Contrasting the direct radiative effect and direct radiative forcing of aerosols

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Abstract. The direct radiative effect (DRE) of aerosols, which is the instantaneous radiative impact of all atmospheric particles on the Earth’s energy balance, is sometimes confused with the direct radiative forcing (DRF), which is the change in DRE from pre-industrial to present-day (not including climate feedbacks). In this study we couple a global chemical transport model (GEOS-Chem) with a radiative transfer model (RRTMG) to contrast these concepts. We estimate a global mean all-sky aerosol DRF of −0.36 Wm−2 and a DRE of −1.83 Wm−2 for 2010. Therefore, natural sources of aerosol (here including fire) affect the global energy balance over four times more than do present-day anthropogenic aerosols. If global anthropogenic emissions of aerosols and their precursors continue to decline as projected in recent scenarios due to effective pollution emission controls, the DRF will shrink (−0.22 Wm−2 for 2100). Secondary metrics, like DRE, that quantify temporal changes in both natural and anthropogenic aerosol burdens are therefore needed to quantify the total effect of aerosols on climate.

1 Introduction

Atmospheric aerosols are the most uncertain driver of global climate change (IPCC, 2013). These particles can scatter or absorb radiation, thereby cooling or warming the Earth and its atmosphere directly. They also play a pivotal role in cloud formation, acting as nuclei for liquid or ice water clouds, and can thus indirectly cool the planet by increasing its albedo. The overall impact of present-day atmospheric aerosols is estimated to be cooling, globally counterbalancing a significant fraction of the warming associated with greenhouse gases (IPCC, 2013). There is thus a critical need to better quantify the role of aerosols in the climate system. The ability of aerosols to modify climate depends on their atmospheric abundance over time as well as their chemical, physical, and optical properties; uncertainties on all of which are large (Myhre et al., 2013; Kinne et al., 2006).

The direct radiative effect (DRE) is Earth’s instantaneous radiative flux imbalance between incoming net solar radiation and outgoing infrared radiation resulting from the presence of a constituent of the Earth’s atmosphere (Boucher and Tanre, 2000). This is distinct from direct radiative forcing (DRF), a leading climate-relevant metric for aerosol (and other constituents), commonly used to quantify aerosol impacts (e.g., Shindell et al., 2009) and in international assessment (e.g., IPCC, 2013). Generally, DRF quantifies the change in DRE over time which will induce a change in global temperatures. In the context of the IPCC climate assessments, this time horizon has been specified as pre-industrial (1750) to present-day (IPCC, 1990, 1995, 2001, 2007, 2013). In this IPCC framework, the radiative forcing has also been restricted to “denote an externally imposed perturbation” (IPCC, 2001), and so excludes feedbacks resulting from a changing climate itself. Both of these more specific definitions of radiative forcing have been widely adopted by the atmospheric science and climate communities. The most recent IPCC report (IPCC, 2013) also describes an alternate...
effective radiative forcing (ERF), which allows physical variables (e.g. the temperature profile) “to respond to perturbations with rapid adjustments”; this does not generally include feedbacks or responses to climate change, but in the case of aerosols, now includes the “semi-direct effect”. The DRF includes both anthropogenic forcing driven by the rise in human emissions and land use change as well natural forcing associated with changes in solar flux and volcanic emissions. This was not clear in a previous definition of aerosol forcing in Sect. 2.4 of the 2007 IPCC report: “the direct RF only considers the anthropogenic components.” There are very few natural aerosol forcers, with volcanoes, which are sporadic in nature and therefore difficult to compare with other forcers, as the primary example. However, anthropogenic land use change and anthropogenically driven changes in the chemical environment can both affect natural aerosols, thus constituting a forcing ignored by this second, incomplete definition.

Climate feedbacks can also drive changes in natural aerosols; for example, rising carbonaceous aerosol emissions associated with enhanced fire activity (Spracklen et al., 2009; Westerling et al., 2006), the impacts of CO2 fertilization on aerosol precursor emissions (Heald et al., 2009), increases in biogenic aerosol formation associated with temperature-driven increases in biogenic volatile organic compound (VOC) emissions (Tsigeridis and Kanakidou, 2007; Heald et al., 2008), or trends in dust emissions associated with changes in vegetation or wind speed (Mahowald et al., 2006; Ridley et al., 2014). Carslaw et al. (2010) suggest that changes in natural aerosols, largely driven by climate feedbacks, may result in radiative perturbations of up to ±1 W m⁻². Previous studies have attempted to differentiate the role of climate feedbacks on biogeochemical cycles and atmospheric chemistry (Raes et al., 2010; Carslaw et al., 2010); however, these investigations require characterization of the climate response via a coupled chemistry–climate model.

