Multi-model study of mercury dispersion in the atmosphere: atmospheric processes and model evaluation

The MIT Faculty has made this article openly available. Please share how this access benefits you. Your story matters.
Multi-model study of mercury dispersion in the atmosphere: atmospheric processes and model evaluation

Oleg Travnikov, Hélène Angot, Paulo Artaxo, Mariantonia Bencardino, Johannes Bieser, Francesco D’Amore, Ashu Dastoor, Francesco De Simone, María del Carmen Diéguez, Aurélien Dommergue, Ralf Ebinghaus, Xin Bin Feng, Christian N. Gencarelli, Ian M. Hedgecock, Olivier Magand, Lynwill Martin, Volker Matthias, Nikolay Mashyanov, Nicola Pirrone, Ramesh Ramachandran, Katie Alana Read, Andrei Ryjkov, Noelle E. Selin, Fabrizio Sena, Shaojie Song, Francesca Sprovieri, Dennis Wip, Ingvar Wångberg, and Xin Yang

1 Meteorological Synthesizing Centre – East of EMEP, Moscow, Russia
2 University Grenoble Alpes, CNRS, IRD, IGE, Grenoble, France
3 University of Sao Paulo, Sao Paulo, Brazil
4 CNR Institute of Atmospheric Pollution Research, Rende, Italy
5 Institute of Coastal Research, Helmholtz-Zentrum Geesthacht, Geesthacht, Germany
6 Air Quality Research Division, Environment and Climate Change Canada, Dorval, Québec, Canada
7 INIBIOMA-CONICET-UNComa, Bariloche, Argentina
8 CNRS, Laboratoire de Glaciologie et Géophysique de l’Environnement, Grenoble, France
9 Institute of Geochemistry, State Key Laboratory of Environmental Geochemistry, Chinese Academy of Sciences, Guiyang, China
10 Cape Point GAW Station, Climate and Environment Research & Monitoring, South African Weather Service, Stellenbosch, South Africa
11 St. Petersburg State University, St. Petersburg, Russia
12 CNR Institute of Atmospheric Pollution Research, Rome, Italy
13 Institute for Ocean Management, Anna University, Chennai, India
14 NCAS, University of York, York, UK
15 Department of Earth, Atmospheric and Planetary Sciences, Massachusetts Institute of Technology, Cambridge, MA, USA
16 Institute for Data, Systems, and Society, Massachusetts Institute of Technology, Cambridge, MA, USA
17 Joint Research Centre, Ispra, Italy
18 Department of Physics, University of Suriname, Paramaribo, Suriname
19 IVL Swedish Environmental Research Institute, Göteborg, Sweden
20 British Antarctic Survey, Cambridge, UK

*now at: Institute for Data, Systems and Society, Massachusetts Institute of Technology, Cambridge, MA, USA

Correspondence to: Oleg Travnikov (oleg.travnikov@msceast.org)

Received: 14 October 2016 – Discussion started: 31 October 2016
Revised: 3 March 2017 – Accepted: 14 March 2017 – Published: 24 April 2017
Abstract. Current understanding of mercury (Hg) behavior in the atmosphere contains significant gaps. Some key characteristics of Hg processes, including anthropogenic and geogenic emissions, atmospheric chemistry, and air–surface exchange, are still poorly known. This study provides a complex analysis of processes governing Hg fate in the atmosphere involving both measured data from ground-based sites and simulation results from chemical transport models. A variety of long-term measurements of gaseous elemental Hg (GEM) and reactive Hg (RM) concentration as well as Hg wet deposition flux have been compiled from different global and regional monitoring networks. Four contemporary global-scale transport models for Hg were used, both in their state-of-the-art configurations and for a number of numerical experiments to evaluate particular processes. Results of the model simulations were evaluated against measurements. As follows from the analysis, the interhemispheric GEM gradient is largely formed by the prevailing spatial distribution of anthropogenic emissions in the Northern Hemisphere. The contributions of natural and secondary emissions enhance the south-to-north gradient, but their effect is less significant. Atmospheric chemistry has a limited effect on the spatial distribution and temporal variation of GEM concentration in surface air. In contrast, RM air concentration and wet deposition are largely defined by oxidation chemistry. The Br oxidation mechanism can reproduce successfully the observed seasonal variation of the RM/GEM ratio in the near-surface layer, but it predicts a wet deposition maximum in spring instead of in summer as observed at monitoring sites in North America and Europe. Model runs with OH chemistry correctly simulate both the periods of maximum and minimum values and the amplitude of observed seasonal variation but shift the maximum RM/GEM ratios from spring to summer. O_3 chemistry does not predict significant seasonal variation of Hg oxidation. Hence, the performance of the Hg oxidation mechanisms under study differs in the extent to which they can reproduce the various observed parameters. This variation implies possibility of more complex chemistry and multiple Hg oxidation pathways occurring concurrently in various parts of the atmosphere.

1 Introduction

Mercury (Hg) is widely recognized as a toxic pollutant that is capable of long-range transport, bioaccumulation in ecosystems and biota, and adverse effects on human health and the environment. Although it is a natural element, its concentrations in the environment have been considerably enriched by human activities since pre-industrial times (Fitzgerald et al., 1998; Mason and and Sheu, 2002; Krabbenhoft and Sunderland, 2013). Once emitted to the atmosphere, Hg can be dispersed globally, impacting remote regions through deposition to aquatic ecosystems, transformation to a potent neurotoxic form (methylmercury), and bioaccumulation in food chains (Mahaffey et al., 2004; Sunderland et al., 2010; Mason et al., 2012). The character of Hg transport and fate in the atmosphere is largely determined by the properties of its various chemical forms. Mercury is emitted into the atmosphere from anthropogenic sources in the form of both gaseous elemental mercury (GEM) and oxidized Hg chemical compounds (Pirrone et al., 2010). The latter are typically divided into two operationally defined forms: gaseous oxidized mercury (GOM) and particle-bound mercury (PBM). In addition, GEM can also originate from natural geogenic and secondary sources (Mason, 2009). Reactive mercury (RM = GOM + PBM) can also be produced in the atmosphere from gas- and aqueous-phase oxidation of GEM (Lindberg and Stratton, 1998). Relatively stable and slightly soluble GEM can drift in the atmosphere for months, providing transport of Hg mass around the globe (Schroeder and Munthe, 1998). In contrast, RM is easily removed from the air by precipitation scavenging (wet deposition) or surface uptake (dry deposition) (Schroeder and Munthe, 1998; Gustin et al., 2012; Sather et al., 2013; Wright et al., 2014). GEM can also contribute to Hg dry deposition through air–surface exchange with various terrestrial and aquatic compartments (Zhang et al., 2009; Wang et al., 2014, 2016). Previously deposited Hg can also be reduced to its elemental form and re-emitted back to the atmosphere (Gustin, 2012; Qureshi et al., 2012).

