5	

Porto Velho Basin - Monthly Surface																							
OI	Runoff	Re	Storage	Precipitation	Absolute Error	Bias	Abias	Slope	Oslope	Intercept	R <sup>2</sup>	R <sup>2</sup> 1.1	Corr Coef	RMSE	AER	BR	SR	CCR	RMR	TR	BR	Plot	
1	Surface	1	Yes	reanalysis	46.37	31.21	31.21	0.3832	2.6096	2.48	0.5829	-1.7068	0.7634	58.13	40	88	107	42	44	321	32	Plot	
2	Surface	1	Yes	gpcp	47.90	38.72	38.72	0.5703	1.7535	-14.80	0.4993	-0.2638	0.7066	62.86	51	120	82	59	72	384	53		
3	Surface	1	Yes	gpcp	47.15	40.33	40.33	0.5064	1.9747	-14.91	0.6080	-0.7316	0.7797	60.80	44	143	92	34	59	372	49		
4	Surface	1	Yes	ghcn cams grid	40.59	26.09	26.09	0.6375	1.5686	-6.29	0.5747	0.1945	0.7581	53.12	13	63	70	44	27	217	16		
5	Surface	1	Yes	gpi	41.96	-24.82	24.82	1.0783	1.0783	20.79	0.7343	0.6535	0.8569	52.69	18	45	7	14	24	108	5		
6	Surface	1	Yes	ssmi	31.44	7.89	7.89	0.8114	1.2324	1.49	0.7100	0.6583	0.8426	39.90	1	10	22	18	2	53	1	Plot	
7	Surface	1	Yes	opi2	56.72	53.05	53.05	0.4709	2.1236	-23.59	0.5603	-1.6292	0.7485	70.66	118	192	98	47	122	577	116		
8	Surface	1	Yes	opi3	55.95	47.72	47.72	0.5004	1.9984	-19.90	0.4001	-0.7566	0.6325	72.64	111	170	93	77	139	590	121		
9	Surface	1	Yes	cmapp	50.82	40.38	40.38	0.5511	1.8146	-15.38	0.4215	-0.3297	0.6492	67.81	73	144	85	70	101	473	79		
10	Surface	1	Yes	cmapp2	50.97	40.19	40.19	0.5537	1.8060	-15.35	0.4123	-0.3083	0.6421	68.32	74	141	84	72	104	475	80		
11	Surface	1	No	reanalysis	38.78	31.18	31.18	0.5366	1.8636	-5.88	0.2231	-0.7220	0.4723	46.36	9	86	87	90	7	279	24		
12	Surface	1	No	gpcp	47.19	38.63	38.63	0.2891	1.2891	-54.70	0.5318	0.0277	0.7292	55.13	45	119	30	56	32	282	25		
13	Surface	1	No	gpcp	43.51	39.98	39.98	1.1438	1.1438	-47.33	0.5540	-0.2035	0.7443	50.69	27	136	13	49	19	244	20		
14	Surface	1	No	ghcn cams grid	40.39	26.06	26.06	1.4581	1.4581	-51.07	0.5870	0.3351	0.7661	48.26	11	62	55	41	13	182	9		
15	Surface	1	No	gpi	64.00	-25.17	25.17	2.1437	2.1437	-33.32	0.5183	0.2917	0.7199	75.34	168	49	99	57	157	530	98		
16	Surface	1	No	ssmi	46.03	8.00	8.00	1.4364	1.4364	-29.77	0.4047	0.3536	0.6361	54.88	38	12	51	75	30	206	14		
17	Surface	1	No	opi2	53.81	52.95	52.95	1.0123	1.0123	-53.64	0.5399	-0.9368	0.7348	60.65	83	189	1	53	58	394	58		
18	Surface	1	No	opi3	53.38	47.62	47.62	1.1954	1.1954	-58.48	0.4759	-0.2918	0.6899	62.29	87	169	18	65	70	409	60		
19	Surface	1	No	cmapp	49.26	40.28	40.28	1.3009	1.3009	-57.01	0.4897	-0.0058	0.6998	58.97	62	142	33	62	47	346	35		
20	Surface	1	No	cmapp2	49.51	40.10	40.10	1.3231	1.3231	-58.06	0.4908	0.0109	0.7006	59.41	64	138	36	61	51	350	36		
21	Surface	1	mon	reanalysis	34.93	31.19	31.19	0.7723	1.2948	-18.77	0.4621	-0.3553	0.6798	41.19	4	87	31	66	4	192	11		
22	Surface	1	mon	gpcp	45.67	38.52	38.52	1.4490	1.4490	-63.37	0.6651	0.1284	0.8155	52.30	37	118	54	23	22	254	22		
23	Surface	1	mon	gpcp	42.24	39.92	39.92	1.3247	1.3247	-56.42	0.7348	-0.0536	0.8572	47.50	20	135	37	13	11	216	15		
24	Surface	1	mon	ghcn cams grid	38.03	26.15	26.15	1.6540	1.6540	-61.84	0.7560	0.4430	0.8695	44.21	7	65	75	9	5	161	8		
25	Surface	1	mon	gpi	59.47	-24.86	24.86	2.6491	2.6491	-58.94	0.7845	0.4035	0.8857	69.17	138	46	109	5	112	410	61		
26	Surface	1	mon	ssmi	40.70	7.73	7.73	1.9156	1.9156	-52.91	0.6928	0.5218	0.8323	47.41	14	9	89	19	10	141	6		
27	Surface	1	mon	opi2	53.62	52.96	52.96	1.1700	1.1700	-62.37	0.7177	-0.7767	0.8472	58.04	90	190	16	16	43	355	39		
28	Surface	1	mon	opi3	53.39	47.50	47.50	1.3180	1.3180	-65.10	0.5726	-0.2090	0.7567	60.38	88	166	35	45	56	390	55		
29	Surface	1	mon	cmapp	48.32	40.18	40.18	1.4424	1.4424	-64.67	0.5959	0.0747	0.7719	56.67	54	140	53	38	37	322	33		
30	Surface	1	mon	cmapp2	48.50	39.99	39.99	1.4824	1.4824	-65.58	0.5935	0.0877	0.7704	57.16	57	137	56	39	40	329	34		
31	Surface	1	mon	reanalysis	34.12	31.24	31.24	0.7901	1.2857	-19.80	0.4845	-0.3287	0.6960	40.80	3	89	28	64	3	187	10		
32	Surface	1	mon	gpcp	48.52	38.50	38.50	1.2458	1.2458	-52.06	0.4885	-0.0018	0.6989	56.13	58	117	23	63	35	296	28		
33	Surface	1	mon	gpcp	43.75	39.83	39.83	1.1632	1.1632	-48.06	0.5573	-0.1926	0.7485	50.60	28	133	15	48	18	242	19		
34	Surface	1	mon	ghcn cams grid	42.57	26.22	26.22	1.3960	1.3960	-47.80	0.5387	0.2999	0.7340	49.61	22	66	45	54	15	202	12		
35	Surface	1	mon	gpi	61.89	-24.39	24.39	2.4558	2.4558	-49.02	0.6675	0.3585	0.8170	71.58	151	43	105	22	127	448	71		
36	Surface	1	mon	ssmi	42.08	7.92	7.92	1.8991	1.8991	-51.80	0.6421	0.4847	0.8013	49.03	19	11	88	26	14	158	7		
37	Surface	1	mon	opi2	53.98	52.91	52.91	1.0826	1.0826	-55.82	0.5774	-0.8978	0.7599	60.05	96	187	3	43	53	382	51		
38	Surface	1	mon	opi3	54.39	47.43	47.43	1.1117	1.1117	-53.59	0.4043	-0.3431	0.6358	63.75	100	165	10	78	77	428	66		
39	Surface	1	mon	cmapp	50.37	40.12	40.12	1.2208	1.2208	-52.29	0.4236	-0.0523	0.6509	60.54	69	139	20	69	57	354	38		
40	Surface	1	mon	cmapp2	50.56	39.92	39.92	1.2322	1.2322	-52.72	0.4181	-0.0400	0.6466	61.14	72	134	21	71	62	360	42		
41	Surface	1	mon	reanalysis	37.82	31.32	31.32	0.6008	1.6644	-9.58	0.2813	-0.6287	0.5304	45.09	5	91	78	83	6	263	23		
42	Surface	1	mon	gpcp	54.80	38.19	38.19	0.7359	1.3589	-23.69	0.1675	-0.3156	0.4092	64.48	102	116	42	95	80	435	69		
43	Surface	1	mon	gpcp	48.99	39.22	39.22	0.7214	1.3862	-25.32	0.2057	-0.5363	0.4535	57.64	60	125	44	92	42	363	46		
44	Surface	1	mon	ghcn cams grid	52.03	26.11	26.11	0.8000	1.2500	-15.22	0.1766	-0.0277	0.4203	60.20	79	64	25	93	54	315	31		
45	Surface	1	mon	gpi	71.35	-24.63	24.63	1.6607	1.6607	-8.34	0.2933	0.1714	0.5415	81.61	204	44	76	82	181	587	119		
46	Surface	1	mon	ssmi	51.38	7.28	7.28	1.3662	1.3662	-24.75	0.3052	0.2720	0.5524	58.56	75	8	43	81	46	253	21		
47	Surface	1	mon	opi2	57.04	52.50	52.50	0.6495	1.5396	-33.27	0.2164	-1.2933	0.4652	66.10	121	184	63	91	88	547	102		
48	Surface	1	mon	opi3	58.83	47.06	47.06	0.6331	1.5795	-28.93	0.1289	-0.6431	0.3590	70.66	128	164	71	105	123	591	122		
49	Surface	1	mon	cmapp	55.91	39.74	39.74	0.6997	1.4292	-23.26	0.1368	-0.3399	0.3698	68.45	110	132	49	101	105	497	89		
50	Surface	1	mon	cmapp2	56.10	39.53	39.53	0.7022	1.4241	-23.19	0.1335	-0.3235	0.3654	69.12	114	129	48	103	110	504	93		
51	Surface	1	mon	reanalysis	43.41	31.31	31.31	0.2645	3.7807	8.71	0.0545	-1.1502	0.2335	51.89	25	90	123	133	20	391	57		
52	Surface	1	mon	gpcp	63.21	37.89	37.89	0.0472	21.864	14.21	0.0007	-0.7305	0.0262	74.09	162	115	160	171	150	758	169		
53	Surface	1	mon	gpcp	55.98	38.79	38.79	0.0656	15.2439	7.56	0.0017	-1.0301	0.0410	66.48	112	155	167	92	647	140			
54	Surface	1	mon	ghcn cams grid	62.68	25.99	25.99	0.0289	34.8021	28.86	0.0002	-0.4511	0.0152	71.62	154	61	167	173	128	683	149		
55	Surface	1	mon	gpi	83.33	-25.43	25.43	0.3816	2.6205	58.10	0.0153	-0.1051	0.1238										



Porto Velho Basin - Monthly Surface

Ol	Runoff	Re	Storage	Precipitation	Absolute Error	Bias	Abias	Slope	Oslope	Intercept	R <sup>2</sup>	R <sup>2</sup> 1.1	Corr Coef	RMSE	AER	BR	SR	CCR	RMR	TR	BR
102	Surface	2	3 mon	gpcp	65.67	52.57	52.57	0.6808	1.4689	-35.06	0.1299	-0.6915	0.3604	76.79	175	185	58	104	162	684	150
103	Surface	2	3 mon	gpcp	59.22	52.70	52.70	0.6392	1.5645	-35.86	0.1347	-1.0277	0.3671	72.49	135	186	68	102	137	628	134
104	Surface	2	3 mon	ghcn cams grid	62.57	43.18	43.18	0.7720	1.2953	-30.67	0.1493	-0.3416	0.3864	72.35	153	146	32	98	134	563	110
105	Surface	2	3 mon	gpi	72.99	-11.15	11.15	1.6205	1.6205	-19.82	0.2645	0.2111	0.5143	81.82	209	14	73	85	182	563	111
106	Surface	2	3 mon	ssmi	56.35	20.96	20.96	1.3438	1.3438	-37.36	0.2757	0.1706	0.5251	64.68	117	38	40	84	81	360	43
107	Surface	2	3 mon	opi2	69.70	66.89	66.89	0.5944	1.6824	-44.63	0.1601	-1.9885	0.4001	80.29	194	230	79	96	174	773	171
108	Surface	2	3 mon	opi3	70.72	61.44	61.44	0.5780	1.7301	-38.29	0.0978	-1.0854	0.3127	83.43	200	223	81	111	188	803	178
109	Surface	2	3 mon	cmcp	67.24	54.12	54.12	0.6445	1.5516	-34.62	0.1063	-0.6929	0.3260	80.42	183	202	66	109	176	736	163
110	Surface	2	3 mon	cmcp2	67.42	53.92	53.92	0.6471	1.5454	-34.56	0.1039	-0.6653	0.3224	80.98	184	201	65	110	179	739	164
111	Surface	2	4 mon	reanalysis	76.84	-8.83	8.83	0.5549	1.8021	31.17	0.0357	0.0072	0.1890	91.27	222	6	83	140	207	658	143
112	Surface	2	4 mon	gpcp	72.87	52.25	52.25	-0.0780		6.58	0.0016	-1.1009	-0.0401	85.74	208	183	201	187	191	970	207
113	Surface	2	4 mon	gpcp	65.00	52.22	52.22	-0.1230		2.31	0.0049	-1.4989	-0.0702	80.73	172	182	225	197	178	954	206
114	Surface	2	4 mon	ghcn cams grid	71.68	42.83	42.83	-0.0531		14.74	0.0007	-0.7438	-0.0265	82.62	205	145	197	184	185	916	200
115	Surface	2	4 mon	gpi	84.65	-12.00	12.00	0.2753	3.6324	47.95	0.0076	-0.0618	0.0870	95.08	233	18	121	159	220	751	167
116	Surface	2	4 mon	ssmi	70.69	20.27	20.27	0.7543	1.3257	-8.63	0.0842	-0.0050	0.2901	71.56	199	37	38	119	126	519	95
117	Surface	2	4 mon	opi2	74.17	66.48	66.48	-0.0300		-10.17	0.0004	-2.5269	-0.0202	87.24	214	229	188	182	198	1011	219
118	Surface	2	4 mon	opi3	75.71	61.13	61.13	-0.1070		-0.61	0.0033	-1.4682	-0.0577	90.94	219	222	216	193	206	1056	227
119	Surface	2	4 mon	cmcp	73.50	53.79	53.79	-0.1027		6.50	0.0027	-1.0609	-0.0518	88.89	212	199	214	189	204	1018	221
120	Surface	2	4 mon	cmcp2	73.76	53.58	53.58	-0.1075		6.97	0.0028	-1.0260	-0.0534	89.48	213	197	217	190	205	1022	222
121	Reanalysis	1	Yes	reanalysis	53.80	-34.87	34.87	0.5191	1.9264	-23.52	0.1482	-0.3353	0.3850	67.50	92	102	90	99	86	481	85
122	Reanalysis	1	Yes	gpcp	57.18	-27.27	27.27	0.3610	2.7701	-16.73	0.1098	-0.4429	0.3314	71.73	123	72	112	108	129	544	101
123	Reanalysis	1	Yes	gpcp	49.37	-17.52	17.52	0.6013	1.6631	-13.07	0.2260	0.0368	0.4754	57.35	63	27	77	89	41	297	29
124	Reanalysis	1	Yes	gpcp trmm	62.96	-32.13	32.13	0.3674	2.7218	-23.97	0.0729	-0.3757	0.2699	78.17	156	92	110	123	167	650	142
125	Reanalysis	1	Yes	ghcn cams grid	62.20	-37.62	37.62	0.3512	2.8474	-20.71	0.1216	-0.7089	0.3487	76.28	152	114	114	106	159	645	139
126	Reanalysis	1	Yes	gpi	90.42	-82.38	82.38	0.3355	2.9806	-35.65	0.2316	-2.4784	0.4812	114.48	239	246	115	87	244	931	204
127	Reanalysis	1	Yes	trmm 3a46	70.86	-66.31	66.31	0.4982	2.0072	-48.64	0.2275	-1.0730	0.4769	92.31	201	228	94	88	210	821	184
128	Reanalysis	1	Yes	trmm 3b43	68.91	-34.34	34.34	0.2504	3.9836	-22.74	0.0403	-0.5905	0.2008	83.42	181	96	126	137	167	727	159
129	Reanalysis	1	Yes	ssmi	64.79	-50.93	50.93	0.4739	2.1101	-31.69	0.2506	-0.7380	0.5006	81.44	171	180	97	86	180	714	154
130	Reanalysis	1	Yes	opi2	51.67	-14.83	14.83	0.4470	2.2371	-12.90	0.1139	-0.1226	0.3374	63.00	77	24	101	107	74	383	52
131	Reanalysis	1	Yes	opi3	54.28	-19.12	19.12	0.3106	3.2196	-13.75	0.0778	-0.4088	0.2790	70.57	99	36	116	121	120	492	88
132	Reanalysis	1	Yes	cmcp	56.92	-25.49	25.49	0.3091	3.2352	-15.53	0.0870	-0.5306	0.2950	73.70	119	59	117	118	147	560	108
133	Reanalysis	1	Yes	cmcp2	57.24	-25.52	25.52	0.2988	3.3467	-15.39	0.0833	-0.5582	0.2887	74.39	124	60	119	120	152	575	114
134	Reanalysis	1	No	reanalysis	47.40	-34.70	34.70	-0.2443		-5.34	0.2955	-10.5427	-0.5436	66.16	48	101	250	270	89	758	168
135	Reanalysis	1	No	gpcp	55.17	-27.45	27.45	-0.1451		-8.54	0.1218	-8.9193	-0.3490	71.76	105	75	232	244	130	786	174
136	Reanalysis	1	No	gpcp	45.41	-17.87	17.87	-0.1150		-5.43	0.3985	-41.5915	-0.6313	54.92	36	30	222	276	31	595	124
137	Reanalysis	1	No	gpcp trmm	58.77	-30.85	30.85	-0.3041		-14.01	0.1472	-3.1900	-0.3836	79.47	129	84	260	253	171	897	195
138	Reanalysis	1	No	ghcn cams grid	59.36	-37.22	37.22	-0.1546		-7.13	0.2110	-15.1984	-0.4594	78.49	138	109	234	262	168	909	197
139	Reanalysis	1	No	gpi	102.13	-82.17	82.17	-0.1105		-4.09	0.1402	-24.0167	-0.3744	129.98	259	244	220	251	259	1233	261
140	Reanalysis	1	No	trmm 3a46	86.42	-60.83	60.83	-0.3471		-13.38	0.2485	-5.5222	-0.4985	109.13	236	221	264	267	239	1227	260
141	Reanalysis	1	No	trmm 3b43	61.41	-33.08	33.08	-0.2710		-13.40	0.1387	-3.6484	-0.3725	83.20	147	95	253	250	188	931	237
142	Reanalysis	1	No	ssmi	76.30	-50.33	50.33	-0.1610		-7.88	0.1380	-10.2079	-0.3715	94.67	220	179	236	248	219	1102	237
143	Reanalysis	1	No	opi2	46.89	-13.97	13.97	-0.1900		-9.82	0.1418	-5.6031	-0.3766	59.06	42	19	240	252	49	602	129
144	Reanalysis	1	No	opi3	49.65	-18.26	18.26	-0.1390		-9.40	0.1074	-7.7571	-0.3278	67.01	65	31	230	239	95	660	144
145	Reanalysis	1	No	cmcp	53.37	-24.96	24.96	-0.1223		-8.79	0.0937	-9.0087	-0.3061	71.84	86	47	224	234	131	722	157
146	Reanalysis	1	No	cmcp2	53.58	-24.99	24.99	-0.1183		-8.84	0.0900	-9.1589	-0.2999	72.37	89	48	223	233	135	728	160
147	Reanalysis	1	1 mon	reanalysis	47.31	-34.65	34.65	-0.2117		-6.12	0.2218	-10.2293	-0.4709	65.28	47	99	247	265	86	744	166
148	Reanalysis	1	1 mon	gpcp	54.17	-27.24	27.24	-0.0985		-9.35	0.0559	-8.3219	-0.2365	69.69	98	71	213	220	115	717	156
149	Reanalysis	1	1 mon	gpcp trmm	44.31	-17.56	17.56	-0.0794		-5.80	0.1899	-39.2376	-0.4358	53.49	30	28	203	259	28	548	103
150	Reanalysis	1	1 mon	ghcn cams grid	55.50	-29.12	29.12	-0.1849		-13.25	0.0666	-3.3518	-0.2582	73.58	107	79	239	228	145	798	176
151	Reanalysis	1	1 mon	gpi	58.88	-37.26	37.26	-0.1139		-8.14	0.1147	-14.5111	-0.3368	76.75	131	110	221	243	161	866	191
152	Reanalysis	1	1 mon	trmm 3a46	83.85	-59.31	59.31	-0.2932		-12.01	0.2145	-6.2818	-0.4632	105.01	232	220	257	264	233	1206	257
153	Reanalysis	1	1 mon	trmm 3b43	56.16	-29.72	29.72	-0.1674		-13.02	0.0633	-3.7358	-0.2517	76.28	115	80	238	227	158	818	183
154	Reanalysis	1	1 mon	ssmi	75.13	-49.80	49.80	-0.1441		-8.60	0.1083	-9.7767	-0.3290	93.11	218	175	231	240	213	1077	234
155	Reanalysis	1	1 mon	opi2	45.41	-14.12	14.12	-0.1317		-10.01	0.0880	-5.3393	-0.2608	57.11	35	20	228	229	39	551	104
156	Reanalysis	1	1 mon	opi3	48.40	-18.43	18.43	-0.0880		-9.79	0.0430	-7.1849	-0.2073	64.89	55	32	206	217	85	595	125
157	Reanalysis	1	1 mon	cmcp	52.08	-25.20	25.20	-0.0756		-9.46	0.0357	-8.4123	-0.1889	69.78	80	50	200	218	110	662	146
158	Reanalysis	1	1 mon	cmcp2	52.25	-25.23	25.23	-0.0727		-9.50	0.0338	-8.5810	-0.1839	70.33	82	51	199	215	119	666	147
159	Reanalysis	1	1 mon	reanalysis	46.80	-34.64	34.64	-0.1471		-7.66	0.1071	-9.5922	-0.3272	63.43	41	97	233	238	76	885	151
160	Reanalysis	1	2 mon	gpcp	52.57	-27.13	27.13	-0.0440		-10.28	0.0111	-7.6711	-0.1055	67.30	84	89	194	204	97	648	141
161	Reanalysis	1	2 mon	gpcp	42.43	-17.33	17.33	-0.0431		-6.27	0.0561	-36.9920	-0.2369	51.99	21	25	193	221	21	481	84
162	Reanalysis	1	2 mon	gpcp trmm	52.50	-27.89	27.89	-0.0967		-12.67	0.0210	-3.3851	-0.1448	69.29	83	77	212	211	113	696	152
163	Reanalysis	1	2 mon	ghcn cams grid	57.06	-37.29	37.29	-0.0471		-9.85	0.0196	-13.3431	-0.1402	73.80	122	111	195	210	148	786	175

Porto Velho Basin - Monthly Surface

QI	Runoff	Re	Storage	Precipitation	Absolute Error	Bias	Abias	Slope	Oslope	Intercept	Rv2	Rv2 1.1	Corr Coef	RMSE	AER	BR	SR	CCR	RMR	TR	BR
203	Reanalysis	2	Yes	gpcn cams grid	67.44	-57.77	57.77	-0.0132		-46.87	0.0003	-3.3624	-0.0175	96.14	185	216	182	181	221	985	213
204	Reanalysis	2	Yes	gpi	110.45	-105.46	105.46	0.0339	29.4965	-50.88	0.0043	-8.6288	0.0657	144.53	261	275	166	164	266	1132	244
205	Reanalysis	2	Yes	trmm 3a46	98.59	-96.23	96.23	0.0050	200.0000	-79.42	0.0000	-6.0870	0.0067	122.96	248	259	179	179	251	1116	240
206	Reanalysis	2	Yes	trmm 3b43	88.40	-83.33	83.33	-0.1255		-84.28	0.0182	-4.3032	-0.1350	113.59	238	251	226	209	243	1167	251
207	Reanalysis	2	Yes	ssmi	78.40	-71.92	71.92	0.1100	9.9099	-52.02	0.0268	-4.2231	0.1636	103.99	226	235	138	143	232	974	210
208	Reanalysis	2	Yes	opi2	53.70	-35.47	35.47	-0.0964		-48.12	0.0098	-1.8474	-0.0989	77.67	91	107	211	201	165	775	172
209	Reanalysis	2	Yes	opi3	56.02	-39.61	39.61	0.0097	103.0928	-46.94	0.0001	-2.1821	0.0118	82.11	113	130	177	174	184	778	173
210	Reanalysis	2	Yes	cmap	60.30	-45.91	45.91	0.0207	48.3092	-48.99	0.0007	-2.5868	0.0267	87.18	140	159	171	170	197	837	186
211	Reanalysis	2	Yes	cmap2	60.55	-45.96	45.96	0.0209	47.8469	-46.99	0.0007	-2.6284	0.0273	87.68	141	160	170	169	200	840	187
212	Reanalysis	2	No	reanalysis	124.78	-105.45	105.45	-0.2867		-30.36	0.2653	-9.4844	-0.5151	162.67	274	274	256	269	275	1348	275
213	Reanalysis	2	No	gpcp	78.93	-48.26	48.26	-0.3639		-46.67	0.1732	-3.1822	-0.4162	102.74	228	174	266	258	230	1156	248
214	Reanalysis	2	No	gpc	66.18	-32.90	32.90	-0.4809		-36.63	0.3272	-3.4188	-0.5720	86.21	178	94	271	274	193	1010	218
215	Reanalysis	2	No	gpc trmm	99.86	-83.65	83.65	-0.6916		-88.71	0.3619	-3.8884	-0.8016	128.08	254	252	276	275	256	1313	268
216	Reanalysis	2	No	gpcn cams grid	84.81	-57.85	57.85	-0.3817		-42.99	0.2130	-3.9037	-0.4615	111.25	234	217	268	263	241	1223	259
217	Reanalysis	2	No	gpi	128.67	-105.68	105.68	-0.2785		-33.57	0.2349	-8.9793	-0.4846	162.73	276	276	254	266	276	1348	276
218	Reanalysis	2	No	trmm 3a46	114.81	-98.14	98.14	-0.5279		-72.33	0.3226	-5.3438	-0.5680	143.59	265	260	273	273	264	1335	273
219	Reanalysis	2	No	trmm 3b43	101.11	-85.80	85.80	-0.6150		-87.16	0.3176	-4.0664	-0.5636	130.39	257	257	275	272	260	1321	270
220	Reanalysis	2	No	ssmi	99.30	-72.47	72.47	-0.3238		-42.85	0.1939	-5.1681	-0.4403	123.57	250	238	262	260	252	1282	265
221	Reanalysis	2	No	opi2	70.59	-35.55	35.55	-0.5480		-53.42	0.2649	-2.3500	-0.5147	91.95	197	108	274	268	209	1058	228
222	Reanalysis	2	No	opi3	73.32	-39.70	39.70	-0.3336		-49.56	0.1973	-2.6820	-0.3706	96.40	210	131	263	247	222	1073	232
223	Reanalysis	2	No	cmip	77.22	-46.00	46.00	-0.3085		-47.43	0.1327	-3.0925	-0.3643	101.63	224	161	261	246	228	1120	241
224	Reanalysis	2	No	cmip2	77.35	-46.04	46.04	-0.2994		-47.41	0.1281	-3.1252	-0.3579	102.04	225	162	259	245	229	1120	242
225	Reanalysis	2	mon	reanalysis	123.49	-104.64	104.64	-0.2282		-33.66	0.1674	-9.0309	-0.4092	158.88	272	272	249	257	274	1324	271
226	Reanalysis	2	mon	gpcp	76.44	-47.77	47.77	-0.2507		-48.62	0.0823	-2.8734	-0.2869	98.73	221	171	251	232	225	1100	236
227	Reanalysis	2	mon	gpc	63.15	-31.02	31.02	-0.2784		-34.54	0.1385	-3.5021	-0.3722	77.90	159	85	255	249	164	912	198
228	Reanalysis	2	mon	gpc trmm	99.59	-83.06	83.06	-0.4638		-88.61	0.1578	-3.4498	-0.3973	122.80	253	249	270	255	250	1277	267
229	Reanalysis	2	mon	gpcn cams grid	82.27	-57.48	57.48	-0.2652		-44.03	0.1033	-3.5602	-0.3214	107.13	230	215	252	236	236	1169	252
230	Reanalysis	2	mon	gpi	127.09	-104.17	104.17	-0.2116		-36.27	0.1489	-9.1538	-0.3558	157.82	275	269	246	254	273	1317	269
231	Reanalysis	2	mon	trmm 3a46	119.20	-100.47	100.47	-0.5108		-72.75	0.2980	-5.3720	-0.5459	144.89	269	261	272	271	267	1340	274
232	Reanalysis	2	mon	trmm 3b43	101.03	-85.25	85.25	-0.3649		-88.37	0.1091	-3.5625	-0.3303	124.34	252	237	258	256	249	1252	262
233	Reanalysis	2	mon	ssmi	99.56	-72.20	72.20	-0.2962		-44.00	0.1597	-4.9849	-0.3997	122.31	252	207	258	249	196	998	216
234	Reanalysis	2	mon	opi2	67.87	-35.15	35.15	-0.3515		-50.96	0.1094	-1.9985	-0.3307	86.87	190	105	265	242	211	1007	217
235	Reanalysis	2	mon	opi3	70.45	-39.25	39.25	-0.2240		-48.55	0.0821	-2.4046	-0.2492	92.56	198	128	248	226	211	1007	217
236	Reanalysis	2	mon	cmip	74.63	-45.56	45.56	-0.2107		-47.12	0.0621	-2.8140	-0.2492	97.97	218	155	245	225	223	1084	230
237	Reanalysis	2	mon	cmip2	74.76	-45.61	45.61	-0.2037		-47.10	0.0596	-2.8469	-0.2441	98.39	217	157	244	223	224	1065	231
238	Reanalysis	2	mon	reanalysis	119.89	-104.28	104.28	-0.1301		-39.27	0.0543	-8.3532	-0.2330	153.83	270	270	227	219	272	1258	263
239	Reanalysis	2	mon	gpcp	72.75	-47.55	47.55	-0.1029		-46.67	0.1039	-2.4753	-0.1178	93.64	207	167	215	208	215	1012	220
240	Reanalysis	2	mon	gpc	59.15	-30.16	30.16	-0.1915		-33.70	0.0682	-3.2729	-0.2611	74.34	133	81	241	230	151	836	185
241	Reanalysis	2	mon	gpc trmm	98.01	-82.60	82.60	-0.0812		-88.34	0.0048	-2.8175	-0.0692	114.86	247	247	204	195	245	1138	245
242	Reanalysis	2	mon	gpcn cams grid	78.63	-57.20	57.20	-0.0884		-45.83	0.0115	-3.0229	-0.1070	100.75	227	212	207	206	227	1079	235
243	Reanalysis	2	mon	gpi	123.52	-103.57	103.57	-0.1323		-40.39	0.0587	-8.6358	-0.2422	153.35	273	264	229	222	271	1259	264
244	Reanalysis	2	mon	trmm 3a46	121.03	-102.55	102.55	-0.4226		-73.89	0.1968	-5.1532	-0.4437	144.03	271	262	269	261	265	1328	272
245	Reanalysis	2	mon	trmm 3b43	99.46	-84.55	84.55	-0.0405		-88.04	0.0013	-2.9393	-0.0364	116.68	251	254	190	186	246	1127	243
246	Reanalysis	2	mon	ssmi	96.77	-71.95	71.95	-0.2002		-45.85	0.0740	-4.6750	-0.2721	118.62	246	238	243	231	247	1203	255
247	Reanalysis	2	mon	opi2	63.44	-35.05	35.05	-0.1830		-48.86	0.0235	-1.6619	-0.1535	81.96	165	103	237	212	183	900	196
248	Reanalysis	2	mon	opi3	66.92	-39.03	39.03	-0.0957		-47.49	0.0113	-2.0780	-0.1064	88.13	182	122	209	205	202	920	201
249	Reanalysis	2	mon	cmip	71.13	-45.37	45.37	-0.0837		-48.87	0.0098	-2.4494	-0.0990	93.30	202	151	205	202	214	974	209
250	Reanalysis	2	mon	cmip2	71.24	-45.42	45.42	-0.0793		-48.86	0.0090	-2.4809	-0.0950	93.72	203	152	202	198	216	971	208
251	Reanalysis	2	mon	reanalysis	115.61	-104.15	104.15	-0.0197		-45.60	0.0012	-7.6176	-0.0353	147.73	266	266	183	185	270	1172	254
252	Reanalysis	2	mon	gpcp	67.51	-47.60	47.60	0.0987	10.1317	-48.81	0.0128	-1.9454	0.1129	85.36	187	168	144	154	195	848	189
253	Reanalysis	2	mon	gpc	53.34	-30.32	30.32	0.0413	24.2131	-33.00	0.0032	-2.4070	0.0562	66.64	85	82	164	166	93	590	120
254	Reanalysis	2	mon	gpc trmm	93.78	-83.06	83.06	0.2415	4.1408	-86.87	0.0423	-2.3285	0.2056	108.42	243	250	129	136	237	995	215
255	Reanalysis	2	mon	gpcn cams grid	73.49	-57.23	57.23	0.0901	11.0988	-47.67	0.0119	-2.4956	0.1092	94.09	211	157	149	156	217	946	205
256	Reanalysis	2	mon	gpi	117.93	-103.58	103.58	0.0115	86.9565	-48.47	0.0004	-7.8517	0.0210	145.67	268	265	175	172	268	1148	246
257	Reanalysis	2	mon	trmm 3a46	116.76	-103.48	103.48	-0.1088		-80.18	0.0130	-4.4301	-0.1138	136.							