The DRF metric is relatively simple to estimate in atmospheric models (in that it does not require quantification of the climate response) and enables quantitative comparison of various anthropogenic forcing mechanisms on the Earth’s energy budget. The aerosol DRF reflects both the change in primary aerosol emissions from anthropogenic activity and the impacts of the changing chemical environment (due to anthropogenic emissions) on secondary aerosol formation. The radiative impacts of natural aerosol are typically reflected in DRE, not DRF. Observations (from satellite or surface sun photometers) characterize a total DRE of present-day aerosol; to estimate DRF the anthropogenic fraction is assumed (Yu et al., 2006; Bellouin et al., 2005). Indeed, as a result of this, observational studies such as Bellouin et al. (2013) have provided a clear contrast between DRE and DRF. Nevertheless, this distinction between DRE and DRF is sometimes confused in the literature (Jo et al., 2013; Heald et al., 2005; Liao et al., 2004; Kim et al., 2006; Artaxo et al., 2009; Massoli et al., 2009; Ma et al., 2012; Athanasopoulou et al., 2013; Goto et al., 2008), where the presence of any aerosol is assumed to imply a DRF. The radiative imbalance associated with the presence of these aerosols is only a DRF if pre-industrial concentrations were zero. However, the distinction between DRE and DRF remains somewhat murky, particularly when considering secondary aerosol formation. For example, changes in the chemical formation of biogenic secondary organic aerosol (SOA) due to changes in anthropogenic nitrogen oxide (NOx) emissions qualifies as a DRF, but similar changes induced by changes in lightning NOx sources (due to a climate feedback) do not. In this study our objective is to globally quantify and contrast these two metrics.

2 Model description

Reducing the uncertainty associated with aerosol radiative forcing requires models that are well tested against observations and that include the capacity to simulate radiative impacts. The temporal matching of observations and simulation, which is only possible using a chemical transport model (CTM) driven by assimilated meteorology (or a GCM nudged towards analyzed meteorology), is critical to the accurate evaluation of a simulation of short-lived species. Therefore, a CTM with an online coupled radiative transfer model is the most appropriate model configuration for consistently evaluating aerosol loading and direct radiative impacts.

Here we integrate the rapid radiative transfer model for GCMs (RRTMG) online within the global GEOS-Chem chemical transport model (www.geos-chem.org), a configuration referred to as GC–RT, to calculate the radiative fluxes associated with atmospheric aerosols. RRTMG uses the correlated-k method to calculate long-wave (LW) and shortwave (SW) atmospheric fluxes. Further details on the aerosol simulation, emissions, optical properties, RRTMG, and the implementation in GEOS-Chem are provided in what follows.

2.1 GEOS-Chem

We use v9-01-03 of GEOS-Chem driven by GEOS-5 assimilated meteorology from the Global Model and Assimilation Office for the year 2010 at a horizontal resolution of 2° × 2.5° and 47 vertical levels.

The GEOS-Chem oxidant-aerosol simulation includes H₂SO₄–HNO₃–NH₃ aerosol thermodynamics coupled to an ozone–NOx–hydrocarbon–aerosol chemical mechanism (Park et al., 2004, 2006). We use the standard bulk aerosol scheme, where all aerosols are described with one or more log-normal size bins. The ISORROPIA II thermodynamic equilibrium model (Fountoukis and Nenes, 2007) calculates the partitioning of total ammonia and nitric acid between the gas and particle (fine mode only) phases. The model scheme
also includes organic aerosol (OA) and black carbon (BC) (Park et al., 2003), sea salt aerosol (2 size bins) (Alexander et al., 2005; Jaegle et al., 2010), and soil dust (4 size bins) (Fairlie et al., 2007; Ridley et al., 2012). The organic matter to organic carbon ratio for primary organic aerosol (POA) is assumed to be 2. SOA is produced from the oxidation of biogenic hydrocarbons following the Chung and Seinfeld (2002) two-product model scheme, with the addition of isoprene SOA (Henze and Seinfeld, 2006) and aromatic SOA (Henze et al., 2008). Of these three categories of SOA, only yields assumed to be 2. SOA is produced from the oxidation of biogenic hydrocarbons following the Chung and Seinfeld (2002) two-product model scheme, with the addition of isoprene SOA (Henze and Seinfeld, 2006) and aromatic SOA (Henze et al., 2008). Of these three categories of SOA, only yields

Global anthropogenic emissions for 2010 are based on the year 2000 EDGAR v3.2 inventory (Olivier et al., 2001) for SO\textsubscript{X}, NO\textsubscript{X}, and CO, and the RETRO inventory (Schultz, 2007) for VOCs. BC and primary OC emissions are taken from Bond et al. (2007). Both natural and anthropogenic (largely agricultural) ammonia emissions follow the global inventory of Bouwman et al. (1997) with seasonal variation specified by Park et al. (2004). Global anthropogenic emissions are overwritten by regional inventories as described by van Donkelaar et al. (2008). We scale all regional and global anthropogenic emissions from their respective base year to 2006, the last year of available statistics (van Donkelaar et al., 2008). We use global biomass burning emissions from the monthly GFED3 inventories (van der Werf et al., 2010). Emissions associated with biofuel use and agricultural waste, are globally fixed (Yevich and Logan, 2003) with seasonality but no interannual variability. Biogenic VOC emissions are predicted interactively in GEOS-Chem using the MEGAN2 scheme (Guenther et al., 2006). Emissions of DMS, NO\textsubscript{X}, from lightning and soils, dust, and sea salt depend on meteorology and are computed online in the model as described by Pye et al. (2009). Tropospheric methane concentrations are fixed at 2007 values from the NOAA CCGC cooperative air sampling network (4 latitude bands, mean concentrations ranging from 1733 to 1856 ppb).