Atmospheric redox chemistry plays an important role in Hg long-range dispersion and deposition. However, the particular mechanisms of Hg oxidation in the atmosphere are not well understood (Lin et al., 2006; Subir et al., 2011, 2012; Gustin et al., 2015; Ariya et al., 2015). Gaseous reactive halogens, in particular atomic bromine (Br), are believed to play a major role in atmospheric oxidation of GEM (Goodsite et al., 2004, 2012; Donohoue et al., 2006; Hynes et al., 2009). Observational evidence exists that Br-initiated chemistry is a dominant GEM oxidation pathway in some atmospheric environments, including the marine boundary layer, the polar regions, and the upper troposphere and lower stratosphere (Hedgecock and Pirrone, 2004; Holmes et al., 2009; Lyman and Jaffe, 2010; Obrist et al., 2011; Gratz et al., 2015). However, very limited data exist with respect to this mechanism in the global atmosphere (Kos et al., 2013). Nevertheless, use of the Br chemistry as the only oxidation pathway in a chemical transport model enables simulation of the Hg atmospheric cycle and reproduction of available observations (Holmes et al., 2010; Soerensen et al., 2010; Amos et al., 2012; Shah et al., 2016). However, in spite of theoretical doubts about the viability and significance of direct GEM oxidation by ozone (O_3) and hydroxyl radical (OH) under atmospheric conditions (Calvert and Lindberg, 2005; Hynes et al., 2009), numerous modeling studies using these reactions as the main GEM oxidation pathways in the free troposphere also demonstrate reasonable results compared with observed GEM concentration and wet deposition flux (Christensen et al., 2004; Travnikov and Ilyin, 2009; Pan et al.,...
different chemical oxidation mechanisms on the levels and spa-
tropogenic and natural/secondary emissions as well as dif-
ced numerical experiments have tried to evaluate the effect of an-
theses governing Hg fate in the atmosphere. A number of
ments from different monitoring networks to analyze pro-
ological and transport of atmospheric Hg as well as total Hg deposi-
hen et al. (2016) and provided important insights into the fate
sphere. They found that standard model simulations signifi-
cient evidence is still incomplete as to whether GEM or
2 Methods
2.1 Measurements
A variety of measured data were used to evaluate the model experiments. The measured dataset was based on the
global GMOS monitoring network for Hg (Sprovieri et al., 2016; 2017; GMOS, 2016), complemented by data from
the European Monitoring and Evaluation Programme (EMEP) regional network for Europe (Tørseth et al., 2012;
EMEP, 2016) and by data from the Mercury Deposition
Network of the National Atmospheric Deposition Program
(NADP/MDN) (Prestbo and Gay, 2009; NADP/MDN,
2016), the Atmospheric Mercury Network (AMNet) (Gay et
al., 2013; AMNet, 2016), and the Canadian National Atmo-
spheric Chemistry Database (NAtChem) (Cole et al., 2013;
Steffen et al., 2015; NatChem, 2016) for North America.
Available measurements of GEM, GOM, and PBM concentra-
tions in air were compiled, as well as wet deposition flux
measurements performed at ground-based sites in 2013. At
most of the sites of interest, the unspeciated measurements of
atmospheric Hg were performed as GEM (Gay et al., 2013;
Sprovieri et al., 2016; Angot et al., 2016). However, the sci-
ten observations of GEM and TGM commonly does not exceed a few percentage
points (Slemr et al., 2015), all unspeciated Hg measurements
were interpreted as GEM. Measured values of RM are used in
the study instead of observations of individual species GOM
and PBM. RM appears to be more valuable for the analysis
because measurements of the individual species are associ-
ated with higher uncertainties (Gustin et al., 2015; Weiss-
Penzi et al., 2015). Therefore, only sites with co-located
observations of GOM and PBM were used in the study.
The original measured data with high temporal resolution
were processed to obtain monthly and yearly mean values. Accord-
ing to accepted criteria, monthly averages were used for
the analysis when the original data covered at least 15
days of the month. Monthly averages were used both to gen-
rate yearly mean values and to characterize the seasonal
variation of the observed parameter. In both cases, only sites
with temporal coverage of at least 7 months were selected.
The characteristics of the selected sites measuring GEM,
RM, and wet deposition are given, respectively, in Tables S1,
2.2 Models

The model ensemble used in this study includes four chemical transport models simulating mercury on a global scale (Table 1). The models differ considerably in their general formulation, spatial resolution, and physical and chemical process parameterizations. The horizontal spatial resolution of the models ranges from 1 to 2.8$^\circ$ in latitude and longitude. The upper boundaries of the model domains vary from 10 hPa ($\sim$ 30 km) to 0.01 hPa ($\sim$ 80 km). Hence, the models cover the entire troposphere and at least a significant part of the stratosphere, which is enough to take into account most processes governing mercury dispersion in the atmosphere. Two of the models (GLEMOS, and GEOS-Chem) use offline meteorological data prepared by an external pre-processor, whereas the other two (GEM-MACH-Hg and ECHMERIT) generate the meteorological fields along with the simulation of pollutant transport. All the models used the same Hg anthropogenic emissions dataset (AMAP/UNEP, 2013a, b), with somewhat different speciations of mercury forms in the base case. In the other model experiments, the emissions speciation was harmonized by assuming all emissions to be a form of GEM (Table 2). By contrast, total values of natural and secondary emissions differed significantly among the models. Parameterization of these processes is an essential element of the model setup describing Hg cycling between the Earth’s surface and the atmosphere, which cannot be easily changed or removed without disturbing the entire model balance. The major chemical mechanisms in the standard model configuration in the base case were also essentially different. The base-case reactions of GLEMOS and ECHMERIT included Hg oxidation by O$_3$ and OH. The GEM-MACH-Hg chemical scheme was based on the reaction with OH, with Br chemistry applied in the polar regions. GEOS-Chem considered Br chemistry as the only Hg oxidation pathway in the gas phase. In addition, two of the four models (GLEMOS and ECHMERIT) also included Hg redox chemistry in the aqueous phase in cloud water. However, it should be noted that the numerical experiments in this study (see Sect. 2.4) were formulated to investigate particular processes when all models used the same chemical mechanisms. Considering these differences, the results, whether the models succeed or fail in reproducing the observed dependencies, appear to be particularly valuable. A more detailed description of the model parameterizations is given subsequently.

2.2.1 GLEMOS

GLEMOS is a multi-scale chemical transport model developed to simulate environmental dispersion and cycling of various chemicals including Hg based on the older MSCE-HM-Hem hemispheric model (Travnikov, 2005; Travnikov and Ilyin, 2009; Travnikov et al., 2009). The model simulates atmospheric transport, chemical transformation, and deposition of three Hg species (GEM, GOM, and PBM). Atmospheric transport of the tracers is driven by meteorological fields generated by the Weather Research and Forecast modeling system (WRF) (Skamarock et al., 2007) fed by operational analysis data from the European Centre for Medium-Range Weather Forecasts (ECMWF) (ECMWF, 2017). In the base configuration, the model grid had a horizontal resolution of 1$^\circ$ $\times$ 1$^\circ$. Vertically, the model domain reached 10 hPa and consisted of 20 irregular terrain-following sigma layers. The atmospheric chemical scheme included Hg oxidation and reduction chemical reactions in both the gaseous phase and aqueous phase of cloud water. The major chemical mechanisms in the gas phase included Hg oxidation by O$_3$ and OH radical, with the reaction rate constants from Hall (1995) and Sommar et al. (2001), respectively. The latter mechanism was scaled down by a factor of 0.1 in the cloud environment and below clouds to account for reduced photochemical activity (Seigneur et al., 2001). The O$_3$ and OH concentration fields were imported from MOZART (Emmons et al., 2010). Two-step gas-phase oxidation of GEM by Br was included as an option. Aqueous-phase reactions include oxidation by O$_3$, OH, and Cl and reduction by decomposition of sulfite complexes (Van Loon et al., 2000). The model distinguished between in-cloud and sub-cloud wet deposition of PBM and GOM based on empirical data. The dry deposition scheme was based on the resistance analogy approach (Wesely and Hicks, 2000). Prescribed fluxes of Hg natural and secondary emissions from soil and seawater were generated depending on Hg soil concentration, soil temperature, and solar radia-
Table 1. Characteristics of the participating global chemistry transport models.

<table>
<thead>
<tr>
<th>Model</th>
<th>GLEMOS</th>
<th>GEOS-Chem</th>
<th>GEM-MACH-Hg</th>
<th>ECHMERIT</th>
</tr>
</thead>
<tbody>
<tr>
<td><strong>Spatial resolution</strong></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Horizontal</td>
<td>1° × 1°</td>
<td>2.5° × 2°</td>
<td>1° × 1°</td>
<td>T42 (~2.8° × 2.8°)</td>
</tr>
<tr>
<td>Vertical</td>
<td>20 levels, top 10 hPa</td>
<td>47 levels, top 0.01 hPa</td>
<td>58 levels, top 7 hPa</td>
<td>19 levels, top 10 hPa</td>
</tr>
<tr>
<td><strong>Driving meteorology</strong></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Data support type</td>
<td>off-line</td>
<td>off-line</td>
<td>on-line</td>
<td>on-line</td>
</tr>
<tr>
<td>Meteorological driver</td>
<td>WRF/ECMWF</td>
<td>GEOS-FP</td>
<td>GEM</td>
<td>ECHAM5</td>
</tr>
<tr>
<td><strong>Anthropogenic emission</strong></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Global emission, t/y</td>
<td>1875</td>
<td>1875</td>
<td>1875</td>
<td>1875</td>
</tr>
<tr>
<td>Average speciation (base case)</td>
<td>81 : 15 : 4</td>
<td>81 : 19 : 0 (^a)</td>
<td>96 : 3 : 1</td>
<td>81 : 15 : 4</td>
</tr>
<tr>
<td><strong>Natural and re-emission</strong></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Global emission, t/y (base case)</td>
<td>3995</td>
<td>5070</td>
<td>3660</td>
<td>8600</td>
</tr>
<tr>
<td><strong>Gaseous chemistry (base-case reactions are in bold)</strong></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Reaction rates ((s^{-1}))</td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>(Hg^0 + Br \rightarrow HgBr)</td>
<td>(3.7 \times 10^{-13} ) g</td>
<td>(3.7 \times 10^{-13} ) g</td>
<td>(3.7 \times 10^{-13} ) g</td>
<td>(3.7 \times 10^{-13} ) g</td>
</tr>
<tr>
<td>(HgBr \rightarrow Hg^0 + Br)</td>
<td>(9.4 \times 10^{-2} ) s(^{-1})</td>
<td>(9.4 \times 10^{-2} ) s(^{-1})</td>
<td>(1.7 \times 10^{-1} ) s(^{-1})</td>
<td>(9.4 \times 10^{-2} ) s(^{-1})</td>
</tr>
<tr>
<td>(HgBr + Br \rightarrow Hg^0 + Br_2)</td>
<td>(3.9 \times 10^{-11} ) j</td>
<td>(3.9 \times 10^{-11} ) j</td>
<td>—</td>
<td>—</td>
</tr>
<tr>
<td>(HgBr + Y \rightarrow HgBrY, Y = Br, OH)</td>
<td>(2.5 \times 10^{-10} ) k</td>
<td>(2.5 \times 10^{-10} ) k</td>
<td>(2.5 \times 10^{-10} ) k</td>
<td>(2.5 \times 10^{-10} ) k</td>
</tr>
<tr>
<td>(Hg^0 + O_3 \rightarrow Hg(II))</td>
<td>(3.0 \times 10^{-20} ) l</td>
<td>—</td>
<td>(3.0 \times 10^{-20} ) l</td>
<td>(3.0 \times 10^{-20} ) l</td>
</tr>
<tr>
<td>(Hg^0 + OH \rightarrow Hg(II))</td>
<td>((0.9 – 8.7) \times 10^{-14} ) m</td>
<td>—</td>
<td>(3.0 \times 10^{-14} ) n</td>
<td>(8.7 \times 10^{-14} ) n</td>
</tr>
<tr>
<td><strong>Aqueous chemistry (in cloud water)</strong></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Oxidation agents</td>
<td>O_3, OH, HOCl/OCI(^-)</td>
<td>—</td>
<td>—</td>
<td>O_3, OH</td>
</tr>
<tr>
<td>Reduction agents</td>
<td>SO_3(^2)</td>
<td>—</td>
<td>—</td>
<td>—</td>
</tr>
</tbody>
</table>