Porto Velho - Yearly Surface																							
OI	Runoff	Re	Storage	Precipitation	Absolute Error	Bias	Abias	Slope	Oslope	Intercpt	R <sup>2</sup>	R <sup>2</sup> 1.1	Corr Coef	RMSE	AER	BR	SR	CCR	RMR	TR	BR	Plot	
1	Surface	1	Yes	reanalysis	32.18	32.18	32.18	0.3940	2.5381	1.51	0.1007	-7.1327	0.3174	34.70	27	27	40	40	26	160	26	15	
2	Surface	1	Yes	gpcp	39.75	39.75	39.75	0.7323	1.3656	-24.58	0.5204	-12.9044	0.7214	40.56	41	42	16	4	34	137	15		
3	Surface	1	Yes	gpcp	43.29	43.29	43.29	0.2604	3.8402	-3.01	0.0994	-51.4138	0.3152	44.01	48	48	63	41	42	242	51		
4	Surface	1	Yes	ghcn cams grid	27.07	27.07	27.07	0.7035	1.4215	-13.49	0.4368	-6.7000	0.6609	31.65	23	23	19	11	23	99	5		
5	Surface	1	Yes	gpi	21.87	-21.87	21.87	0.3034	3.2960	59.80	0.1550	-15.5277	0.3937	23.06	14	14	52	37	9	126	10		
6	Surface	1	Yes	ssmi	15.00	14.58	14.58	0.3875	2.5806	19.40	0.1837	-4.9685	0.4286	16.44	5	9	42	34	3	93	3		
7	Surface	1	Yes	opi2	54.08	54.08	54.08	0.1403	7.1276	-5.33	0.0227	-30.2290	0.5873	49.95	59	59	18	17	52	205	41		
8	Surface	1	Yes	opi3	48.75	48.75	48.75	0.7046	1.4192	-32.00	0.3449	-14.0934	0.5873	49.95	59	59	18	17	52	205	41		
9	Surface	1	Yes	cmapp	41.41	41.41	41.41	0.7580	1.3193	-27.68	0.3605	-9.0455	0.6004	42.88	47	47	13	15	40	162	29		
10	Surface	1	Yes	cmapp2	41.22	41.22	41.22	0.7584	1.3186	-27.52	0.3366	-8.3607	0.5802	42.85	46	46	12	18	39	161	28		
11	Surface	1	No	reanalysis	31.18	31.18	31.18	0.1955	5.1151	12.74	0.0195	-6.8785	0.1396	34.15	26	26	66	66	24	208	42		
12	Surface	1	No	gpcp	38.63	38.63	38.63	0.7397	1.3519	-24.16	0.4083	-12.2525	0.6390	39.60	39	39	14	14	30	136	14		
13	Surface	1	No	gpcp	39.98	39.98	39.98	0.4135	2.4184	-9.99	0.3164	-43.5767	0.5625	40.58	42	43	38	24	35	182	35		
14	Surface	1	No	ghcn cams grid	26.06	26.06	26.06	0.6814	1.4676	-11.47	0.3151	-6.2987	0.5614	30.81	21	21	22	25	22	111	8		
15	Surface	1	No	gpi	25.17	-25.17	25.17	0.4109	2.4337	55.30	0.3588	-20.0819	0.5990	26.04	18	18	39	16	105	6			
16	Surface	1	No	ssmi	11.97	9.74	9.74	0.2911	3.4352	26.16	0.1569	-2.8684	0.3961	13.23	3	3	53	36	2	97	4		
17	Surface	1	No	opi2	52.95	52.95	52.95	0.5441	1.8379	-27.61	0.2627	-28.1102	0.5126	53.81	61	61	27	28	58	235	49		
18	Surface	1	No	opi3	47.62	47.62	47.62	0.9248	1.0813	-43.44	0.4569	-13.2680	0.6759	46.56	57	57	5	9	49	177	34		
19	Surface	1	No	cmapp	40.28	40.28	40.28	0.9538	1.0487	-37.71	0.4388	-8.4294	0.6624	41.54	45	45	45	3	10	38	141	16	
20	Surface	1	No	cmapp2	40.10	40.10	40.10	1.0160	1.0160	-40.99	0.4846	-7.7324	0.6816	41.39	43	44	2	8	37	134	12		
21	Surface	2	Yes	reanalysis	20.40	-5.71	5.71	0.3809	2.6254	40.26	0.0865	-0.1925	0.2941	27.73	13	1	43	42	15	174	9	Plot	
22	Surface	2	Yes	gpcp	53.35	53.35	53.35	0.2742	3.6470	-12.85	0.2641	-27.5967	0.5139	55.94	63	63	57	27	66	276	63		
23	Surface	2	Yes	gpcp	53.84	53.84	53.84	0.0240	41.6667	-3.75	0.0044	-82.2493	0.0665	56.75	65	65	75	74	68	347	81		
24	Surface	2	Yes	ghcn cams grid	44.20	43.90	43.90	0.2515	3.9761	-2.13	0.2068	-18.0245	0.4548	47.29	52	52	64	33	47	248	52		
25	Surface	2	Yes	gpi	17.02	-11.31	11.31	-0.0217	0.1170	63.75	0.0048	-14.8839	-0.0691	21.82	8	4	92	77	7	188	37		
26	Surface	2	Yes	ssmi	11.47	11.47	11.47	0.1170	8.5470	19.59	0.0421	-28.3117	0.2052	34.29	1	5	71	58	25	160	25		
27	Surface	2	Yes	opi2	67.68	67.68	67.68	0.0391	25.5754	-14.06	0.0063	-53.1847	0.0797	70.90	82	82	74	73	79	390	83		
28	Surface	2	Yes	opi3	62.35	62.35	62.35	0.3487	2.8678	-26.00	0.3198	-27.3936	0.5655	64.42	74	74	48	21	73	290	70		
29	Surface	2	Yes	cmapp	55.01	55.01	55.01	0.3732	2.6795	-20.03	0.3294	-19.2164	0.5739	57.32	70	70	45	19	70	274	62		
30	Surface	2	Yes	cmapp2	54.82	54.82	54.82	0.3803	2.6295	-20.24	0.3185	-17.7492	0.5644	57.20	69	69	44	22	69	273	61		
31	Surface	2	No	reanalysis	19.37	-5.93	5.93	1.1082	1.1082	-0.08	0.1682	0.1120	0.4101	23.93	11	2	7	35	10	85	2		
32	Surface	2	No	gpcp	53.13	53.13	53.13	0.7178	1.3931	-37.44	0.4157	-25.4419	0.6447	53.79	62	62	17	13	57	211	43		
33	Surface	2	No	gpcp	53.66	53.66	53.66	0.4871	2.0530	-27.43	0.4193	-74.4767	0.6475	54.04	64	64	33	12	61	234	48		
34	Surface	2	No	ghcn cams grid	43.68	43.68	43.68	0.6595	1.5163	-24.76	0.3267	-15.9932	0.5716	44.69	49	51	23	20	45	188	38		
35	Surface	2	No	gpi	11.49	-11.49	11.49	0.4945	2.0640	37.86	0.5455	-4.5624	0.7386	12.79	2	6	34	3	1	46	1	Plot	
36	Surface	2	No	ssmi	24.18	24.18	24.18	0.3901	2.5634	6.71	0.3178	-15.0274	0.5638	25.36	15	17	41	23	13	109	7		
37	Surface	2	No	opi2	67.45	67.45	67.45	0.5221	1.9153	-40.90	0.2594	-48.9989	0.5094	68.11	79	79	29	29	76	292	72		
38	Surface	2	No	opi3	62.12	62.12	62.12	0.9029	1.1075	-56.73	0.4925	-25.9208	0.7018	62.72	73	73	6	6	71	229	47		
39	Surface	2	No	cmapp	54.78	54.78	54.78	0.9317	1.0733	-50.99	0.4716	-17.9969	0.6867	55.57	68	68	4	7	64	211	44		
40	Surface	2	No	cmapp2	54.60	54.60	54.60	0.9941	1.0059	-54.27	0.4999	-16.5820	0.7070	55.39	67	67	1	5	63	203	40		
41	Reanalysis	1	Yes	reanalysis	35.43	-33.78	33.78	-0.1997	0.1997	-5.48	0.0922	-9.1032	-0.3037	44.33	32	32	89	83	44	280	66		
42	Reanalysis	1	Yes	gpcp	28.34	-26.34	26.34	0.5114	1.9554	-18.28	0.1147	-2.9784	0.3387	30.40	22	22	31	39	20	134	11		
43	Reanalysis	1	Yes	gpcp	14.61	-14.56	14.56	-0.2983	0.2983	-0.07	0.1216	-10.0276	-0.3488	17.27	4	8	87	86	4	189	39		
44	Reanalysis	1	Yes	gpcp trmm	36.27	-36.27	36.27	3.5048	3.5048	-76.13	0.9488	-1.9915	0.9740	40.04	35	35	55	2	31	158	21		
45	Reanalysis	1	Yes	ghcn cams grid	38.64	-36.84	36.84	0.1673	5.9773	-14.83	0.0189	-7.3047	0.1300	40.19	36	36	69	68	33	242	50		
46	Reanalysis	1	Yes	gpi	79.73	-79.73	79.73	0.7862	1.2719	-84.70	0.2484	-22.5453	0.4984	81.07	84	84	9	30	84	291	71		
47	Reanalysis	1	Yes	trmm 3a46	62.45	-62.45	62.45	-2.9977	2.9977	75.86	0.3387	-8.2280	-0.5820	67.23	75	75	78	89	75	392	84		
48	Reanalysis	1	Yes	trmm 3b43	31.05	-31.05	31.05	0.6090	1.6420	-25.00	0.0514	-1.8220	0.2268	38.33	25	25	26	53	28	157	20		
49	Reanalysis	1	Yes	ssmi	47.26	-47.26	47.26	0.4594	2.1768	-27.98	0.0477	-6.8075	0.2184	50.88	56	56	53	56	53	256	56		
50	Reanalysis	1	Yes	opi2	16.76	-13.64	13.64	0.1900	5.2632	-10.56	0.0182	-1.1145	0.1351	22.18	7	7	67	67	8	156	18		
51	Reanalysis	1	Yes	opi3	19.46	-17.78	17.78	0.2357	4.2427	-11.71	0.0347	-1.8915	0.1862	25.00	12	11	65	60	12	160	24		
52	Reanalysis	1	Yes	cmapp	24.46	-24.09	24.09	0.3187	3.1377	-14.38	0.0686	-2.7428	0.2619	29.48	16	15	50	49	18	148	17		
53	Reanalysis	1	Yes	cmapp2	24.53	-24.13	24.13	0.2903	3.4447	-13.99	0.0612	-2.8118	0.2474	29.75	17	16	54	51	19	157	19		
54	Reanalysis	1	No	reanalysis	36.08	-34.70	34.70	-0.2249	0.2249	-5.80	0.2365	-19.2996	-0.4863	44.18	34	33	88	87	43	285	67		
55	Reanalysis	1	No	gpcp	27.45	-27.45	27.45	0.3141	3.1837	-16.12	0.0681	-5.3619	0.2610	30.64	24	24	51	50	21	170	32		
56	Reanalysis	1	No	gpcp	17.87	-17.87	17.87	0.0850	11.7647	-7.66	0.0253	-33.2118	0.1591	16.99	9	12	72	62	5	160	23		
57	Reanalysis	1	No	gpcp trmm	38.81	-38.81	38.81	2.9186	2.9186	-69.34	0.9499	-3.5192	0.9746	40.96	40	40	49	1	36	186	31		
58	Reanalysis	1	No	ghcn cams grid	37.57	-37.57	37.57	0.0495	20.2020	-12.67	0.0031	-16.1701	0.0553	40.14	37	38	73	75	32	255	54		
59	Reanalysis	1	No	gpi	82.17	-82.17	82.17	0.6398	1.5630	-56.84	0.2295	-33.5857	0.4790	83.19	85	85	24	32	85	311	75		
60	Reanalysis	1	No	trmm 3a46	66.34	-66.34	66.34	-2.1680	2.1680	43.26	0.4768	-24.7297	-0.6905	68.42	78	78	80	81	97	404	86		
61	Reanalysis	1	No	trmm 3b43	33.08	-33.08	33.08	0.5241	1.9080	-25.71	0.0484	-2.6651	0.2201	38.73	30	30	28	54	29	171	33		
62	Reanalysis	1	No	ssmi	50.25	-50.25	50.25	0.2691	3.7161	-24.19	0.0235	-11.1585	0.1532	52.81	60</								

# Appendix B

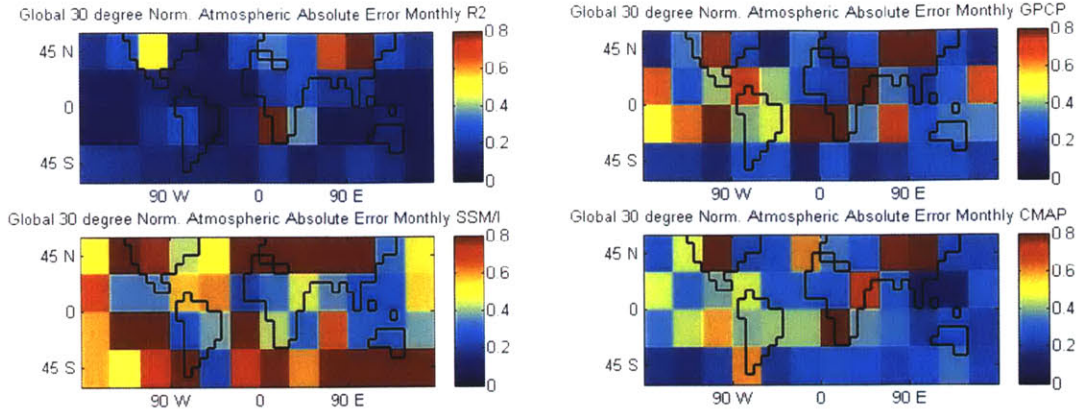


Figure B-1: 30 degree Normalized Absolute Error Different Precipitation Global Monthly

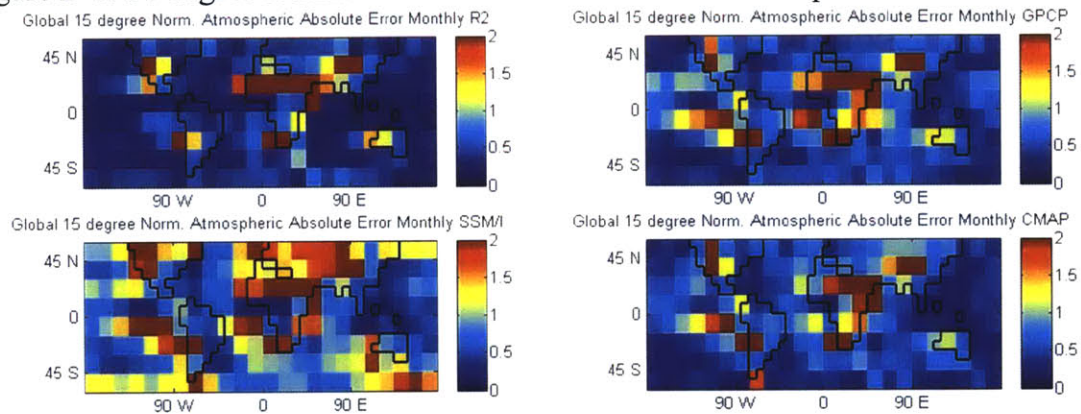


Figure B-2: 15 degree Normalized Absolute Error Different Precipitation Global Monthly

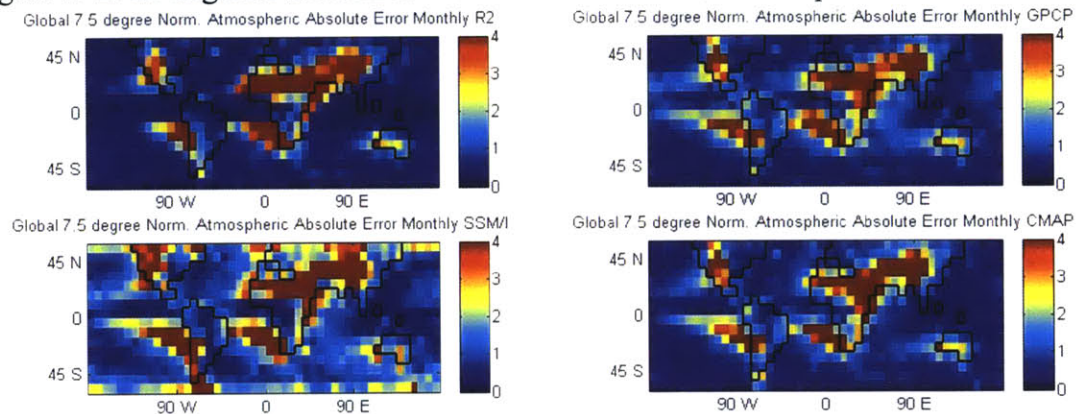


Figure B-3: 7.5 degree Normalized Absolute Error Different Precipitation Global Monthly

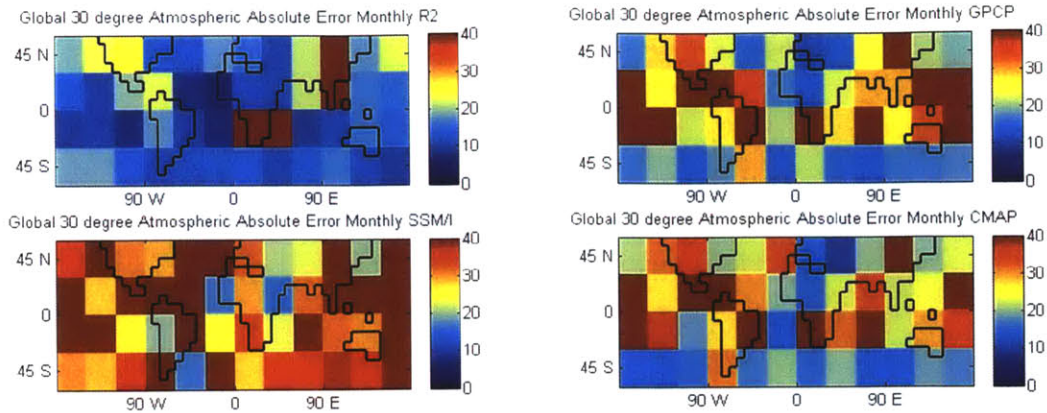


Figure B-4: 30 degree Absolute Error Different Precipitation Global Monthly

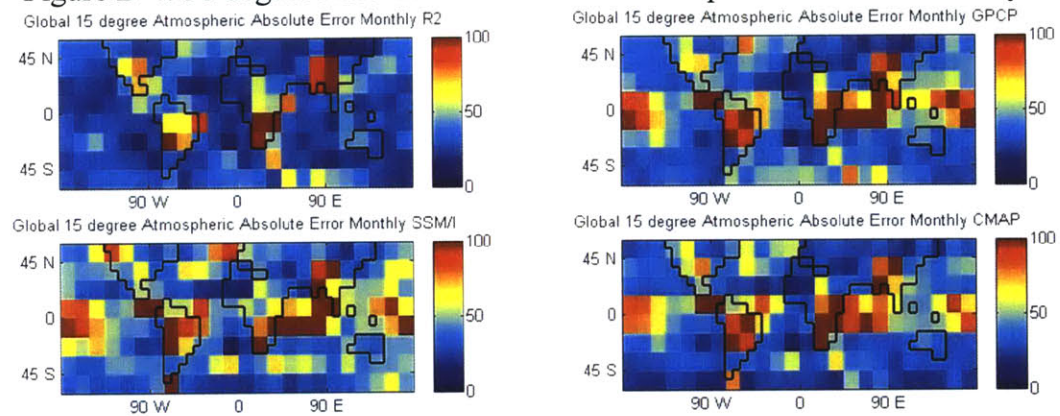


Figure B-5: 15 degree Absolute Error Different Precipitation Global Monthly

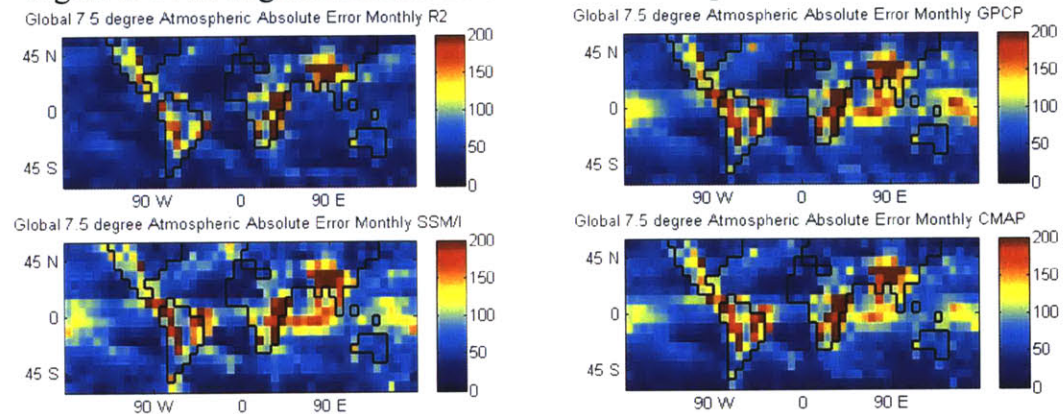


Figure B-6: 7.5 degree Absolute Error Different Precipitation Global Monthly

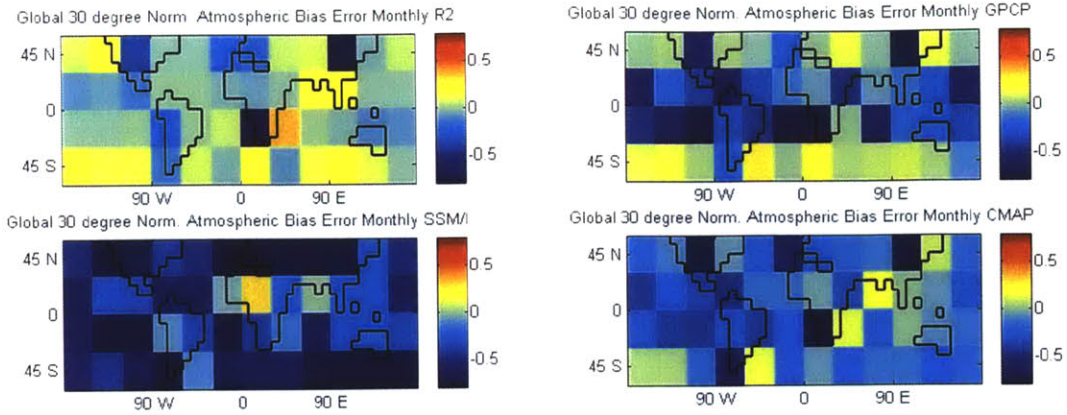


Figure B-7: 30 degree Normalized Bias Error Different Precipitation Global Monthly

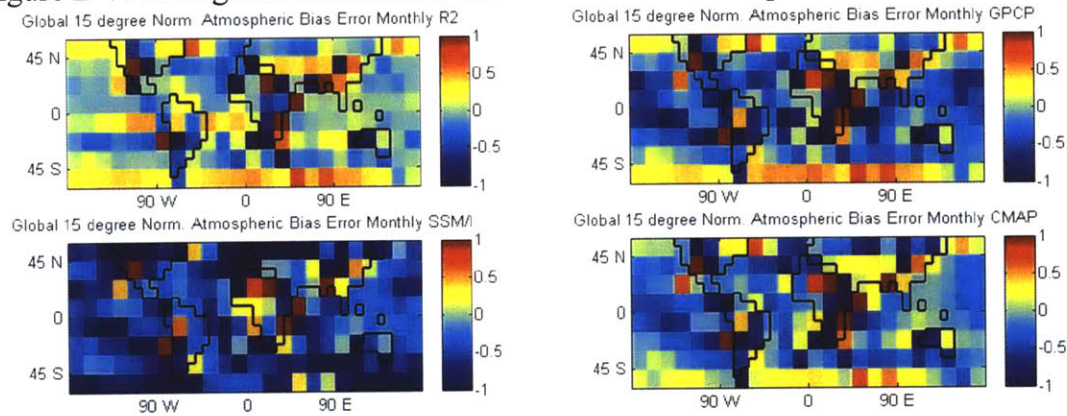


Figure B-8: 15 degree Normalized Bias Error Different Precipitation Global Monthly

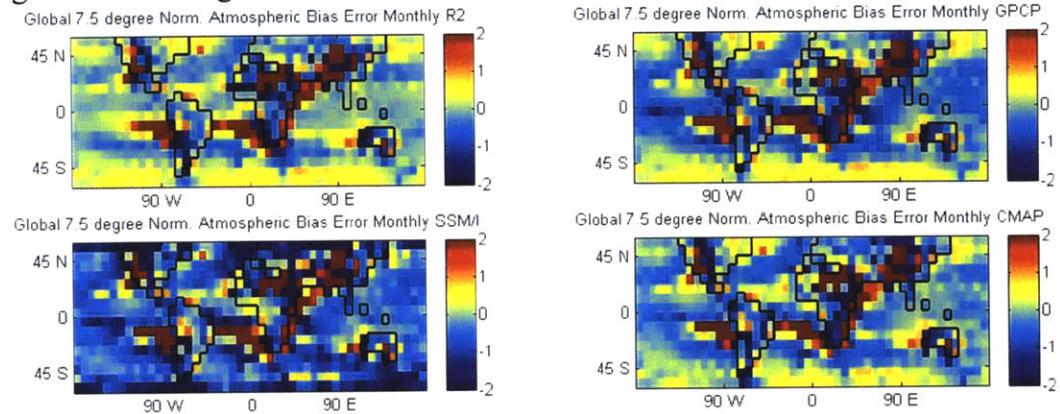


Figure B-9: 7.5 degree Normalized Bias Error Different Precipitation Global Monthly

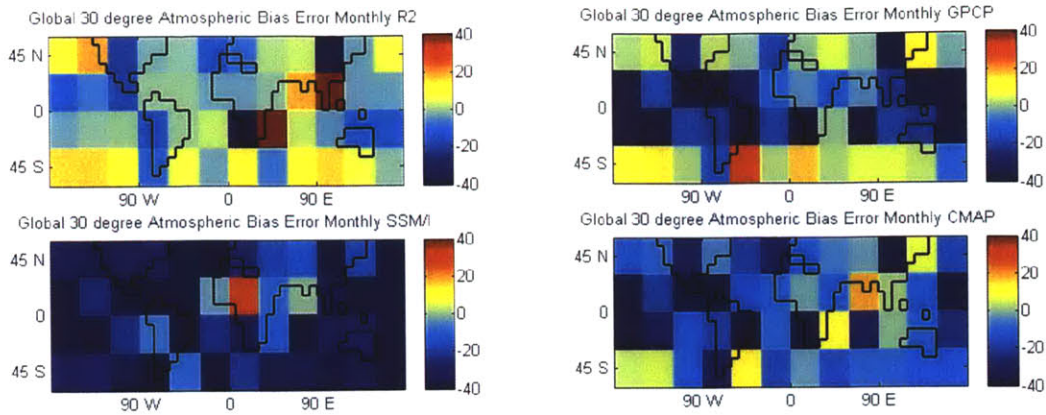


Figure B-10: 30 degree Bias Error Different Precipitation Global Monthly

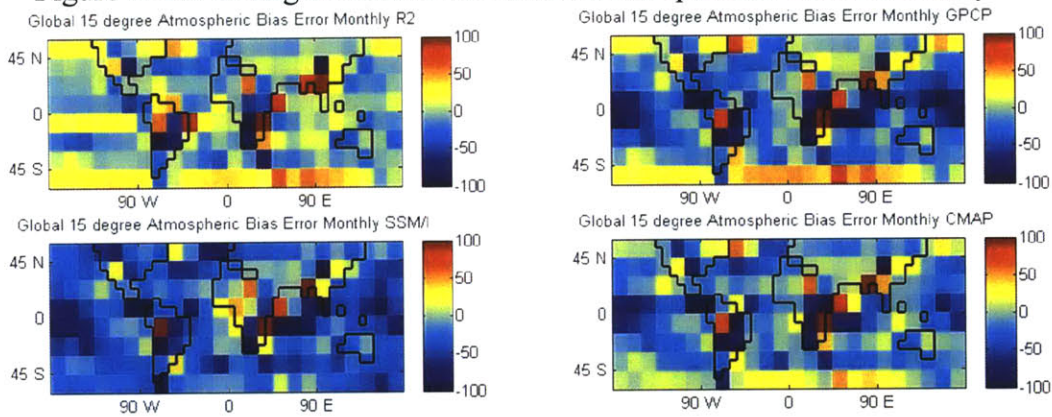


Figure B-11: 15 degree Bias Error Different Precipitation Global Monthly

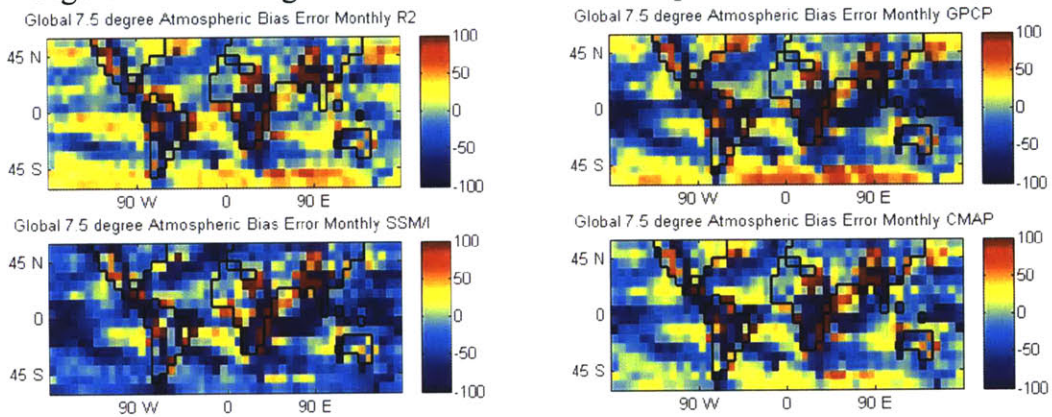


Figure B-12: 7.5 degree Bias Error Different Precipitation Global Monthly



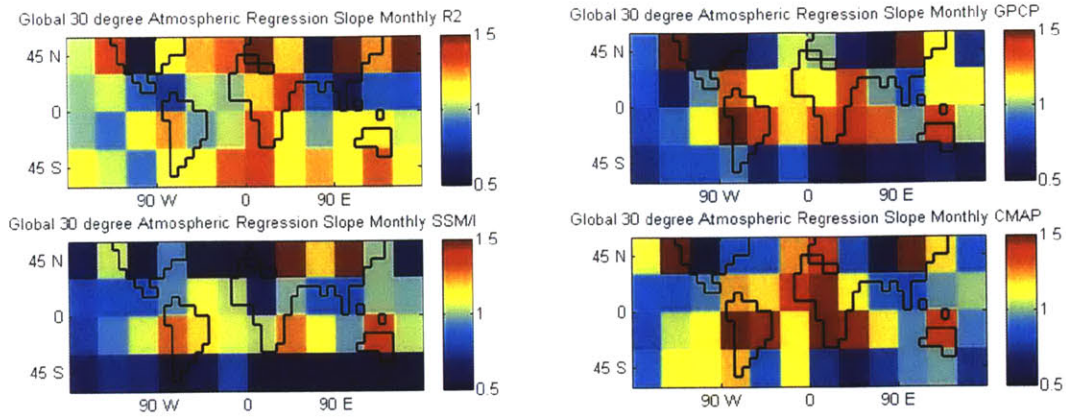


Figure B-13: 30 degree Regression Slope Different Precipitation Global Monthly

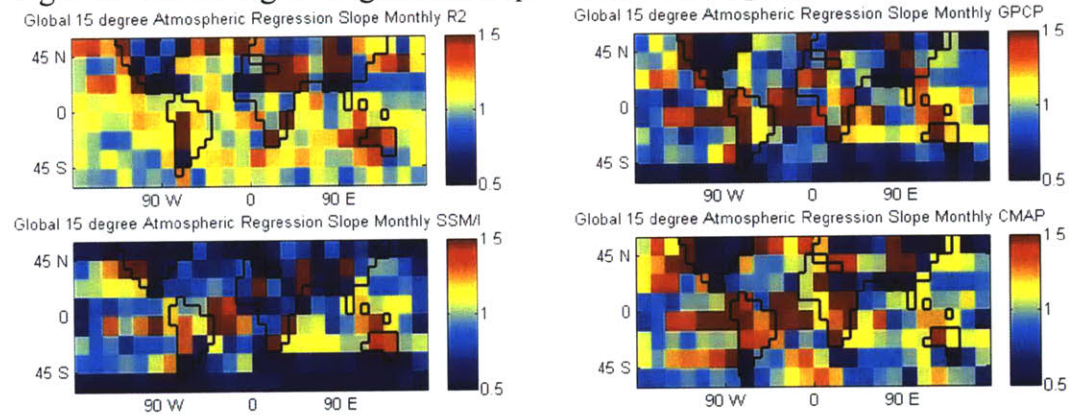


Figure B-14: 15 degree Regression Slope Different Precipitation Global Monthly

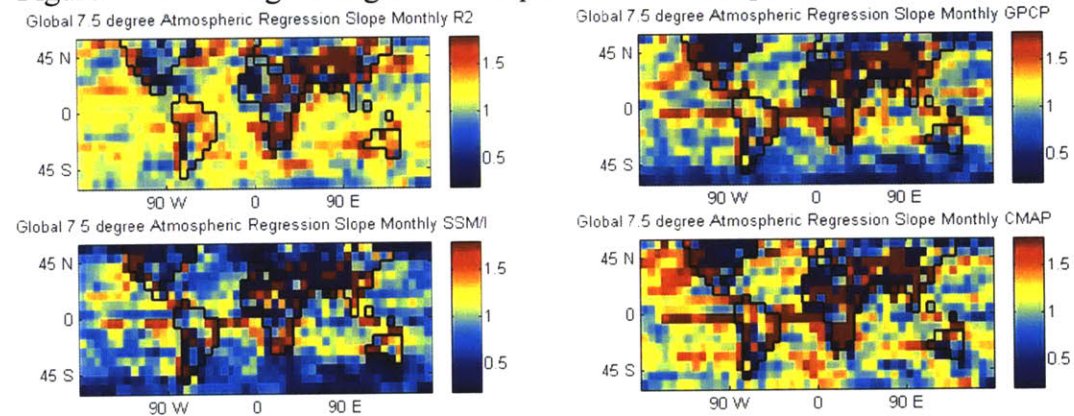


Figure B-15: 7.5 degree Regression Slope Different Precipitation Global Monthly

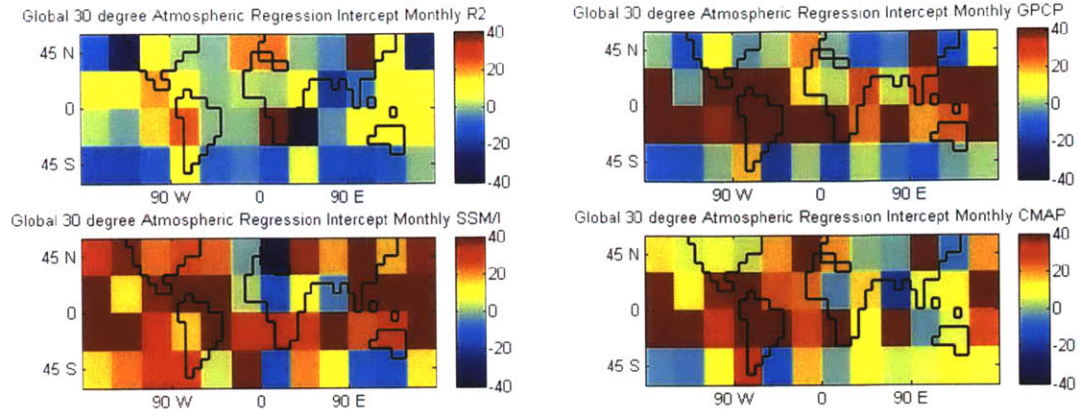


Figure B-16: 30 degree Regression Intercept Different Precipitation Global Monthly