Anthropogenic emissions of ozone and aerosol precursors (SO\textsubscript{X}, NO\textsubscript{X}, NH\textsubscript{3}, BC, OC, CO, and VOCs) for the year 2100 follow the RCP 4.5 scenario as implemented by Holmes et al. (2013). These include fossil fuel, biofuel, and agricultural emissions; all other natural and fire emissions, as well as methane concentrations, are identical in the 2100 and 2010 simulations performed here. We note that projections of aerosol precursor emissions show similar trajectories in all four of the RCP scenarios, with the exception of ammonia which is projected to remain reasonably constant in the RCP 4.5 scenario but projected to rise in all other scenarios. Global emission totals for aerosols and their precursors in 2010 and 2100 are given in Table 1.

Recent versions of the GEOS-Chem standard aerosol simulation have been extensively tested against airborne (van Donkelaar et al., 2008; Heald et al., 2011; Wang et al., 2011), shipborne (Lapina et al., 2011), and surface-site (Zhang et al., 2012; Heald et al., 2012) mass-concentration measurements as well as aerosol deposition measurements (Fisher et al., 2011) and satellite and ground-based observations of AOD (Ridley et al., 2012; Jaegle et al., 2010; Ford and Heald, 2012).

### Table 1. Annual global aerosol or aerosol precursor emissions for used in GC–RT.

<table>
<thead>
<tr>
<th></th>
<th>Total emissions (2010)</th>
<th>Anthropogenic emissions\textsuperscript{a} and percent of total in brackets (2010)</th>
<th>Total emissions (2100)</th>
</tr>
</thead>
<tbody>
<tr>
<td>SO\textsubscript{X} (Tg Syr\textsuperscript{−1})</td>
<td>63.8</td>
<td>53.3 (83 %) 19.5</td>
<td></td>
</tr>
<tr>
<td>NO\textsubscript{X} (Tg Nyr\textsuperscript{−1})</td>
<td>49.4</td>
<td>31.8 (64 %) 28.0</td>
<td></td>
</tr>
<tr>
<td>NH\textsubscript{3} (Tg Nyr\textsuperscript{−1})</td>
<td>55.9</td>
<td>37.9 (68 %) 56.3</td>
<td></td>
</tr>
<tr>
<td>POA (Tg Cyr\textsuperscript{−1})</td>
<td>29.3</td>
<td>9.3 (32 %) 20.4</td>
<td></td>
</tr>
<tr>
<td>BC (Tg Cyr\textsuperscript{−1})</td>
<td>6.8</td>
<td>4.5 (66 %) 4.4</td>
<td></td>
</tr>
<tr>
<td>Sea Salt (Tg yr\textsuperscript{−1})</td>
<td>3544</td>
<td>– 3544</td>
<td></td>
</tr>
<tr>
<td>Dust (Tg yr\textsuperscript{−1})</td>
<td>1563</td>
<td>312 (20 %) 1563</td>
<td></td>
</tr>
</tbody>
</table>

\textsuperscript{a} Includes fossil fuel, biofuel, and agriculture (but not biomass burning).

### 2.2 Rapid radiative transfer model for GCMs (RRTMG)

RRTMG (Iacono et al., 2008) is a fast radiative transfer code that calculates long-wave and shortwave atmospheric fluxes using the correlated-k method (Lacis and Oinas, 1991). The absorption coefficients that are used to develop the code’s k-distributions are attained directly from the line-by-line radiative transfer model (LBLRTM) (Clough et al., 1992, 2005; Alvarado et al., 2013), which connects the spectroscopic foundation of RRTMG to high-spectral resolution validations done with atmospheric radiance observations. There are 16 bands in the long-wave RRTMG code and 14 bands in the shortwave code (extending from 230 nm through 56 µm); for boundaries see Mlawer et al. (1997) and Mlawer and Clough (1998). Modeled sources of extinction in RRTMG include H\textsubscript{2}O, O\textsubscript{3}, long-lived greenhouse gases, aerosols, ice and liquid clouds, and Rayleigh scattering. RRTMG was successfully incorporated into a number of GCMs (Iacono et al., 2003, 2008), including the ECMWF IFS, the NCEP GFS and the NCAR CAM5.

For cloudy cases, vertical overlap of cloudy layers is handled using the Monte Carlo independent column approximation (McICA; Pincus et al., 2003), which reduces the
Table 2. GC–RT speciated dry aerosol size and optical properties.

<table>
<thead>
<tr>
<th></th>
<th>Geometric radius ($r_g$) (µm)</th>
<th>Geometric SD ($σ_g$, log(µm))</th>
<th>Refractive index (550 nm)</th>
<th>Density (g cm$^{-3}$)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Sulfate, Nitrate and Ammonium</td>
<td>0.070</td>
<td>1.6</td>
<td>1.53–0.006i</td>
<td>1.7</td>
</tr>
<tr>
<td>OA</td>
<td>0.064</td>
<td>1.6</td>
<td>1.53–0.006i</td>
<td>1.8</td>
</tr>
<tr>
<td>BC</td>
<td>0.020</td>
<td>1.6</td>
<td>1.95–0.79i</td>
<td>1.8</td>
</tr>
<tr>
<td>Sea salt</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Accumulation</td>
<td>0.085</td>
<td>1.5</td>
<td>1.50–0.00000001i</td>
<td>2.2</td>
</tr>
<tr>
<td>Coarse</td>
<td>0.40</td>
<td>1.8</td>
<td>1.50–0.00000001i</td>
<td>2.2</td>
</tr>
<tr>
<td>Dust</td>
<td>7 bins: 0.015, 0.25, 0.40, 0.80, 1.5, 2.5, 4.0</td>
<td>2.2</td>
<td>1.56–0.0014i</td>
<td>2.5</td>
</tr>
</tbody>
</table>

Computational load for the treatment of complex vertically overlapping cloud to that of a model run for a simpler configuration (e.g., clear, single-layer clouds) by assigning statistically appropriate combinations of cloud layers to each spectral element in the calculation. This requires that comparisons between RRTMG and observed fluxes and heating rates average a sufficiently large number of data points (in space and/or time) to remove the unbiased noise introduced by the McICA approximation. This random noise in the radiative fluxes and heating rates introduced by McICA has been shown to have statistically insignificant effects on GCM simulations (e.g., Barker et al., 2008).