<table>
<thead>
<tr>
<th><strong>Model</strong></th>
<th>GLEMOS</th>
<th>GEOS-Chem</th>
<th>GEM-MACH-Hg</th>
<th>ECHMERIT</th>
</tr>
</thead>
</table>

\(^a\) Dynamic gas–particle partitioning of RM in the atmosphere according to Amos et al. (2012). \(^b\) Prescribed fluxes from terrestrial and aquatic surfaces as a function of temperature and solar radiation; dynamic re-emission from snow. \(^c\) Prescribed fluxes from terrestrial surfaces as a function of temperature and solar radiation; dynamic fluxes from aquatic surfaces based on multi-media modeling. \(^d\) Prescribed fluxes from terrestrial surfaces as a function of solar radiation and leaf area index; dynamic re-emission from snow and aquatic surfaces. \(^e\) Prescribed fluxes from terrestrial surfaces as a function of temperature and solar radiation; dynamically calculated ocean emissions. \(^f\) Temperature and pressure dependence applied to most reactions; the reaction rates are given at 298 K and 1 atm. \(^g\) Donohue et al. (2006). \(^h\) Goodsite et al. (2012). \(^i\) Dibble et al. (2012). \(^j\) Balabanov et al. (2005). \(^k\) Goodsite et al. (2004). \(^l\) Hall (1995). \(^m\) Sommar et al. (2001) scaled down by a factor 0.1 in the cloud environment and below clouds to account for reduction of photochemical activity (Seigneur et al., 2001). \(^n\) Sommar et al. (2001) scaled down by a factor 0.34 to take into account possible dissociation/reduction reactions. \(^o\) Sommar et al. (2001). \(^p\) Parrella et al. (2012). \(^q\) Yang et al. (2005, 2010). \(^r\) Emmons et al. (2010).
Goodsite et al. (2012), Donohoue et al. (2006), and Babanov et al. (2005). The GEM surface fluxes included anthropogenic sources, biomass burning, and geogenic activities as well as bidirectional fluxes in atmosphere–terrestrial and atmosphere–ocean exchanges (Song et al., 2015). Biomass burning emissions were estimated using a global CO emissions database and an Hg/CO volume ratio of 1 × 10⁻⁷. Geogenic activities were spatially distributed based on the locations of Hg mines. For atmosphere–terrestrial exchange, GEOS-Chem treated evasion and dry deposition of GEM separately (Selin et al., 2008). Dry deposition was parameterized by a resistance-in-series scheme (Wesely, 1989). In addition, effective GOM uptake by sea-salt aerosols was also included over the ocean (Holmes et al., 2010). GEM evasion included volatilization from soil and rapid recycling of newly deposited Hg. The former was estimated as a function of soil Hg content and solar radiation. The latter was modeled by recycling a fraction of wet/dry deposited RM to the atmosphere as GEM immediately after deposition (60 % for snow-covered land and 20 % for all other land surfaces) (Selin et al., 2008). GEOS-Chem estimated atmosphere–ocean GEM exchange using a standard two-layer diffusion model. Ocean Hg in the mixed layer was assumed to interact not only with the atmospheric boundary layer but also with subsurface waters through entrainment and detrainment of the mixed layer and wind-driven Ekman pumping (Soerensen et al., 2010).

### 2.2.3 GEM-MACH-Hg

GEM-MACH-Hg is a new chemical transport model for Hg based on the GRAHM model developed by Environment and Climate Change Canada (Dastoor and Larocque, 2004; Dastoor et al., 2008; Durnford et al., 2010, 2012; Kos et al., 2013; Dastoor et al., 2015). GEM-MACH-Hg uses a newer version of the operational meteorological model of the Environment and Climate Change Canada. The horizontal resolution of the model is 1° × 1°. GEM is oxidized in the atmosphere by the OH radical. The reaction rate constant was obtained from Sommar et al. (2001) but was scaled down by a coefficient of 0.34 to account for possible dissociation and reduction reactions (Tossell et al., 2003; Goodsite et al., 2004). Gaseous oxidation of Hg by Br was modeled in the polar regions using reaction rate constants from Donohoue et al. (2006), Dibble et al. (2012), and Goodsite et al. (2004). Atmospheric mercury depletion events (AMDE) parameterization was based on Br production and chemistry and snow re-emission of GEM (Dastoor et al., 2008). OH fields were obtained from MOZART (Emmons et al., 2010), and BrO was derived from 2007–2009 satellite observations of BrO vertical columns. The associated Br concentration was then calculated from a photochemical steady state (Platt and Janssen, 1995). Dry deposition in GEM-MACH-Hg was based on the resistance approach (Zhang, 2001; Zhang et al., 2003). In the wet deposition scheme, GEM and GOM were partitioned between cloud droplets and air using a temperature-dependent Henry’s law constant. Total global emissions from natural sources and re-emissions of previously deposited Hg (from land and oceans) in GEM-MACH-Hg were based on global Hg budgets from Gbor et al. (2007), Shetty et al. (2008), and Mason (2009). Land-based natural emissions were spatially distributed according to the natural enrichment of Hg. Land re-emissions were spatially distributed according to historic Hg deposition and land-use type and were dependent on solar radiation and the leaf area index. Oceanic emissions depended on the distributions of primary production and atmospheric deposition.

#### 2.2.4 ECHMERIT

ECHMERIT is a global online chemical transport model derived from the ECHAM5 global circulation model, with a highly flexible chemistry mechanism designed to facilitate the investigation of atmospheric Hg chemistry (Jung et al., 2009; De Simone et al., 2014, 2015, 2017). The model uses the same spectral grid as ECHAM. The standard horizontal resolution of the model is T42 (approximately 2.8° × 2.8°), whereas in the vertical dimension the model is discretized with a hybrid-sigma pressure system with non-equidistant levels up to 10 hPa. The base chemical mechanism includes GEM oxidation by OH and O₃ in the gaseous and aqueous phases. Reaction rate constants were obtained from Sommar et al. (2001), Hall (1995), and Munthe (1992). OH and O₃ concentration fields were imported from MOZART.
Hg emissions from natural and secondary sources were fully or partly represented in the models as bi-directional air–surface exchange fluxes or as re-emissions of previously deposited Hg (Table 1), simple exclusion of this emission type from the simulations was not feasible without disturbing the whole Hg cycle in the model. In contrast, assuming additivity of Hg processes in the atmosphere with respect to contributions from different sources, the effect of natural and secondary emissions (NoNAT) could be estimated by subtracting the NoANT results from the BASE case. Four additional model experiments were intended to evaluate different chemical mechanisms of GEM oxidation in the atmosphere. To avoid the influence of direct anthropogenic emissions on simulated RM concentrations, all emissions were assumed to be a form of GEM. The BrCHEM1 and BrCHEM2 model runs included a single mechanism of GEM oxidation by atomic Br, but using two different datasets of Br concentration in the atmosphere: one simulated by GEOS-Chem (Parrealla et al., 2012) and the other by p-TOMCAT (Yang et al., 2005, 2010). A comparison of the spatial and temporal variations in Br concentration from these two datasets is given in Figs. S7 and S10 in the Supplement. Two other experiments, O3CHEM and OHCHEM, were based on O3- and OH-initiated oxidation chemistry. The models used the same O3 and OH concentration datasets extracted from MOZART model simulations results (Emmons et al., 2010). The spatial gradients and seasonal variation of the reactants are shown, respectively, in Figs. S8–S9 and S11–S12 in the Supplement. Note that not all models were used to perform the whole simulation program. The results presented below are based on available simulations for each particular experiment.