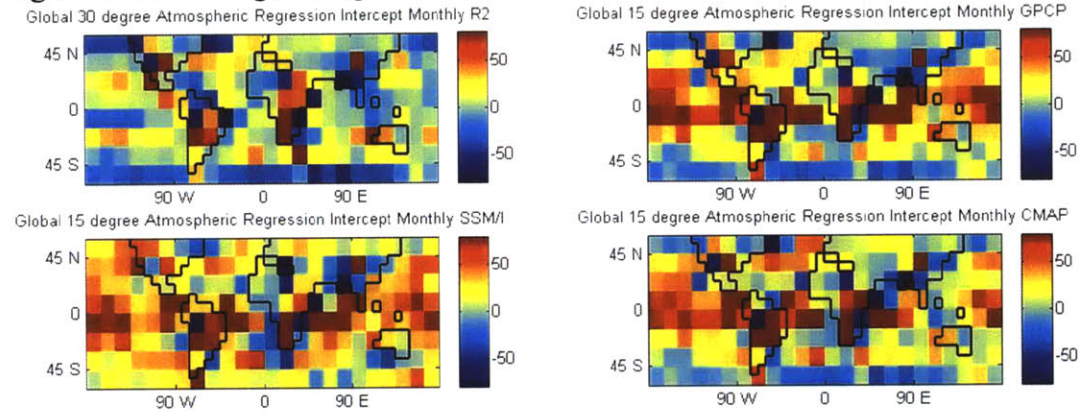


Figure B-17: 15 degree Regression Intercept Different Precipitation Global Monthly

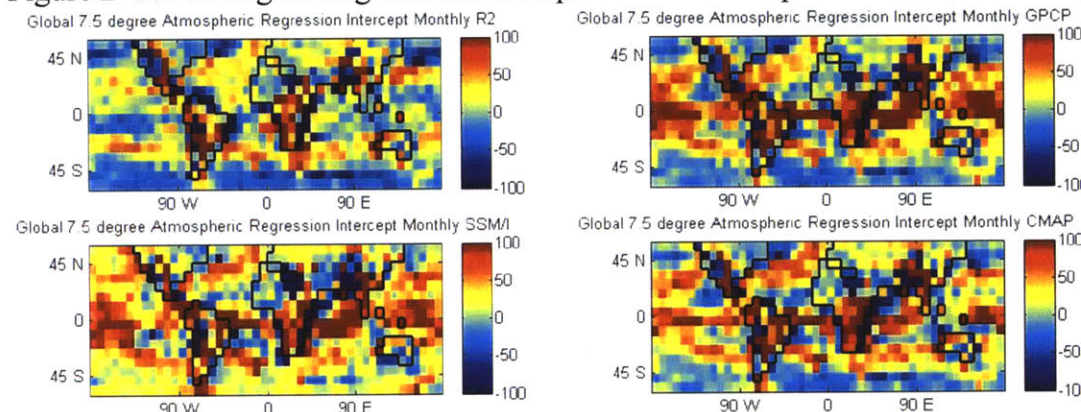


Figure B-18: 7.5 degree Regression Intercept Different Precipitation Global Monthly

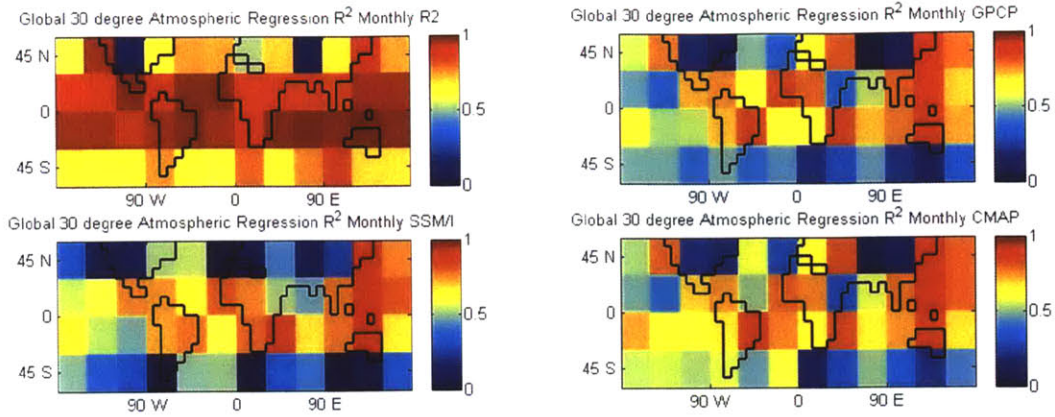


Figure B-19: 30 degree Regression  $R^2$  Different Precipitation Global Monthly

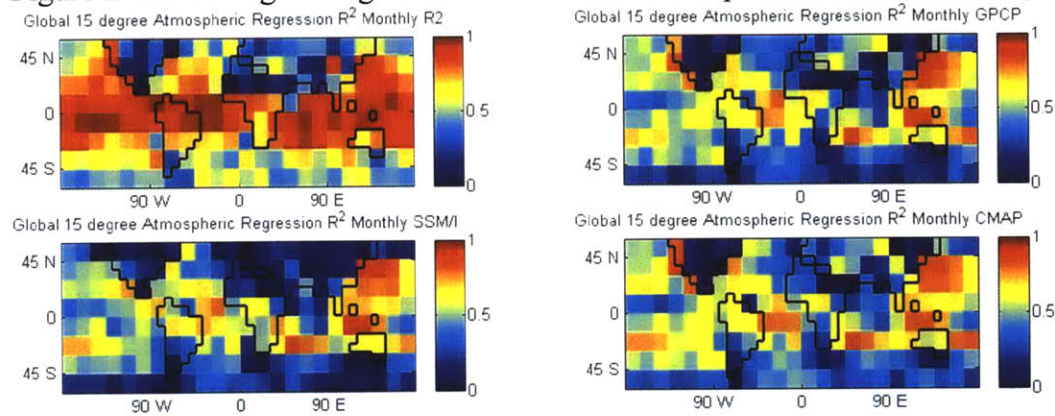


Figure B-20: 15 degree Regression  $R^2$  Different Precipitation Global Monthly

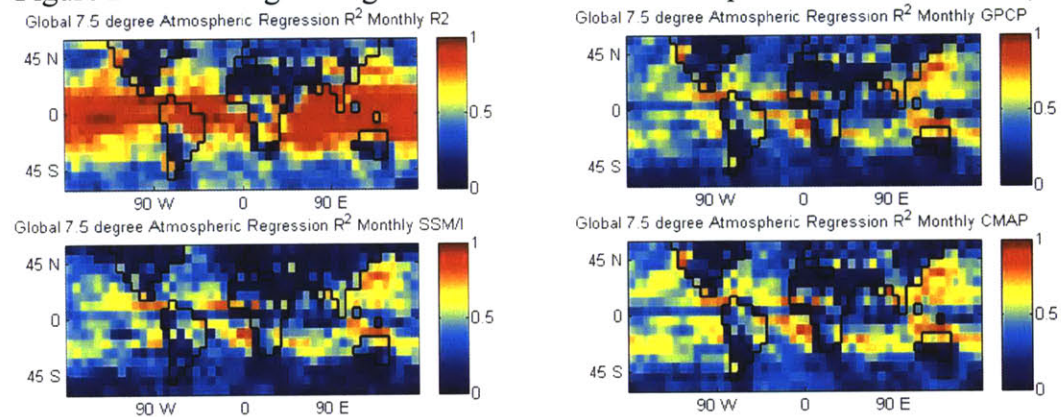


Figure B-21: 7.5 degree Regression  $R^2$  Different Precipitation Global Monthly

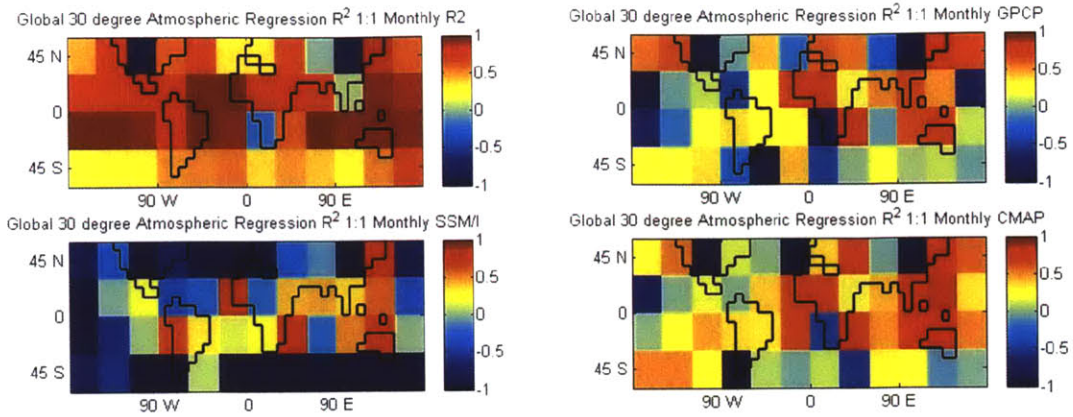


Figure B-22: 30 degree Regression  $R^2$  1:1 Different Precipitation Global Monthly

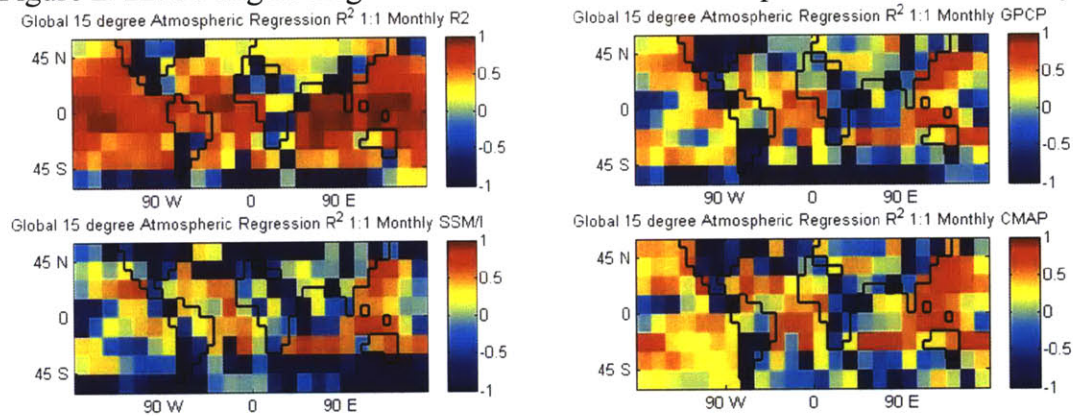


Figure B-23: 15 degree Regression  $R^2$  1:1 Different Precipitation Global Monthly

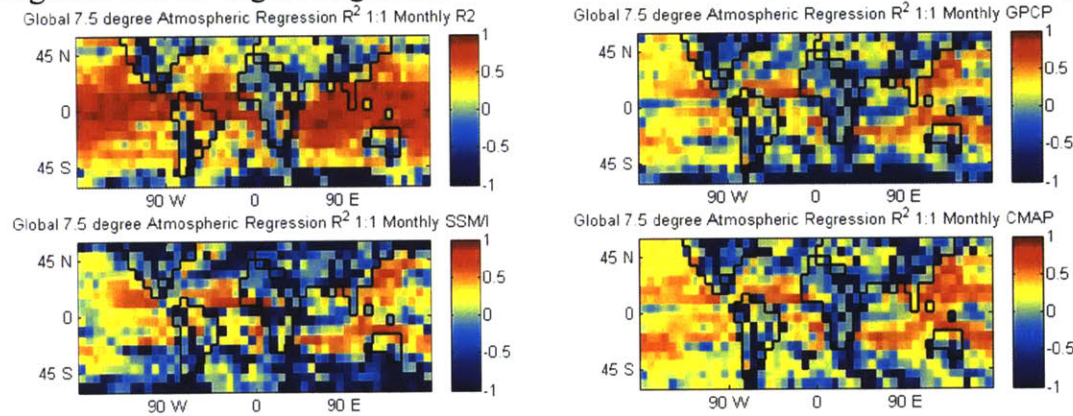


Figure B-24: 7.5 degree Regression  $R^2$  1:1 Different Precipitation Global Monthly

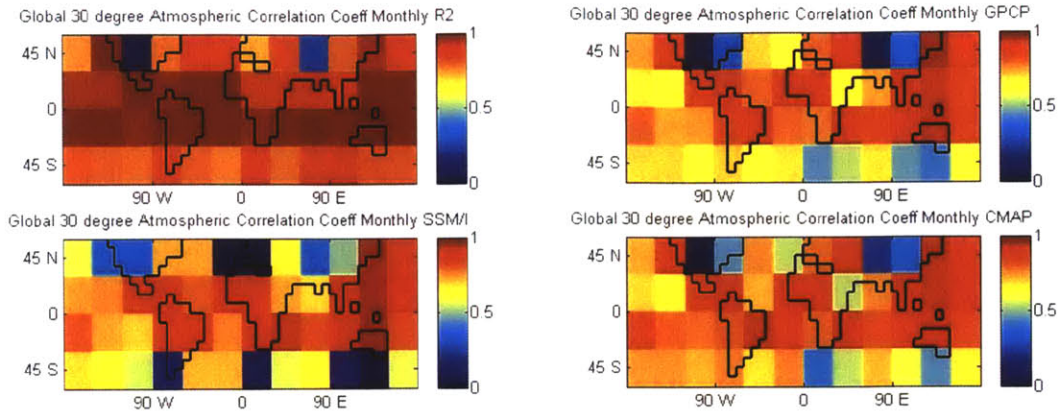


Figure B-25: 30 degree Correlation Coefficient Different Precipitation Global Monthly

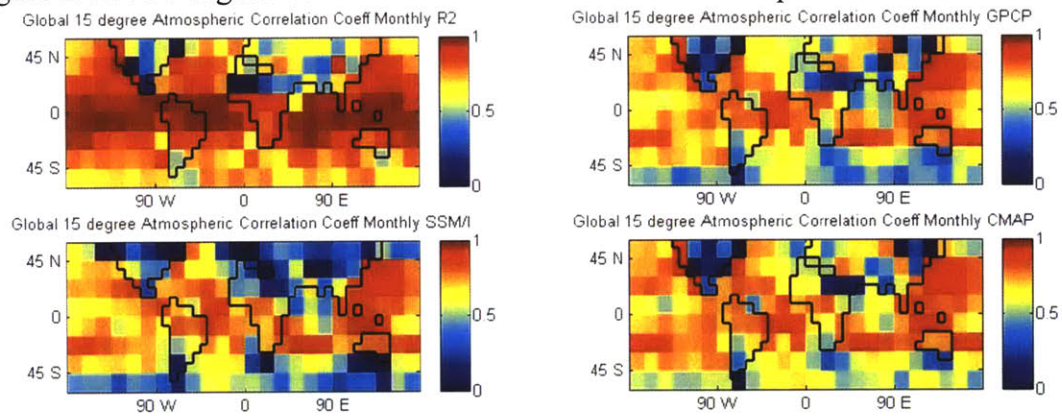


Figure B-26: 15 degree Correlation Coefficient Different Precipitation Global Monthly

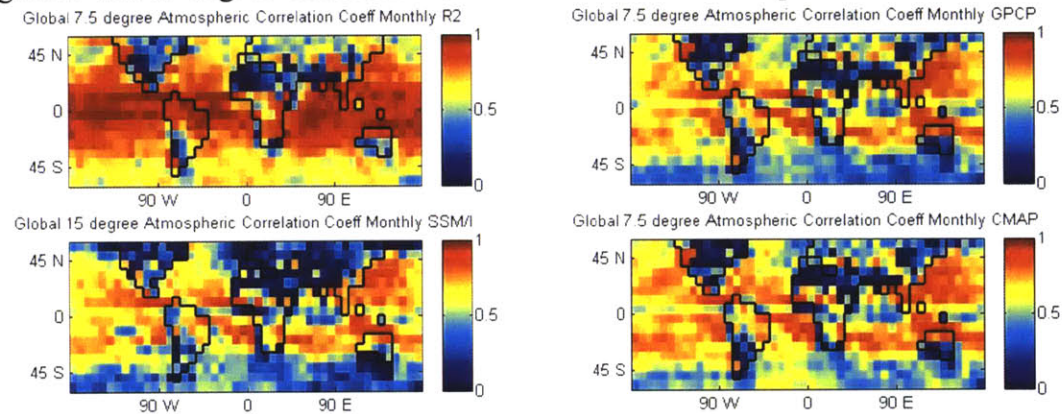


Figure B-27: 7.5 degree Correlation Coefficient Different Precipitation Global Monthly

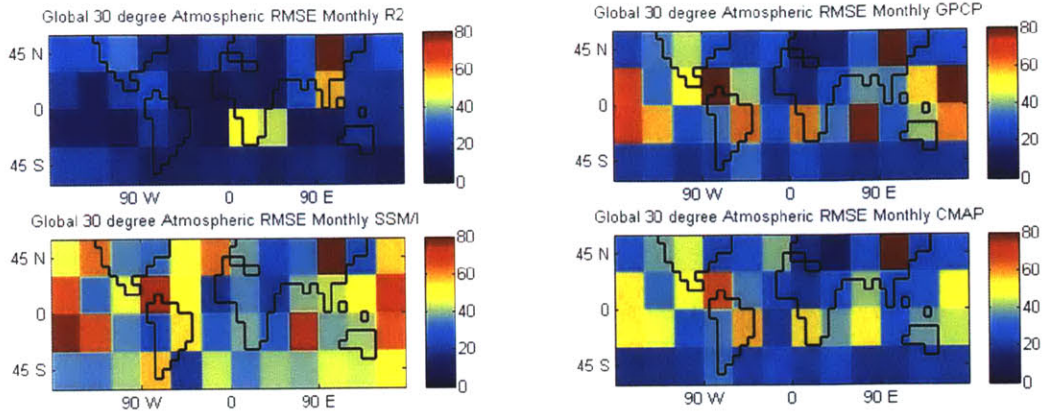


Figure B-28: 30 degree RMSE Different Precipitation Global Monthly

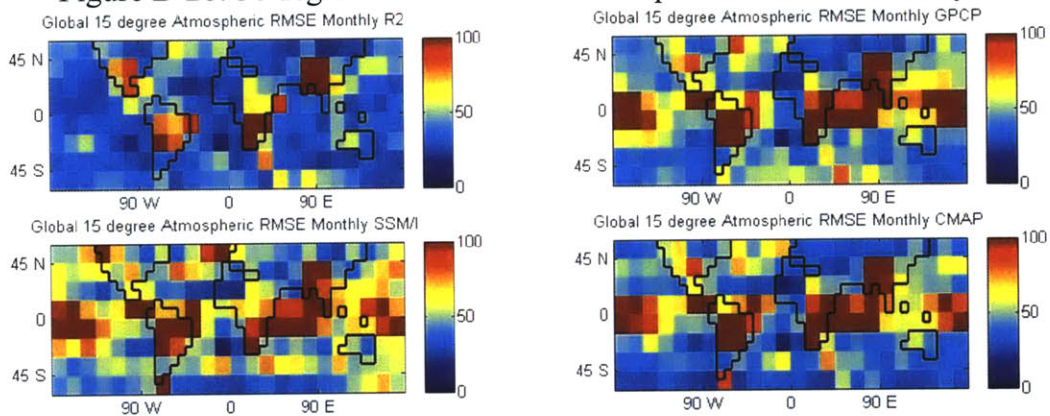


Figure B-29: 15 degree RMSE Different Precipitation Global Monthly

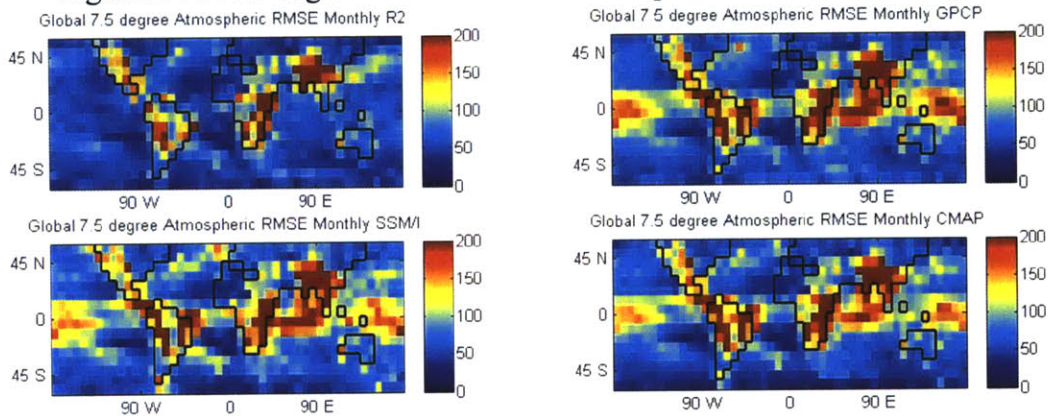


Figure B-30: 7.5 degree RMSE Different Precipitation Global Monthly

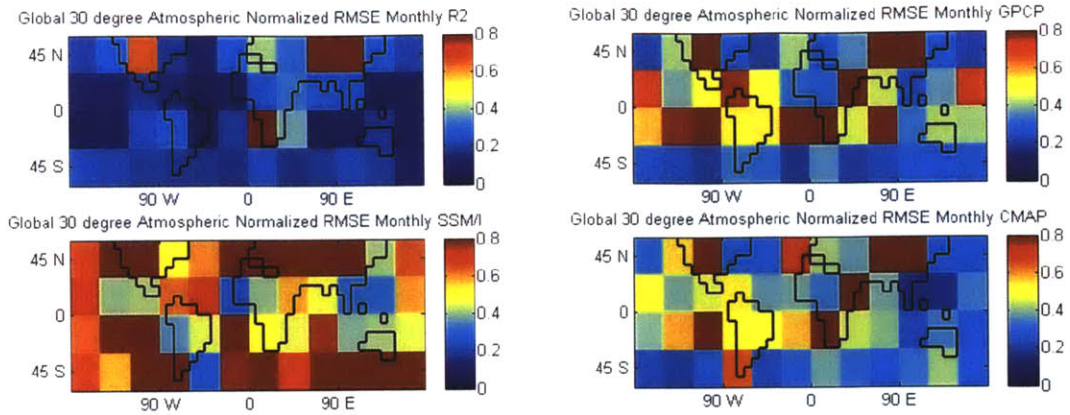


Figure B-31: 30 degree Normalized RMSE Different Precipitation Global Monthly

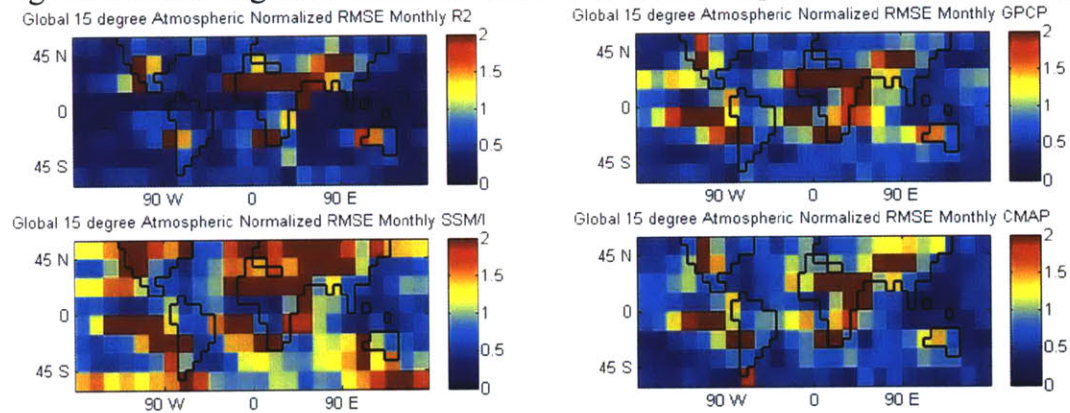


Figure B-32: 15 degree Normalized RMSE Different Precipitation Global Monthly

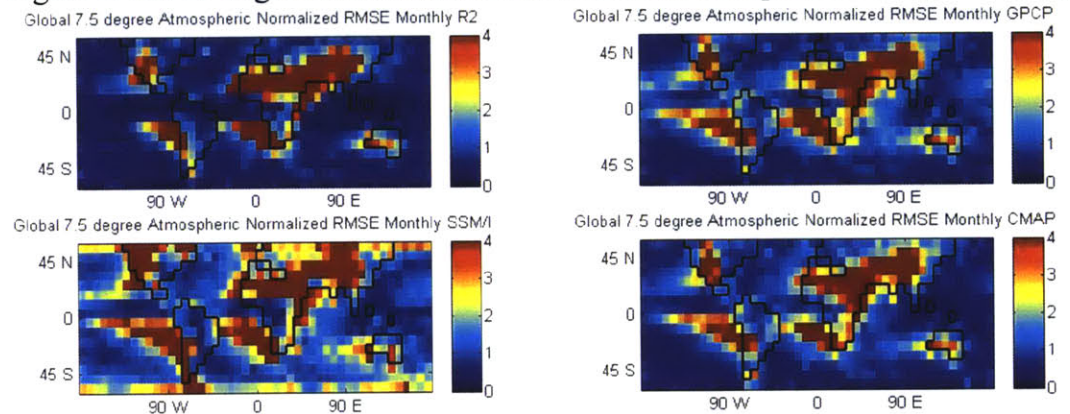


Figure B-33: 7.5 degree Normalized RMSE Different Precipitation Global Monthly

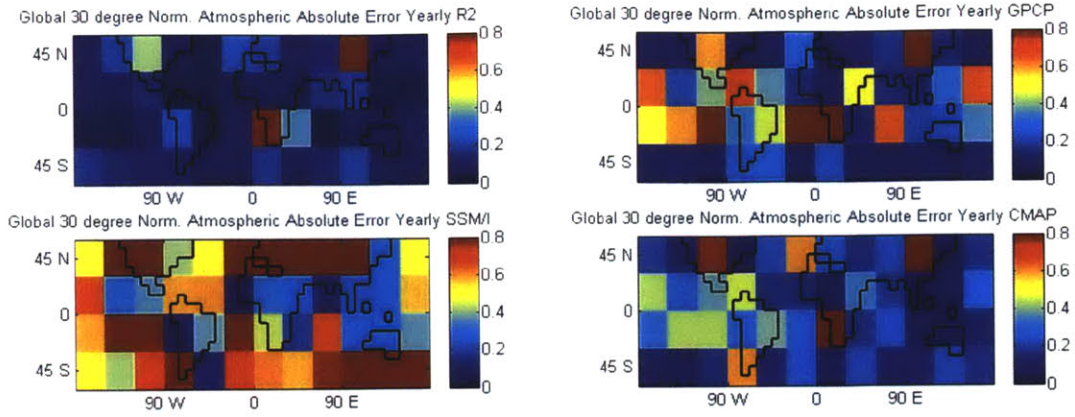


Figure B-34: 30 degree Normalized Absolute Error Different Precipitation Global Yearly

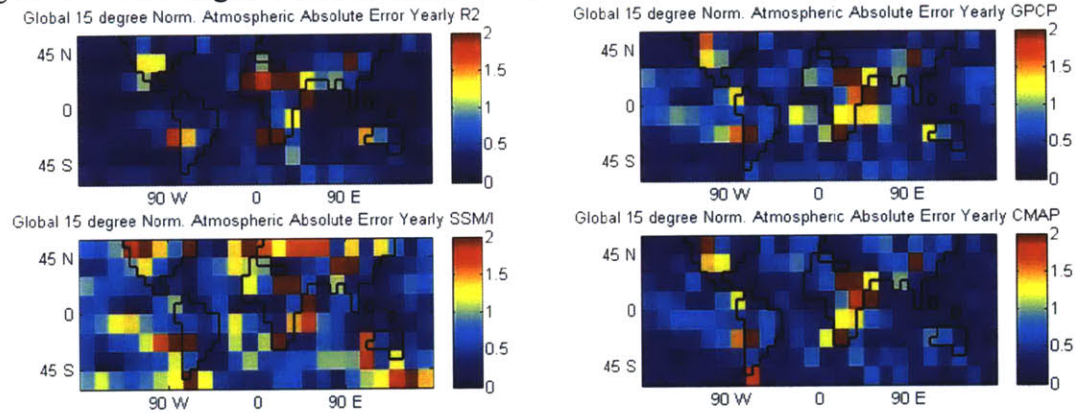


Figure B-35: 15 degree Normalized Absolute Error Different Precipitation Global Yearly

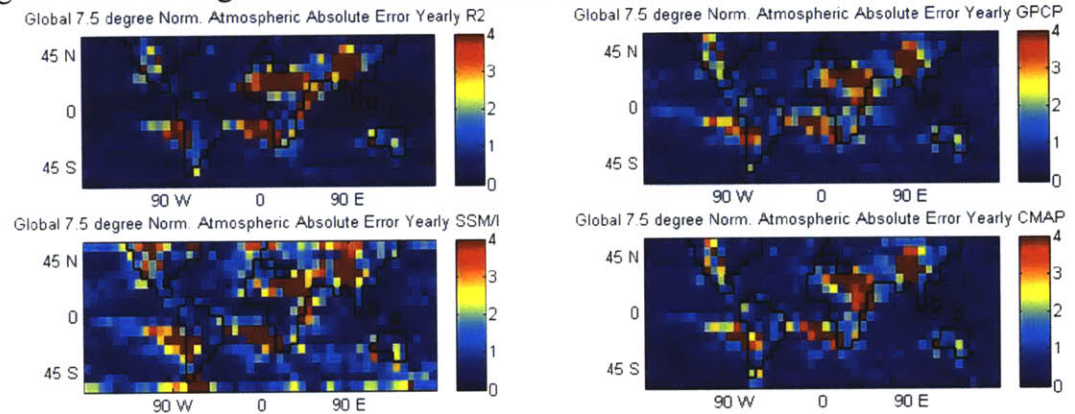


Figure B-36: 7.5 degree Normalized Absolute Error Different Precipitation Global Yearly



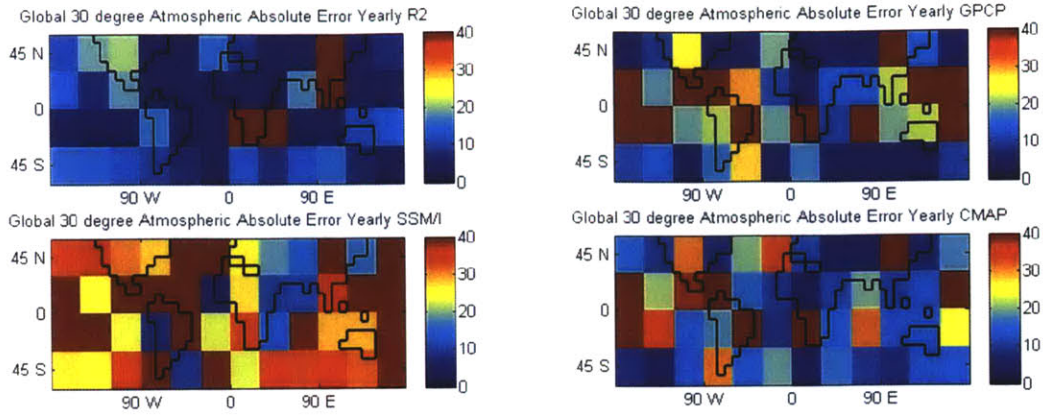


Figure B-37: 30 degree Absolute Error Different Precipitation Global Yearly

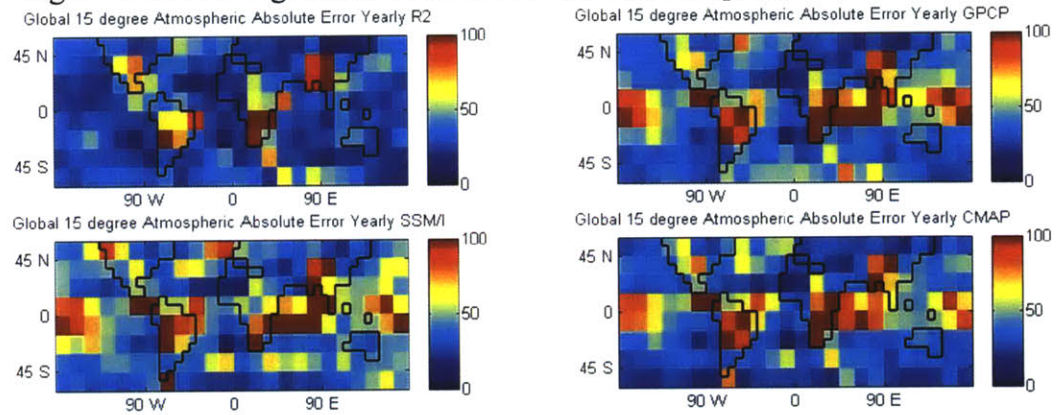


Figure B-38: 15 degree Absolute Error Different Precipitation Global Yearly

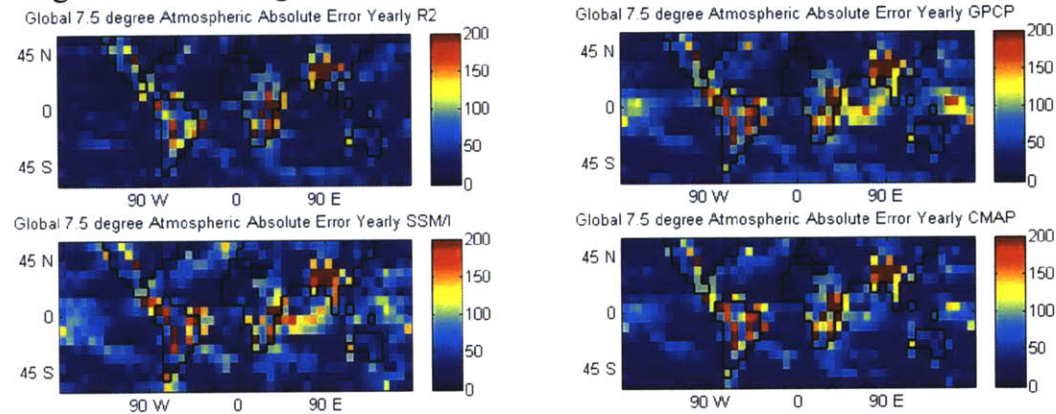


Figure B-39: 7.5 degree Absolute Error Different Precipitation Global Yearly

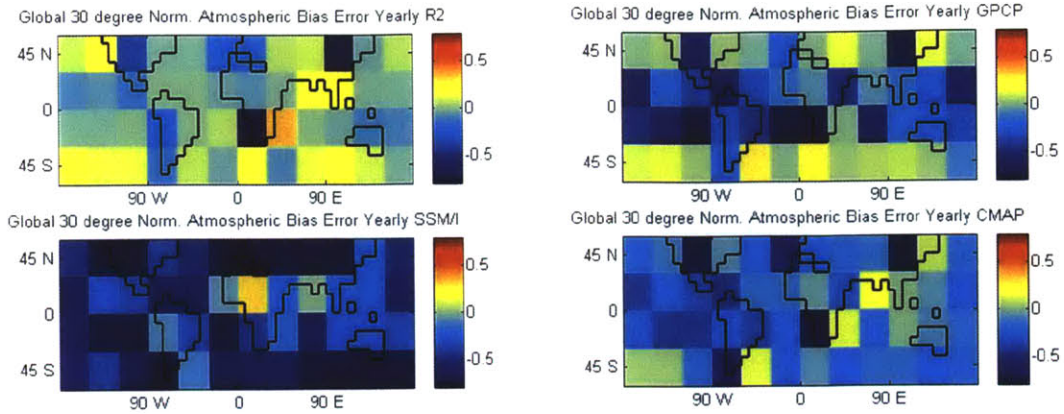


Figure B-40: 30 degree Normalized Bias Error Different Precipitation Global Yearly