RRTMG has been shown to be highly accurate in tests against reference radiative transfer calculations as part of the Continual Intercomparison of Radiation Codes (CIRC) project (Oreopoulos and Mlawer, 2010; Oreopoulos et al., 2012), which included numerous evaluations of the radiative effects due to aerosols.

2.3 Integration of GEOS-Chem and RRTMG (GC–RT)

RRTMG is called at a user-specified temporal frequency (3 h here) within GEOS-Chem and calculates instantaneous radiative fluxes in both the shortwave and long wave. In addition to total fluxes, GC–RT can calculate the SW or LW flux associated with a specific constituent of the troposphere (ozone, methane, sulfate, nitrate, ammonium, BC, OA, sea salt, dust, or total particulate matter) by calling RRTMG again with zero constituent concentration for the species of interest and differing the result. We describe the specification of a suite of relevant surface and atmospheric composition properties here.

All aerosols in GEOS-Chem are treated as externally mixed with log-normal size distributions and optical properties (including refractive indices and hygroscopic growth factors) defined by the Global Aerosol Data Set (GADS) database (Kopke et al., 1997), with recent updates (Drury et al., 2010; Jaegle et al., 2010; Ridley et al., 2012). We do not include any absorption enhancement from the coating of BC in these simulations. We update the BC density and refractive index to follow recommendations of Bond and Bergstrom (2006) and the shortwave refractive indices for dust following Sinyuk (2003). We also link simulated sulfate with the GADS water-soluble-type aerosol properties, not the sulfuric acid properties previously applied in the GEOS-Chem simulation, as recommended by Hess et al. (1998). Table 2 gives the relevant optical and size properties for dry aerosol. Mie code is used to calculate the resulting optical properties (including mass extinction efficiency) at 7 discrete relative humidities (RH) for each wavelength. The optical properties generated at 61 GADS wavelengths (from 250 nm to 40 µm) are spline interpolated to the 30 RRTMG wavelengths (230 nm to 56 µm) and stored in a look-up table which includes the mass extinction efficiency, the single scattering albedo (SSA) and asymmetry parameter ($g$). Aerosol properties at the two RRTMG wavelengths that fall outside of the range of the GADS wavelengths are fixed at the values of the shortest and longest wavelengths in GADS. The AOD at a specific wavelength is calculated within GEOS-Chem as a function of local relative humidity from the mass concentration and mass extinction efficiency according to the formulation of Tegen and Lacis (1996). RRTMG uses the AOD, SSA and asymmetry parameter for each aerosol type to calculate aerosol impacts on radiative fluxes in both the shortwave and long wave.

Figure 1. Annual mean surface direct albedo in the UV visible (left) and near infrared (right) from MODIS 2002–2007 observations.
Vertical profiles of a suite of greenhouse gases are fixed in GC–RT for both long-wave and shortwave flux calculations. These include: \( \text{N}_2\text{O} \) and stratospheric \( \text{CH}_4 \) (following the July 2012 zonal mean climatology from the TES instrument), CFCs (CFC-11, CFC-12, and CCL from UARS climatology and CFC-22 from the MIPAS climatology, all scaled to match surface values provided in IPCC, 2007) and \( \text{CO}_2 \) (set to 390 ppm globally). Zonally averaged tropospheric methane concentrations are fixed as described in Sect. 2.1. Tropospheric ozone is simulated interactively in GEOS-Chem; stratospheric concentrations are calculated based on the method used in FAST-J (Wild et al., 2000). Water vapor concentrations are specified according to the GEOS-5 assimilated meteorology. All gas-phase optical properties follow HITRAN (Rothman et al., 2005), as described by Iacono et al. (2008) and Mlawer and Clough (Mlawer and Clough, 1998).

Cloud optical properties required in RRTMG are calculated based on the liquid and ice visible optical depths from the GEOS-5 assimilated meteorology. We assume fixed effective radii of 14.2 \( \mu \text{m} \) for water droplets and 24.8 \( \mu \text{m} \) for ice particles based on a near-10-year average of MODIS-Aqua data (L. Oreopoulos, personal communication, 12 December 2012). From the optical depths and radii, we calculate the liquid and ice water path (LWP, IWP), the later based on the Fu (1996) treatment of ice cloud particles. The optical properties (extinction, SSA and asymmetry parameter) are then calculated from the LWP/IWP, the effective radii and a series of wavelength-specific parameters for liquid water (Hu and Stammes, 1993) and ice (Fu, 1996; Fu et al., 1998). Bosilovich et al. (2011) provide an overview of the water budget in the MERRA re-analysis (based on the same GEOS-5 assimilation system used here).