2.5 Statistical analysis

Modelling results were compared with observations using the following statistical parameters. Both spatial and temporal correlations of simulated and observed values were characterized by the Pearson correlation coefficient:

$$R_{\text{corr}} = \frac{\sum_i (M_i - \bar{M})(O_i - \bar{O})}{\sqrt{\sum_i (M_i - \bar{M})^2 \sum_i (O_i - \bar{O})^2}},$$

where $M_i$ and $O_i$ are monthly or annual mean simulated and observed values, respectively, and $\bar{M}$ and $\bar{O}$ are average values. The averaging and summing were performed over monthly values to calculate temporal correlations at particular sites or over annual mean values for all sites to calculate spatial correlation coefficients. An arithmetic mean of all temporal correlation coefficients for individual sites was then used in the analysis. The discrepancy between simulated and observed values was characterized by a symmetric relative bias:

$$\text{RBIAS} = 2 \left( \frac{\bar{M} - \bar{O}}{\bar{M} + \bar{O}} \right) \times 100\%,$$
where RBIAS varies within ±200 % and small deviations between model results and observations are characterized by values close to zero.

3 Results and discussion

3.1 Gaseous elemental mercury

The GEM concentration in air is a parameter representing the balance between Hg global emissions and sinks by chemical transformation to other Hg forms or direct interaction with the surface. Given the long residence time of GEM in the atmosphere, its spatiotemporal gradients likely characterize the distribution of global emission regions as well as long-range atmospheric dispersion and cycling in the atmosphere (Selin, 2009; Travnikov, 2012; Ariya et al., 2015). Figure 2 shows the global distribution of GEM concentration in surface air as simulated by four global models according to the BASE case, along with ground-based observations represented by colored circles in the same color palette. The models predicted spatial Hg concentration patterns with a pronounced gradient between the Southern Hemisphere (ca. 0.9–1.1 ng m\(^{-3}\)) and the Northern Hemisphere (ca. 1.1–1.6 ng m\(^{-3}\)) and elevated concentrations in the major industrial regions in East and South Asia, Europe, and North America (above 1.4 ng m\(^{-3}\)).

Elevated concentrations were also predicted in tropical areas of South America, central Africa, and Southeast Asia, where considerable Hg emissions from artisanal and small-scale gold mining are expected (AMAP/UNEP, 2013a). The models generally agreed with ground-based observations, as shown in Fig. 2. The measurements also provided evidence of a statistically significant interhemispheric gradient and relatively high concentrations in industrial regions (Sprovieri et al., 2016). A more detailed comparison of modeling results with measurements is given in Fig. S2 in the Supplement. The model–measurement divergence commonly did not exceed ±30 %. In general, the models exhibited lower spatial variations in annual GEM concentration than did the measurements. This observation relates mostly to sites that are compactly located in North America and Europe. It can be explained by the relatively low spatial resolution of the model grids (1–2 hundreds of kilometers), which made it difficult for the models to reproduce the influence of local emission sources and specific meteorological conditions at measurement sites. Another reason can be associated with possible overestimation of GEM residence time in the atmosphere, leading to additional smoothing of the simulated concentration patterns. The possibility of a shorter GEM lifetime with respect to oxidation was recently proposed in Horowitz et al. (2017), but this would require the existence of a fast atmospheric reduction process to match observations.

The model-to-model difference in simulated GEM concentrations did not exceed ±20 % (see also Fig. S5 in the Supplement). Nevertheless, pronounced distinctions could be observed among the spatial patterns simulated by different models because of deviations in physical and chemical process parameterizations. As mentioned in Sect. 2.2, the various models used quite different chemical mechanisms of GEM oxidation in the atmosphere in their base configurations. Moreover, even if they used the same anthropogenic emissions data, they differed in their estimates of natural and secondary emissions and Hg air–surface exchange. As shown in Fig. 2 and Fig. S5 in the Supplement, all four models were in relatively agreement when simulating GEM concentrations in regions affected by direct anthropogenic emissions (e.g., eastern North America, western and central Europe, East Asia). However, the differences between the models became larger as the distance from emission regions increased. For instance, GLEMOS and ECHMERIT predicted lower GEM concentrations (below 1.3 ng m\(^{-3}\)) over North America (mostly Canada) and eastern Siberia (Fig. 2a and d). This phenomenon was caused by two factors: use of O\(_3\) oxidation chemistry with elevated O\(_3\) concentrations in the northern midlatitudes (Fig. S8 in the Supplement) and use of relatively low levels of natural/secondary emissions over these regions. In particular, this led to some underestimation of observed GEM levels at North American sites. In contrast, GEM-MACH-Hg simulated higher GEM concentrations (above 1.4 ng m\(^{-3}\)) than other models over the North Atlantic and Pacific oceans due to relatively high Hg evasion from seawater in these aquatic regions and use of the OH oxidation mechanism. The latter provided weak oxidative capacity in the temperate and high latitudes of both hemispheres due to the pronounced meridional profile of OH concentrations, with high values in the tropics (Fig. S9 in the Supplement). The elevated GEM concentrations over the Pacific generally agreed with observations at Mauna Loa site in Hawaii and at coastal sites in Europe (Fig. 2c). However, the former is a high-elevation site (3384 m a.s.l.), where conditions are more like those in the free troposphere. Besides, it can be affected by Hg emissions from the active Kilauea volcano (Gay et al., 2013). Note that GLEMOS and ECHMERIT also used the OH mechanism in their base configurations. However, the effect described above diminished when used in combination with O\(_3\) chemistry. Simultaneous use of these two oxidation mechanisms can probably lead to surplus GEM oxidation or imply the existence of an Hg reduction mechanism in the atmosphere. In particular, it can explain some underestimation of observed GEM concentrations over the oceans by ECHMERIT. The effect was less pronounced in the case of GLEMOS because it used a reduced rate constant for the reaction with OH (Table 1). Use of Br chemistry by GEOS-Chem provided reasonable agreement with most observations (Fig. 2b). Unlike GEM-MACH-Hg, it simulated lower GEM concentrations over the North Atlantic and Pacific oceans, somewhat underestimating available observations in these regions. Note that the Br oxidation mechanism predicted strong GEM oxidation in the southern high latitudes due to high Br concentrations.
Figure 2. Spatial distribution of GEM air concentration in 2013 simulated according to the BASE case by four global models: (a) GLEMOS, (b) GEOS-Chem, (c) GEM-MACH-Hg, and (d) ECHMERIT. Circles show observed values in the same color scale.

(Fig. S7 in the Supplement), which was compensated for by large re-emission fluxes from seawater, leading to intensive air–surface exchange in this region. All models tended to overpredict GEM concentrations in the northern part of South America as measured at two sites: Manaus (Brazil) and Nieuw Nickerie (Surinam). As mentioned earlier, the high simulated concentrations in this region were caused by large expected emissions from artisanal and small-scale gold mining, which were probably overestimated.

Figure 3 illustrates an analysis of the interhemispheric GEM concentration gradient. The figure shows the meridional distribution of both observed and model-predicted concentrations in surface air. The latter are split into two fractions contributed by anthropogenic and natural/secondary sources. All four models reproduced the observed GEM concentration differences between the Southern and Northern hemispheres. The lowest concentrations (below 1 ng m$^{-3}$) were typical of the high and temperate latitudes of the Southern Hemisphere. There was a weak maximum of zonal-mean GEM concentration (1.4–1.6 ng m$^{-3}$) in the temperate latitudes of the Northern Hemisphere, corresponding to the location of most anthropogenic emission sources. Hence, the interhemispheric gradient is largely formed by the contribution of direct anthropogenic emissions, which is larger in the Northern Hemisphere. The contribution of natural and secondary emissions also increases northward, but the gradient is commonly smaller. The models predicted some decrease in concentrations further northward, which was not evident from the observations. This decrease was mostly connected with underestimation of Hg re-emissions from snow and seawater. More detailed analysis of the modeling results performed in the Arctic and Antarctica is given in the companion paper (Angot et al., 2016). In particular, three of the four models successfully reproduced the springtime GEM concentration minimum caused by occurrence of AMDEs. However, the models tended to underestimate the summertime maximum, which is determined by revolatilization from snowpack and meltwater and evasion from the ocean.

