## The GFDL Global Atmospheric Chemistry-Climate Model AM4.1: Model Description and Simulation Characteristics

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#### Abstract

We describe the baseline model configuration and simulation characteristics of GFDL's Atmosphere Model version 4.1 (AM4.1), which builds on developments at GFDL over 2013–2018 for coupled carbon-chemistry-climate simulation as part of the sixth phase of the Coupled Model Intercomparison Project. In contrast with GFDL's AM4.0 development effort, which focused on physical and aerosol interactions and which is used as the atmospheric component of CM4.0, AM4.1 focuses on comprehensiveness of Earth system interactions. Key features of this model include doubled horizontal resolution of the atmosphere (~200 km to ~100 km) with revised dynamics and physics from GFDL's previous-generation AM3 atmospheric chemistry-climate model. AM4.1 features improved representation of atmospheric chemical composition, including aerosol and aerosol precursor emissions, key land-atmosphere interactions, comprehensive land-atmosphere-ocean cycling of dust and iron, and interactive ocean-atmosphere cycling of reactive nitrogen. AM4.1 provides vast improvements in fidelity over AM3, captures most of AM4.0's baseline simulations characteristics and notably improves on AM4.0 in the representation of sudden stratospheric warmings in the coldest months. Distributions of reactive nitrogen and sulfur species, carbon monoxide, and ozone are all substantially improved over AM3. Fidelity concerns include degradation of upper atmosphere equatorial winds and of aerosols in some regions.

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## 7

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## 17

## 18 Key Points:

- A new atmospheric chemistry-climate model (AM4.1) has been developed for the
- 20 Geophysical Fluid Dynamics Laboratory (GFDL) 4<sup>th</sup>-generation suite.
- AM4.1 includes an advanced dynamical core and physical parameterizations, with enhanced vertical resolution and revised aerosol and chemistry interactions.
- The AM4.1 model exhibits substantially enhanced fidelity compared to previous-
- 24 generation GFDL atmospheric models.

## 25 Abstract

- 26 We describe the baseline model configuration and simulation characteristics of GFDL's
- 27 Atmosphere Model version 4.1 (AM4.1), which builds on developments at GFDL over 2013–
- 28 2018 for coupled carbon-chemistry-climate simulation as part of the sixth phase of the Coupled
- 29 Model Intercomparison Project. In contrast with GFDL's AM4.0 development effort, which
- 30 focused on physical and aerosol interactions and which is used as the atmospheric component of
- 31 CM4.0, AM4.1 focuses on comprehensiveness of Earth system interactions. Key features of this
- 32 model include doubled horizontal resolution of the atmosphere ( $\sim 200$  km to  $\sim 100$  km) with
- 33 revised dynamics and physics from GFDL's previous-generation AM3 atmospheric chemistry-
- 34 climate model. AM4.1 features improved representation of atmospheric chemical composition,
- 35 including aerosol and aerosol precursor emissions, key land-atmosphere interactions,
- 36 comprehensive land-atmosphere-ocean cycling of dust and iron, and interactive ocean-
- 37 atmosphere cycling of reactive nitrogen. AM4.1 provides vast improvements in fidelity over
- 38 AM3, captures most of AM4.0's baseline simulations characteristics and notably improves on
- 39 AM4.0 in the representation of aerosols over the Southern Ocean, India, and China—even with
- 40 its interactive chemistry representation—and in its manifestation of sudden stratospheric
- 41 warmings in the coldest months. Distributions of reactive nitrogen and sulfur species, carbon
- 42 monoxide, and ozone are all substantially improved over AM3. Fidelity concerns include
- 43 degradation of upper atmosphere equatorial winds and of aerosols in some regions.

## 44 Plain Language Summary

- 45 GFDL has developed a coupled chemistry-climate Atmospheric Model (AM4.1) as part of its 4<sup>th</sup>-
- 46 generation coupled model development activities. AM4.1 includes comprehensive chemistry for
- 47 representing ozone and aerosols, and has been developed for use in chemistry and air quality
- 48 applications, including advanced land-atmosphere-ocean coupling. With fidelity near to that of
- 49 AM4.0, AM4.1 features vastly improved representation of climate mean patterns and variability
- 50 from previous GFDL atmospheric chemistry-climate models.

## 51 **1. Introduction**

- 52 Atmospheric chemistry and composition are intrinsically coupled to the Earth's climate system.
- 53 The pivotal role of chemistry-climate interactions in regulating climate has been recognized for
- 54 over 30 years (Ramanathan et al., 1987). Advances in fundamental knowledge of atmospheric
- 55 chemistry and availability of long-term measurements, combined with advances in computing,
- 56 have provided an opportunity to enhance the comprehensiveness of our representation of

57	atmospheric chemistry-climate interactions. For the fifth phase of the Coupled Model
58	Intercomparison Project (CMIP5), GFDL contributed its first coupled chemistry-climate model,
59	CM3 (Donner et al., 2011), which allowed us to represent interactive aerosols and ozone, rather
60	than prescribing concentrations from offline models (e.g., Horowitz et al., 2003). Much has been
61	learned in the preceding years as to the strengths and weaknesses of CM3 chemistry, including
62	the critical role of prognostic aerosol interactions (e.g., Naik et al., 2013a; Levy et al., 2013). As
63	such, interactive aerosols were included in all of GFDL's 4 <sup>th</sup> -generation atmospheric model
64	development efforts, targeted for the sixth phase of the Coupled Model Intercomparison Project
65	(CMIP6). The high computational cost of interactive atmospheric chemistry, however, was
66	avoided in GFDL's recent CM4.0 development (Held et al., 2019) by prescribing ozone and
67	other oxidants. A full interactive chemistry representation, along with a comprehensive carbon
68	cycle, was reserved in this generation of GFDL models for Earth System Model development in
69	ESM4.1 (Dunne et al., submitted). Thus, CM4 included a focus on ocean resolution, while
70	ESM4.1 focused on a comprehensive representation of the Earth system.
71	The overall goal of AM4.1 development was to merge a suite of mostly parallel sets of updates
72	and innovations into GFDL's 4 <sup>th</sup> -generation atmospheric model. These updates include a revised
73	chemical mechanism from AM3 to AM4.1 to take advantage of new laboratory kinetic data (e.g.,
74	Sander et al., 2011; Mao et al., 2013b; as implemented by Li et al., 2016), in particular for
75	photooxidation of biogenic volatile organic compounds (BVOCs). Analysis of deficiencies in
76	AM3 chemistry has pointed out improper treatment of nitrate aerosols and gas-aerosol
77	interactions and biases in wet and dry deposition. We also wanted to leverage successful follow-
78	on development efforts from AM3 targeted to implement reduced nitrogen cycling (Paulot et al.,
79	2016, 2017b) and improved representation of the seasonal cycle in sulfate (Paulot et al., 2017a).