A climatology of surface albedo and emissivity is based on the multi-wavelength 8-day land composites from MODIS for 2002 through 2007 available at \( 0.05^\circ \) horizontal resolution (Schaaf et al., 2002; Wan et al., 2002). The MODIS land emissivity product (MOD11C2) provided at 6 long-wave wavelengths is interpolated to the 16 RRTMG wavelengths (values at wavelengths outside of the observed range are held fixed at the edge values). The emissivity of the ocean is set to 0.98. From Sidran (1981) this appears to be relatively constant between 0.2 and 12 \( \mu \text{m} \), after which it decreases slightly to 0.95 until beyond 40 \( \mu \text{m} \). The broadband direct and diffuse albedos for both the UV visible and visible IR are specified from the MODIS land albedo data (MCD43C3). Ocean albedo is held constant at 0.07 across all wavelengths (Stier et al., 2007). We do not include diurnal variation in surface albedos. Annual mean surface albedos used here are shown in Fig. 1. We note that aerosols produce a very modest

### Table 3. Global annual mean aerosol budget and impacts simulated for 2010 using GC–RT (comparisons with AEROCOM II means from Myhre et al. (2013) in round brackets; comparisons with AEROCOM I medians from Kinne et al. (2006) in square brackets). Note anthropogenic here does not include biomass burning.

<table>
<thead>
<tr>
<th></th>
<th>Total</th>
<th>Sulfate</th>
<th>Nitrate</th>
<th>Ammonium(^a)</th>
<th>BC</th>
<th>OA(^b)</th>
<th>Sea Salt</th>
<th>Dust</th>
</tr>
</thead>
<tbody>
<tr>
<td>Burden [Tg]</td>
<td>1.27</td>
<td>0.26</td>
<td>0.35</td>
<td>0.10</td>
<td>2.01</td>
<td>3.94</td>
<td>22.9</td>
<td></td>
</tr>
<tr>
<td>Anthropogenic</td>
<td>0.60</td>
<td>0.82</td>
<td>0.82</td>
<td>0.57</td>
<td>0.21</td>
<td>0.0</td>
<td>0.20</td>
<td></td>
</tr>
<tr>
<td>burden [Tg]</td>
<td>0.76</td>
<td>(0.91 ± 0.24)</td>
<td>0.21</td>
<td>(0.29 ± 0.14)</td>
<td>0.29</td>
<td>0.057</td>
<td>(0.071 ± 0.036)</td>
<td>0.44</td>
</tr>
<tr>
<td>AOD, 550 nm</td>
<td>0.092</td>
<td>0.0154</td>
<td>0.0031</td>
<td>0.0041</td>
<td>0.0012</td>
<td>0.0147</td>
<td>0.032</td>
<td>0.021</td>
</tr>
<tr>
<td>Anthro AOD, 550 nm</td>
<td>0.023</td>
<td>(0.030 ± 0.01)</td>
<td>0.0092</td>
<td>(0.021 ± 0.009)</td>
<td>0.0025</td>
<td>(0.0056 ± 0.0027)</td>
<td>0.0034</td>
<td>0.0007</td>
</tr>
<tr>
<td>MIE = AOD/burden [m(^2)g(^{-1})]</td>
<td>6.3</td>
<td>(12.7 ± 8.6)</td>
<td>5.7</td>
<td>(9.8 ± 2.0)</td>
<td>5.9</td>
<td>5.9 (10.5 ± 3.9)</td>
<td>(8.9)</td>
<td>3.8</td>
</tr>
<tr>
<td>TOA DRE, all-sky [Wm(^{-2})]</td>
<td>-2.75</td>
<td>-0.54</td>
<td>-0.095</td>
<td>-0.14</td>
<td>0.10</td>
<td>-0.61</td>
<td>-1.10</td>
<td>-0.37</td>
</tr>
<tr>
<td>LW</td>
<td>-3.01</td>
<td>-0.55</td>
<td>-0.097</td>
<td>-0.14</td>
<td>0.10</td>
<td>-0.63</td>
<td>-1.16</td>
<td>-0.53</td>
</tr>
<tr>
<td>TOA DRE, all-sky [Wm(^{-2})]</td>
<td>-1.83</td>
<td>-0.35</td>
<td>-0.067</td>
<td>-0.095</td>
<td>0.14</td>
<td>-0.42</td>
<td>-0.77</td>
<td>-0.26</td>
</tr>
<tr>
<td>LW</td>
<td>-2.03</td>
<td>-0.36</td>
<td>-0.069</td>
<td>-0.097</td>
<td>0.14</td>
<td>-0.43</td>
<td>-0.81</td>
<td>-0.40</td>
</tr>
<tr>
<td>TOA DRE, all-sky [Wm(^{-2})]</td>
<td>-0.57</td>
<td>(±0.67 ± 0.18)</td>
<td>-0.29</td>
<td>-0.079</td>
<td>-0.11</td>
<td>0.06</td>
<td>-0.075</td>
<td>0.0</td>
</tr>
<tr>
<td>TOA DRE, all-sky [Wm(^{-2})]</td>
<td>-0.36</td>
<td>(±0.32 ± 0.16)</td>
<td>-0.20</td>
<td>(±0.32 ± 0.11)</td>
<td>-0.055</td>
<td>0.078</td>
<td>-0.055</td>
<td>-0.003</td>
</tr>
</tbody>
</table>

\(^a\) Contributions from ammonium are included in sulfate and nitrate in AeroCom II. \(^b\) We sum POA and SOA mean model values form AeroCom II. \(^c\) Taken from Fig. 7 of Myhre et al. (2013).
warming in the shortwave over the most highly reflective hot spots over North Africa (El Djouf, Erg of Bilma, and Great Sand Sea), the Middle East, and Greenland (Fig. 1).