Statistics for simulated and observed GEM concentrations for different model experiments (Table 2) are illustrated in Fig. 4 in terms of spatial and temporal correlation coefficients and relative bias. Details of the statistics used are given in Sect. 2.5. In the BASE simulation, all models produced concentration distributions that agreed well with measurements (the spatial correlation coefficient was about 0.7 and the bias approximately zero). However, the models differed in their ability to reproduce temporal GEM variations in surface air. The temporal correlation coefficient between simulated and observed monthly mean values varied between $-0.3$ and $0.5$. (Fig. 4c). Sprovieri et al. (2016) found a consistent seasonal cycle of observed GEM concentration at most measurement sites in both Northern and Southern hemispheres, with higher concentrations during winter and spring and lower concentrations in summer and fall. However, it should be noted that the seasonal variation of monthly mean concentration was not significant at temperate and low latitudes, where most of the sites were located, and commonly did not exceed $\pm20\%$.

Therefore, reproduction of GEM temporal variation is a challenging task for models in the absence of data on seasonal variation in the anthropogenic emissions used in this study (AMAP/UNEP, 2013b).

Switching off anthropogenic emissions (NoANT) led to a decrease in GEM levels in the atmosphere (the bias was...
−40 %) and some decrease in the spatial correlation with measurements. It is worth noting that the spatial distribution of Hg concentrations in this experiment was largely determined by model-specific natural and secondary emissions and that therefore the change in spatial correlation differed considerably among the models. Removing anthropogenic emissions from the model simulations did not affect the temporal variation of the modeling results. In contrast, the results of the experiment with no natural and secondary emissions (NoNAT) demonstrated significant improvement in temporal correlation with measurements for the models that showed poor correlations in the base run. Besides, the exclusion of natural and secondary emissions led to some decrease in spatial correlation and a large negative bias (≈100 %). Simulations with different chemical mechanisms (BrCHEM1, BrCHEM2, O3CHEM, and OHCHEM) provided somewhat different spatial GEM concentration patterns in surface air. The model run with the first Br dataset (BrCHEM1) showed practically unbiased results, whereas the GEM concentrations from the BrCHEM2 run based on the second dataset were biased high (Fig. 4b) due to the lower Br concentrations provided by this dataset compared to the first one (Fig. S7 in the Supplement). The O3CHEM and OHCHEM simulations also showed a positive bias, indicating the smaller oxidation capacities of these mechanisms. The spatial correlation coefficient varied among the model runs with different chemical mechanisms (0.61−0.73). Somewhat higher spatial correlations were obtained for the oxidation reactions with Br (BrCHEM1, 0.72 ± 0.01) and OH radical (OHCHEM, 0.71 ± 0.02), which better reproduced the meridional profile of GEM concentration. The reaction with O3 (O3CHEM) provided poorer spatial correlation (0.65 ± 0.04) because it leveled the intercontinental GEM gradient. As mentioned earlier, O3 concentrations are considerably higher in the northern midlatitudes than in the Southern Hemisphere (Fig. S8 in the Supplement), leading to more intensive oxidation and decreased GEM levels in the Northern Hemisphere.

3.2 Reactive mercury

The oxidized Hg species composing RM originate in the atmosphere both from direct anthropogenic emissions and through oxidation of GEM in the gas phase, in the aqueous phase of cloud water, and in a heterogeneous manner at various atmospheric interfaces (Ariya et al., 2015). Therefore, simulating RM with contemporary models is a much more challenging task given the incomplete state of current knowledge of Hg atmospheric chemistry as well as the sparsity and uncertainty of measurements (Gustin et al., 2015). Figure 5 shows the global distributions of RM concentration in surface air as simulated by the models for the BASE case. The models predicted considerably different spatial RM concentration patterns. The concentrations were comparable in industrial regions, which were affected by direct anthropogenic emissions, but differed significantly in remote regions where the influence of emissions was weaker. Hence, the simulated patterns depended strongly on the chemical mechanisms and removal process parameterizations used. Indeed, the models that used O3 and/or OH oxidation chemistry in the BASE case (Figs. 5a, c, d) predicted elevated RM concentrations at low latitudes (the tropics and the equatorial zone) due to high concentrations of these photo-oxidants (mainly the OH radical) in these regions (see Fig. S9 in the Supplement). In

Figure 3. Global zonal-mean distribution of GEM air concentration in 2013 simulated by four models: (a) GLEMOS, (b) GEOS-Chem, (c) GEM-MACH-Hg, and (d) ECHMERIT. Black dots are the same observations as in Fig. 2 and the dotted line is a polynomial approximation.
The scarcity of long-term RM observations did not permit reconstruction of reliable spatial trends on a global scale. Annual mean RM observations for the year under study are available only at nine sites in North America, two sites in Europe, one site in the Arctic, and two sites in the Southern Hemisphere (Fig. 1). Considering the short lifetime of RM in the atmosphere with respect to its deposition (Gustin et al., 2015; Ariya et al., 2015) this limited observational dataset cannot characterize spatial RM variation around the globe. Nevertheless, the measurements can be used to evaluate modeling results at particular locations. The models showed variable performance in reproducing measured values. The scatter plots of the model-to-measurement comparison shown in Fig. S3 in the Supplement demonstrate significantly poorer model agreement with observations than in the case of GEM. From 30 to 90 % of the simulated values fell outside the range of agreement within a factor of 3. In addition, there was a general tendency to overestimate observed concentrations. The level of overestimation varied among sites and among models and can be explained by a number of factors, including measurement uncertainties associated with losses due to interference of oxidants and incomplete capture of GOM (Lyman et al., 2010; Huang and Gustin, 2015; Gustin et al., 2015), incorrect emissions speciation (Zhang et al., 2012; Amos et al., 2012; Kos et al., 2013; Bieser et al., 2014), and the uncertainties of atmospheric chemistry (Weiss-Penzias et al., 2015; Ariya et al., 2015; Shah et al., 2016).

Figure 6 shows statistics for model-to-measurement comparisons of RM air concentration for different model experiments. As mentioned earlier, the models considerably overestimated the observed values in the BASE case simulation. Similar overestimation was observed by Kos et al. (2013) when simulating Hg oxidized forms in a series of model sensitivity runs. This observation was attributed to a significant extent to incorrect speciation of anthropogenic emissions, with a too-high proportion of oxidized Hg forms. This was also confirmed by the NoANT experiment in the current study, where zeroed-out anthropogenic emissions led to a significantly lower positive or even negative bias (Fig. 6a).

To reduce the effect of this uncertainty in the current study, a modified speciation of emissions data was used for the model experiments focused on comparing chemical mechanisms, with all Hg emissions treated as GEM (Sect. 2.4). The overprediction of observed RM concentrations by a factor of 2.5 was also found by Weiss-Penzias et al. (2015) for a number of high- and mid-elevation sites and was found to be connected with the collection inefficiency of the KCl denuder used for the GOM measurements (Gustin et al., 2013). The models differed in their ability to reproduce the temporal variation in RM concentration (the correlation coefficient varied from −0.5 to 0.6 in the BASE case) (Fig. 6b). This variation was connected both with the different chemical mechanisms used in the standard model configurations (Table 1) and with deviations in model treatment of the removal processes responsible for dry and wet deposition. Ex-
Figure 5. Spatial distribution of annual mean RM air concentration in 2013 simulated according to the BASE case by four global models: (a) GLEMOS, (b) GEOS-Chem, (c) GEM-MACH-Hg, and (d) ECHMERIT. Circles show observed values in the same color scale.

Figure 6. Relative bias (a) and spatial correlation coefficient (b) of simulated and observed annual mean RM air concentration for different model experiments.

cclusion of anthropogenic and natural/secondary emissions (NoANT and NoNAT) only slightly affected the temporal correlations of the modeling results with observations. However, it should be pointed out once again that the emissions inventory used for this study (AMAP/UNEP, 2013b) did not resolve the intra-annual variability of anthropogenic emissions. Hence, one can expect a stronger effect of anthropogenic emissions on RM temporal variation. Among the chemical mechanisms, the best correlations between modeled and observed values were obtained for reactions with Br (BrCHEM1 and BrCHEM2), followed by the OH oxidation mechanism (OHCHEM). Inclusion of the reaction with O$_3$ led to a negative correlation with observations.