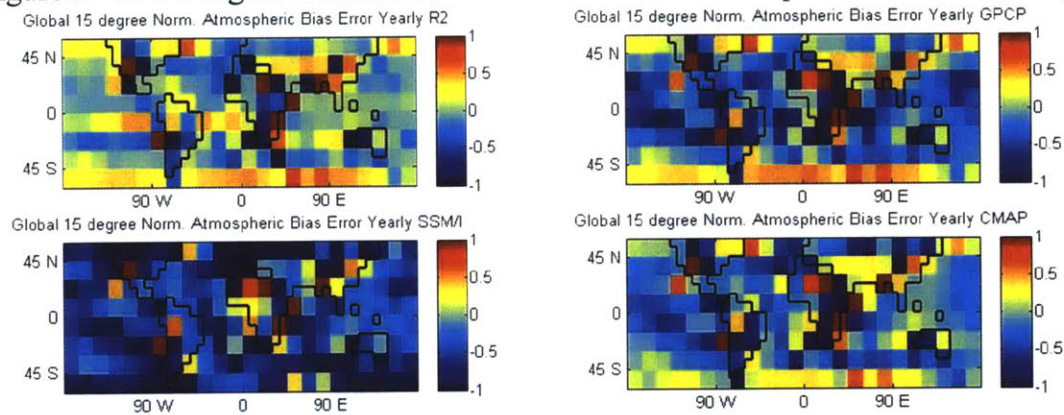


Figure B-41: 15 degree Normalized Bias Error Different Precipitation Global Yearly

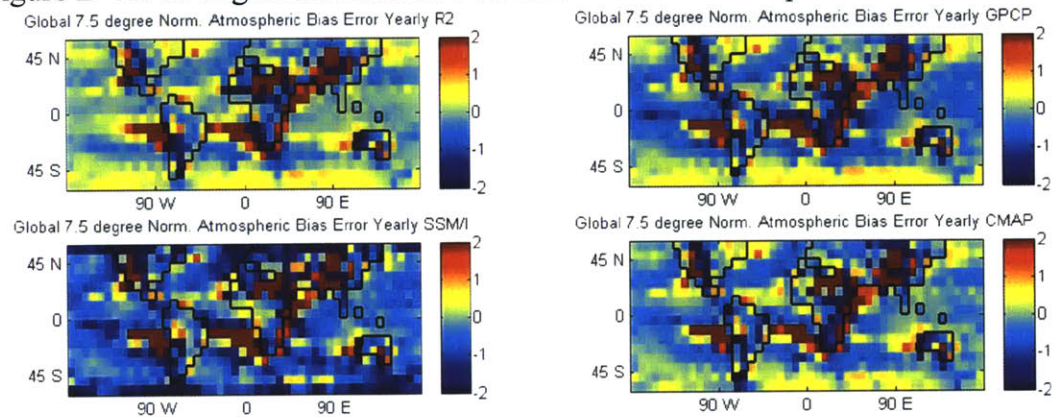


Figure B-42: 7.5 degree Normalized Bias Error Different Precipitation Global Yearly

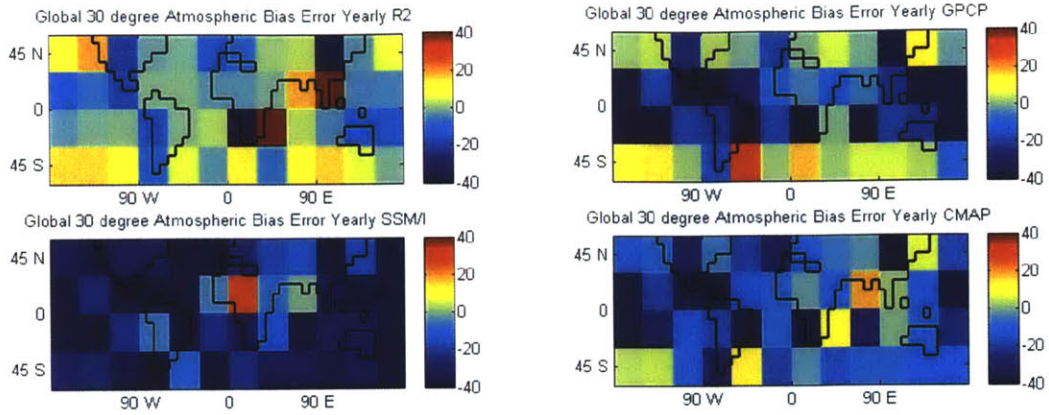


Figure B-43: 30 degree Bias Error Different Precipitation Global Yearly

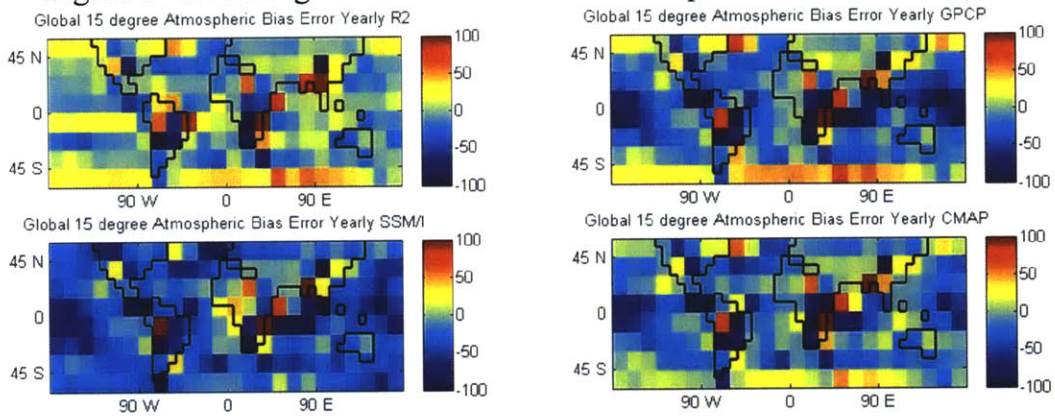


Figure B-44: 15 degree Bias Error Different Precipitation Global Yearly

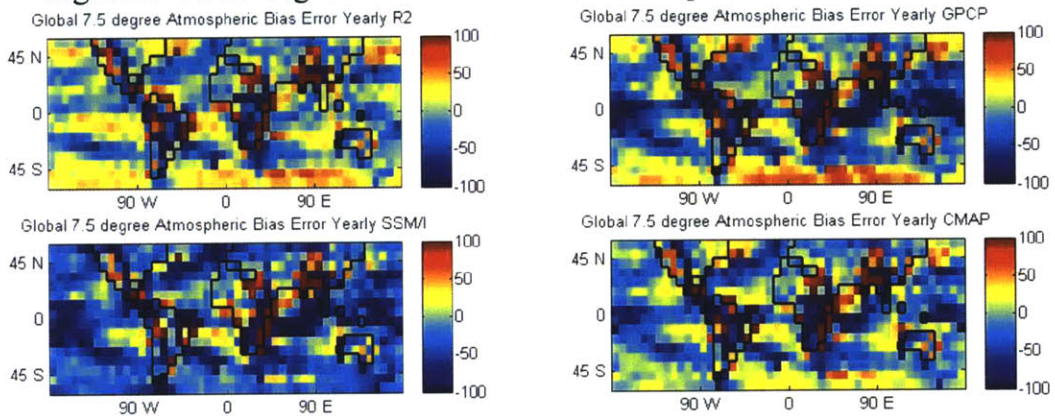


Figure B-45: 7.5 degree Bias Error Different Precipitation Global Yearly

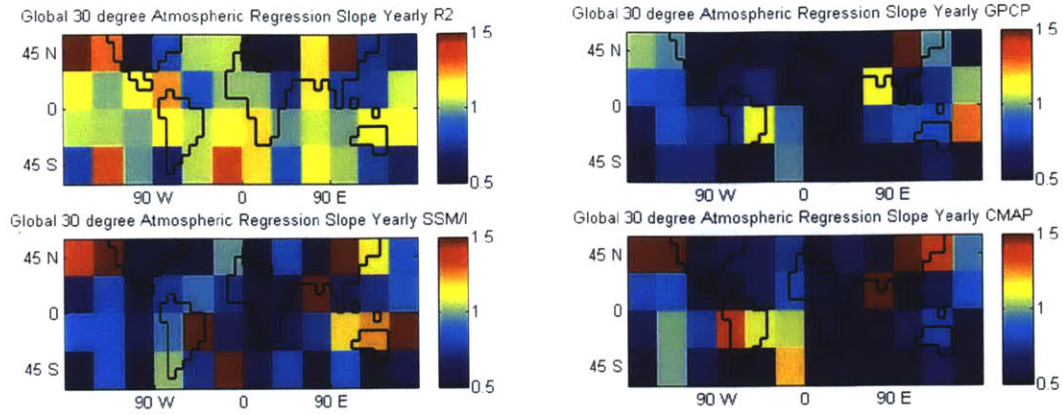


Figure B-46: 30 degree Regression Slope Different Precipitation Global Yearly

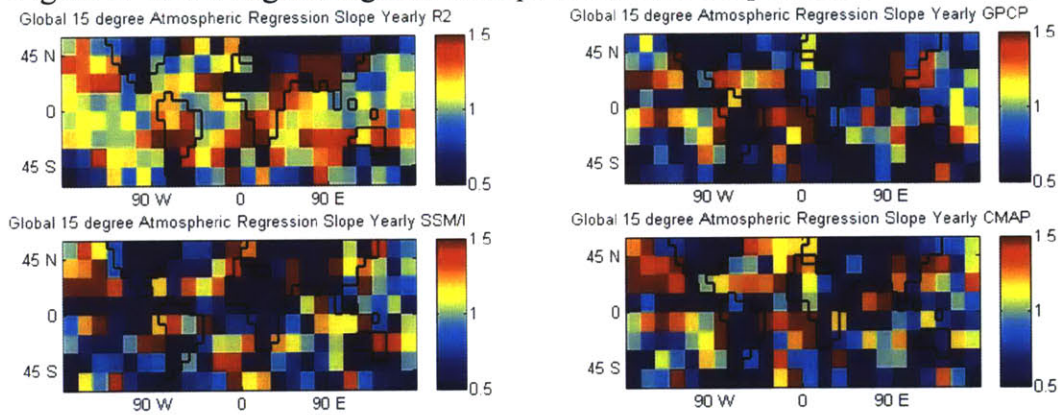


Figure B-47: 15 degree Regression Slope Different Precipitation Global Yearly

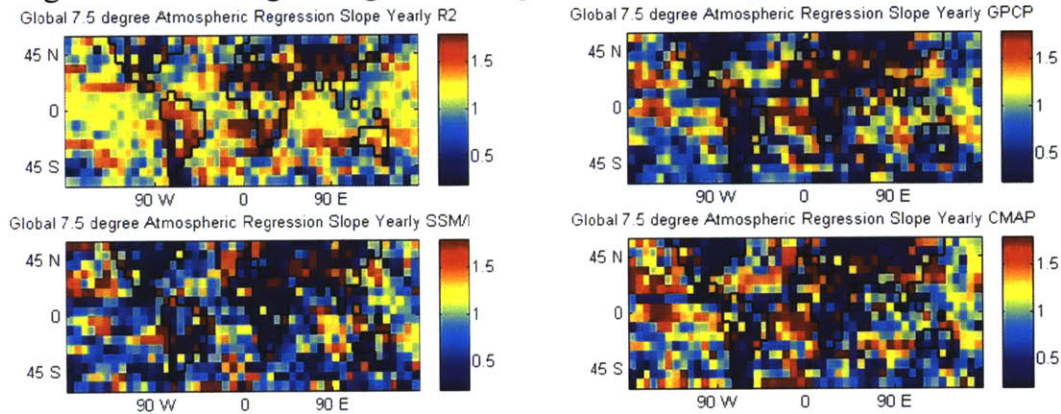


Figure B-48: 7.5 degree Regression Slope Different Precipitation Global Yearly

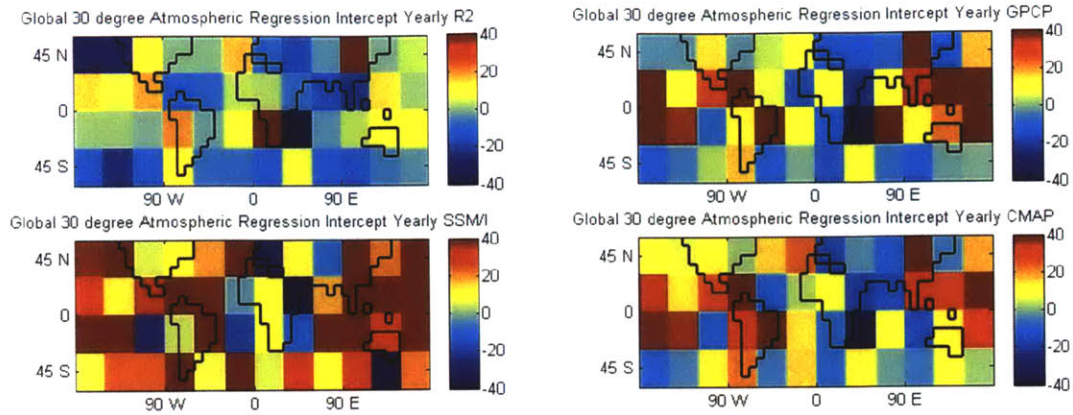


Figure B-49: 30 degree Regression Intercept Different Precipitation Global Yearly

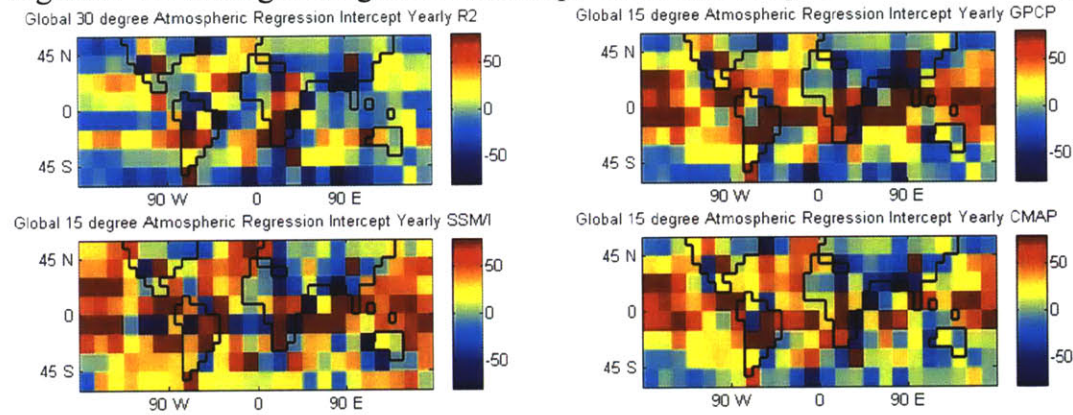


Figure B-50: 15 degree Regression Intercept Different Precipitation Global Yearly

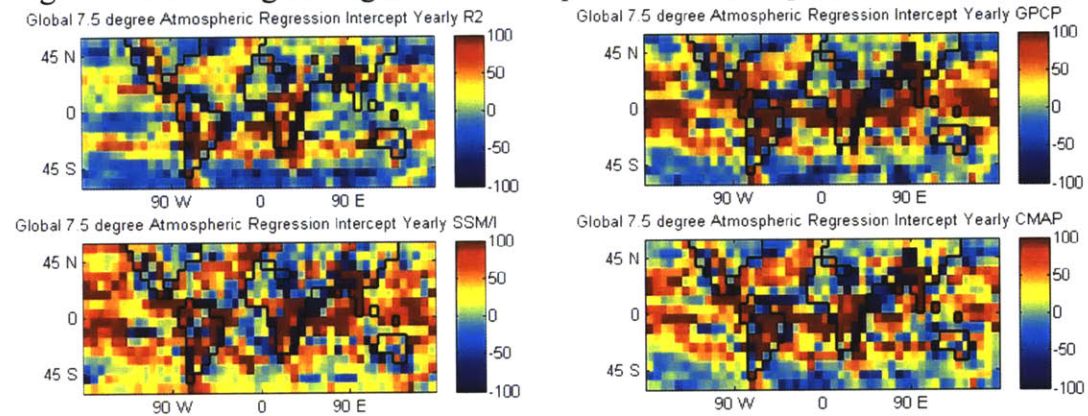


Figure B-51: 7.5 degree Regression Intercept Different Precipitation Global Yearly

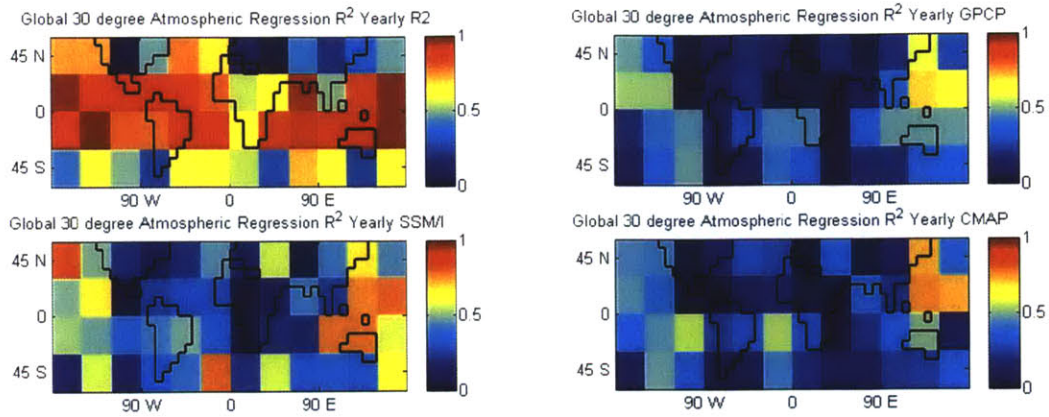


Figure B-52: 30 degree Regression  $R^2$  Different Precipitation Global Yearly

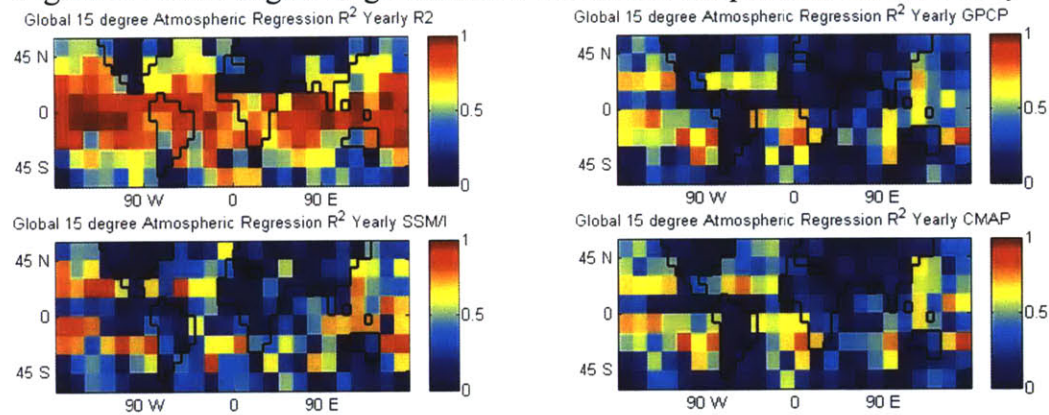


Figure B-53: 15 degree Regression  $R^2$  Different Precipitation Global Yearly

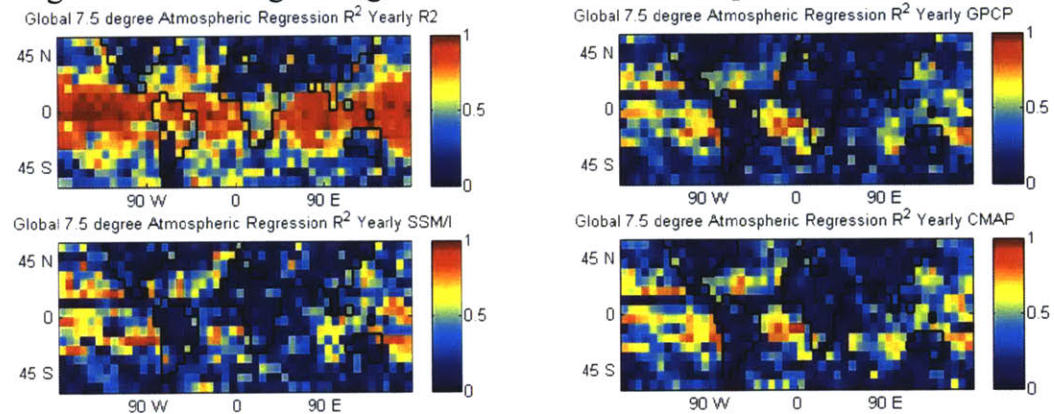


Figure B-54: 7.5 degree Regression  $R^2$  Different Precipitation Global Yearly

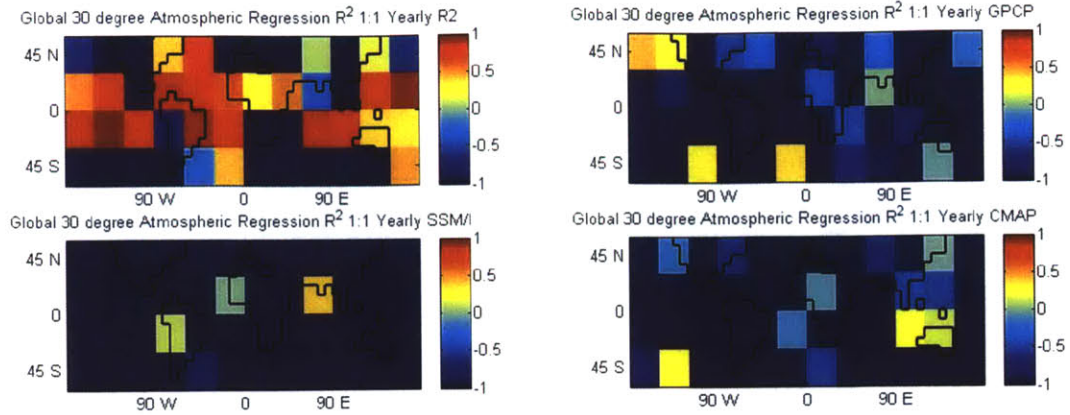


Figure B-55: 30 degree Regression  $R^2$  1:1 Different Precipitation Global Yearly

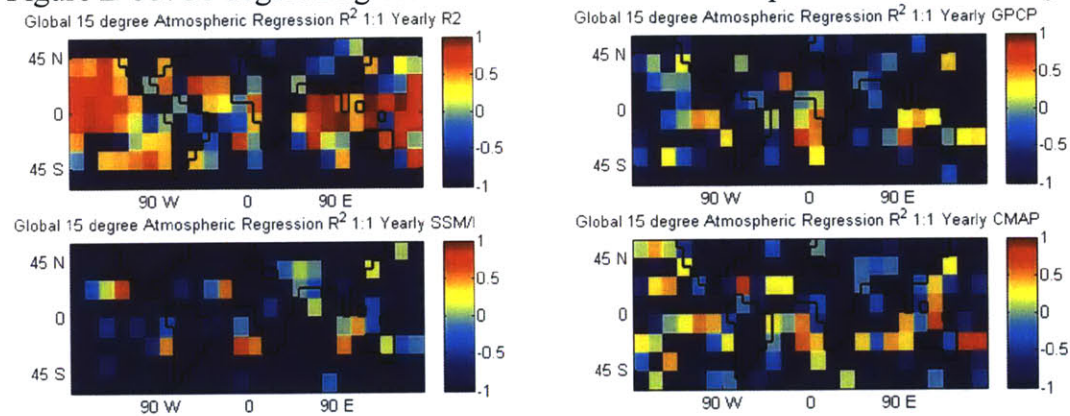


Figure B-56: 15 degree Regression  $R^2$  1:1 Different Precipitation Global Yearly

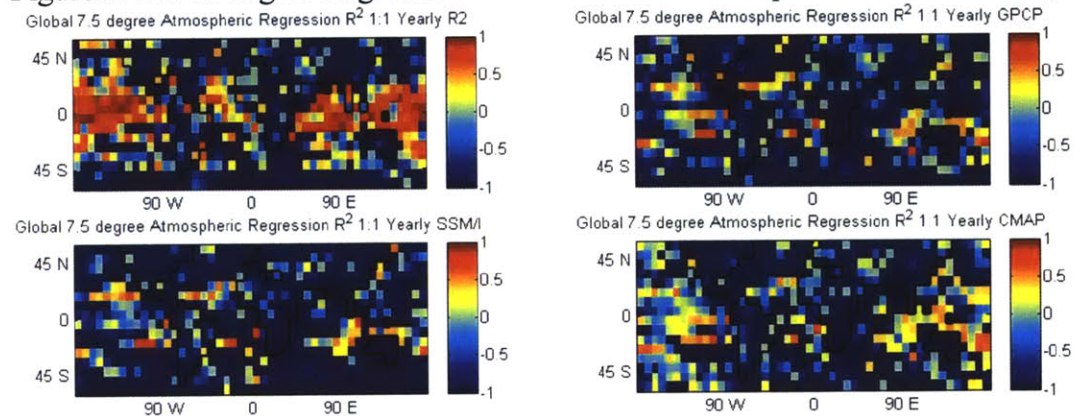


Figure B-57: 7.5 degree Regression  $R^2$  1:1 Different Precipitation Global Yearly

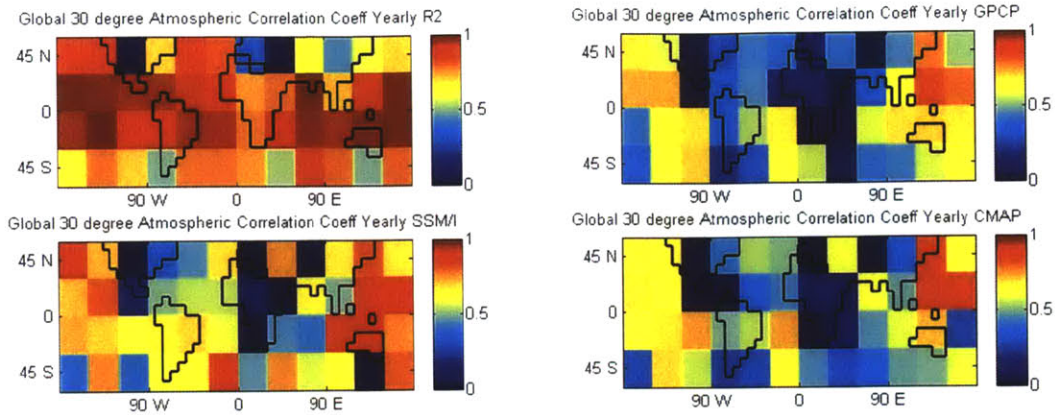


Figure B-58: 30 degree Correlation Coefficient Different Precipitation Global Yearly

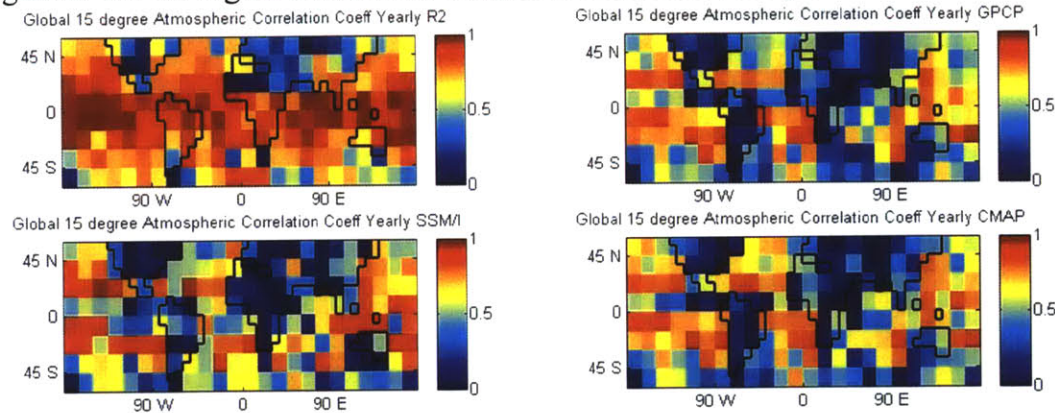


Figure B-59: 15 degree Correlation Coefficient Different Precipitation Global Yearly

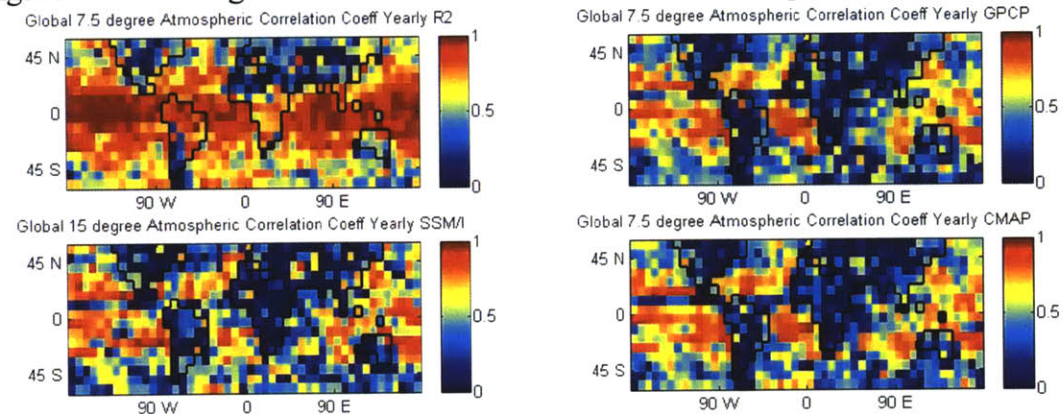


Figure B-60: 7.5 degree Correlation Coefficient Different Precipitation Global Yearly