2.4 GC–RT simulations

To characterize the radiative impact of present-day aerosols, we perform two simulations: a baseline 2010 simulation and a second identical simulation with zero anthropogenic emissions (biomass burning emissions are not included as anthropogenic, see details in Sect. 3). We zero out anthropogenic emissions.

![Diagram showing annual mean AOD, shortwave TOA direct radiative effect, and long-wave TOA direct radiative effect for different aerosol types](image)

**Figure 2.** Annual mean AOD (left), shortwave TOA clear-sky direct radiative effect (center) and long-wave TOA clear-sky direct radiative effect (right) simulated by GC–RT for 2010. Color bars are saturated at respective values.
emissions for 1750 following Dentener et al. (2006) who suggest that fossil fuel emissions are negligible for this time period. We neglect here the biofuel emissions for 1750, which are small for carbonaceous aerosols (12% of present-day emissions) and negligible for all other species according to Dentener et al. (2006). The difference between these simulations provides an estimate of the anthropogenic contribution to the burden, AOD, and ultimately the DRF (which is defined under a fixed climate). While DRF for greenhouse gases is calculated at the tropopause after stratospheric temperature adjustment, “for most aerosol constituents, stratospheric adjustment has little effect on the RF, and the instantaneous RF at either the top of the atmosphere or the tropopause can be substituted” (IPCC, 2007). Estimates of the anthropogenic fraction of global dust emissions due to agricultural activities (e.g., plowing, grazing, irrigation) vary widely (e.g., Tegen et al., 2004; Mahowald and Luo, 2003); here we assume that 20% of all dust is of anthropogenic origin; the anthropogenic contribution likely varies regionally, we assume a fixed fraction here.

3 Results

Table 3 summarizes the GC–RT aerosol simulation and compares these results to the AeroCom I and AeroCom II model means where available (Myhre et al., 2013; Kinne et al., 2006). Global annual mean burdens for OA and dust are similar (within 20 and 15%, respectively) to the model medians from the AeroCom I models for the year 2000 (Kinne et al., 2006), whereas, the GC–RT burden of BC is about half of the AeroCom I model median value. Koch et al. (2009) show that the AeroCom I models are on average a factor of 8 times larger than BC concentrations measured aboard aircraft over the Americas, but underestimate BC at high northern latitudes and in Asia. Schwarz et al. (2010) show that the AeroCom I models generally overestimate BC (by an average a factor of 5) in remote regions based on comparisons with the HIPPO airborne observations over the Pacific. Sulfate is also 32% lower than the AeroCom I median; however, the sum of inorganic aerosol (sulfate, nitrate and ammonium) in GEOS-Chem is within 5% of the sulfate-only value for AeroCom I. The burden of sea salt is significantly lower (∼40%) than AeroCom I; however, the sea salt in GEOS-Chem has recently been evaluated and refined based on comparisons with satellite and in situ observations (Jaegle et al., 2010). The calculated global mean mass extinction efficiency (MEE) for all aerosol are lower than the AeroCom II model means; however, the wide range of model estimates given by Myhre et al. (2013) suggests that these are not well constrained in the AeroCom II models. In fact this large range in model MEE suggests that differences in model treatments of aerosol removal, size, and optical properties (including water uptake) can lead to at least a factor of 2 difference in model estimates of aerosol radiative fluxes.

Lower aerosol MEE in GC–RT translate to a lower global mean mid-visible AOD of 0.092 compared to the AeroCom I model mean of 0.127 and satellite-based estimates (∼0.15) (Kinne et al., 2006). Figure 2 shows the geographical distribution of annual mean AOD for present-day, over half of which is attributed to dust and sea salt. Sulfate and OA contribute a further 33%, with nitrate, ammonium, and BC making minor near-source contributions to the global AOD. The GC–RT anthropogenic AOD is 77% of the mean of the
AeroCom II models. Similarly, the GC–RT estimate of clear-sky (−0.57 W m\(^{-2}\)) global mean TOA aerosol direct radiative forcing (DRF) is 15 % lower (less cooling) than the AeroCom II model mean.

The IPCC (2007) estimates the aerosol DRF at −0.5 W m\(^{-2}\) (with a range of −0.9 to −0.1 W m\(^{-2}\)) in 2005 based on both models and satellite measurement. The model-only value (adjusted for dust and nitrate which is not included in all models) is lower at −0.4 W m\(^{-2}\). Our global-mean all-sky value for 2010 (−0.36 W m\(^{-2}\)) is similar. If we double BC absorption, as a crude approximation of internal mixing, then the 2010 DRF would be −0.28 W m\(^{-2}\), a 20 % change from our base estimate. We do not include any biomass burning sources in our DRF calculation due to the uncertainty in attributing the anthropogenic fraction. Anthropogenic modulation of biomass burning emissions is driven by changing agricultural practices, land clearing, and human fire suppression but not by climate change (this constitutes a feedback). We use an additional simulation to estimate a DRE of −0.19 W m\(^{-2}\) from all biomass burning particles in 2010 (−0.23 W m\(^{-2}\) from OA, +0.06 W m\(^{-2}\) from BC and −0.02 W m\(^{-2}\) from inorganic aerosol). The IPCC (2007) estimate of the aerosol DRF of biomass burning linked with human activities is +0.03 W m\(^{-2}\), indicating that the net impact of biomass burning aerosol is treated as more absorbing than our estimate or that the amount and spatial distribution of the aerosol relative to underlying reflecting clouds and areas with high-surface albedo differ.