A more detailed analysis of chemical oxidation mechanisms is illustrated in Fig. 7 in terms of simulated and observed RM / GEM ratios. Indeed, atmospheric RM originates either from direct emissions from anthropogenic sources or as a product of GEM oxidation in the atmosphere (Selin, 2009; Travnikov, 2012; Kos et al., 2013; Ariya et al., 2015). Hence, in the immediate vicinity of emission sources, the RM / GEM ratio reflects the speciation of Hg emissions, whereas in remote regions far away from any emissions, it largely quantifies the oxidative capability of the atmosphere. Given the short lifetime of RM in the atmosphere with respect to deposition, the influence of direct emissions on the RM / GEM ratio should quickly weaken with distance from sources. Following the methodology suggested by Kos et al. (2013) the sites used for the following analysis were classified with respect to their remoteness from significant emission sources, based on the model sensitivity run with Hg atmospheric chemistry turned off. The simulated RM concentrations showed (Fig. S6 in the Supplement) that all the
selected sites (except one) could be classified as being far from sources (0–30 pg m\(^{-3}\)). This observation agrees with the characteristics of the North American sites as given by Lan et al. (2012). The only site that was probably directly affected by anthropogenic emissions was Waldhof, Germany (Weigelt et al., 2013). Nevertheless, because both the mean levels and the seasonal variation of RM concentrations measured at this site did not differ significantly from other sites, it was retained in the dataset. However, it should be noted that this analysis essentially depended upon the emissions data used and could translate any data uncertainties into the classification results.

Figure 7 shows a comparison of simulated and observed annual mean RM / GEM ratios for different chemical mechanisms. The whiskers show the standard deviation of monthly mean simulated and observed values. It should be pointed out that the observed values of RM / GEM (1–10 pg ng\(^{-1}\)) correspond to the background conditions of the continental boundary layer and are considerably lower than those from mountain sites analyzed by Weiss-Penzias et al. (2015) (10–100 pg ng\(^{-1}\)). Exceptions are the site at Alert, Canada, located in the High Arctic (86 pg ng\(^{-1}\)) and the elevated site at Salt Lake City, United States (21 pg ng\(^{-1}\)). It is interesting to note that the other elevated site (Longobucco, Italy) did not show similarly increased RM / GEM values (9.5 pg ng\(^{-1}\)). As shown in Fig. 7, the best qualitative agreement between model and measurements was found in the BrCHEM1 experiment with the Br chemistry and one of the Br concentration datasets (Fig. 7a). Three of the four models demonstrated good performance in this experiment, reproducing observations at most sites with a deviation within a factor of 3. The fourth model (ECHMERIT) significantly overestimated the observed values, which was also typical of the other model experiments (except for O3CHEM). This overestimation was caused not by the chemistry used but by underestimation of wet scavenging as shown below. Use of Br chemistry with the other Br concentration dataset (BrCHEM2) led to less consistent results (Fig. 7b). The RM / GEM ratios simulated by the models varied from moderate underestimation to overestimation of observed values. The inter-model differences could have been caused both by discrepancies in removal process formulation and by particular implementations of the Br chemical mechanism (Table 1). Somewhat similar results were obtained in the experiment with OH chemistry.

Figure 7. Scatter plots of simulated vs. observed ratios of annual mean RM concentration to GEM concentration in 2013 for different model experiments: (a) BrCHEM1, (b) BrCHEM2, (c) O3CHEM, and (d) OHCHEM. Whiskers show standard deviation of monthly mean simulated and observed values. Dotted red line depicts the 1:1 ratio; dotted black lines show deviation by a factor of 5.
(OHCHEM, Fig. 7d). The deviations between modeled and measured RM/GEM were mostly within a factor of 5, and the models tended to overestimate the observed values. Application of the O$_3$-initiated chemistry led very consistent results (O3CHEM, Fig. 7c). The models also somewhat overestimated the measured RM/GEM ratios, with minimum scattering of modeling results. However, the models tended to underestimate considerably the intra-annual variation of monthly values, as shown by the whiskers. None of the chemical mechanisms could reproduce the high annual RM/GEM ratios (above 80 pg ng$^{-1}$) observed at the Arctic site Alert, Canada (Fig. 7a–c). These high annual values are connected with intensive Hg oxidation during springtime AMDEs. Analysis of specific processes typical of the polar regions is beyond the scope of this paper. A discussion of the results of the study that focused on the polar regions can be found elsewhere (Angot et al., 2016).

More insight into the effects of different chemical mechanisms can be obtained by analyzing RM/GEM seasonal variation. Figure 8 shows both measured and simulated variation in the monthly mean RM/GEM ratio averaged over selected sites. Because seasonal variation of both RM and GEM differs in the Northern and Southern hemispheres and because most of the study sites are located in North America and Europe, two sites situated north of the Equator were selected for this purpose. In addition, the Arctic site (Alert) and the two high-elevation sites (Salt Lake City and Longobucco) were excluded to avoid the effects of specific conditions in the polar regions and the free troposphere, respectively. Hence, the collection of sites characterizes the seasonality of Hg oxidation in the continental boundary layer in northern temperate latitudes. The observed values demonstrated pronounced seasonal changes in RM/GEM, with a maximum in March and a minimum in September (Fig. 8). Similar seasonal variations of Hg oxidized forms at background sites were observed in previous studies (Poissant et al., 2005; Sigler et al., 2009; Nair et al., 2012; Weigelt et al., 2013). The chemical oxidation mechanisms differed in their ability to reproduce the observed seasonal variation.

Use of Br chemistry with both Br concentration datasets (BrCHEM1 and BrCHEM2) provided the best agreement with measurements. Three of the four models (GEOS-Chem, GEM-MACH-Hg, and GLEMOS) reproduced the maximum RM/GEM ratios during the spring months (Fig. 8a and b). High levels of oxidized Hg in spring are evidently caused by high Br concentrations in both the free troposphere and the boundary layer in the Northern Hemisphere (Fig. S10 in the Supplement). However, only GEOS-Chem reproduced correctly the time of the maximum in March. Two other models moved the highest RM/GEM ratio to April and May. This discrepancy can be explained by partitioning the GEM oxidation products. GEOS-Chem is the only model that used dynamic gas–particle partitioning of RM (Table 1). This meant that oxidized Hg originated from GEM oxidation was dynamically fractionated between GOM and PBM. This equilibrium is sensitive to air temperature shifting to GOM under warmer conditions (Amos et al., 2012). Furthermore, GOM is more efficiently removed from the atmosphere than PBM by both precipitation and surface uptake. This phenomenon leads to higher deposition and lower RM concentrations in late spring (April and May). Two other models predicted a fixed share of the products of GEM oxidation by Br, thus missing changes in RM concentration during the season. The fourth model (ECHMERIT) predicted the highest ratios during late summer independently of the chemical mechanism used, a phenomenon that was determined by meteorological conditions and removal processes. In particular, the model simulated extremely low precipitation in late summer and early fall (Fig. 12i and j). This led to significant underestimation of wet RM removal and ultimately to overestimation of oxidized Hg concentrations during this period.

Model simulations with OH chemistry (OHCHEM) predicted maximum RM/GEM ratios during the summer months (Fig. 8d) in accordance with the seasonal variation of OH concentration, which was also highest in summer (Fig. S12 in the Supplement). Use of the O$_3$-initiated chemistry did not lead to significant variation of Hg oxidation during the year (Fig. 8c).

### 3.3 Wet deposition

Wet deposition is one of the major removal mechanisms responsible for Hg exchange between the atmosphere and the Earth’s surface (Travnikov, 2012; Swartzenzuber and Jaffe, 2012). It is largely determined by precipitation events on the one hand and by availability of soluble Hg forms in the atmosphere on the other. Given the poor solubility of GEM (Clever et al., 1985; Ariya et al., 2015), Hg wet deposition consists mostly of scavenging of Hg oxidized forms (GOM and PBM). Therefore, Hg concentration in precipitation and ultimately wet deposition flux largely depends upon three factors: direct emissions of oxidized Hg from anthropogenic sources, Hg oxidation in the atmosphere, and precipitation amount. Figure 9 shows the spatial patterns of annual mean Hg wet deposition as simulated by the four models according to the BASE case. Available measurements are also shown in the same color palette. Generally, the simulated deposition maps have similar spatial distributions, reflecting the influence of global precipitation patterns and major emission regions. High deposition fluxes are characteristic of Asia, Europe, and North America, where significant anthropogenic sources are located, as well as of regions with intensive precipitation (e.g., the Intertropical Convergence Zone). The lowest wet deposition fluxes occurred in dry regions (e.g., in northern Africa, Greenland, and Antarctica). Divergences among the modeling results could be mostly explained by the different chemical mechanisms used by the models in the BASE case. For instance, GEOS-Chem predicted elevated wet deposition in the high latitudes of the Southern Hemisphere, where high Br concentrations...
Figure 8. Normalized seasonal variation of monthly ratio of annual mean RM concentration to GEM concentration. Black line with dots shows observations averaged over selected sites (whiskers are standard deviation). Colored lines present model simulations averaged over the same sites for different model experiments: (a) BrCHEM1, (b) BrCHEM2, (c) O3CHEM, and (d) OHCHEM.