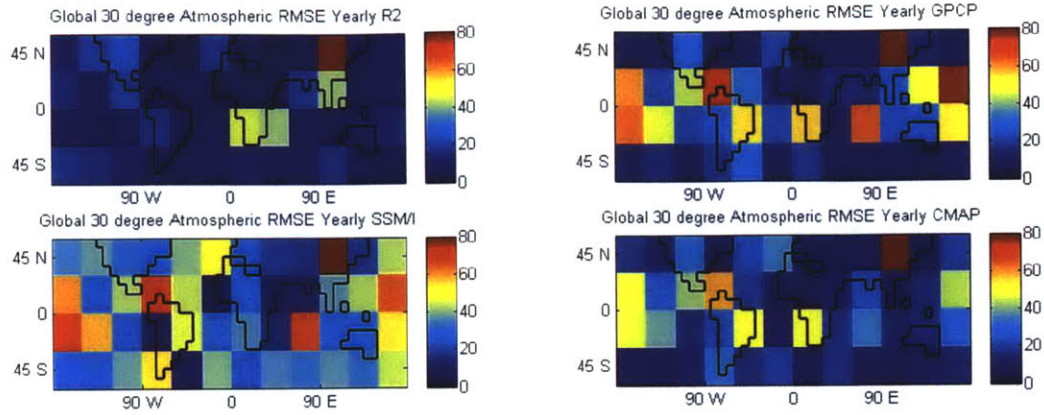


Figure B-61: 30 degree RMSE Different Precipitation Global Yearly

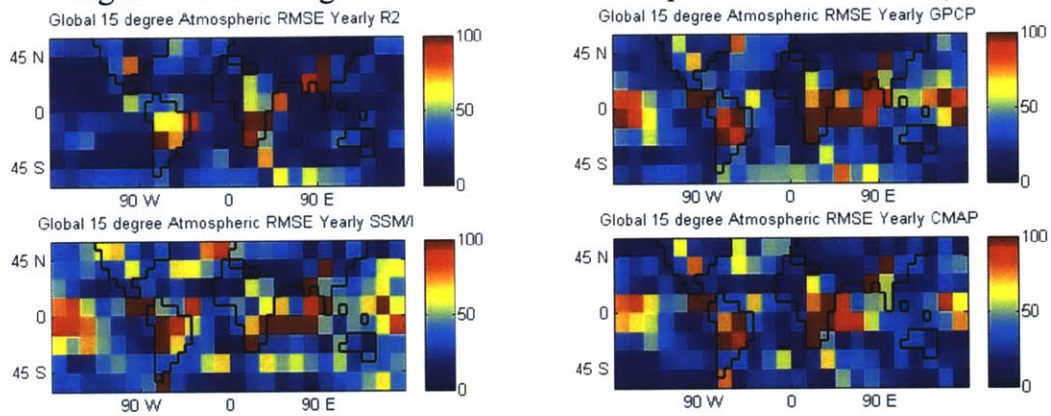


Figure B-62: 15 degree RMSE Different Precipitation Global Yearly

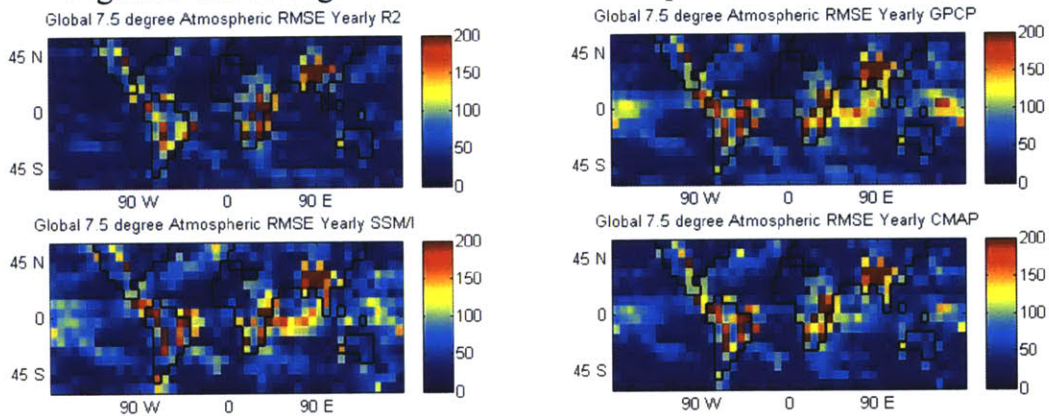


Figure B-63: 7.5 degree RMSE Different Precipitation Global Yearly

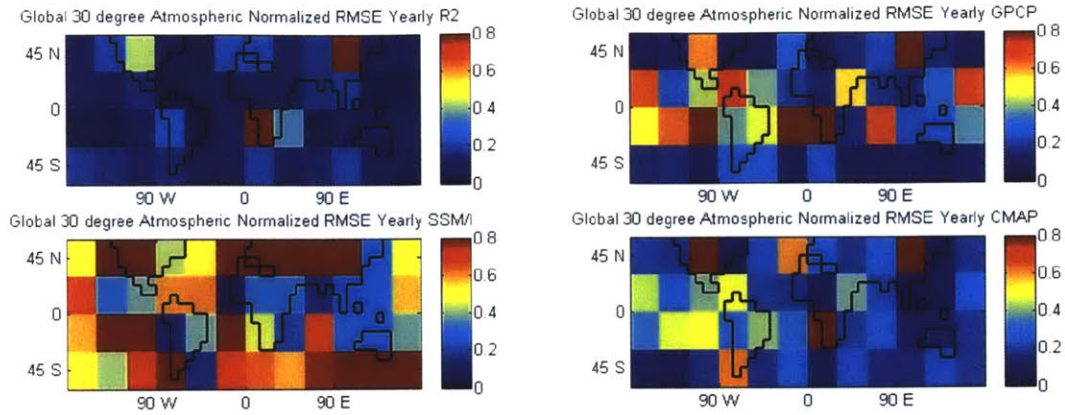


Figure B-64: 30 degree Normalized RMSE Different Precipitation Global Yearly

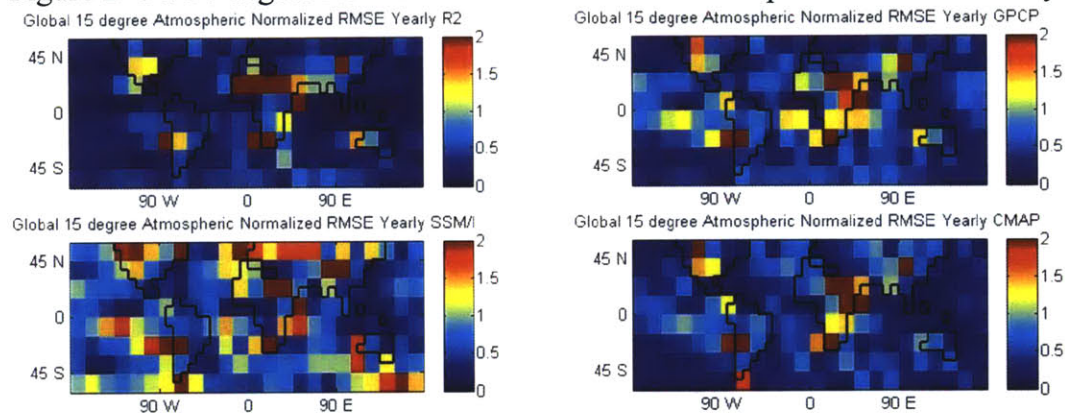


Figure B-65: 15 degree Normalized RMSE Different Precipitation Global Yearly

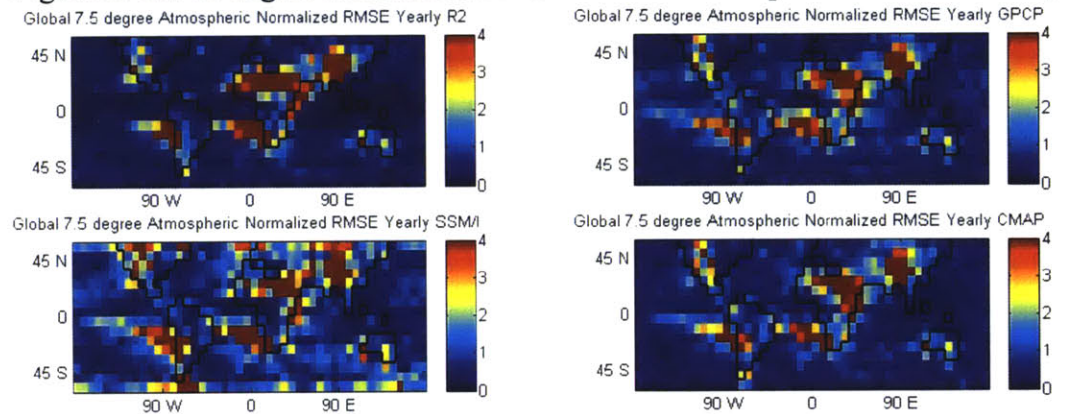


Figure B-66: 7.5 degree Normalized RMSE Different Precipitation Global Yearly

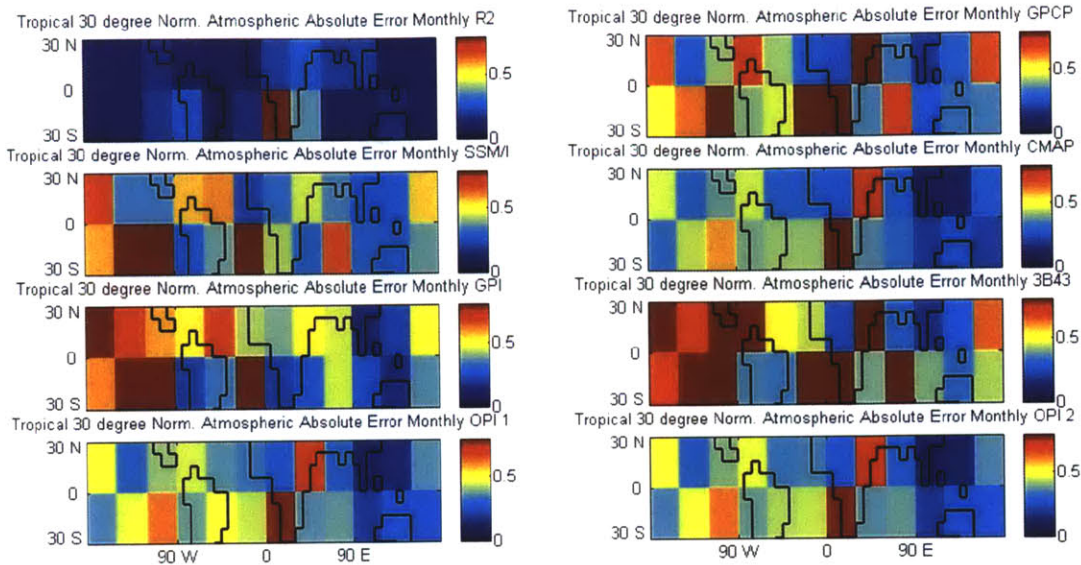


Figure B-67: 30 degree Normalized Absolute Error Different Precipitation Tropical Monthly

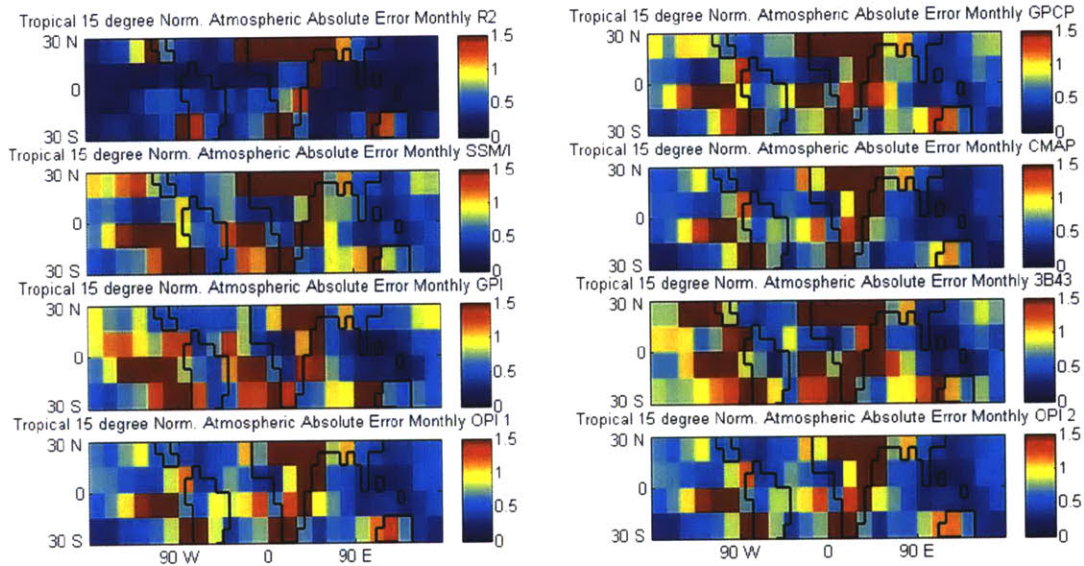


Figure B-68: 15 degree Normalized Absolute Error Different Precipitation Tropical Monthly

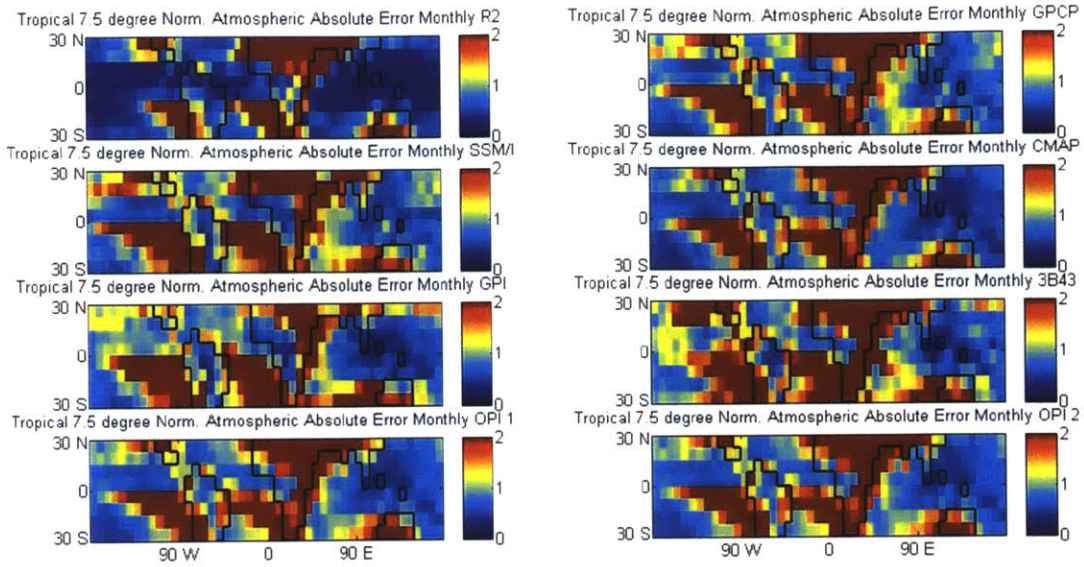


Figure B-69: 7.5 degree Normalized Absolute Error Different Precipitation Tropical Monthly

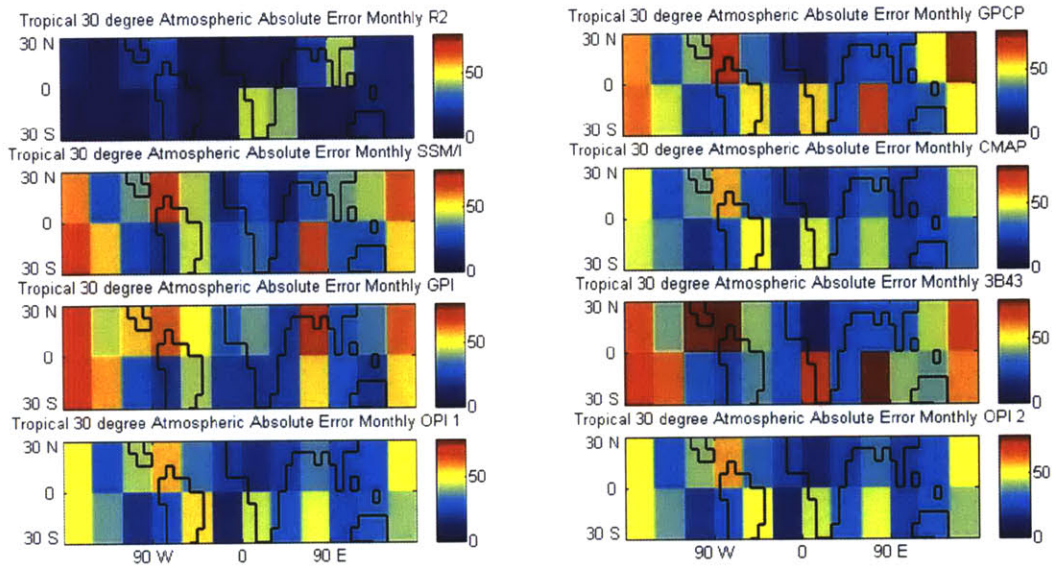


Figure B-70: 30 degree Absolute Error Different Precipitation Tropical Monthly

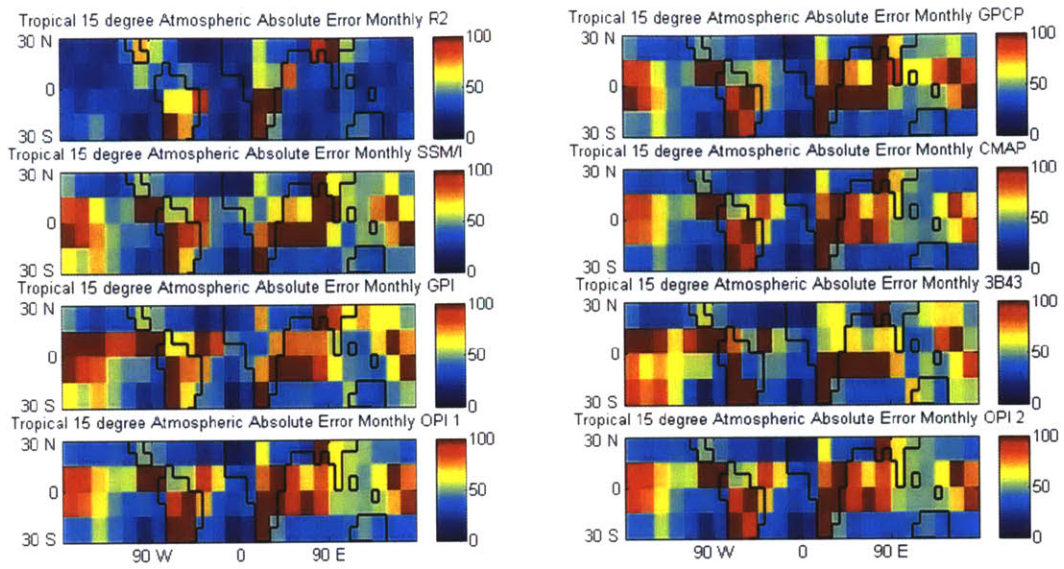


Figure B-71: 15 degree Absolute Error Different Precipitation Tropical Monthly

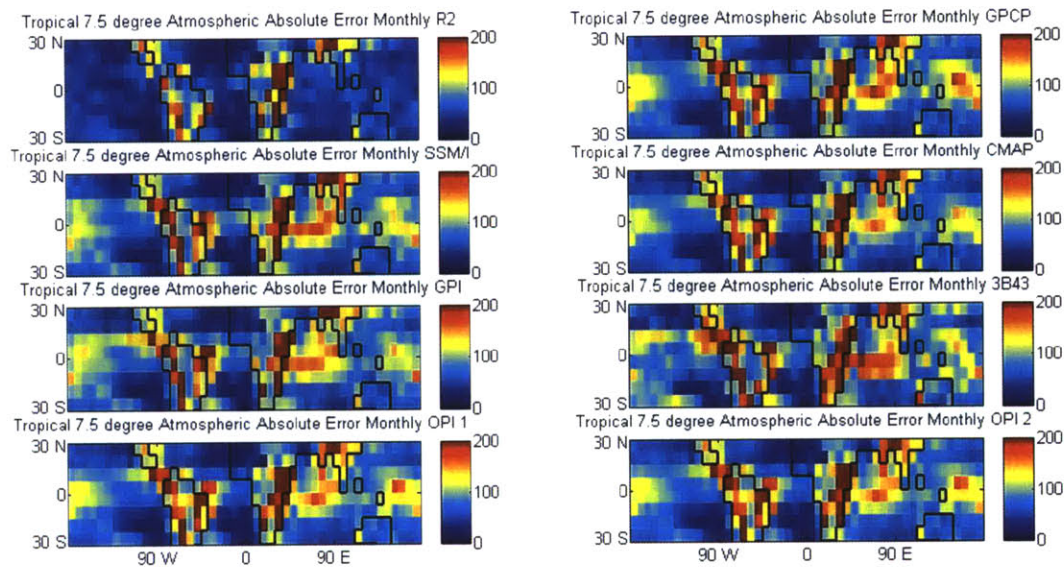


Figure B-72: 7.5 degree Absolute Error Different Precipitation Tropical Monthly

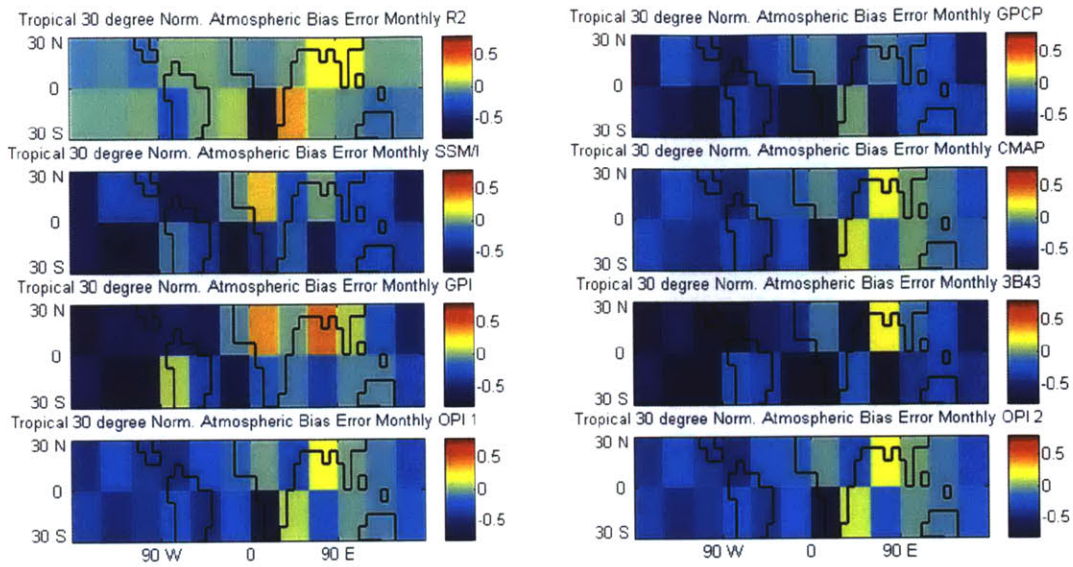


Figure B-73: 30 degree Normalized Bias Error Different Precipitation Tropical Monthly

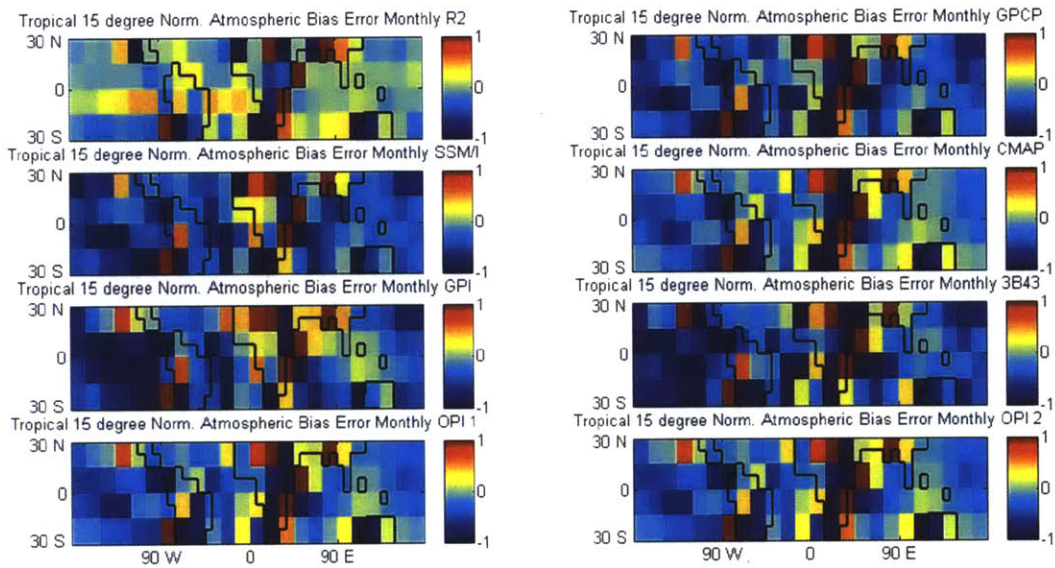


Figure B-73: 15 degree Normalized Bias Error Different Precipitation Tropical Monthly

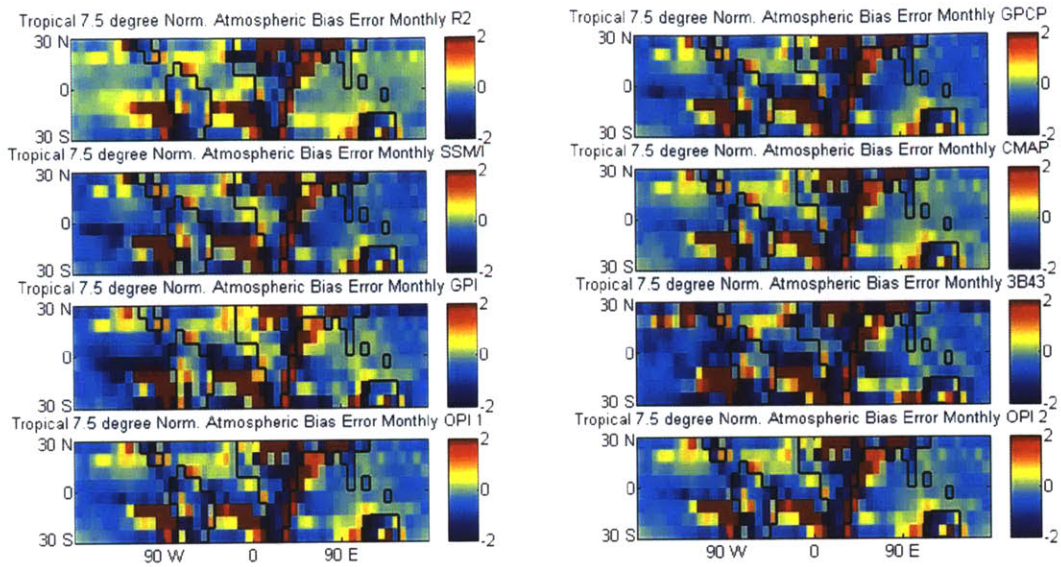


Figure B-75: 7.5 degree Normalized Bias Error Different Precipitation Tropical Monthly

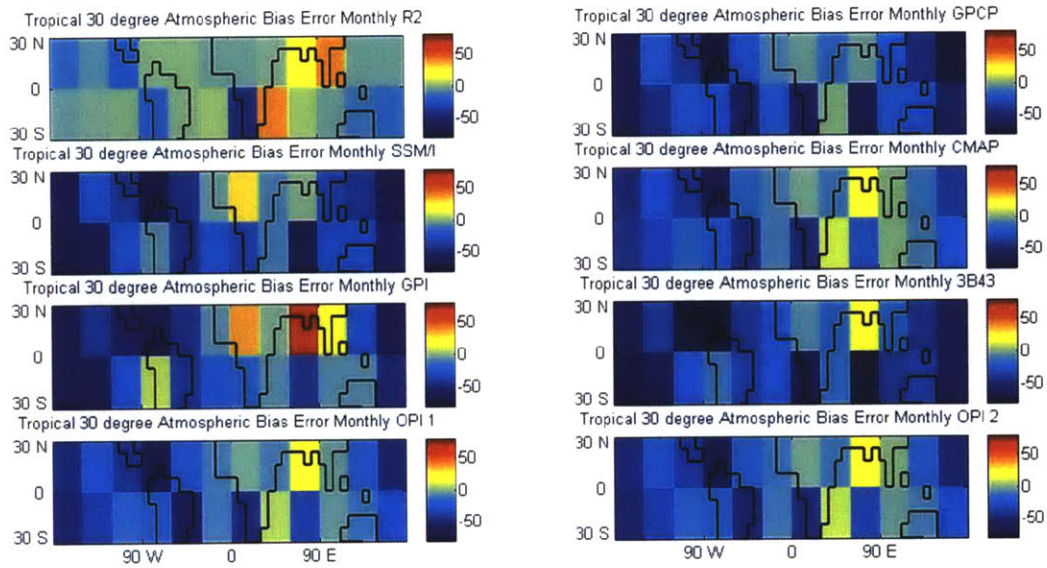


Figure B-76: 30 degree Bias Error Different Precipitation Tropical Monthly

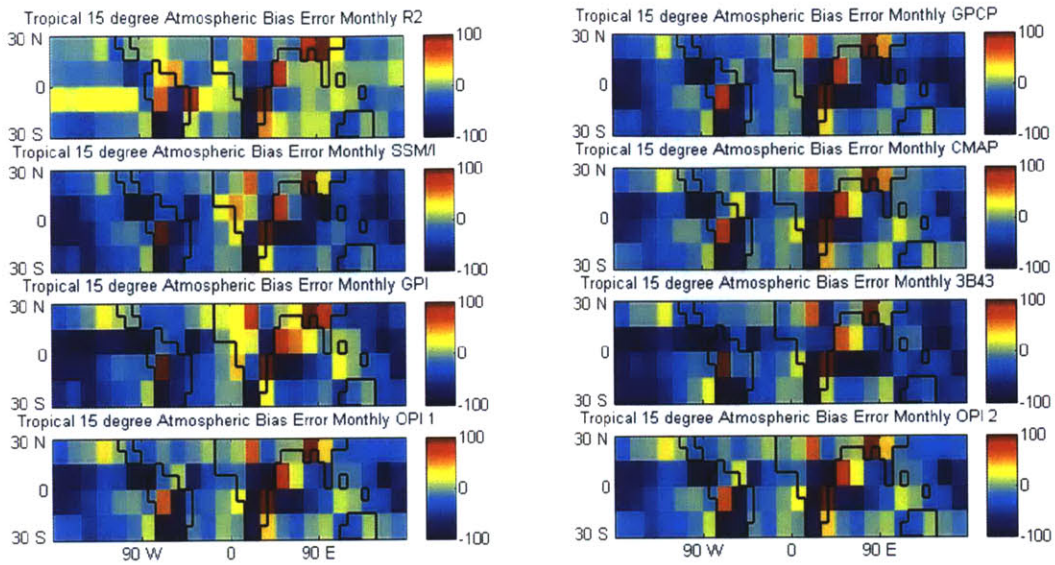


Figure B-77: 15 degree Bias Error Different Precipitation Tropical Monthly

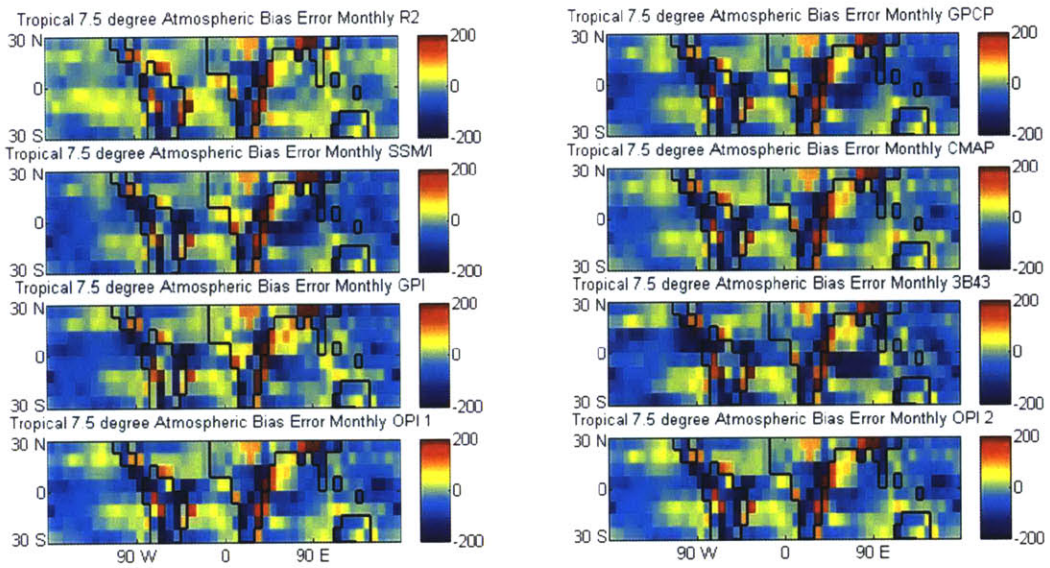


Figure B-78: 7.5 degree Bias Error Different Precipitation Tropical Monthly



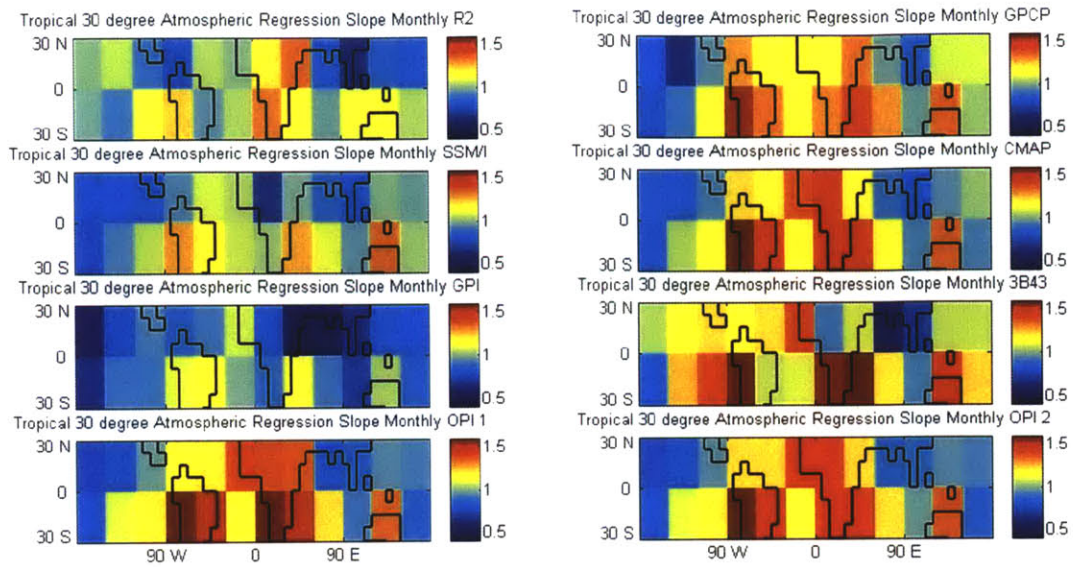


Figure B-79: 30 degree Regression Slope Different Precipitation Tropical Monthly

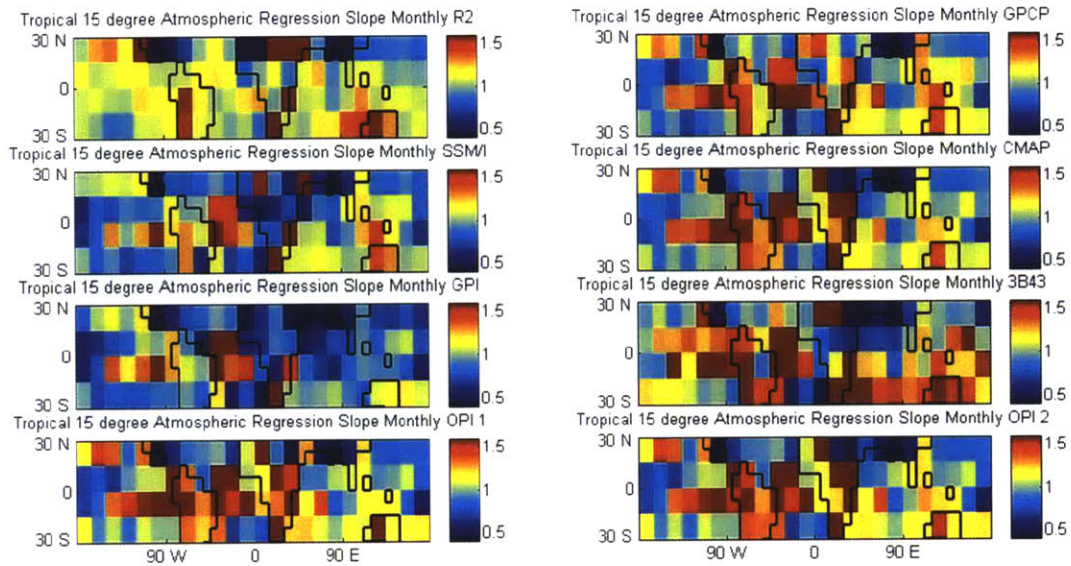


Figure B-80: 15 degree Regression Slope Different Precipitation Tropical Monthly

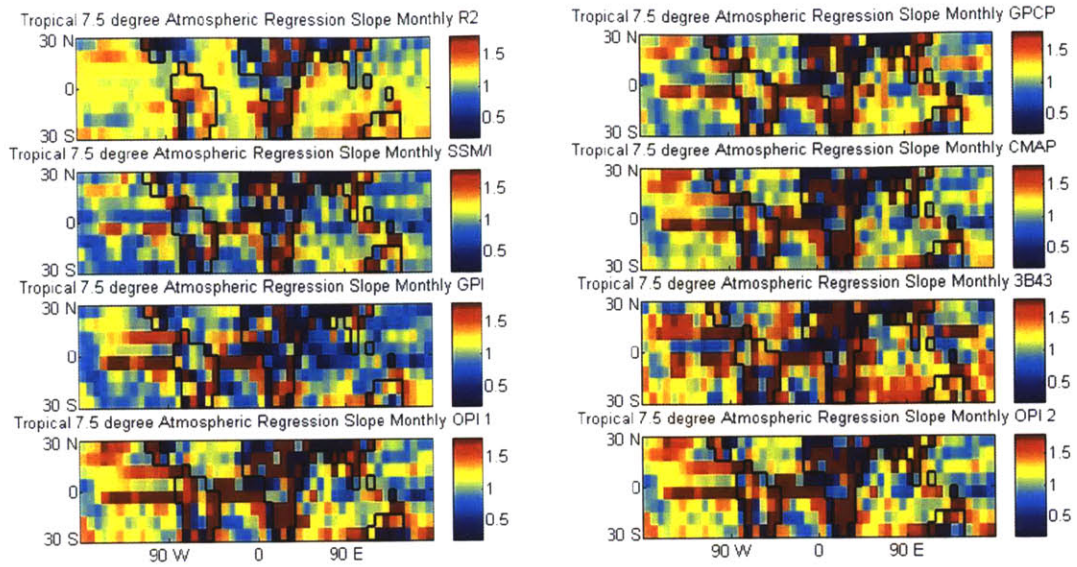


Figure B-81: 7.5 degree Regression Slope Different Precipitation Tropical Monthly