Finally, we wanted to provide the ability for the atmospheric model to handle a more diverse
suite of land-atmosphere and ocean-atmosphere linkages for comprehensive Earth system
representation of not only heat and hydrology but also CO<sub>2</sub>, dust, reactive nitrogen, and organic
carbon.

84 The focus of the present study is to document the atmosphere physics and chemistry developed 85 for AM4.1, as distinguished from the physical climate-focused AM4.0 (Zhao et al., 2018ab), for 86 standalone atmospheric applications. A more comprehensive discussion of coupled atmosphere-87 ocean-land Earth system interactions in ESM4.1 is provided by Dunne et al. (submitted). We 88 focus our analysis on evaluating the AMIP configuration of AM4.1 used for CMIP6, and 89 document differences in results between AM4.1 and other GFDL CMIP models, including 90 AM4.0 (CMIP6) and AM3 (CMIP5). In the case of comparisons with AM3, the differences in 91 results reflect updates to both the model configuration and the emissions (as discussed in Section 92 2.4).

## 93 2. Model description

A general schematic of AM4.1 forcing, dynamics, physics, aerosol, and chemistry interactions is
provided in Figure 1. The following sections discuss the formulations for these components in
reference to their AM4.0 (Zhao et al., 2018ab) counterparts.

#### 97 2.1 Physical model formulation

- 98 The physical formulation of AM4.1 is similar to that of AM4.0, but the model top has been
- 99 raised from 100 Pa (~45 km) to 1 Pa (~80 km), and the number of vertical levels has been
- 100 increased from 33 to 49, similar to the 48-level structure of AM3. This enhanced vertical extent
- 101 and resolution allows AM4.1 to represent stratospheric dynamics and chemistry, and

stratosphere-troposphere coupling. The time step used in the dynamical core for gravity wave
and the Lagrangian dynamics is reduced from 150 s in AM4.0 to 130 s in AM4.1 for increased
numerical stability.

105 Like AM4.0, AM4.1 includes 5 tracers for water (specific humidity, liquid water, ice water, 106 cloud amount, and liquid droplet number concentration), and uses the same large-scale and 107 convective cloud parameterizations as in AM4.0. Cloud parameterizations in AM4.1 were re-108 tuned slightly compared to AM4.0 in order to improve agreement with observed top-of-109 atmosphere shortwave and longwave radiative fluxes, in response to initially excessive reflection 110 from convective clouds over sub-Saharan Africa, North Indian Ocean, and the western tropical 111 Pacific. In particular, the scale factor applied to the fall speed of ice clouds (*c1* in Zhao et al., 112 2018b) was reduced from 0.90 in AM4.0 to 0.85 in AM4.1 to increase ice water path and 113 decrease outgoing longwave radiation. The cloud erosion timescale ( $\tau_{eros}$ ) in convectively active 114 regions is decreased slightly from 6.9 hours to 5.6 hours to increase the absorbed shortwave 115 radiation. The cloud erosion timescale under other conditions is unchanged from AM4.0. 116 As described by Zhao et al. (2018b), non-orographic gravity wave drag in AM4.0/AM4.1 is 117 parameterized following Alexander and Dunkerton (1999), but the parameters used in AM4.1 are 118 modified from those in AM4.0 to improve stratospheric circulation. In AM4.1, the magnitudes of 119 the non-orographic gravity wave flux above 350 hPa for the tropics, northern extratropics, and southern extratropics (St, Sn, and Ss) are set to 0.004, 0.005, and 0.0035 m<sup>2</sup> s<sup>-2</sup>, respectively. 120 121 Land hydrology and ecosystem dynamics are represented in AM4.1 by the GFDL Land Model 122 version 4.1 (LM4.1; Shevliakova et al., submitted), replacing the LM4.0 model used in AM4.0

123 (Zhao et al., 2018b). LM4.1 includes advanced vegetation and canopy competition, fire, land use

124 representation, and dynamic atmospheric dust coupling.

125 2.2 Atmospheric chemistry and aerosol scheme

126 AM4.1 includes interactive tropospheric and stratospheric gas-phase and aerosol chemistry. The 127 bulk aerosol scheme, including 18 transported aerosol tracers (see Supplementary material, Table 128 S1), is similar to that in AM4.0 (Zhao et al., 2018b), with the following updates: (1) ammonium 129 and nitrate aerosols are treated explicitly, with the sulfate-nitrate-ammonia thermodynamic 130 equilibrium simulated using ISORROPIA (Fountoukis and Nenes, 2007), as described by Paulot 131 et al. (2016); (2) oxidation of sulfur dioxide and dimethyl sulfide to produce sulfate aerosol is 132 driven by the gas-phase oxidant concentrations (OH,  $H_2O_2$ , and  $O_3$ ) and cloud pH simulated by 133 the online chemistry scheme (Paulot et al., 2016), and (3) the rate of aging of black and organic 134 carbon aerosols from hydrophobic to hydrophilic forms varies with calculated concentrations of 135 hydroxyl radical (OH), as described by Liu et al. (2011).