(Fig. S7 in the Supplement) caused intensive GEM oxidation in the atmosphere (Fig. 9b). In contrast, significant deposition fluxes were simulated in the High Arctic by GEM-MACH-Hg (Fig. 9c) because parameterizations of physical and chemical processes occurring during AMDEs were used. The models agreed relatively well with available long-term observations of Hg wet deposition. Model-to-measurement deviations commonly did not exceed a factor of 2 (Fig. S4 in the Supplement). However, it should be noted that available observations of Hg wet deposition are still mostly restricted to two regions: North America and Europe. Only a few measurements are available in other regions, particularly in the Southern Hemisphere.

Figure 10 shows statistics of the comparison of simulated and observed wet deposition fluxes. Results of the BASE case simulation are characterized by significant temporal correlation with measurements (0.4–0.6) and some slight bias (±40%) that varies among models. Direct anthropogenic emissions of oxidized Hg contributed considerably to wet deposition, and hence their elimination (NoANT) resulted in a noticeable deposition decrease characterized by negative bias. In contrast to anthropogenic emissions, natural/secondary sources emit Hg mostly as GEM. Nevertheless, turning off natural/secondary emissions (NoNAT) also led to a substantial decrease in wet deposition, indicating their indirect effect through GEM oxidation on soluble Hg forms, with subsequent scavenging by precipitation. The temporal correlation of wet deposition was not sensitive to emission changes. This is not surprising because the anthropogenic emissions inventory used in this study did not contain information on temporal emissions variation. Oxidation chemistry considerably affected both the general level and the temporal variation of Hg wet deposition. The Br oxidation mechanism provided a relatively high correlation with observations, but the results for the two Br concentration datasets (BrCHEM1 and BrCHEM2) differed widely in terms of relative bias. The highest correlation was obtained for OH oxidation chemistry (OHCHEM). Note that, unlike the other models used for this study, ECHMERIT is based on the ECHAM climate model, which is expected to reproduce actual weather behavior, in particular precipitation events, over a relatively longer temporal period and a wider area than the other models, but may diverge on shorter timescales and over smaller regional areas (see, for example, Angálil et al., 2016). Because simulated Hg wet deposition is largely driven by model-generated precipitation, the results of the climate-based ECHMERIT model can be significantly biased with regard to other wet deposition models and observations (Fig. 10).

Similarly to RM concentration, Hg wet deposition is strongly determined by atmospheric oxidation chemistry (Selin and Jacob, 2008; Kos et al., 2013). Therefore, analysis of wet deposition can also be used to evaluate the chemical mechanisms of Hg oxidation in the atmosphere. Unlike near-surface RM observations, wet deposition measurements characterize processes occurring in the free troposphere because scavenging of soluble Hg takes place both in the cloud environment and below clouds along the whole precipitation pathway. Figure 11 shows a comparison of simulated and observed wet deposition fluxes for different model experiments. Both measured and simulated values are averaged over different groups of sites, including seven groups in North America following the latitudinal ranges suggested by Selin and Jacob (2008), three groups in Europe (southern Europe, western Europe, and northern Europe), and one group per region in Asia, Australia, and the Indian Ocean (see Table S3 in the Supplement). Note that the highest observed
wet deposition values (30–45 ng m$^{-2}$ day$^{-1}$) are associated with the southern United States, whereas the lowest values (below 10 ng m$^{-2}$ day$^{-1}$) are characteristic of sites located in East Asia and in the Southern Hemisphere (see Fig. 9). As shown in Fig. 11, simulations with the Br oxidation mechanism and the first set of Br concentration data (BrCHEM1) satisfactorily reproduced observations (Fig. 11a). The models agreed relatively well with each other, and the model-to-measurement deviations generally did not exceed a factor of 2. However, all models overpredicted low deposition fluxes at Asian and southern sites. The overestimation of Hg wet deposition at two high-altitude Asian sites (Mt. Waliguan and Mt. Ailao) can be connected with the inability of the coarse-spatial-resolution global models to reproduce complex meteorological conditions in mountain regions. Overprediction at southern sites (Cape Grim and Amsterdam Island) can be explained by the very high Br concentrations predicted by the first dataset at temperate latitudes in the Southern Hemisphere (Fig. S7 in the Supplement). Use of the same mechanism with the other Br dataset led to considerably lower wet deposition levels (Fig. 11b) due to use of much lower Br concentrations, particularly in the free troposphere (Fig. S7 in the Supplement). Hence, uncertainties in the available estimates of Br atmospheric concentration strongly affected simulation results for Hg cycling in the atmosphere. Model simulations with the O$_3$ and OH oxidation mechanisms (O3CHEM and OHCHEM) provided reasonable agreement between modeling results and measurements (Fig. 11c and d). Two of the three models (GEM-MACH-Hg and GLEMOS) demonstrated fairly good correlations with observations, but again tended to overestimate lower observed values. The OH oxidation chemistry pro-

---

**Figure 9.** Spatial distribution of wet deposition flux in 2013 simulated according to the BASE case by four global models: (a) GLEMOS, (b) GEOS-Chem, (c) GEM-MACH-Hg, and (d) ECHMERIT. Circles show observed values in the same color scale.

**Figure 10.** Relative bias (a) and spatial correlation coefficient (b) of simulated and observed annual mean wet deposition flux for different model experiments.
provided somewhat better agreement in terms of the slope of the regression line, which was closer to the reference 1 : 1 line, indicating better reproduction of both low and high wet deposition fluxes. The third model (ECHMERIT) showed wider divergence of simulation results because of the uncertainties described above in its reproduction of precipitation events in some locations. In particular, these uncertainties led to significant underestimation of Hg wet deposition in the southern part of North America.

More information on the performance of different chemical mechanisms can be obtained by analyzing seasonal wet deposition patterns. Because most available wet deposition measurements are obtained at sites in North America and Europe, further discussion will focus on these two regions. Figure 12 shows a comparison of modeled and measured temporal variation of monthly mean wet deposition flux averaged over sites in North America and Europe. The monthly fluxes were normalized by the annual average value to remove absolute differences among the models and reveal peculiarities of seasonal changes. The observations exhibit a pronounced seasonal cycle, with a maximum in summer and a minimum during the cold season (winter and early spring). Similar seasonal variations have been reported in previous studies (Guentzel et al., 2001; Keeler et al., 2005; Choi et al., 2008; Prestbo and Gay, 2009; Sprovieri et al., 2017). Sprovieri et al. (2017) attributed these seasonal changes to variations in meteorological conditions (mainly precipitation amount), more effective Hg scavenging by rain compared to snow, and changes in availability of soluble Hg. As shown in Fig. 12i and j, precipitation amounts measured in North America and Europe do not reveal a similar seasonality to explain the intra-annual variation of wet deposition. Seasonal variation of precipitation amounts in North America exhibits a similar pattern, with a maximum in summer and a minimum in winter, but the amplitude of the variation is much less than that of wet deposition. Average precipitation amounts in Europe do not show an evident seasonal pattern. Note that one of the four models (ECHMERIT) failed to reproduce the seasonal cycle of precipitation amounts in North America and Europe for the reasons mentioned previously. Because precipitation variation directly affects seasonal variation of Hg wet deposition, the simulation results

Figure 11. Scatter plots of simulated vs. observed annual mean wet deposition flux in 2013 averaged over different territorial groups of sites (see Table S3 in the Supplement) for different model experiments: (a) BrCHEM1, (b) BrCHEM2, (c) O3CHEM, and (d) OHCHEM. Solid lines depict linear approximation. Dotted red line depicts the 1 : 1 ratio; dotted black lines show deviation by a factor of 2.
Figure 12. Normalized seasonal variation of monthly mean wet deposition flux in North America (left column) and Europe (right column). Black line with dots shows observations averaged over all sites in the regions (whiskers are standard deviation). Colored lines present model simulations averaged over the same sites for different model experiments: (a, b) BrCHEM1, (c, d) BrCHEM2, (e, f) O3CHEM, and (g, h) OHCHEM. Seasonal variations of precipitation amount in North America and Europe are also shown in panels (i) and (j), respectively.