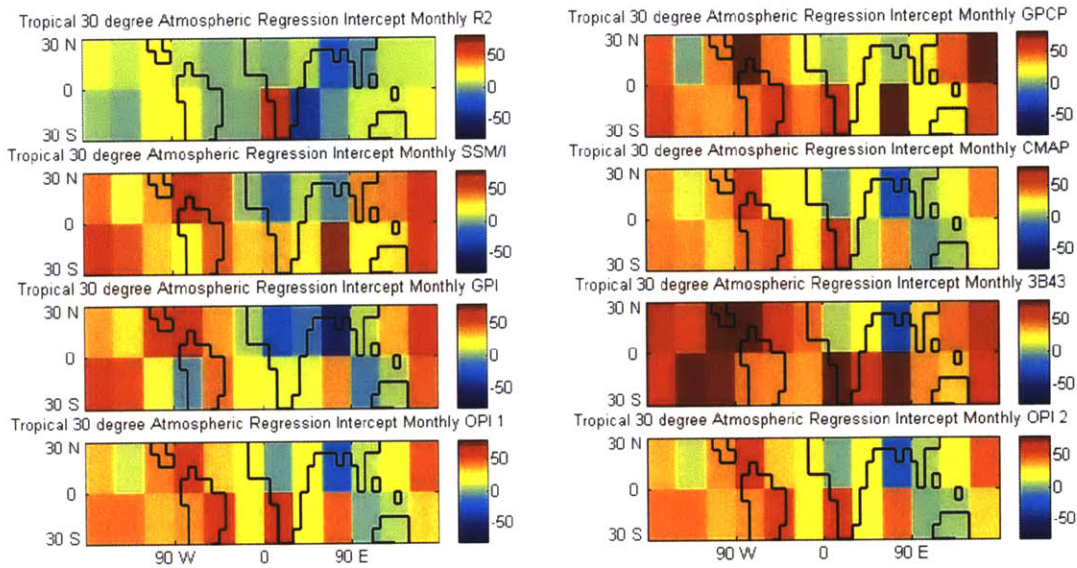


Figure B-82: 30 degree Regression Intercept Different Precipitation Tropical Monthly

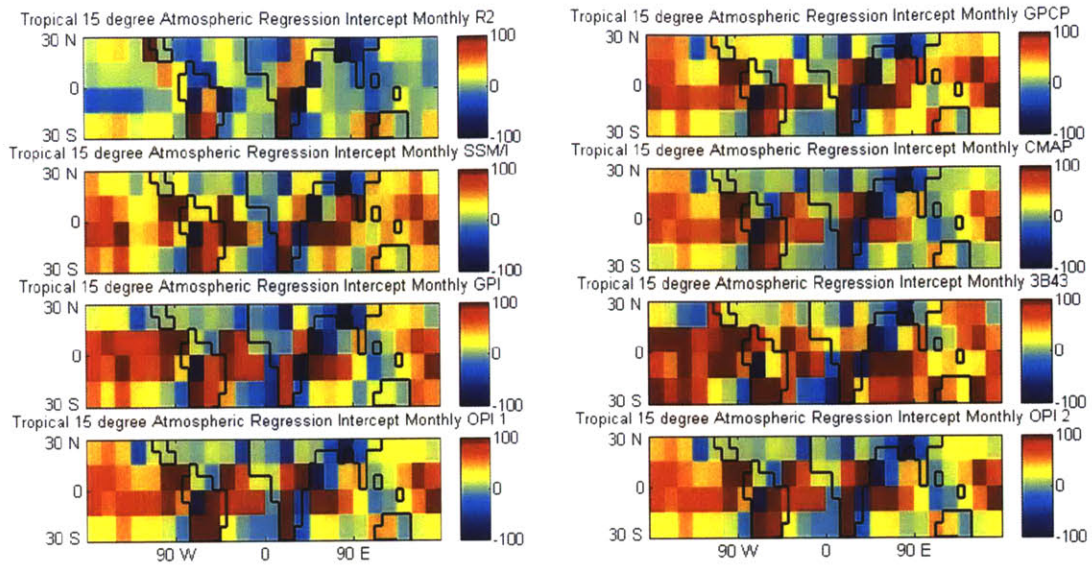


Figure B-83: 15 degree Regression Intercept Different Precipitation Tropical Monthly

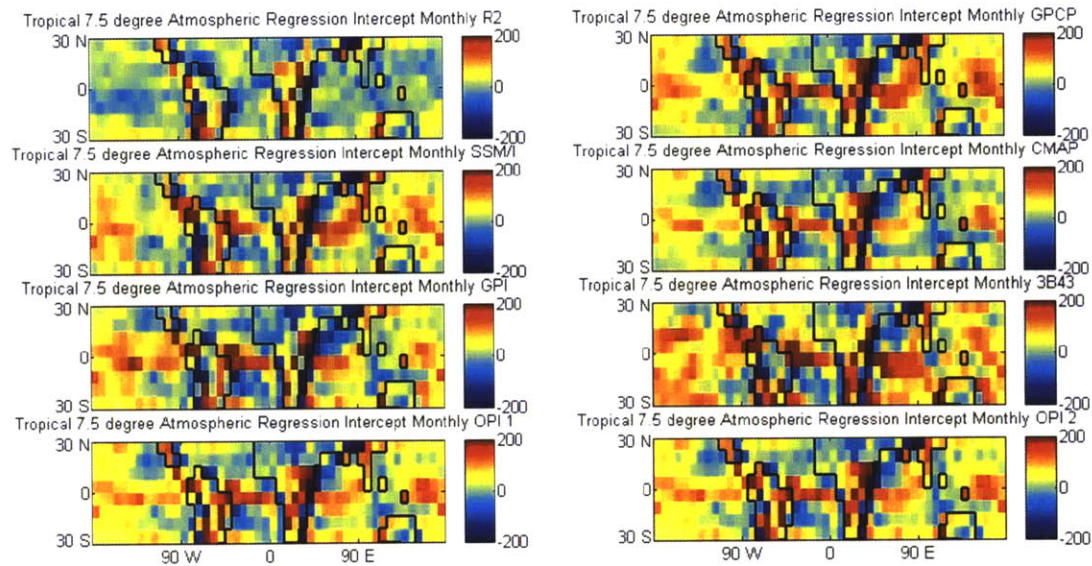


Figure B-84: 7.5 degree Regression Intercept Different Precipitation Tropical Monthly

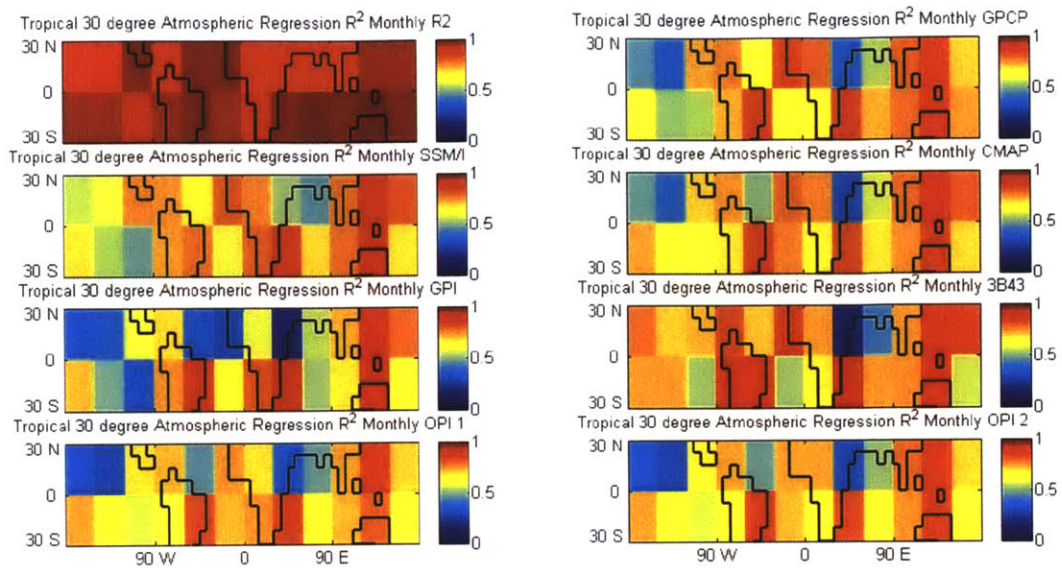


Figure B-85: 30 degree Regression  $R^2$  Different Precipitation Tropical Monthly

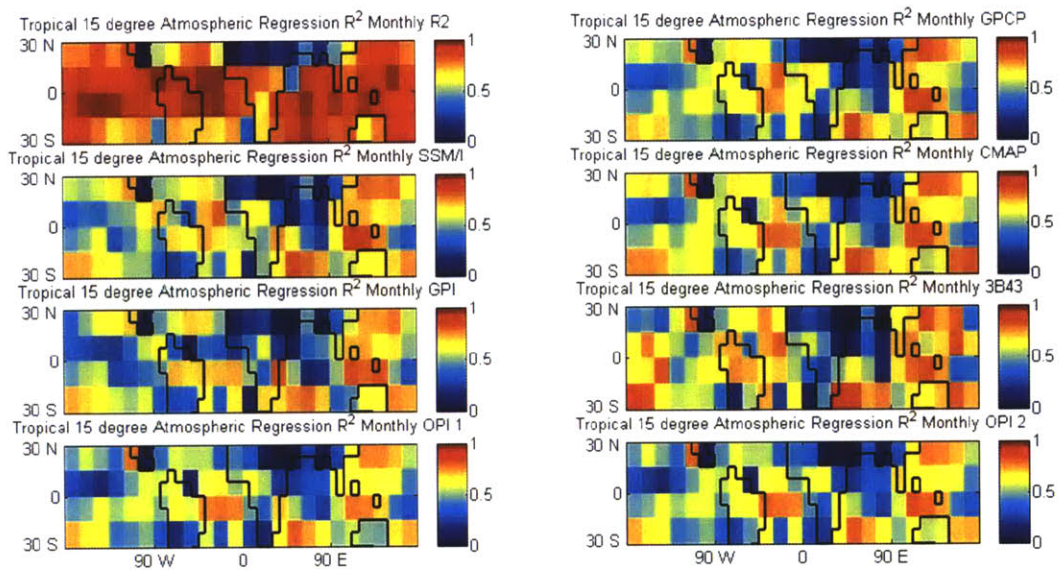


Figure B-86: 15 degree Regression  $R^2$  Different Precipitation Tropical Monthly

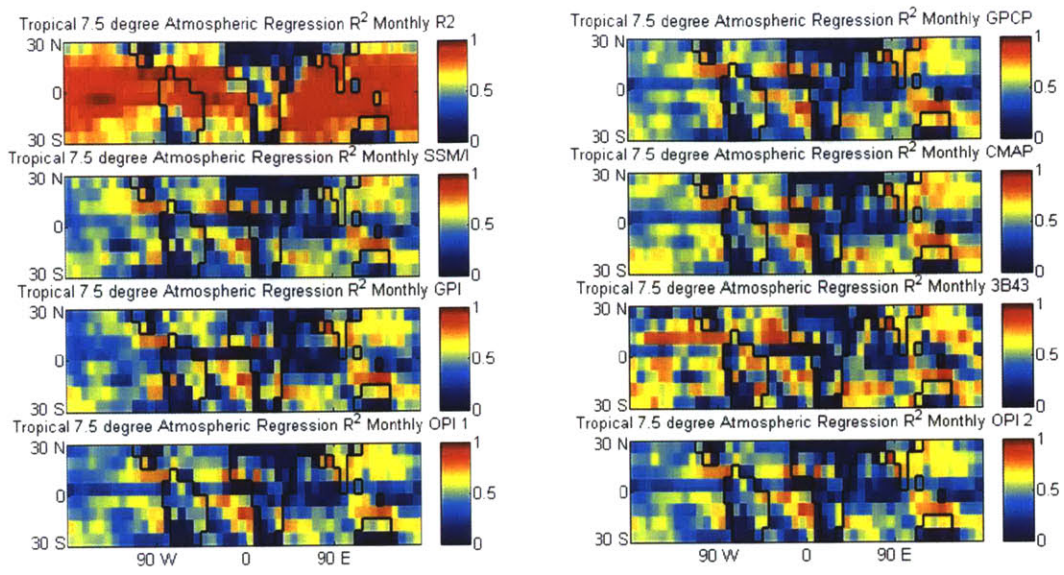


Figure B-87: 7.5 degree Regression  $R^2$  Different Precipitation Tropical Monthly

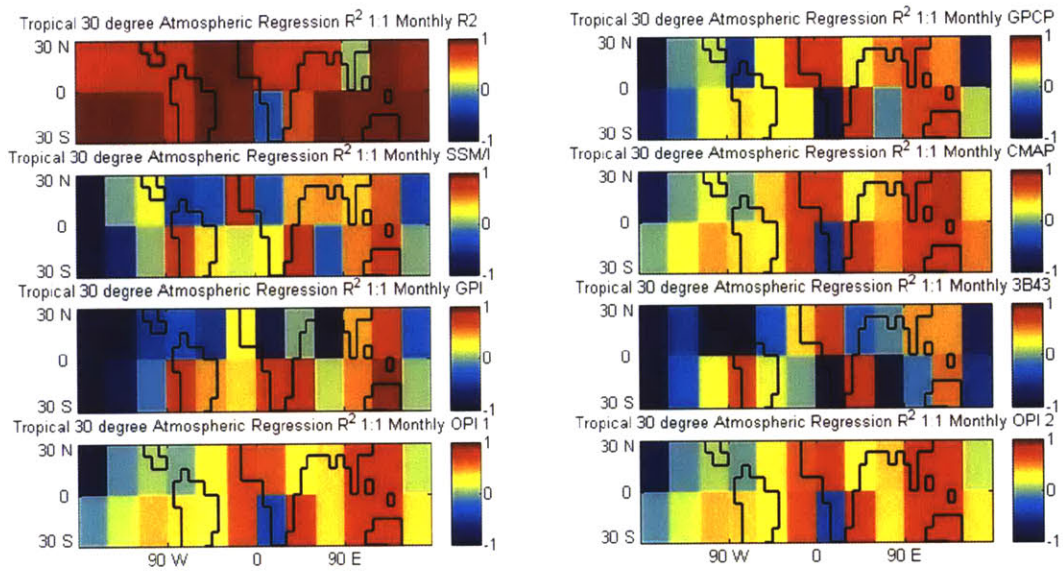


Figure B-88: 30 degree Regression  $R^2$  1:1 Different Precipitation Tropical Monthly

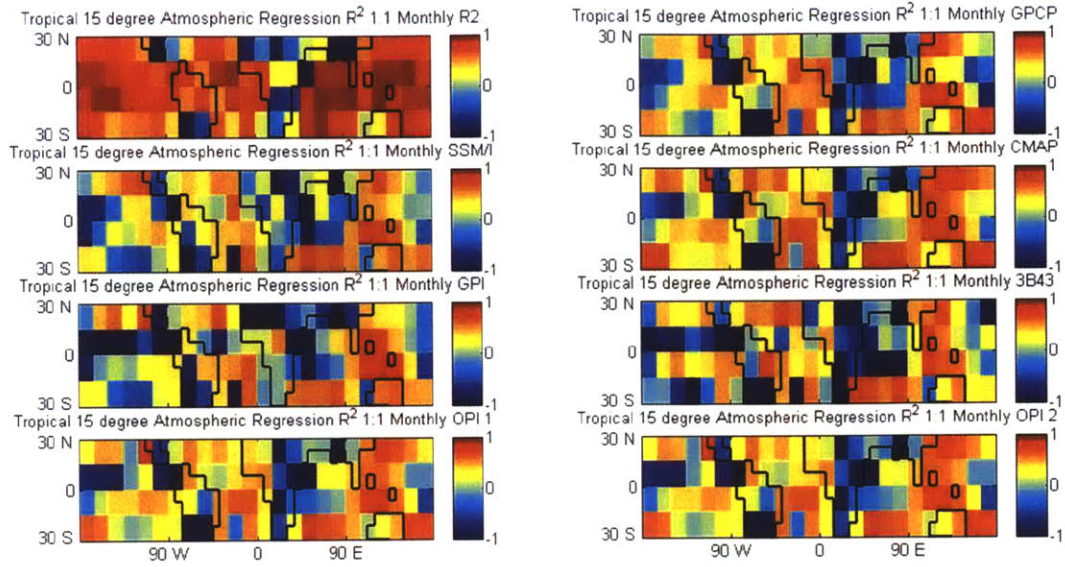


Figure B-89: 15 degree Regression  $R^2$  1:1 Different Precipitation Tropical Monthly

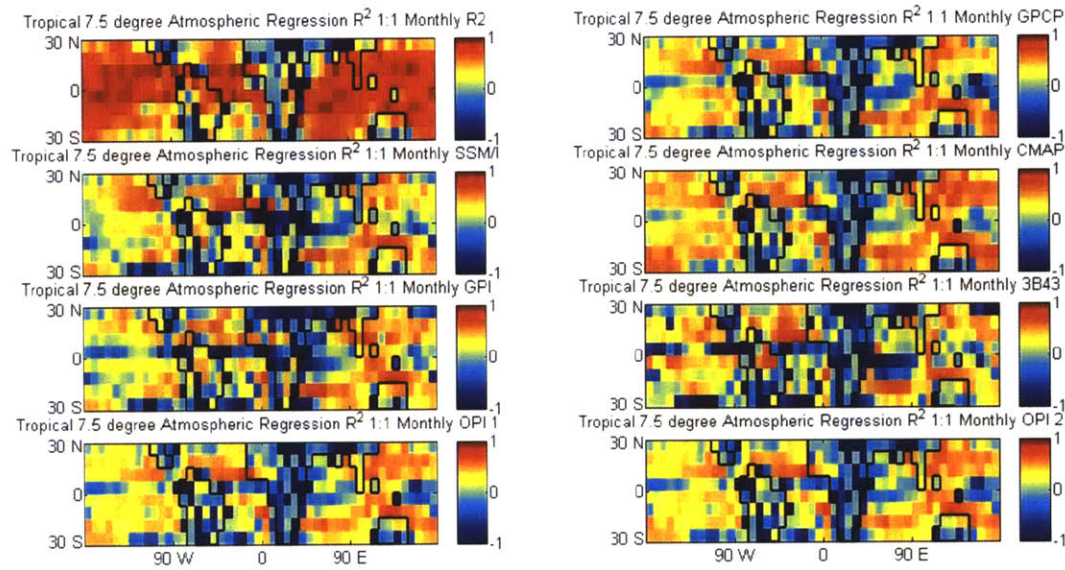


Figure B-90: 7.5 degree Regression  $R^2$  1:1 Different Precipitation Tropical Monthly

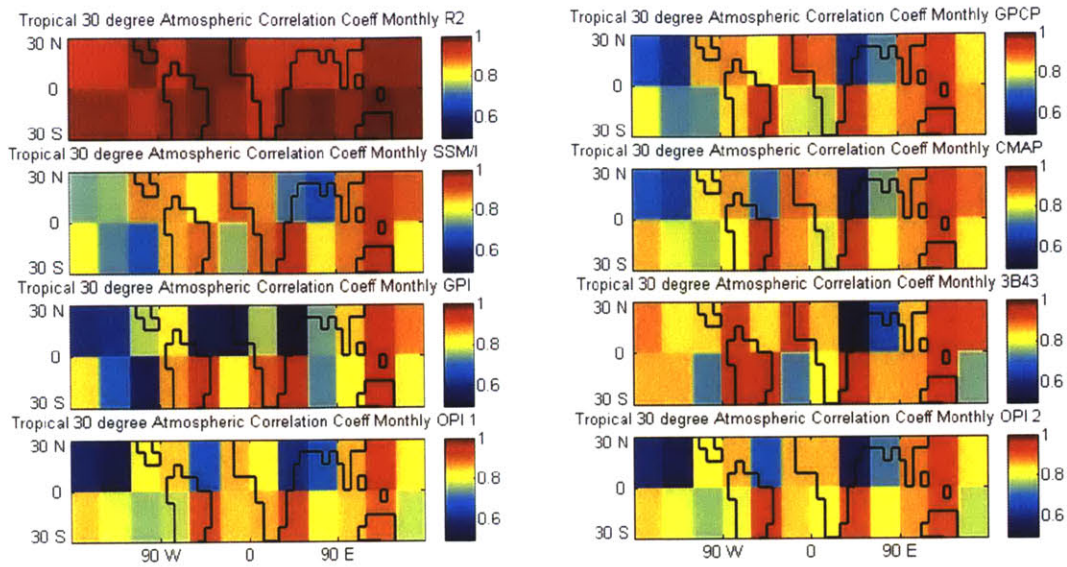


Figure B-91: 30 degree Correlation Coefficient Different Precipitation Tropical Monthly

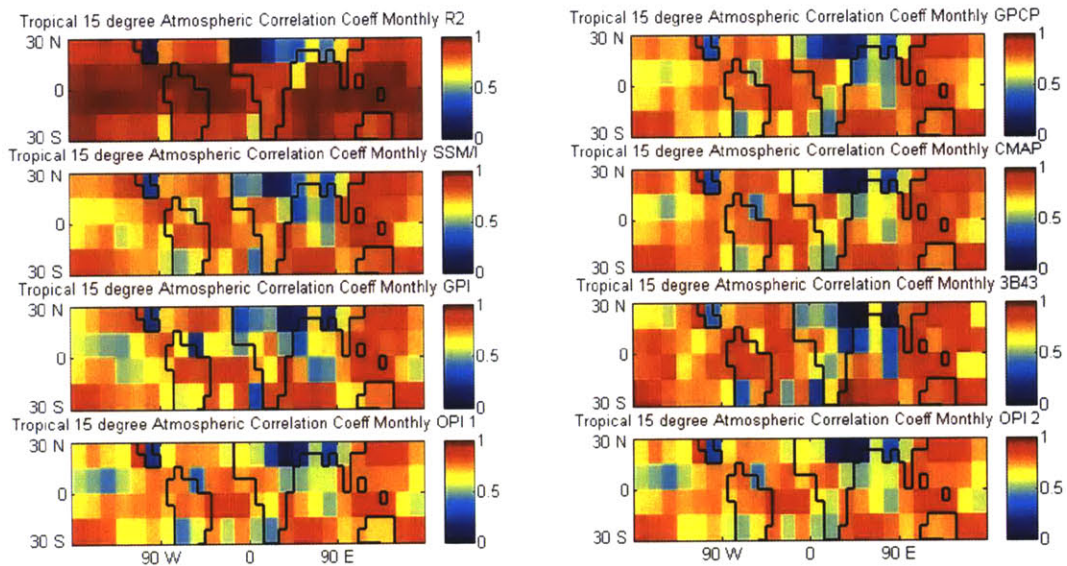


Figure B-92: 15 degree Correlation Coefficient Different Precipitation Tropical Monthly

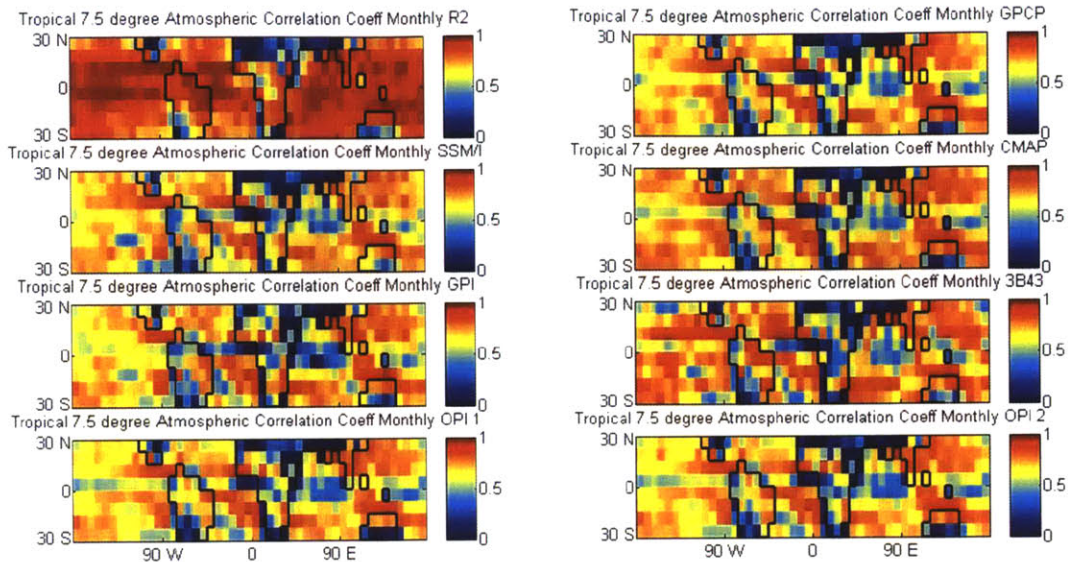


Figure B-93: 7.5 degree Correlation Coefficient Different Precipitation Tropical Monthly

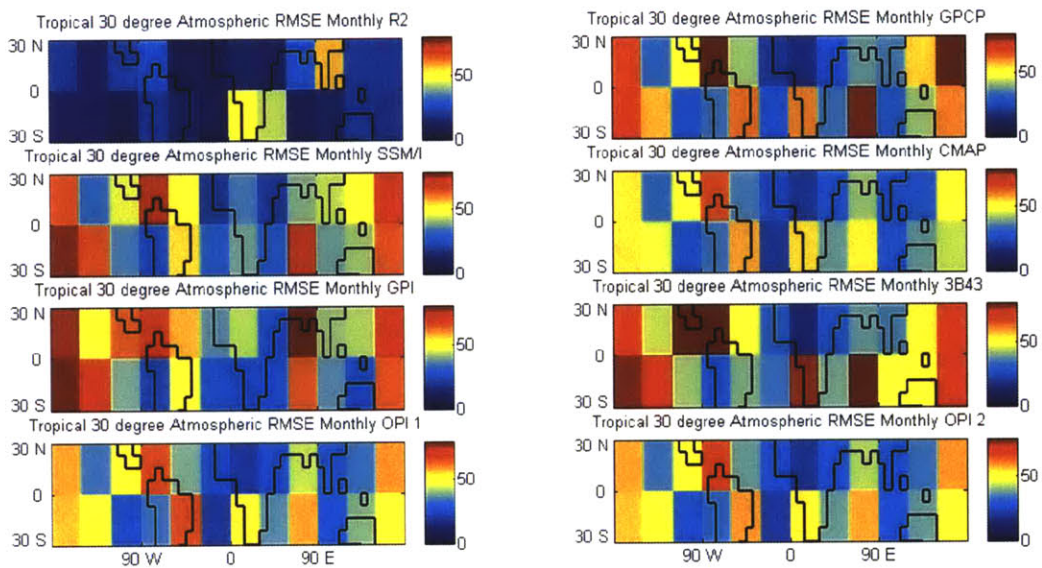


Figure B-94: 30 degree RMSE Different Precipitation Tropical Monthly



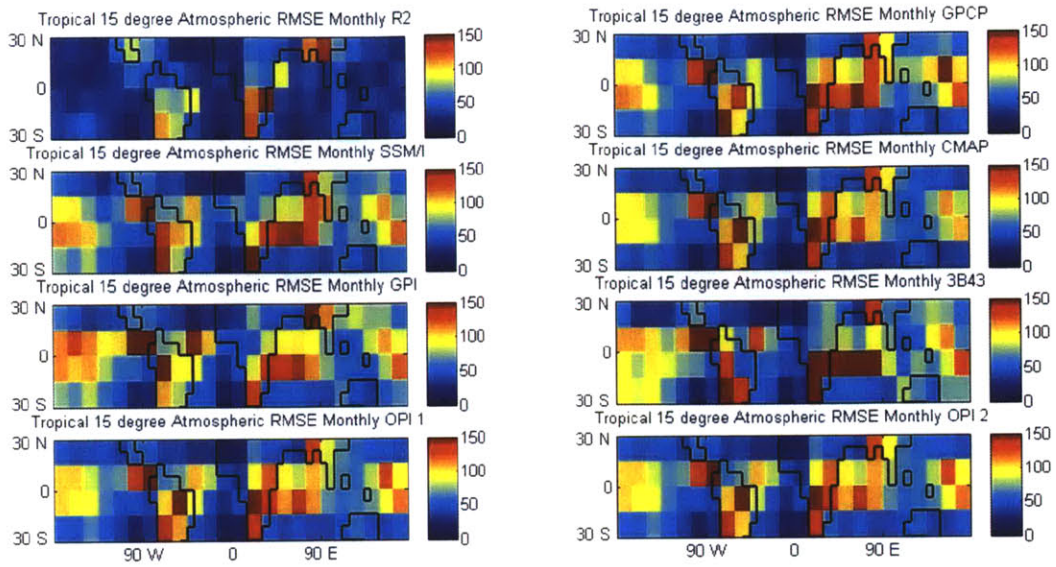


Figure B-95: 15 degree RMSE Different Precipitation Tropical Monthly

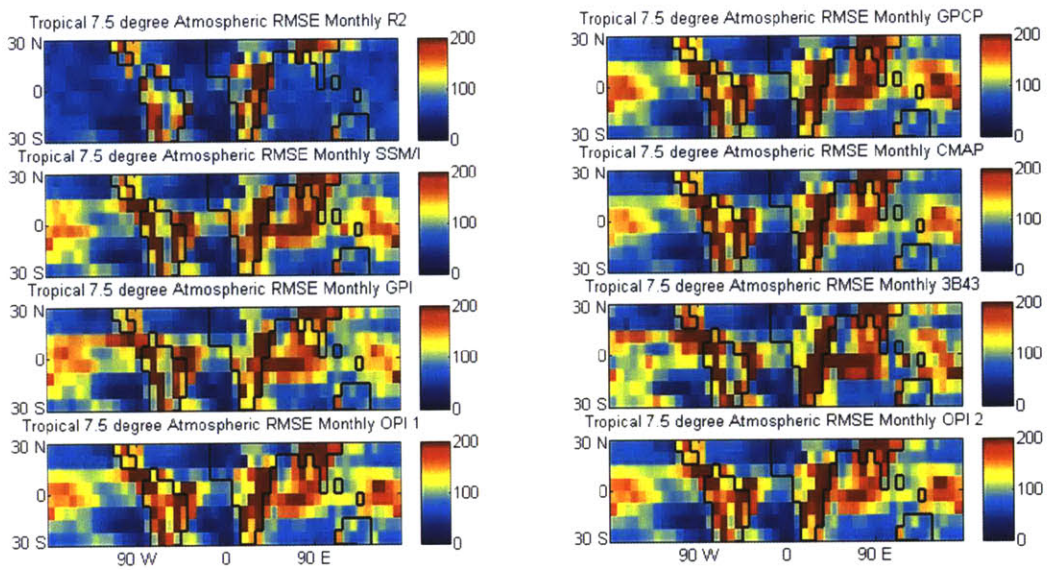


Figure B-96: 7.5 degree RMSE Different Precipitation Tropical Monthly

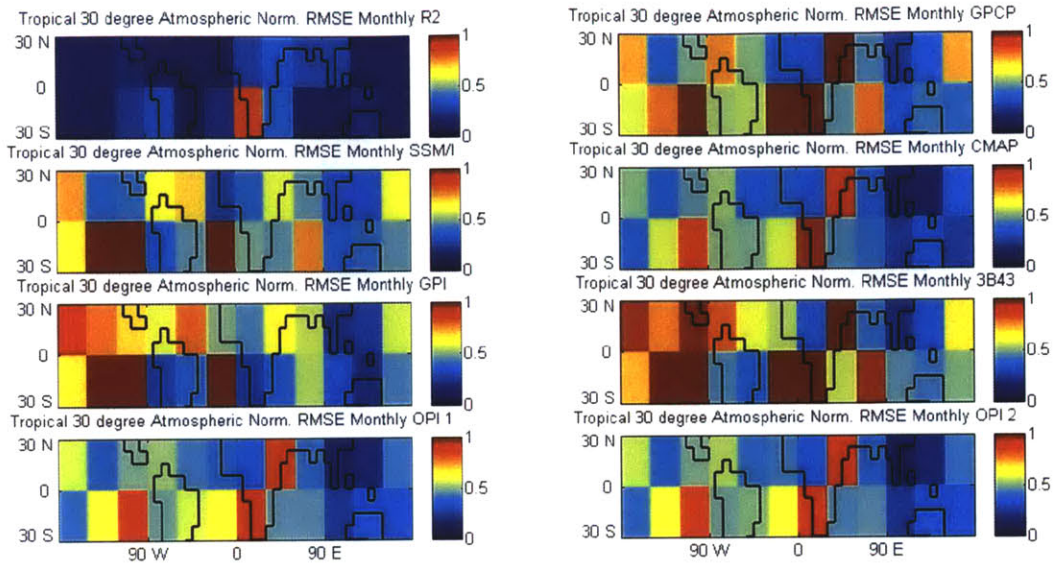


Figure B-97: 30 degree Normalized RMSE Different Precipitation Tropical Monthly

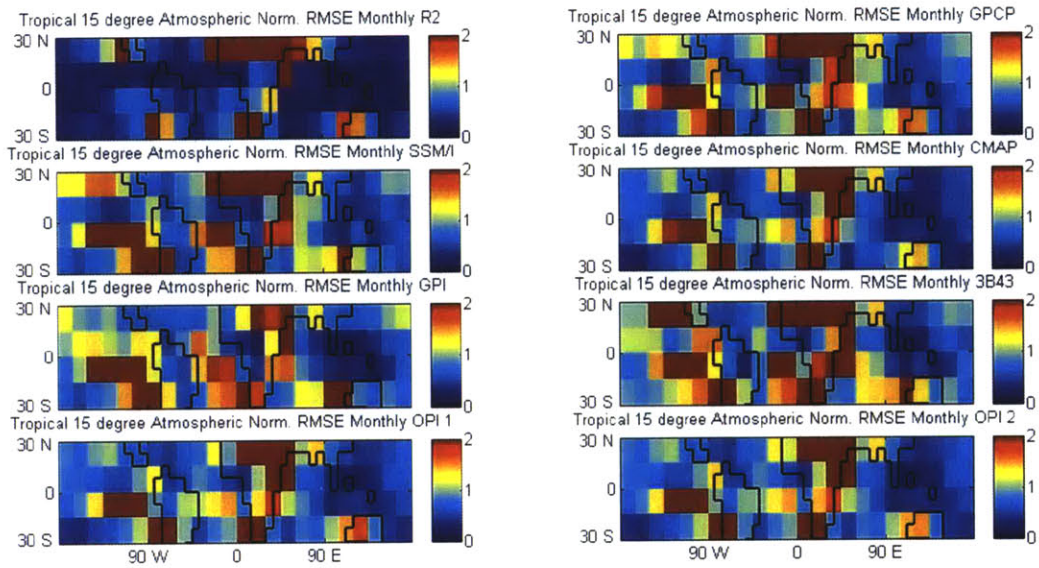


Figure B-98: 15 degree Normalized RMSE Different Precipitation Tropical Monthly

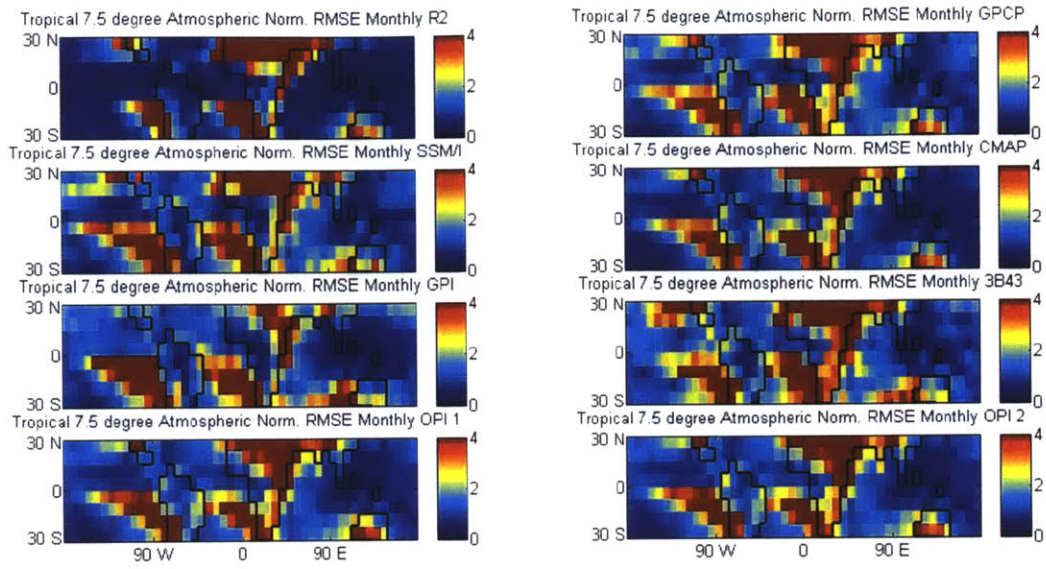


Figure B-99: 7.5 degree Normalized RMSE Different Precipitation Tropical Monthly

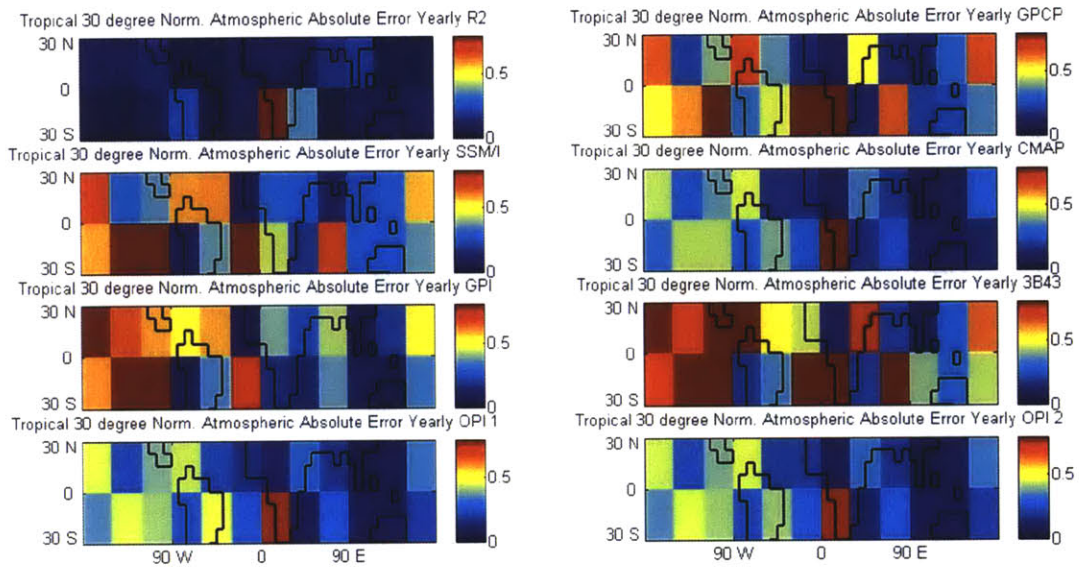


Figure B-100: 30 degree Normalized Absolute Error Different Precipitation Tropical Yearly