136 Unlike AM4.0, the AM4.1 model has an online representation of gas-phase tropospheric and 137 stratospheric chemistry. The combined tropospheric and stratospheric chemistry scheme includes 138 18 prognostic (transported) aerosol tracers, 59 prognostic gas phase, 6 prognostic ideal tracers, 139 and 41 diagnostic (non-transported) chemical tracers (Table S2), with 43 photolysis reactions, 140 190 gas-phase kinetic reactions, and 15 heterogeneous reactions (Table S3). The tropospheric 141 chemistry includes reactions of the NO<sub>x</sub>-HO<sub>x</sub>-O<sub>x</sub>-CO-CH<sub>4</sub> system and oxidation schemes for 142 other non-methane volatile organic compounds. The stratospheric chemistry accounts for the 143 major ozone loss cycles ( $O_x$ ,  $HO_x$ ,  $NO_x$ ,  $ClO_x$ , and  $BrO_x$ ) and heterogeneous reactions on liquid 144 and solid stratospheric aerosols as in Austin et al. (2013). The base chemical mechanism is

145	updated from that in AM3 (Naik et al., 2013a), using gas-phase and heterogeneous chemistry
146	updates from Mao et al. (2013ab), similar to the configuration described by Schnell et al. (2018).
147	We include heterogeneous reactions of HO2, NO2, N2O5, and NO3 on the surfaces of all
148	simulated aerosol types, with specified gamma values (given in Table S3). Note in particular that
149	$\gamma(HO_2)$ is reduced from the value of 1 recommended by Mao et al. (2013a) to 0.2. We also
150	include the heterogeneous oxidation of SO <sub>2</sub> on aerosols following Zheng et al. (2015). The
151	heterogeneous chemistry included in AM4.1 has a much stronger effect on oxidant levels than
152	that in AM3, which used $\gamma(N_2O_5)=0.1$ , $\gamma(NO_3)=0.1$ , $\gamma(NH_3)=0.05$ , $\gamma(HO_2)=0$ , and $\gamma(NO_2)=0$ ,
153	applied only to sulfate aerosols. The chemical system is solved using an implicit Euler backward
154	method with Newton-Raphson iteration, as in Horowitz et al. (2003). Photolysis rates are
155	calculated interactively using the FAST-JX version 7.1 code, as described by Li et al. (2016),
156	accounting for the radiative effects of simulated aerosols and clouds.
157	Dry deposition velocities for all aerosols are calculated interactively using a wind-driven
158	resistance method (Gallagher et al., 2002), accounting for the effect of whitecaps over the ocean
159	(Williams, 1982). The treatment of wet deposition accounts for slower removal by frozen
160	precipitation due to the Bergeron process in mixed-phase clouds (Liu et al., 2011). Dry and wet
161	deposition for gases are as described by Paulot et al. (2016).
162	Ontical properties of aerosols are precalculated as in AM4.0 using Mie theory assuming
163	sphericity. The extinction efficiency, single scattering albedo, and asymmetry parameter are
105	sphericity. The extinction efficiency, single scattering albedo, and asymmetry parameter are
164	tabulated as a function of wavelength, aerosol type, aerosol size (for dust and sea salt), and
165	relative humidity. Sulfate is assumed to be internally mixed with black carbon for the calculation
166	of optical properties. Unlike AM4.0, radiative effects of nitrate aerosol are included in AM4.1

167 (as in Paulot et al., 2017b).

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168 2.3 AMIP (1980–2014) simulation configuration

169 We conduct AMIP simulations with AM4.1 over the period 1979–2014 using observed gridded 170 sea surface temperature (SST) and sea-ice concentration boundary conditions from the 171 reconstructions of Taylor et al. (2000). Historical reconstructions of monthly solar spectral 172 irradiances are from Matthes et al. (2017). For radiation calculations, global monthly mean 173 concentrations of greenhouse gases (GHGs), including nitrous oxide (N<sub>2</sub>O), and ozone depleting 174 substances (ODSs, including CFC-11, CFC-12, CFC-113, and HCFC-22) are specified from 175 Meinshausen et al. (2017). Global mean mixing ratios of methane (CH<sub>4</sub>) and N<sub>2</sub>O are specified at 176 the surface as lower boundary conditions for chemistry. Carbon dioxide (CO<sub>2</sub>) mixing ratio is 177 restored to observed global-mean values with a one-year timescale. The simulated global-mean

CO<sub>2</sub> and CH<sub>4</sub> concentrations are used for radiation calculations.

179 2.4 Emissions

178

Annually varying time series of monthly anthropogenic and biomass burning emissions of ozone 180 181 precursors and aerosols (and their precursors) are from the Community Emissions Data System 182 (CEDS; Hoesly et al., 2018) and the data set of van Marle et al. (2017), respectively, developed 183 in support of CMIP6. Wildfire emissions are distributed vertically between the surface and 6 km, 184 with location- and biome-dependent vertical profiles, as recommended by Dentener et al. (2006), 185 similar to the treatment in AM3 (Donner et al., 2011; Naik et al., 2013a). Natural emissions of 186 NO<sub>x</sub>, CO, non-methane volatile organic compounds (NMVOC), hydrogen (H<sub>2</sub>), and ammonia 187 (NH<sub>3</sub>) are generally the same as those considered by Naik et al. (2013a), namely from the 188 Precursors of Ozone and their Effects in the Troposphere (POET) inventory for present day 189 (corresponding to year 2000) (Granier et al., 2005). Emissions of NH<sub>3</sub> from sea bird colonies, not

accounted for in AM3, are included in AM4.1 following Riddick et al. (2012). The treatment ofmarine ammonia emissions is also revised as described below.