All-sky DRF in GC–RT is 63 % of the estimate of clear-sky DRF. The global annual mean cloud fraction in GC–RT for 2010 is 60 %. This implies a global-mean cloud-sky TOA DRF of −0.22 W m\(^{-2}\), a lower value (less cooling), consistent with a shading of scattering aerosols below clouds and enhanced absorption from BC above clouds. The all-sky to clear-sky ratio is typically ∼ 50 % for the AeroCom II models (Myhre et al., 2013). Our slightly higher fraction is likely the result of the lower estimated BC loading.

Table 3 shows that the total TOA radiative effects are dominated by aerosol impacts in the shortwave (visible UV) wavelengths, where they reflect solar radiation and cool the Earth (with the exception of BC which absorbs solar radiation). Modest warming (< 10 % of the cooling effect) results from scattering (by large particles such as dust) or absorption in the long-wave (IR) wavelengths (Fig. 2). This indicates that a SW-only aerosol DRE estimate would overestimate the cooling effect by ∼ 5–10 %.

Figure 3 contrasts the GC–RT estimates of all-sky DRE and DRF. The DRE represents a change in radiative balance from pre-industrial to present-day (2010) with fixed climate, which reflects the rise in anthropogenic emissions (and anthropogenic land use, not considered here). Conversely, the DRE represents the total present-day radiative impact from all aerosols in the atmosphere, including those of natural origin. As a result, aerosols that are dominated by anthropogenic sources (e.g., nitrate) show a similar DRE and DRF, whereas natural aerosols (e.g., sea salt) have a large DRE but zero DRF. The estimate of total aerosol DRE (−1.83 W m\(^{-2}\)) is more than 5 times the value of the DRF (−0.36 W m\(^{-2}\)) in 2010, thus the global radiative impact of “natural” aerosol is more than 4 times that of anthropogenic aerosol perturbation. Rap et al. (2013) estimated an all-sky natural aerosol DRF of −0.81 W m\(^{-2}\) from DMS sulfate, sea salt, terpene SOA, and wildfire alone.

Figure 4 shows the zonal distribution of the GC–RT simulated radiative impacts. We see that the forcing is highly concentrated at northern mid-latitudes, driven by anthropogenic
emissions in North America, Europe, and Asia, in agreement with previous model studies (Myhre et al., 2013). Modest radiative impacts in the Arctic are the result of spring/summer Eurasian sulfate pollution, during seasons of minimum snow cover. Conversely, the DRE is distributed throughout the world with maxima associated with not only anthropogenic emissions in the NH, but also with sea salt emissions in the Southern Ocean, biomass burning in the tropics and Arctic, and dust from Africa and Asia. Warming dominates over the deserts in the Sahara, the Middle East, and the highly reflective regions at high northern latitudes; shortwave scattering prevails in all other regions in the simulation.

Similarly, Fig. 5 shows that the global mean DRF from aerosols is largest in boreal springtime (due to sulfate), whereas the DRE is more uniform throughout the year, with summertime peaks due to OA compensated by wintertime enhancements in ammonium nitrate and sea salt.

Figure 3 also shows an estimate of the aerosol all-sky DRF in 2100 from a GC–RT simulation based on the RCP 4.5 emissions scenario. Due to the steep decline in SO2 emissions (Table 1), the global mean sulfate DRF drops by ~85% from 2010 to −0.031 Wm−2. As shown by Pinder et al. (2007), reductions in sulfate can lead to enhanced ammonium nitrate formation in environments with abundant ammonia supply. Given that ammonia emissions in 2100 are predicted to be approximately equivalent to those in 2010 in the RCP4.5 scenario, nitrate DRF in 2100 increases by 60% to −0.088 Wm−2. Overall forcing from ammonium decreases (the decline in ammonium sulfate outweighing the increase in ammonium nitrate) to −0.046 Wm−2. Finally, changes in BC and POA are more modest, but the magnitude of DRF decreases for both (to 0.056 and −0.053 Wm−2, respectively). Dust forcing is maintained at 2010 values. Overall the clear-sky TOA DRF in 2100 (−0.22 Wm−2) dropped by 39% from 2010. Smith and Bond (2013) also estimate a precipitous decline (62%) in aerosol (carbonaceous and sulfate only) radiative forcing from 2000 to 2100 when including both the direct and indirect forcing.

4 Discussion and conclusions

We use the newly online coupled GEOS-Chem–RRTMG (GC–RT) model to estimate the direct radiative effect and direct radiative forcing of atmospheric aerosols. The global TOA DRE is over five times the cooling estimated as the DRF. This illustrates that tropospheric aerosols exert a large influence on the global energy balance. However, the quantification of DRE and DRF is highly uncertain and the estimate of the ratio between them is specific to this model configuration and the concurrent assumptions. While the aerosol burden in the GEOS-Chem model is well tested against observations (e.g., Zhang et al., 2012; Fisher et al., 2011; Ridley et al., 2012; Jaegle et al., 2010; Lapina et al., 2011), the uncertainties associated with aerosol properties and radiative effects require further investigation.