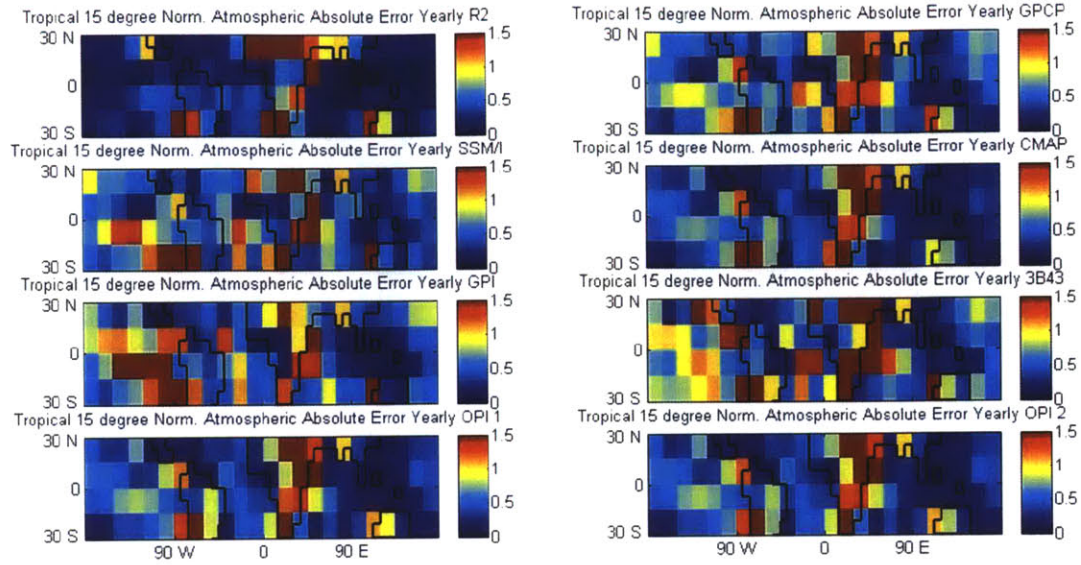


Figure B-101: 15 degree Normalized Absolute Error Different Precipitation Tropical Yearly

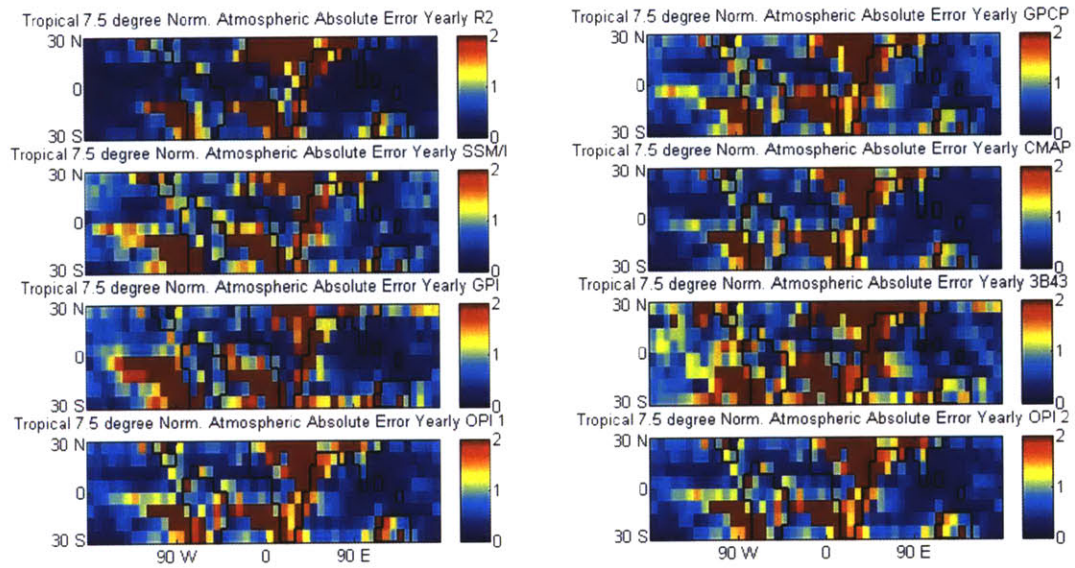


Figure B-102: 7.5 degree Normalized Absolute Error Different Precipitation Tropical Yearly

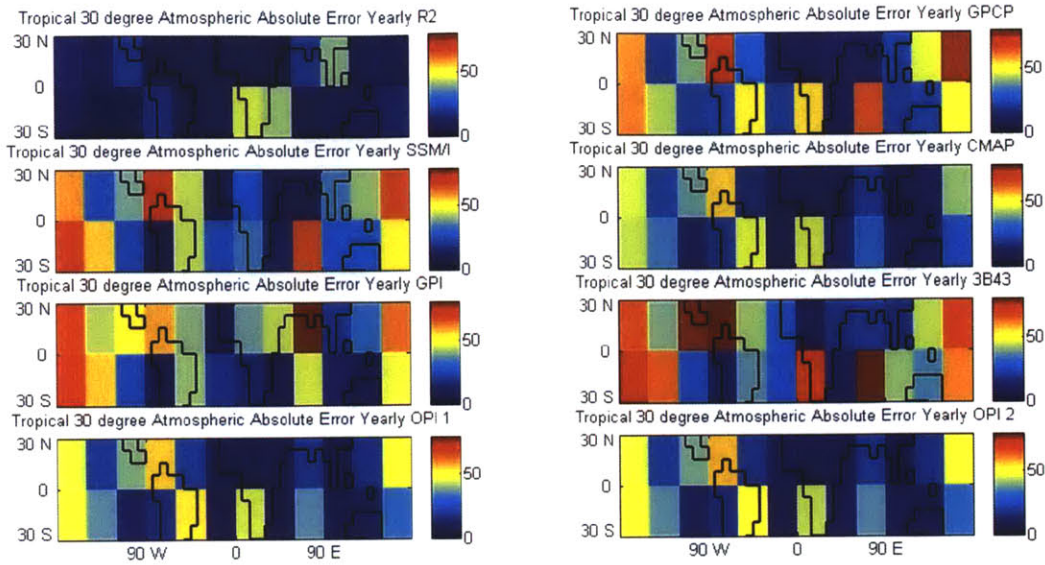


Figure B-103: 30 degree Absolute Error Different Precipitation Tropical Yearly

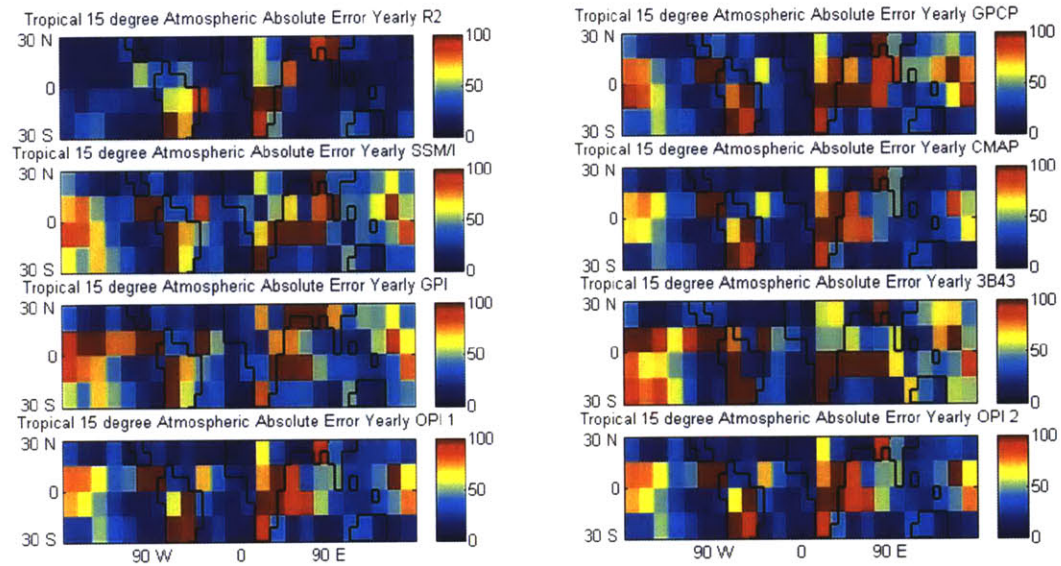


Figure B-104: 15 degree Absolute Error Different Precipitation Tropical Yearly

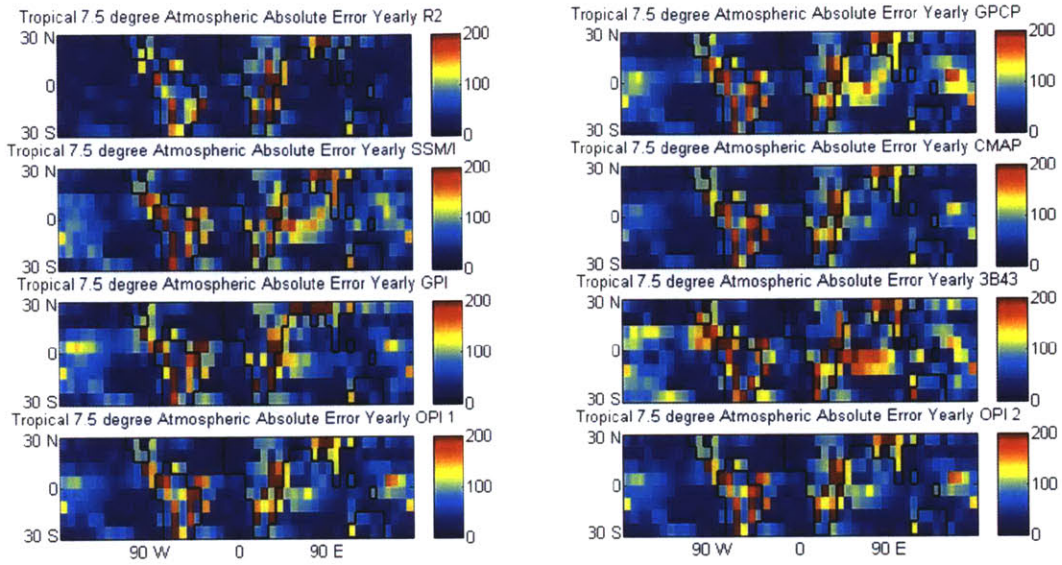


Figure B-105: 7.5 degree Absolute Error Different Precipitation Tropical Yearly

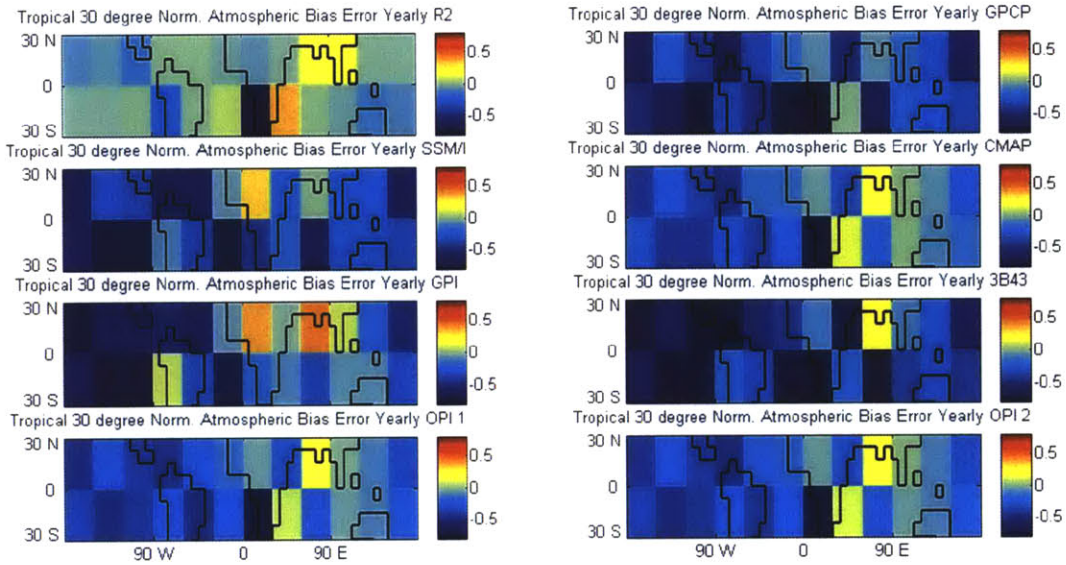


Figure B-106: 30 degree Normalized Bias Error Different Precipitation Tropical Yearly

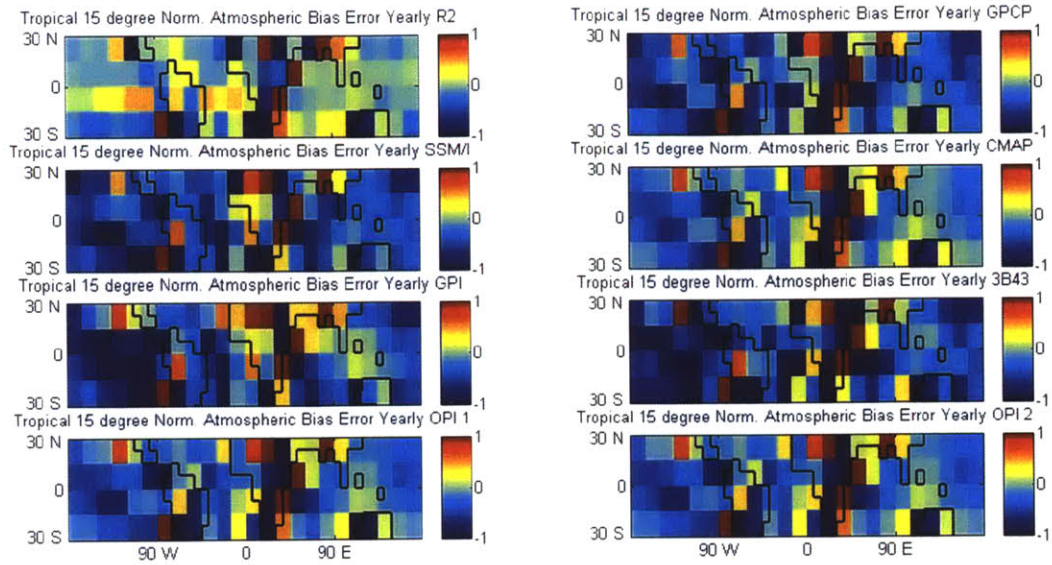


Figure B-107: 15 degree Normalized Bias Error Different Precipitation Tropical Yearly

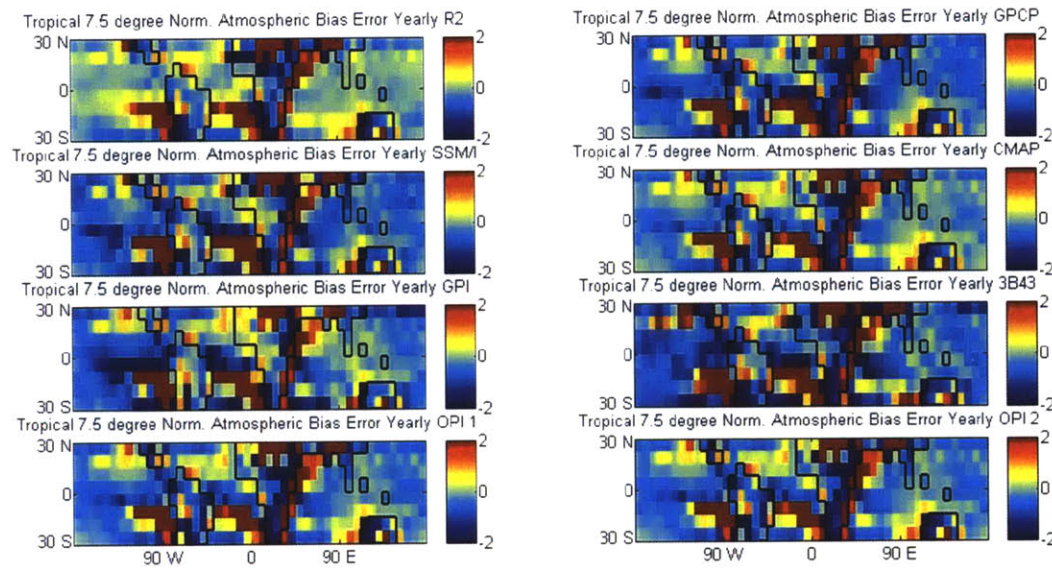


Figure B-108: 7.5 degree Normalized Bias Error Different Precipitation Tropical Yearly

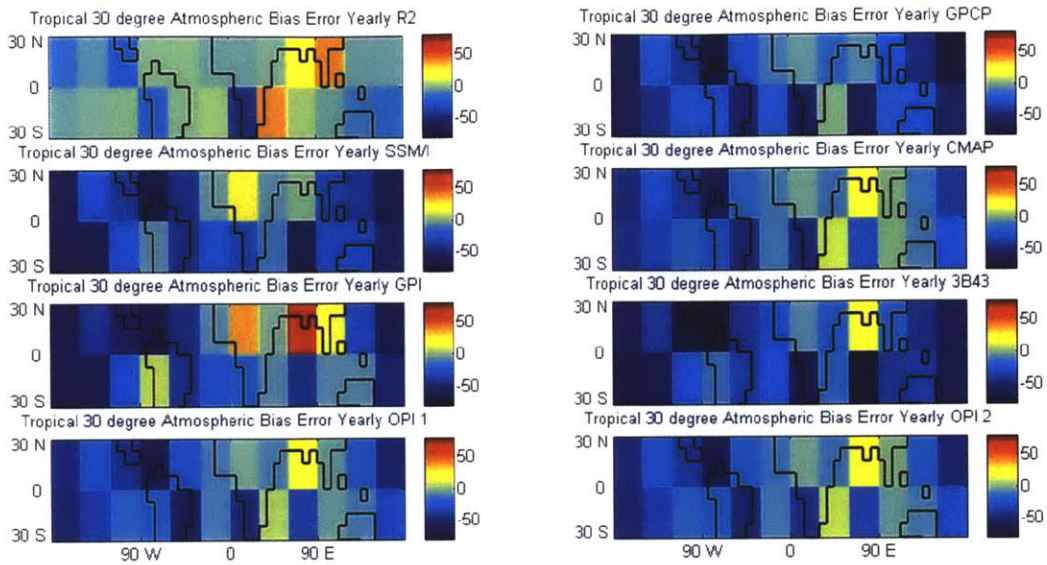


Figure B-109: 30 degree Bias Error Different Precipitation Tropical Yearly

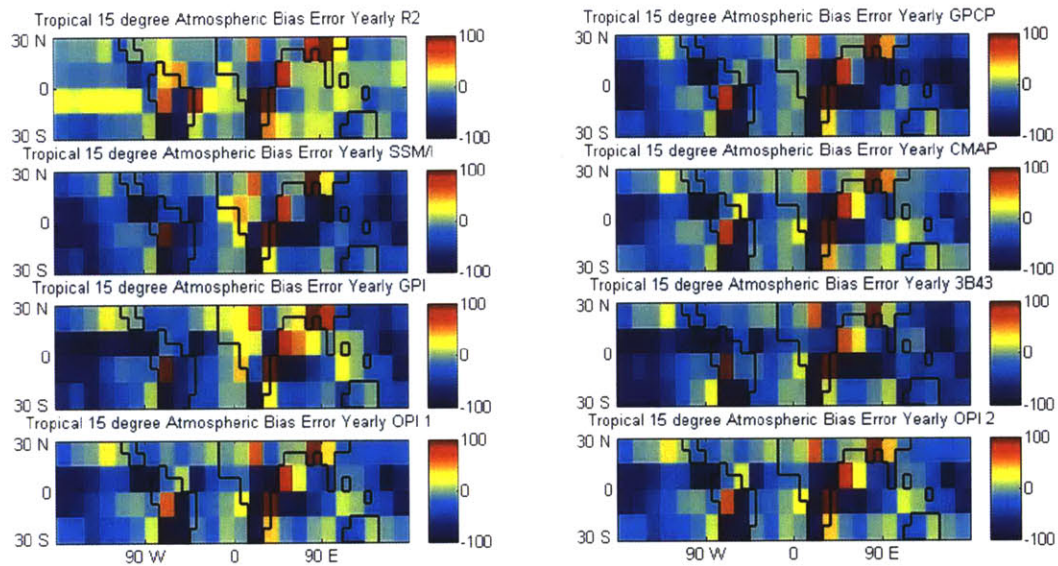


Figure B-110: 15 degree Bias Error Different Precipitation Tropical Yearly



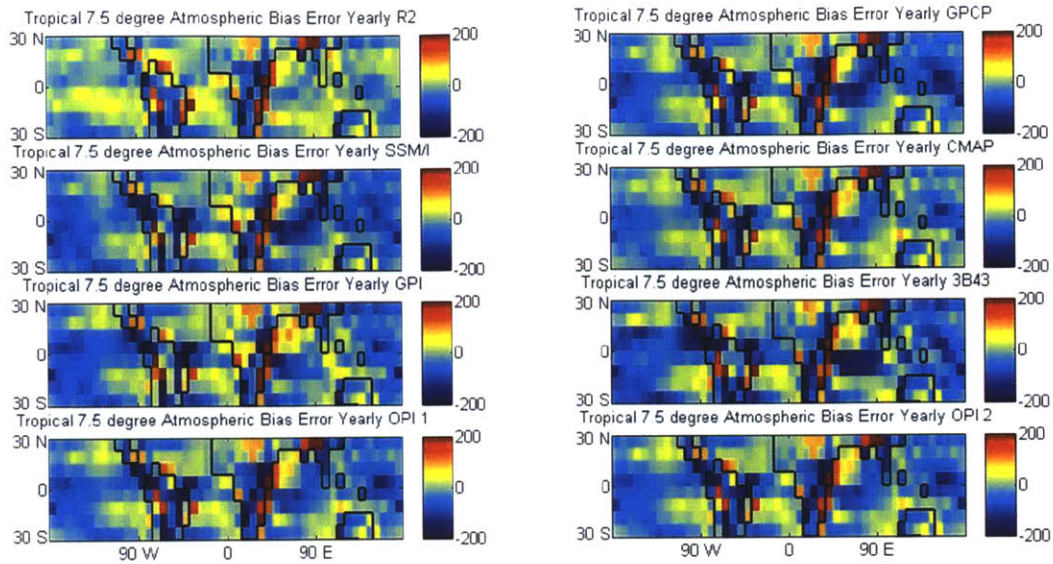


Figure B-111: 7.5 degree Bias Error Different Precipitation Tropical Yearly

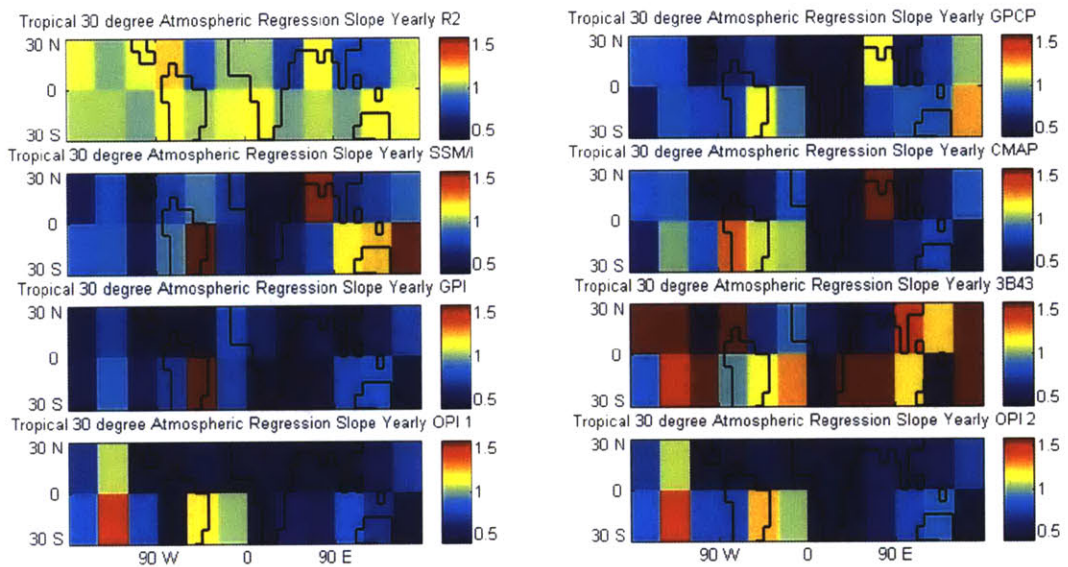


Figure B-112: 30 degree Regression Slope Different Precipitation Tropical Yearly

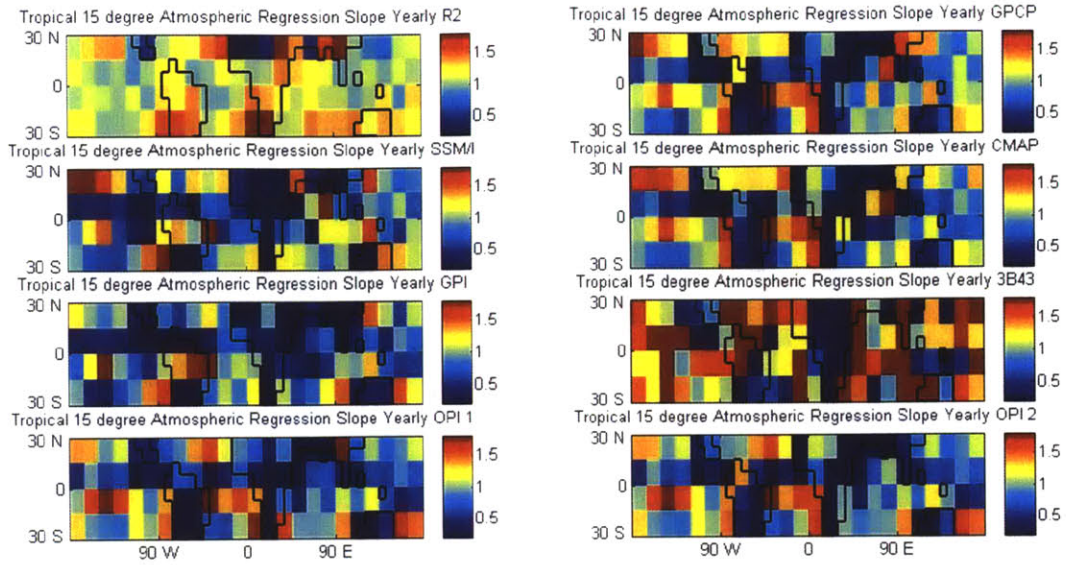


Figure B-113: 15 degree Regression Slope Different Precipitation Tropical Yearly

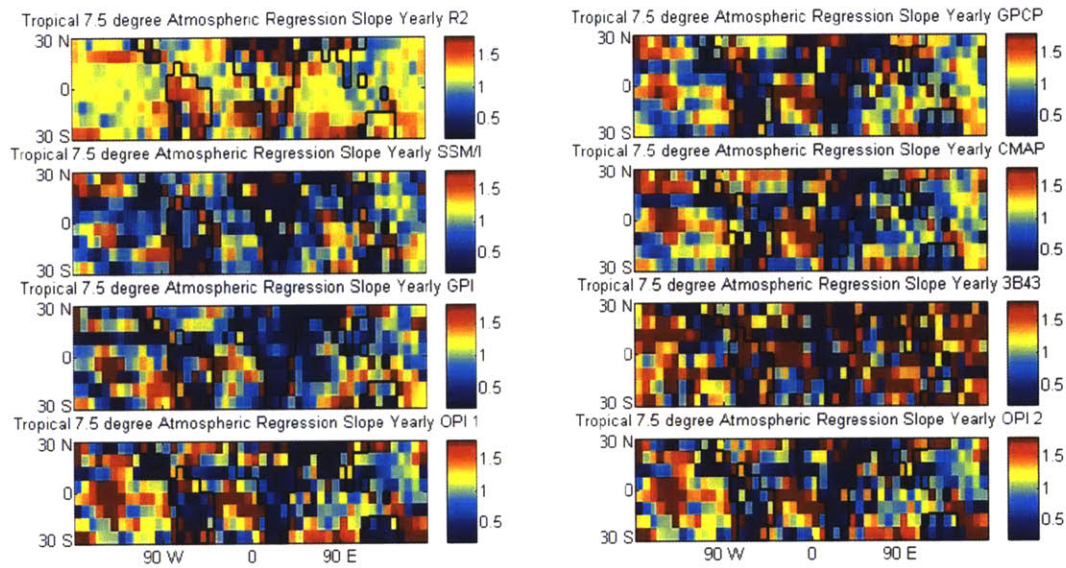


Figure B-114: 7.5 degree Regression Slope Different Precipitation Tropical Yearly

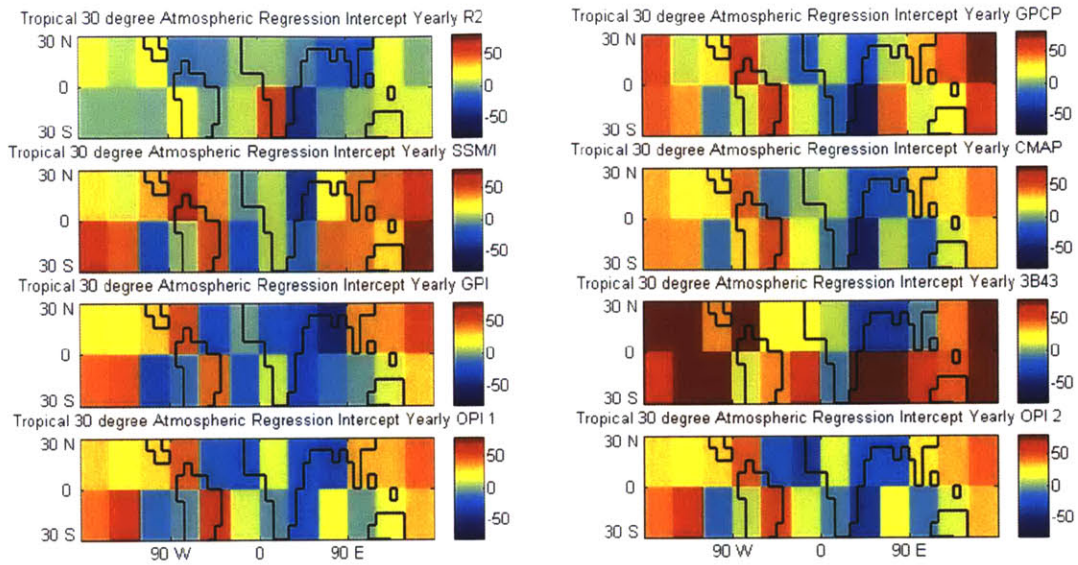


Figure B-115: 30 degree Regression Intercept Different Precipitation Tropical Yearly