192 Biogenic emissions of isoprene and monoterpenes are calculated online using the Parameterized 193 Canopy Environment Emission Activity (PCEEA algorithm; Guenther et al., 2006) in the Model 194 of Emissions of Gases and Aerosols from Nature (MEGAN v2.1; Guenther et al., 2012) as a 195 function of simulated air temperature and shortwave radiative fluxes, implemented as described 196 by Rasmussen et al (2012). Leaf area indices for 17 plant functional types are based on AVHRR 197 and MODIS data and are mapped to five vegetation types (Emmons et al. 2010). These 198 vegetation types and leaf area indices are independent of those simulated by the LM4.1 dynamic 199 vegetation model, due to a lack of coupling between the dynamic vegetation properties simulated 200 by LM4.1 and the atmospheric emissions module. We do not apply the soil moisture or  $CO_2$ 201 responses from Guenther et al. (2012). Future model development plans include coupling 202 biogenic emissions to LM4.1. Sea salt emissions are based on the parameterization of Monahan 203 (1986) as in CM3 (Donner et al., 2011), but are modulated by sea surface temperature following 204 Jaeglé et al. (2011). Ocean ammonia emissions are calculated following Paulot et al. (2015), 205 using the simulated seawater concentration of NH<sub>4</sub><sup>+</sup> in ESM4.1. Other marine emissions, 206 including primary organic aerosols (POA) and dimethyl sulfide (DMS), are calculated similarly 207 to in CM3. DMS emissions are calculated using an empirical formula as a function of a 208 prescribed monthly climatology of DMS concentration in sea water (Lana et al., 2011) and 209 calculated wind speed at 10 m, as described by Chin et al. (2002). Thus, oceanic emissions of 210 POA, DMS, ammonia, and sea salt are dependent on the simulated meteorology in the model.

211	Emission totals for year 2014 are shown in Table 1. Time series of annual global emissions in
212	AM4.1 (using CMIP6 inventories) are shown for select species in Figure 2, and compared with
213	corresponding totals in AM3 (using CMIP5 inventories).

214 Sources of secondary organic aerosols (SOA) include an anthropogenic source from oxidation of

215 the simulated C<sub>4</sub>H<sub>10</sub> hydrocarbon tracer by hydroxyl radical (with a 10% per-carbon yield) and a

216 biogenic pseudo-emission assuming a 10% per-carbon yield from emissions of BVOCs,

217 including isoprene and monoterpenes, from vegetation. This yield is in the range of values

218 suggested by recent studies using more detailed schemes for SOA production (e.g., Bates and

Jacob, 2019; Pai et al., 2020). In year 2014, the sources of SOA are 83.84 Tg a<sup>-1</sup> from BVOCs

and  $3.49 \text{ Tg a}^{-1}$  from anthropogenic hydrocarbon oxidation.

221 Lightning NO<sub>x</sub> emissions are calculated interactively as a function of subgrid convection in

AM4.1, as diagnosed by the double plume convection scheme described by Zhao et al. (2018b).

223 The lightning NO<sub>x</sub> source is calculated as a function of convective cloud-top height, following

the parameterization of Price et al. (1997), and is injected with the vertical distribution of

225 Pickering et al. (1998), as in AM3 (Naik et al., 2013a). The global total production of NO<sub>x</sub> by

lightning is 3.59 Tg N for year 2014.

Dust emissions are calculated dynamically online in the land component, LM4.1, as a function of wind speed, topography, vegetation cover, snow cover, soil moisture, and land type, as described by Evans et al. (2016) and Shevliakova et al. (submitted).

As in AM3, direct injection of SO<sub>2</sub> from volcanic eruptions and emissions of carbonyl sulfide

231 (COS) are not considered in AM4.1. Instead, we specify time series of stratospheric aerosol

232 optical properties, accounting for not only the volcanic contribution to stratospheric aerosol

- abundance but also other natural and anthropogenic contributions. Tropospheric emissions of
- 234 SO<sub>2</sub> from continuously degassing and explosive volcanoes are treated in the same way as in
- AM3 (Donner et al., 2011), with a climatological total of 3.59 Tg S a<sup>-1</sup>.

#### 236 **3. Results: Physical climate simulation (AMIP, 1980–2014)**

#### 237 *Surface air temperature*

- 238 Comparison of surface air temperature over land with observations from CRU TS (Figure 3)
- 239 illustrates the substantial decrease in overall root mean square error (RMSE) achieved in AM4.1
- 240 (RMSE =  $1.92^{\circ}$ C) from the previous generation full-chemistry AM3 (RMSE =  $2.18^{\circ}$ C) and
- similar, if slightly degraded, pattern to AM4.0 (RMSE = 1.85°C). The most notable difference
- from AM3 to AM4.0 and AM4.1 is an improvement in boreal warm biases and South American
- cold biases.

#### 244 Precipitation

- 245 Comparison of precipitation with observations from GPCP v2.3 (Figure 4) also illustrates the
- substantial decrease in overall RMSE achieved in AM4.1 (RMSE =  $0.83 \text{ mm d}^{-1}$ ) from the
- previous generation full-chemistry AM3 (RMSE =  $1.02 \text{ mm d}^{-1}$ ) and a similar pattern to AM4.0
- 248 (RMSE =  $0.85 \text{ mm d}^{-1}$ ). The most notable difference from AM3 to AM4.0 and AM4.1 is an
- 249 improvement in Amazon dry biases, and in wet biases over Australia and the Indian Ocean.