Uncertainty in our estimates of DRF is dominated by the uncertainty in the MEE (as shown by the AeroCom II study of Myhre et al., 2013). Assumptions regarding size, water uptake and absorption efficiency (e.g., the prevalence of brown carbon (Andreae and Gelencser, 2006)) all contribute to this. We do not treat absorption enhancement from the coating of BC (Jacobson, 2000), the importance of which is unclear (Cappa et al., 2012; Lack et al., 2012), therefore both DRE and DRF values may underestimate absorption. In addition to the uncertainty associated with aerosol optics, our simulations likely underestimate the “anthropogenically controlled” SOA in the atmosphere, which is estimated to contribute −0.26 Wm−2 (Spracklen et al., 2011) of direct cooling. The global source of dust arising from anthropogenic activity is also poorly constrained.
Uncertainties in our estimate of DRE are likely even larger than uncertainties on DRF. This is the result of (1) better constraints on anthropogenic sectoral emissions (e.g. mobile sources, power generation); (2) a poor understanding of natural particle emissions from ecosystems, both marine and terrestrial, for example, terrestrial primary biological aerosol particles (e.g., Heald and Spracklen, 2009) marine OA and methane sulfonate (Heintzenberg et al., 2000) – none of which are included in this GC–RT simulation; and (3) the geographical extent and remoteness of regions impacted by natural aerosols and hence a lack of measurement constraint (and thus well-tested models).

We use Fig. 6 to distinguish the definition of DRF and DRE of aerosols. The DRF is linked to human-driven activities and neglects all feedbacks. The DRF accounts for the long-term direct drivers of the climate system, which are primarily anthropogenic. In the case of aerosols, this includes not only emission of primary particles (e.g., soot) and secondary aerosol precursors (e.g., SO2), but also the changes to the chemistry that governs the formation of secondary aerosols. To the extent that this highly uncertain chemistry (including peroxy radical chemistry and particle acidity) is included in models, it also contributes to the DRF (and its uncertainty). This complicates the separation and interpretation of "anthropogenic" and "natural" aerosol forcing. Anthropogenic land use change is also a driver of aerosol DRF (for example via changes in dust, soil/vegetation, and ammonia emissions associated with crop expansion), although it is rarely included in DRE assessments.

Figure 6a also illustrates the potential difference between an activity-based approach and an agent-based approach to estimating RF. When associating a forcing with an agent, in this case an aerosol, all the radiative impacts described in Figure 6a are ascribed to this agent. This mirrors the approach taken in previous IPCC reports (e.g., IPCC, 2007, 2001). Alternatively, an activity-based approach, as suggested by Shindell et al. (2009) and used in the most recent IPCC report (IPCC, 2013) can attribute the forcing with (controllable) behaviors such as emissions or land use change.

Clearly, the DRF provides an incomplete estimate of the global radiative imbalance imposed by changing aerosol abundance. Indeed, as anthropogenic emissions of aerosols and their precursors are expected to continue to decline globally (van Vuuren et al., 2011), the aerosol DRF will continue to decrease as shown here and by Smith and Bond (2013), diminishing its relevance as a climate metric. At the same time feedbacks from climate change on aerosol are likely to grow. This possibility was highlighted in the IPCC (2013): "climate change may alter natural aerosol sources as well as removal by precipitation". It is therefore critical that we expand our set of metrics to address the many factors that the DRF neglects by design. Figure 6b shows a more comprehensive set of drivers for changing aerosol abundance, including climate feedbacks and natural emissions. In addition to the direct effect of aerosols on climate (the focus of this study), the indirect effects of aerosols, both the conventional aerosol–cloud interactions and the less well-constrained effect of aerosols on biogeochemistry (Mahowald, 2011), can feed back on aerosol abundance via climate impacts. The relative importance of these feedbacks and forcings on the global radiative flux imbalance over time is unclear and deserves further investigation. The challenge of attributing a feedback is also non-trivial (e.g., to what agent does one associate a temperature-driven change in aerosol abundance?). Inherently, the calculation of DRE over time (with varying climate) includes the potential to quantify (though not attribute) these effects. Furthermore, estimates of the aerosol indirect effect are very sensitive to the pre-industrial aerosol burden, as originally shown by Menon et al. (2002) and more recently discussed by Carslaw et al. (2013), for which little observational constraint exists. Thus, two climate models with identical aerosol DRFs could provide very different estimates of the aerosol indirect aerosol forcing due to differences in the pre-industrial (natural) aerosol burden. Comparing the pre-industrial DRE simulated in models would provide a first step towards identifying these key differences.

Quantifying and reporting the instantaneous DRE in global models is a simple and necessary first step in going beyond the DRF metric for aerosols. While it may not directly serve as a policy tool, the DRE is more easily tested against observations (e.g., satellites), is a more thorough gauge for model comparisons, and offers a more complete picture of aerosols in the climate system. As such, it is an important complement to the DRF for advancing our understanding and predictions of the global aerosol burden and how it may counteract future trends in greenhouse gas warming.

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