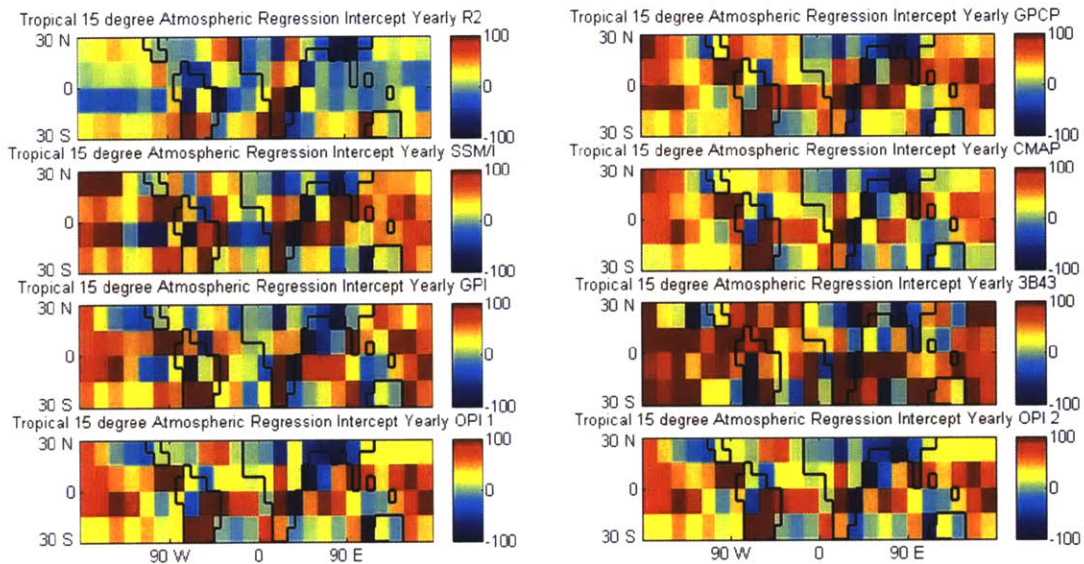


Figure B-116: 15 degree Regression Intercept Different Precipitation Tropical Yearly

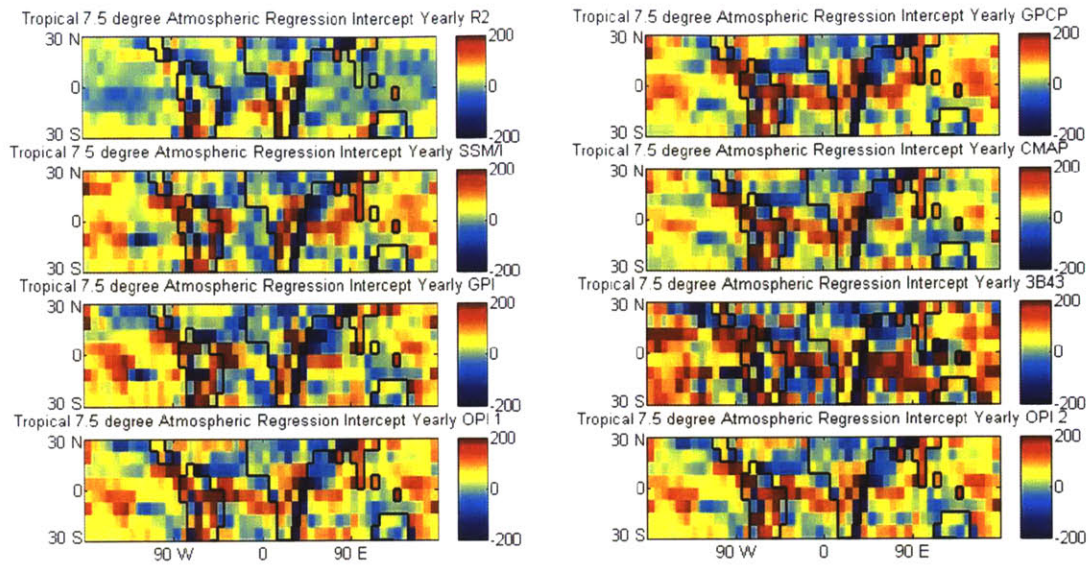


Figure B-117: 7.5 degree Regression Intercept Different Precipitation Tropical Yearly

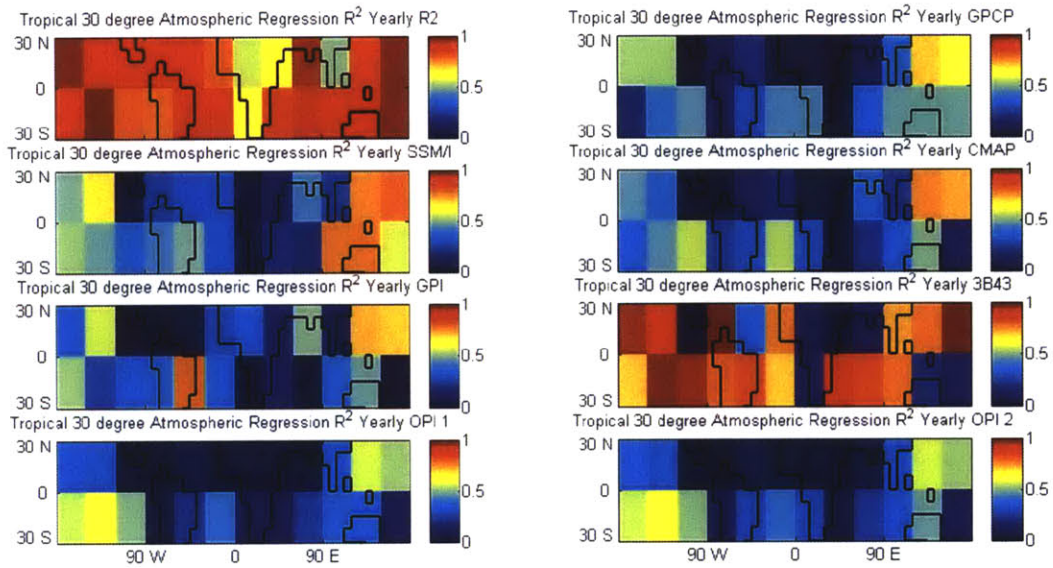


Figure B-118: 30 degree Regression  $R^2$  Different Precipitation Tropical Yearly

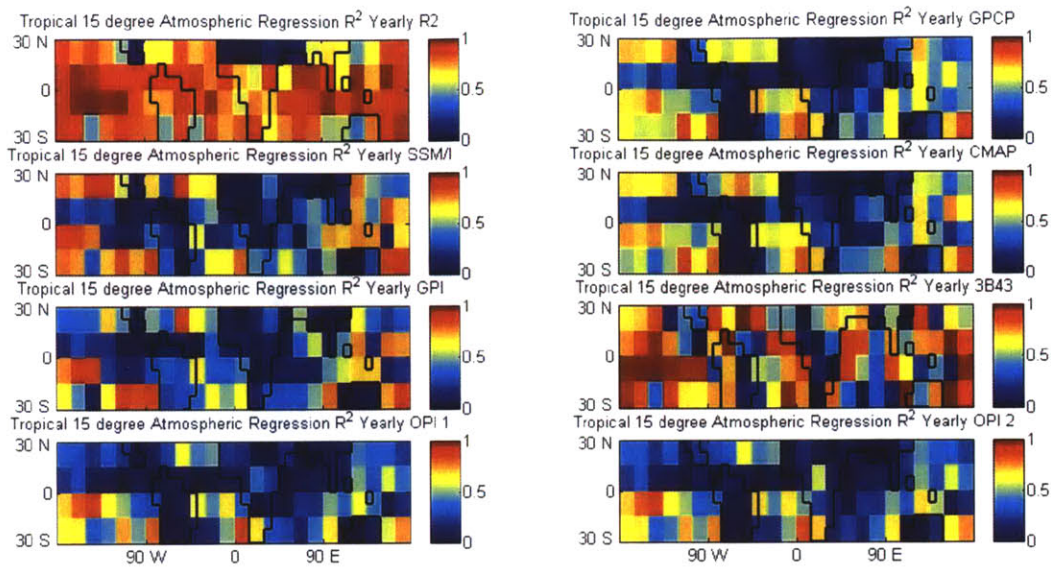


Figure B-119: 15 degree Regression  $R^2$  Different Precipitation Tropical Yearly

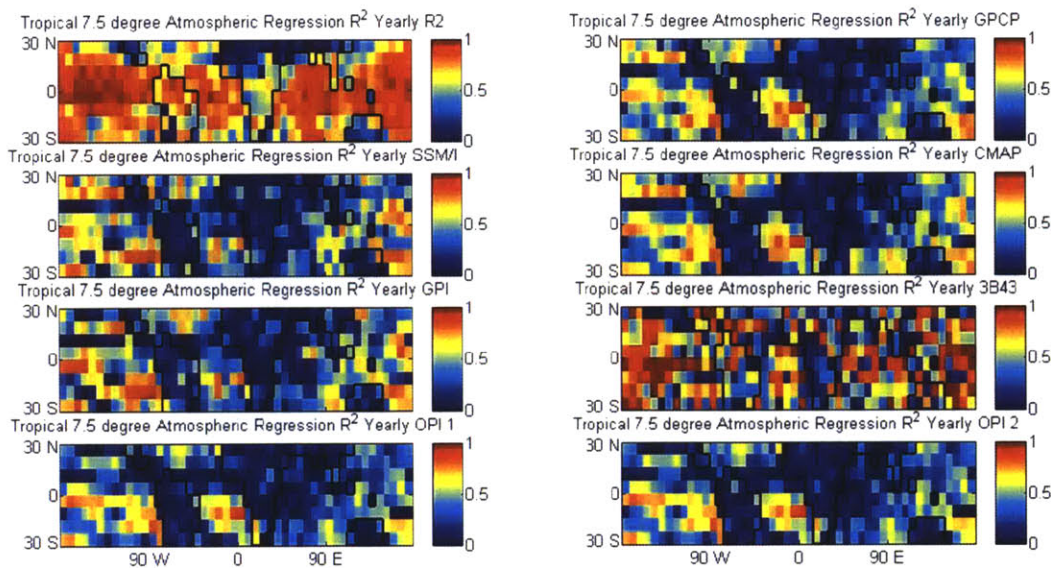


Figure B-120: 7.5 degree Regression  $R^2$  Different Precipitation Tropical Yearly

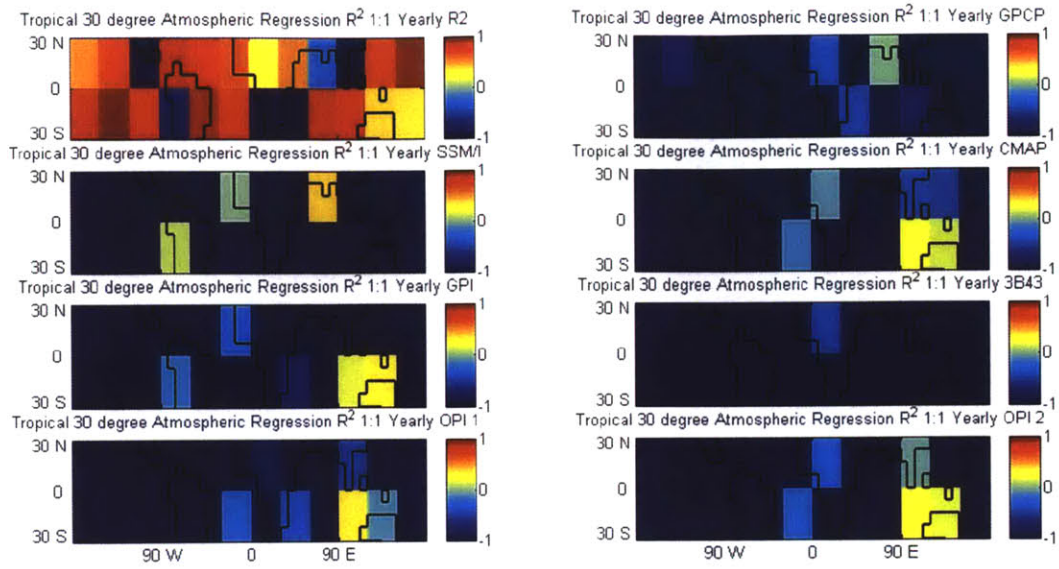


Figure B-121: 30 degree Regression  $R^2$  1:1 Different Precipitation Tropical Yearly

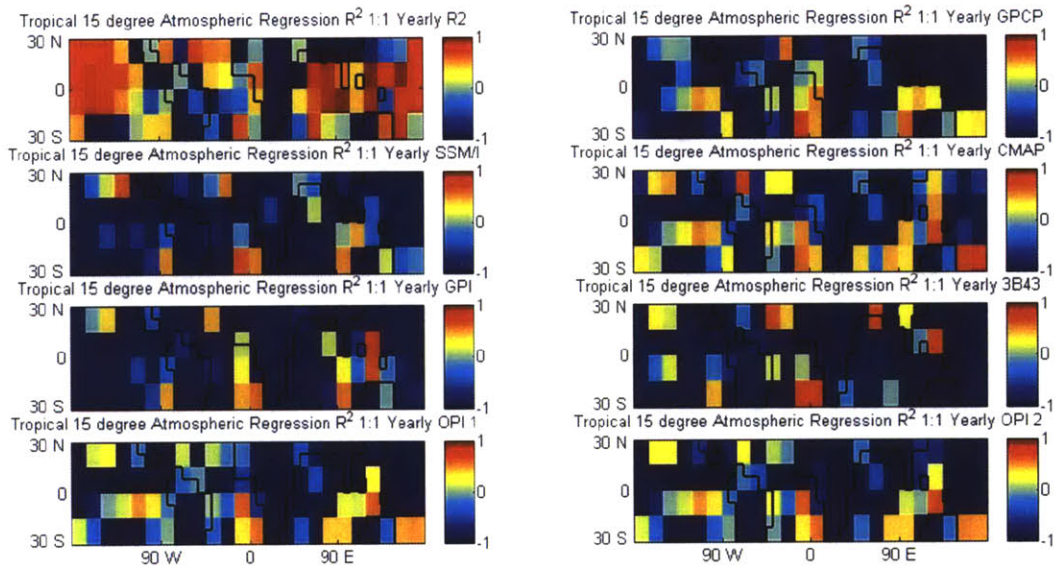


Figure B-122: 15 degree Regression  $R^2$  1:1 Different Precipitation Tropical Yearly

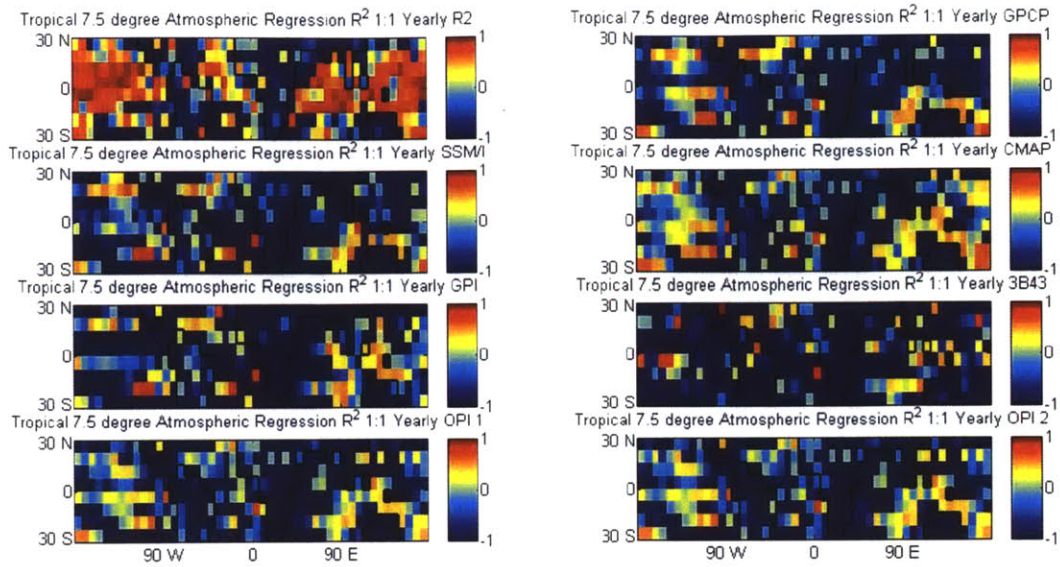


Figure B-123: 7.5 degree Regression  $R^2$  1:1 Different Precipitation Tropical Yearly

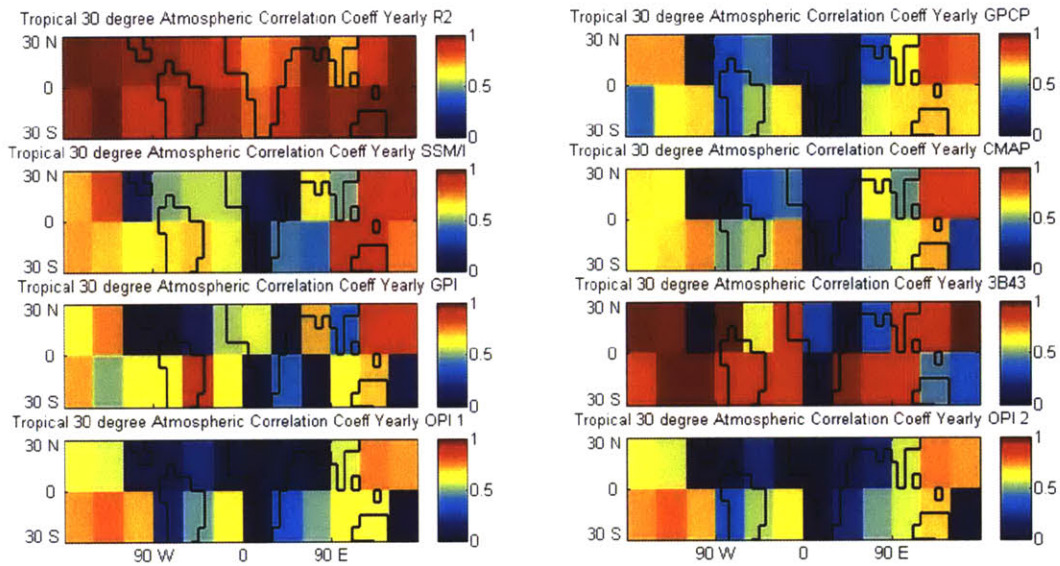


Figure B-124: 30 degree Correlation Coefficient Different Precipitation Tropical Yearly

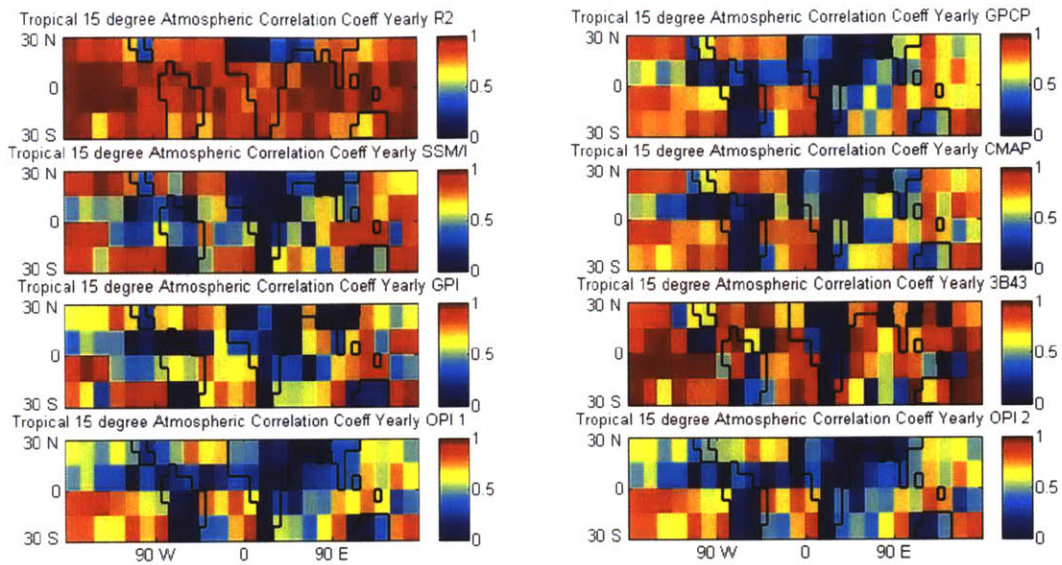


Figure B-125: 15 degree Correlation Coefficient Different Precipitation Tropical Yearly

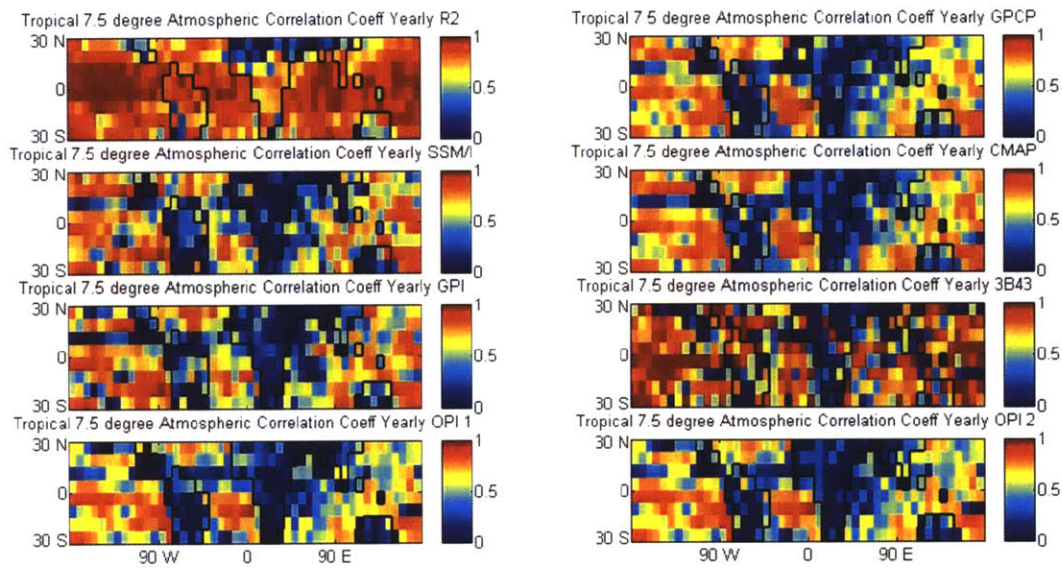


Figure B-126: 7.5 degree Correlation Coefficient Different Precipitation Tropical Yearly



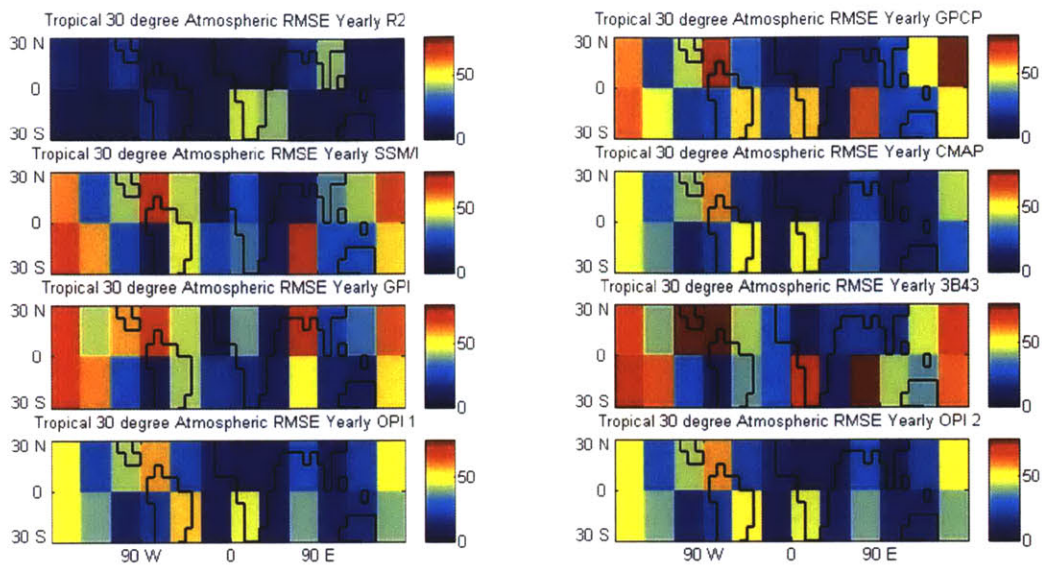


Figure B-127: 30 degree RMSE Different Precipitation Tropical Yearly

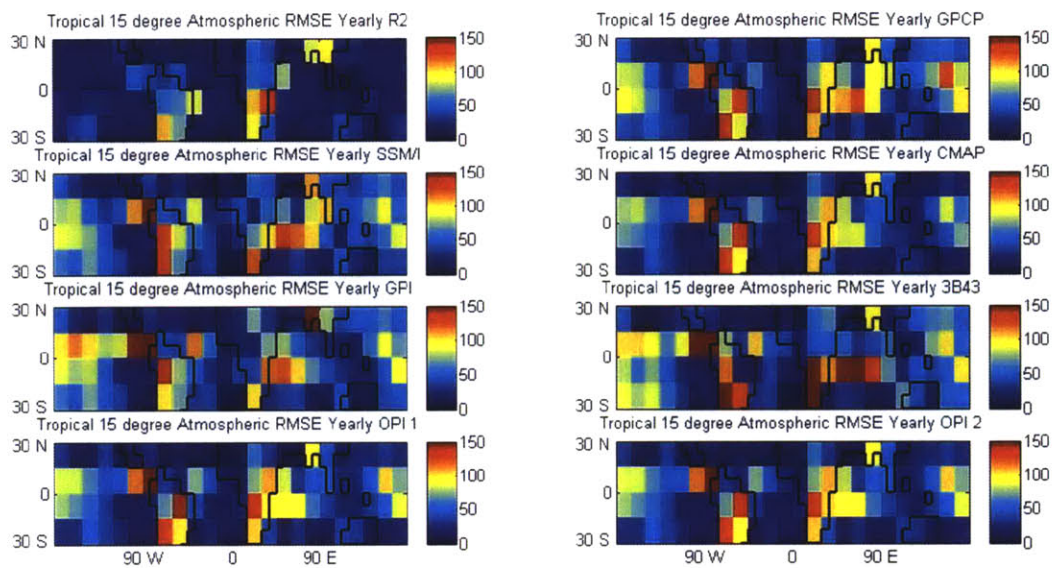


Figure B-128: 15 degree RMSE Different Precipitation Tropical Yearly

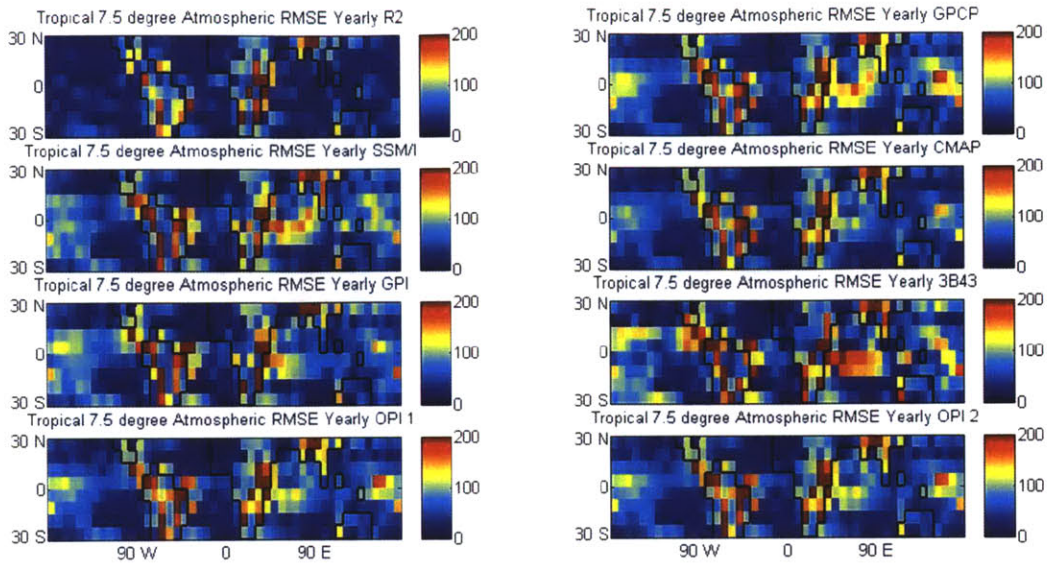


Figure B-129: 7.5 degree RMSE Different Precipitation Tropical Yearly

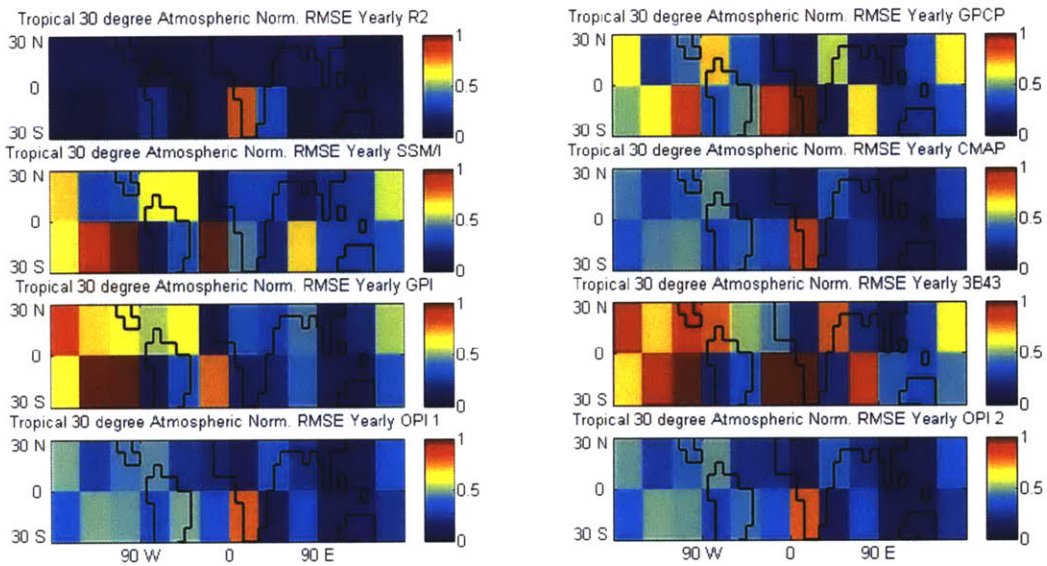


Figure B-130: 30 degree Normalized RMSE Different Precipitation Tropical Yearly

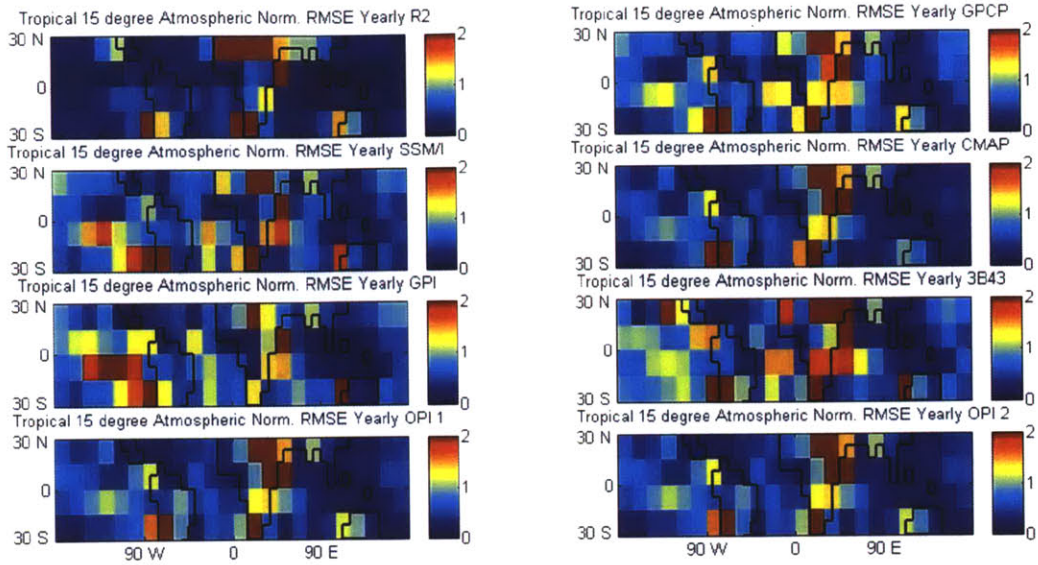


Figure B-131: 15 degree Normalized RMSE Different Precipitation Tropical Yearly

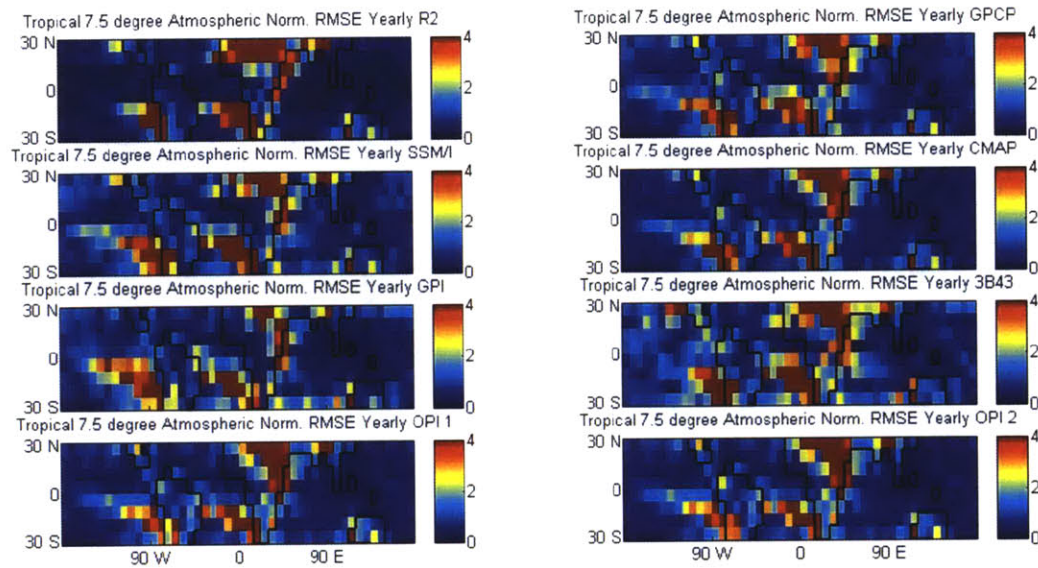


Figure B-132: 7.5 degree Normalized RMSE Different Precipitation Tropical Yearly

## **Acknowledgements**

I would like to thank NASA for sponsoring this work under grant NAG59640, Application of TRMM Products in Hydrologic Studies and the Amazon Region. The opinions expressed here do not represent NASA's positions or policies.

I would also like to thank my advisor, Rafael Bras, as well as the members of my research group: Jingfeng Wang, Frederic Chagnon, Jean Fitzmaurice, and Fotis Fotopoulos, without which this work would not be possible.

And lastly, I would like to thank to my parents and brother for there never ending support and love.