#### 250 *Circulation*

- 251 Comparison of zonal mean zonal winds with the ERA40 reanalysis (Figure 5) illustrates a
- substantial decrease in overall RMSE in AM4.1 (RMSE =  $1.32 \text{ m s}^{-1}$ ) from the previous
- 253 generation high-top full-chemistry AM3 (RMSE =  $1.75 \text{ m s}^{-1}$ ). The AM4.1 RMSE is greater than

254	that in the low-top AM4.0 (RMSE = $1.00 \text{ m s}^{-1}$ ), owing to a westerly wind bias in the equatorial
255	stratosphere, and a weak, equatorward-shifted Arctic stratospheric jet in AM4.1. The
256	representation of the stratospheric wintertime westerly polar jet associated with the Antarctic
257	vortex is significantly improved in AM4.1 (not shown) compared with AM3 (Donner et al.,
258	2011), in which the westerlies were excessively strong (leading to a too-cold Antarctic vortex).
259	We plan to work towards further improving the stratospheric circulation in future versions of
260	AM4.1 through improvement in our representation of parameterized gravity wave drag.
261	Tropospheric circulation patterns in AM4.1 are very similar to those in AM4.0.
262	Stratospheric variability
263	Comparison of statistics for sudden stratospheric warmings with the ERA40 reanalysis (Figure 6)
264	illustrates an improvement in AM4.1 with respect to capturing events in the coldest months
265	(December-January), which were largely missed in AM4.0 (Zhao et al., 2018a), even though
266	AM4.0 already performs quite well among low-top atmospheric models (Charlton-Perez et al.,
267	2013). In the surrounding months (November, February), AM4.1 overestimates warming events,
268	whereas AM4.0 matches the reanalysis data fairly well.
260	Dadiation Annes

#### 269 *Radiation fluxes*

- 270 Comparison of top-of-atmosphere (TOA) net radiation with CERES EBAF observations (Figure
- 271 7) further illustrates the substantial decrease in overall root mean square error (RMSE) achieved
- in AM4.1 (RMSE =  $7.2 \text{ W m}^{-2}$ ) from the previous generation full-chemistry AM3 (RMSE = 8.6
- 273 W m<sup>-2</sup>) and similar, if slightly degraded, pattern to AM4.0 (RMSE = 6.8 W m<sup>-2</sup>). The most
- 274 notable difference from AM3 to AM4.0 and AM4.1 is associated with an improvement in areas
- 275 of tropical convection along the intercontinental convergence zone (ITCZ) that had previously

276	been too absorbing and increased absorption in northern boreal regions that had been previously
277	too reflective, as discussed by Zhao et al. (2018a). The most notable differences between AM4.0
278	and AM4.1 are associated with a decrease in the global TOA from a nearly unbiased AM3 (0.02
279	W m <sup>-2</sup> ) to a slight negative bias in AM4.0 (-0.14 W m <sup>-2</sup> ) and substantial low bias in AM4.1 (-
280	0.80 W m <sup>-2</sup> ). This increase in bias is due in part to increased albedo of northern boreal regions
281	associated with snow masking depth in LM4.1, as discussed by Shevliakova et al. (submitted),
282	and also in part to differences over Antarctic associated with the prescribed albedo of snow on
283	glaciers that was modified late in the development cycle of ESM4.1 to address Southern Ocean
284	dynamics, as discussed by Dunne et al. (submitted).
285	Lightning flash frequency
286	Figure S1 shows the lightning flash frequency retrieved from the spaceborne Optical Transient
287	Detector (OTD) and Lightning Imaging Sensor (LIS) (Cecil et al., 2014), compared with
288	simulated values from AM3 and AM4.1. In both AM3 and AM4.1, lightning flash frequency is

289 parameterized as a function of convective cloud top height, following Price et al. (1997), but the

290 two models use different parameterizations of cumulus convection (Zhao et al., 2018b). While

- the overall correlation between model and observations is lower in AM4.1 than AM3, there are
- some notable areas of improvement in the representation of flash frequency, including a
- 293 reduction of the high biases present in AM3 over the Amazon and the maritime continent,
- 294 improving agreement with observations.

## 295 4. Results: Simulation of atmospheric composition

296 4.1 Ozone

In this section, we evaluate model simulations of ozone, including surface ozone concentrationsrelevant for air quality and column ozone abundances relevant for climate.

299 *Surface ozone* 

300 We focus on the seasonal mean of the maximum daily 8-hour average (MDA8) surface ozone over 301 the period 2005–2014, when observations are available from densely clustered monitoring sites 302 across northern mid-latitude populated regions (Figure 8 for MAM, Figure 9 for JJA). 303 Observations were obtained from the Tropospheric Ozone Assessment Report (TOAR) Database for 2005–2014 (Schultz et al., 2017) and a monitoring network operated since 2013 by China's 304 305 Ministry of Environmental Protection (CNMEP, http://106.37.208.233:20035/). Observations are 306 averaged onto the same 1°x1° grid as AM4.1. We compare simulated ozone from the AM4.1 AMIP 307 simulation with that from the AM3 AMIP simulation.

308 Surface MDA8 ozone in AM3 is biased high by 12 ppb on average during MAM (Figure 8b) and 309 by up to 20 ppb over the eastern U.S. during summer (Figure 9b), as documented in previous 310 studies (Lin et al., 2012a; Lin et al., 2012b; Fiore et al., 2014; Rieder et al., 2015; Lin et al., 2017). 311 AM4.1 shows substantially reduced biases in mean ozone for both spring and summer over the 312 eastern U.S. and Europe (Figures 8c and 9c). This dramatic improvement in the simulation of 313 surface ozone concentrations results from a combination of updates to the chemical mechanism 314 from AM3 to AM4.1, including updates to the isoprene oxidation scheme and the representation 315 of heterogeneous reactions (Mao et al., 2013b), and the change from CMIP5 emissions in AM3 to 316 CMIP6 emissions in AM4.1 (Section 2.3). The shallow surface layer of the model (30 m thick) 317 may also have an impact on the comparison with surface sites. Zhao et al. (2018a) found a 318 significant improvement in diagnosed 2-m temperatures associated with this shallower surface319 layer.