from this model were not used when analyzing the performance of chemical mechanisms. However, the results themselves are shown in Fig. 12 for completeness. Availability of soluble Hg in the free troposphere depends strongly on oxidation chemistry. Therefore, different chemical mechanisms should affect the seasonality of wet deposition differently. Indeed, both model runs with the Br oxidation chemistry (BrCHEM1 and BrCHEM2) predicted a maximum in wet deposition during the spring months instead of in summer (Fig. 12a–d), following the seasonal variation of Br concentration in the atmosphere (Fig. S10 in the Supplement). Simulations with O3-initiated chemistry (O3CHEM) provided a much lower seasonality of deposition flux (Fig. 12e–f). In contrast, use of OH chemistry (OHCHEM) reproduced faithfully the observed seasonal variation of wet deposition in both study regions (Fig. 12g–h). Similar results were ob-
tained by other researchers. Selin and Jacob (2008) simulated Hg wet deposition over the United States using combined OH/O₃ oxidation chemistry and successfully reproduced the measured seasonal variation. They attributed the summer maximum in the northeast to GEM photochemical oxidation and to inefficient scavenging by snow in winter. Holmes et al. (2010) compared the Br mechanism with the OH/O₃ mechanism for simulating the Hg global cycle. They found that the OH/O₃ chemistry better simulated the southeast Hg wet deposition summer maximum. It was postulated that the maximum reflected scavenging of GOM from the free troposphere by deep convection. Kos et al. (2013) also performed a number of sensitivity runs with different chemical process parameterizations and showed that the OH oxidation chemistry improved simulation of the seasonal cycle of wet deposition in North America. Considering that Hg wet deposition is largely defined by GEM oxidation (Selin and Jacob, 2008), a significant effect of OH-initiated chemistry on Hg oxidation in the free troposphere can be expected. Moreover, when comparing this conclusion with the results presented in Sect. 3.2, where it was shown that the seasonal dynamics of the RM/GEM ratio observed at ground-based sites is dominated by Br oxidation chemistry, one can assume the possibility of different Hg oxidation mechanisms occurring concurrently in different parts of the atmosphere.

4 Conclusions

The study presented here provides a complex analysis of the processes governing Hg cycling in the atmosphere, involving both measured data from ground-based sites and application of chemical transport models. A variety of long-term GEM and RM concentration measurements as well as wet deposition flux readings have been compiled from different global and regional monitoring networks. Four contemporary global-scale Hg transport models were used, both in their state-of-the-art configurations and for a number of numerical experiments aimed at evaluation of particular processes. Results of the model simulations were evaluated against measurements. The models predicted similar global spatial GEM concentration patterns in near-surface air. The model-to-model difference in simulated GEM concentrations did not exceed ±20 % and agreed with observed values within ±30 %. Nevertheless, there were pronounced distinctions among the spatial patterns simulated by different models, which resulted from deviations in the parameterizations of physical and chemical processes. All four models reproduced the observed GEM concentration decrease from the Northern to the Southern hemisphere. Analysis revealed that the interhemispheric gradient is largely formed by the spatial distribution of anthropogenic emissions that prevails in the Northern Hemisphere. The contributions of natural and secondary emissions enhanced the south-to-north gradient, but their effect was less significant. Simulations with different chemical mechanisms provided somewhat different spatial GEM concentration patterns in surface air. Higher spatial correlations were obtained for the oxidation reactions with Br and OH radical, enabling better reproduction of the meridional GEM concentration profile. The reaction with O₃ provided poorer spatial correlation because it leveled the intercontinental GEM gradient.

Model simulation of RM is a much more challenging task because of incomplete current knowledge of Hg atmospheric chemistry and the scarcity and uncertainty of measured data. The models differed considerably in their predictions of spatial and temporal RM concentration patterns. The simulated RM levels were comparable in industrial regions, which are affected by direct anthropogenic emissions, but differed significantly in remote regions where the influence of emissions is weaker. Hence, the simulated patterns depended strongly on the chemical mechanisms and removal process parameterizations used. Model-to-measurement comparisons demonstrated significantly poorer model agreement with observations than in the case of GEM. From 30 to 90 % of the simulated values fell beyond the agreement range within a factor of 3. In addition, there was a general tendency to overestimate observed RM concentrations, which can be attributed to incorrect speciation of Hg emissions, the uncertainties of Hg atmospheric chemistry, and incomplete RM capture by measurements. Atmospheric chemistry strongly affected the RM/GEM ratio in the atmosphere. The Br chemistry provided the best agreement with observations, reproducing both general levels and seasonal variation of the RM/GEM ratio in the near-surface layer. However, the global distribution of Br concentration is highly uncertain. Model simulations with the OH chemical mechanism predicted a shift in maximum RM/GEM ratios from spring to summer, but O₃-initiated chemistry did not predict significant seasonal variations in Hg oxidation.

Wet deposition maps simulated by different models had similar spatial distributions, reflecting the influence of global precipitation patterns and the location of major emission regions. High deposition fluxes are characteristic of Asia, Europe, and North America, where regions with significant anthropogenic sources and intensive precipitation are located. The models agreed relatively well with available long-term Hg wet deposition observations. The model-to-measurement deviations commonly did not exceed a factor of 2. However, there was a tendency to overpredict low deposition fluxes measured in Asia and in the Southern Hemisphere. Similarly to RM concentrations, wet deposition of Hg in background regions is strongly determined by atmospheric oxidation chemistry. Model runs with the Br oxidation mechanism predicted a wet deposition maximum in spring, instead of in summer as observed at monitoring sites in North America and Europe. O₃ chemistry did not predict significant seasonal changes of wet deposition flux in these regions. Use of OH chemistry enabled reproduction of both the periods of max-

www.atmos-chem-phys.net/17/5271/2017/ Atmos. Chem. Phys., 17, 5271–5295, 2017
imum and minimum values and the amplitude of observed seasonal variations.

Hence, the performance of the Hg oxidation mechanisms studied differed in the quality of their reproduction of the various observed parameters, which can imply the possibility of more complex chemistry and multiple Hg oxidation pathways occurring concurrently in various parts of the atmosphere. More extensive measurements of atmospheric RM (including identification of Hg chemical species) and wet deposition are needed in various geographic regions and under different climatic conditions to achieve further improvement of Hg chemical transport models.

Data availability. Mercury modeling and measurement data discussed in this paper are reported within the GMOS central database and are available upon request at http://sdi.iiia.cnr.it/geoint/publicpage/GMOS.

The Supplement related to this article is available online at doi:10.5194/acp-17-5271-2017-supplement.

Author contributions. The names after the first author in the above list are in alphabetical order and all authors have made significant contribution. In particular, H. Angot, P. Artaxo, M. Bencardino, F. D’Amore, M. C. Diéguez, A. Dommergues, R. Ebingerhaus, X. B. Feng, O. Magand, L. Martin, N. Mashyanov, N. Pirrone, R. Ramachandran, K. A. Read, F. Sena, F. Sprovieri, D. Wip and I. Wängberg contributed to measurements; J. Bieser, A. Dastoor, F. De Simone, C. N. Gencarelli, I. M. Hedgecock, V. Matiasson, A. Ryjkov, N. E. Selin, S. Song, O. Travnikov, and X. Yang contributed to modeling.

Competing interests. The authors declare that they have no conflict of interest.

Acknowledgements. The authors wish to thank David Gay from the National Atmospheric Deposition Program (NADP) and the principal investigators for the AMNet sites (E. Edgerton, W. Luke, J. Chaffin, R. Callison, B. Call, M. Pendleton, E. Miller, and M. Allen) for providing the Hg measured data used in this study. The authors also gratefully acknowledge the Canadian Air and Precipitation Monitoring Network (CAPMOn) for the data and the National Atmospheric Chemistry (NaChem) Database and Analysis Facility of Environment Canada (www.ec.gc.ca/natchem) for internet provision of the data. This study was financially supported in part by the EU FP7-ENV-2010 project, “Global Mercury Observation System” (GMOS, grant agreement no. 265113). Noelle E. Selin and Shaojie Song also acknowledge the US National Science Foundation Atmospheric Chemistry Program (grant no. 1053648) for their financial support. Aurélien Dommergue and Olivier Magand acknowledge the French Polar Institute, IPEV (GMOstral 1028), for funding and logistics.

References


AMAP/UNEP, Geospatially distributed mercury emissions dataset 2010v1, available at: http://www.amap.no/mercury-emissions (last access: 1 August 2016), 2013b.


Atmos. Chem. Phys., 17, 5271–5295, 2017

www.atmos-chem-phys.net/17/5271/2017/


O. Travnikov et al.: Multi-model study of mercury dispersion in the atmosphere


Platt, U. and Janssen, C.: Observation and role of the free radicals \( \text{NO}_2, \text{ClO}, \text{BrO} \) and \( \text{IO} \) in the troposphere, Faraday Discuss., 100, 175–198, 1995.
Snider, G., Raofie, F., and Ariya, P. A. A.: Effects of relative humidity and CO(g) on the \( \text{O}_3 \)-initiated oxidation reaction of Hg0(g):
O. Travnikov et al.: Multi-model study of mercury dispersion in the atmosphere