To further explore the causes of the differences in surface ozone abundances between AM3 and AM4.1, we conduct two additional simulations—an AM4.1 simulation with nudged meteorology and an additional AM4.1 nudged simulation with AM3-like chemistry (AM4.1\_AM3Chem; Lin et al., 2019). The two experiments use the same CMIP6 emissions and have nearly identical meteorology (as a result of the nudging), allowing us to isolate the influence of changes in chemistry alone. Seasonal-mean MDA8 ozone from these simulations are plotted in Supplementary Figures S2–S5.

Similar to the results from AM3 (Figures 8b and 9b), surface MDA8 ozone in AM4.1\_AM3Chem is biased high by 11 ppb on average during spring (Figure S2b) and by up to 20 ppb over the eastern U.S. during summer (Figure S3b). Switching the chemistry scheme from AM3 to AM4 leads to substantial reductions in mean ozone biases for both spring and summer over the eastern U.S. and Europe (Figures S2c and S3c), but the model underestimates springtime MDA8 ozone over central eastern China by 20 ppb (Figure S2c versus CNMEP observations in Figure S2a).

Observations show more severe springtime ozone pollution over central eastern China and Mexico than in the U.S. and Europe. This regional contrast is not simulated in either of our experiments. Particularly, the enhanced heterogeneous chemistry in AM4 (Section 2.2; Mao et al., 2013ab) likely leads to excessive heterogeneous loss of HO<sub>x</sub> and NO<sub>x</sub> radicals over eastern China and Mexico, where aerosol loadings are high during the spring season. For summer over the southeastern U.S., where high mean-state ozone biases are found in many current-generation CTMs and CCMs (Fiore et al., 2009; Young et al., 2018), the AM4.1 experiment shows remarkable

340 agreement with observations. However, on the basis of analysis conducted for an intensive field 341 campaign, Travis et al. (2016) suggested that the common model biases in simulating summertime 342 ozone over the southeastern U.S. may reflect a combination of excessive NO<sub>x</sub> emissions (too high 343 by 50%) and the deep model surface layer that cannot resolve near-surface ozone gradients. A 344 balanced view is needed to interpret the reduced ozone biases in the AM4.1 experiment.

Our results suggest the complexity of various sources, sinks, transport, and chemistry in influencing the simulation of surface ozone. In the future, process-based assessments, not only for means but also for variability and extreme events, are needed to fully evaluate how the choices of different emission datasets, chemistry, and deposition schemes affect simulations of surface ozone and related tracers.

#### 350 Tropospheric ozone column

351 We compare climatological annual mean tropospheric ozone columns simulated by AM3 (mean 352 over 2000–2008) and AM4.1 (2005–2014) with those derived from the OMI-MLS (Ziemke et 353 al., 2019) (Figure 10). In the analysis shown here, AM3's native ozone output on model levels is 354 used to calculate tropospheric ozone column using the WMO tropopause definition, while for 355 AM4.1, the tropospheric ozone column (tropoz) is diagnosed at every time step, by applying the 356 WMO tropopause definition using model simulated temperature. The global mean tropospheric 357 ozone column simulated by AM3 and AM4.1 are 35 DU and 31 DU, respectively, compared to 358 the OMI/MLS value of 30 DU. While AM3 showed consistent high biases globally except over 359 the Antarctic, AM4.1 shows an interhemispheric pattern in the biases with high values in the 360 Northern Hemisphere mid-latitudes and over continents and low values in the Southern 361 Hemisphere extra-tropics. This pattern is consistent with global chemistry-climate models

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evaluated against the OMI/MLS climatology by Young et al. (2013) for a slightly different time
period. An interesting feature in AM4.1 is the strong positive bias over Oceania, possibly related
to the different biomass burning emissions applied in the two models. AM3 exhibited an average
high bias of 21.7%, which has been reduced to 7.3% in AM4.1; accordingly the RMSE has been
reduced considerably, from 7.1 DU in AM3 to 4.6 DU in AM4.1.

#### 367 Total Ozone Column

368 Figure 11 shows the evaluation of modeled time series of total column ozone against two 369 datasets for 1980–2015, namely Multi-Satellite Merged Total Column NASA and NOAA 370 product from Frith et al. (2013; SBUV; open triangles) and version 3.4 of the National Institute 371 of Water and Atmospheric Research - Bodeker Scientific (NIWA-BS; closed circles) total 372 column ozone database. AM3 results are plotted for 1980-2008 period, while AM4.1 results are 373 for 1980–2014. The comparison is shown for the annual average globally, in the tropics, and in 374 southern and northern midlatitudes, and for March in the Arctic and October in the Antarctic. 375 Globally (Figure 11a), absolute values of total column ozone for AM3 were biased high compared to both datasets, whereas AM4.1, on the other hand, is biased low. Both models 376 377 generally capture the trend in total column ozone, although the evaluation of AM3 is truncated at 378 2008. As suggested by the greater correlation coefficients for AM4.1 compared with AM3, 379 AM4.1 is able to better capture the observed interannual variability and trends of global mean 380 total column ozone. In the tropics (Figure 11b), total ozone column values remain lower than 381 observed in AM4.1, as opposed to higher in AM3. Consistent with observations, both the models 382 simulate negligible trends in total column ozone in the tropics, however AM4.1 exhibits greater 383 skill in capturing the observed evolution of total column ozone. In the northern mid-latitudes 384 (Figure 11c), AM4.1 differs more from observations than AM3 does, although with fairly similar

385	skill in simulating the observed time evolution of total column ozone. The comparison is
386	opposite for the southern mid-latitudes (Figure 11d), where AM4.1 is much closer to observed
387	values than AM3 with similar correlations. In the Arctic in March (Figure 11e), AM4.1
388	reproduces the observed total ozone column values slightly better than AM3, however both have
389	fairly low skill in reproducing the observed evolution. In the Antarctic in October (Figure 11f),
390	AM4.1 exhibits greater skill in simulating ozone depletion compared to AM3 both in terms of
391	trends and absolute values. This improvement likely results from the improved dynamical
392	representation of the Antarctic polar vortex in AM4.1 (Section 3). Overall, AM4.1 compares
393	slightly better against observations of total column ozone than AM3.
394	4.2 Carbon monoxide

The simulated tropospheric CO columns are evaluated against CO retrievals from the MOPITT (Measurements of Pollution in The Troposphere) instrument in Figure 12. We use the MOPITT V8 Joint (NIR+TIR) retrievals (Deeter et al., 2019) during 2001–2014, which are available from the NASA Earthdata archive (https://earthdata.nasa.gov). The model is interpolated to the

399 gridded monthly MOPITT observations and the averaging kernel for each grid is applied to the

400 simulated monthly mean CO profiles.

401 The tropospheric CO columns are in general higher in AM4.1 than AM3, in better agreement

402 with MOPITT retrievals in terms of magnitudes (RMSE reduced from  $(2.6-2.7) \times 10^{17}$  cm<sup>-2</sup> to

403  $(1.6-1.8)x10^{17}$  cm<sup>-2</sup>) and spatial distribution (r<sup>2</sup> increased from 0.7–0.9 to 0.8–0.9). Compared to

404 AM3, AM4.1 reduces the underestimations in column CO in the Northern Hemisphere, but

405 overestimates column CO in the Southern Hemisphere, especially during summer. This is in part

406 due to lower OH levels in AM4.1 than AM3.

- 407 To evaluate surface CO, we use measurements from a globally distributed network of air
- 408 sampling sites maintained by the Global Monitoring Laboratory (GML) of the National Oceanic
- 409 and Atmospheric Administration (NOAA) (Pétron et al., 2019; data available at
- 410 ftp://aftp.cmdl.noaa.gov/data/trace\_gases/co/flask/). Surface CO observations during 1988–2014
- 411 are used to evaluate model performance (Figure 13).
- 412 AM4.1 simulates higher surface CO concentrations than AM3 over the Southern Hemisphere,
- 413 and slightly overestimates surface CO concentrations by < 5 ppb when compared to surface
- 414 observations. Over the Northern Hemisphere, AM4.1 largely reduces the negative biases that
- 415 occurred in AM3, with a mean bias of  $\pm 20$  ppb over most GMD sites. This is consistent with the
- 416 comparisons to the MOPITT retrievals shown above. In addition, compared to AM3, AM4.1
- 417 better captures the seasonal cycles (with correlation coefficient R > 0.5) at most sites and better
- 418 captures the latitudinal gradient as well (R = 1.0 versus R = 0.9).
- 419 Comparisons of surface CO concentrations over pristine sites show significant improvement in
- 420 AM4.1 over AM3 across latitudes from South to North. In the Southern Hemisphere, such as at
- 421 South Pole (SPO), Ushuaia (USH), and Easter Island (EIC) sites, the underestimation of surface
- 422 CO concentrations by AM3 are reduced in AM4.1. In the Northern Hemisphere, such as at
- 423 Mauna Loa (MLO), Barrow (BRW), and Alert (ALT) sites, both surface concentrations and
- 424 monthly variations are improved significantly in AM4.1 compared to AM3. These improvements
- 425 are mainly associated with improved chemistry in AM4.1.

426 4.3 Aerosols

- 427 We first evaluate concentrations of aerosols in surface air. Figure 14 (top panels) compare
- 428 simulated concentrations of sulfate and nitrate aerosols from AM4.1 with observations over the

429	United States from the IMPROVE network. The model successfully captures the wide range of
430	observed sulfate aerosol concentrations. While nitrate concentrations are well correlated with
431	observations (R=0.74), simulated concentrations are generally too high (normalized mean bias
432	(NMB) = +80%). This bias is larger than in Paulot et al. (2016), where nitrate aerosols are
433	assumed to deposit rapidly like nitric acid. Simulated concentrations of sulfate and nitrate in
434	precipitation are compared with observations from the NADP network in the lower panels of
435	Figure 14. The rainwater abundances of sulfate and nitrate are well correlated with observations,
436	but with a low bias for sulfate (NMB = -19%) and a high bias for nitrate (NMB = $+35\%$ ).
437	Figure 15 compares simulated concentrations of sulfate, dust, and sea salt aerosols from AM3,
438	AM4.0, and AM4.1 with observations from the University of Miami network (Savoie and
439	Prospero, 1977). The model successfully captures the wide range of observed sulfate aerosol
440	concentrations. For sulfate, the RMS error versus observations is reduced in AM4.1 (0.20 $\mu g$
441	m <sup>-3</sup> ) from AM3 and AM4.0 (both 0.22 $\mu$ g m <sup>-3</sup> ), and the correlation is improved (r = 0.93 in
442	AM4.1, $r = 0.89$ in AM3 and AM4.0). The agreement between simulated and observed dust
443	improves from AM3 to AM4.0, but then degrades in AM4.1, reflecting the shift from prescribed
444	to interactive source regions for dust in LM4.1 (Shevliakova et al., submitted). The RMS error
445	for simulated sea salt is reduced significantly in AM4.1 (0.35 $\mu g~m^{\text{-}3})$ compared with AM3 (0.47
446	$\mu g$ m^-3) and AM4.0 (0.49 $\mu g$ m^-3), as a result of updates to the emissions and deposition
447	parameterizations in AM4.1.
448	We next evaluate the simulated AOD against measurements from the AERONET sunphotometer

449 network (Holben et al., 1998) in Figure 16. Here we use the quality assured and cloud screened

- 450 level 2 version 2 AOD data (Smirnov et al., 2000). For comparison, we also show the results
- 451 from AM4.0 (middle) and AM3 (bottom). Both AM4.0 and AM4.1 exhibit higher correlation

(0.89 and 0.9) and lower RMS (0.07 and 0.08) with AERONET observations than AM3 (0.81
and 0.09, respectively). In particular, the large positive biases in the tropics and equatorial
regions are reduced, which reflects the more efficient removal of aerosol by convective
precipitation (Paulot et al., 2016). AM4.1 exhibits a greater positive bias than AM4.0 over the
Midwest United States, associated with higher dust loading and nitrate aerosol (not included in

457 AM3 and AM4.0).

458 Figures 17 and S6 and compare the simulated regional monthly mean AOD simulated by AM3,

459 AM4.0, and AM4.1 with observations from the MODIS (Levy et al., 2007) and MISR (Kahn et

460 al., 2009) instruments. AM4.0 and AM4.1 have reduced the seasonal contrast between winter

461 and summer months, in better agreement with observational constraints. The spring maximum

462 over East Asia and the North Pacific is also better captured with AM4.0 and AM4.1. The AM3

463 high bias over the Caribbean Sea and maritime continent are reduced consistent with the

464 comparison against AERONET.

466

465 These improvements primarily reflect changes in the treatment of aerosol removal, including

reduced removal by frozen precipitation formed by the Bergeron process and more efficient

467 scavenging by convective precipitation (Paulot et al., 2016). AM4.1 exhibits greater bias over

468 Asia than AM4.0, which primarily reflects higher optical depth from dust and ammonium nitrate.

469 Uncertainties in Asian SO<sub>2</sub> and NH<sub>3</sub> emissions (Zhang et al., 2009) and aerosol hygroscopic

470 growth may also contribute to the AM4.1 high bias over this region (Paulot et al., 2018).

471 4.4 Hydroxyl Radical (OH) and Methane Lifetime

472 Here, we evaluate the climatological mean hydroxyl (OH) radical simulated by AM4.1, as OH is

473 the primary atmospheric oxidant determining the abundance and lifetime of several short-lived

21

474	climate forcers, including methane. The simulation of OH depends on the chemical mechanism,
475	particularly the representation of isoprene photooxidation (Archibald et al., 2010; Bates and
476	Jacob, 2019). Differences in emissions, meteorology, and photochemical mechanisms across
477	models also lead to differences in OH (Murray et al., submitted 2019; Wild et al., 2020).
478	Climatological mean (1980-2014) global airmass-weighted tropospheric OH simulated by
479	AM4.1 is 10.4x10 <sup>5</sup> molecules cm <sup>-3</sup> , about 18% lower than that simulated by AM3, but is within
480	the range of values reported for ACCMIP models for the 2000s (Naik et al., 2013b).
481	Consequently, the mean whole-atmosphere chemical lifetime of methane (calculated as the
482	global methane burden divided by global total loss) in AM4.1 is 8.5 years; lifetime against loss
483	by reaction with tropospheric OH is 9.7 years, which is 13% greater than the AM3 value of 8.6
484	years (1981–2000), but still lower than the observationally derived estimate of 11.2±1.3 years
485	(Prather et al., 2012). Figure 18 shows the comparison of tropospheric OH distribution for 12
486	regions simulated by AM4.1 with estimates from AM3, ACCMIP ensemble mean, and the
487	climatology of Spivakovsky et al. (2000). AM4.1 simulates reduced OH levels compared to
488	AM3 throughout the troposphere, possibly because of differences in emissions and chemical
489	mechanisms between the two model versions. In particular, the lower lightning $NO_x$ in AM4.1
490	versus AM3 acts to lower OH because of the strong sensitivity of OH to lightning NO <sub>x</sub> emissions
491	(Murray et al., 2013). Relative to the Spivakovsky et al. climatology, AM4.1 exhibits a reduced
492	high bias compared with AM3, but has too low OH, particularly in the tropical upper
493	troposphere.

## 494 5. Sensitivities to Greenhouse Gases, Aerosols, and SST Perturbations

495 Table 2 shows the net radiative flux perturbations that result from historical changes in

496 anthropogenic forcing agents, and from idealized changes in CO<sub>2</sub> and SST. Comparison of these

497	radiative metrics between AM3, AM4.0 and AM4.1 indicates that effective radiative forcings
498	(ERF) from pre-industrial to present-day changes in greenhouse gases and aerosols are nearly
499	identical between AM4.0 and AM4.1. However, the ERF from quadrupling CO <sub>2</sub> is significantly
500	lower in AM4.1, mostly because of the inclusion of interactive ozone (colder stratospheric
501	temperatures reduce the rate of ozone chemical loss), but also partially resulting from increased
502	dust emissions from LM4.1 (related to increased fires under elevated-CO <sub>2</sub> conditions). The Cess
503	feedback, the change in net radiative flux resulting from an increase of SSTs by 2K, is
504	significantly more negative in ESM4.1 (corresponding to a weaker Cess sensitivity), likely
505	resulting from increased emissions of salt, dust, and BVOCs with increasing temperatures in
506	ESM4.1. While comparison with previous-generation models is complicated by changes in the
507	AMIP configuration since the AM3 model simulations were conducted (in particular, updating
508	the "present-day" conditions from representing 1990 conditions to 2014 conditions), some
509	assessment of these differences can be made using AM4.0 simulations conducted for 1990
510	conditions (as in Zhao et al., 2018a). The most important differences between AM3 and AM4.0
511	are a decrease in the magnitude of the negative aerosol ERF from AM3 to AM4.0, an increase in
512	the $4xCO2$ ERF consistent with an update to the treatment of $CO_2$ radiative bands (Zhao et al.,
513	2018b), and a strengthening of the negative Cess feedback. The decrease in the magnitude of the
514	aerosol ERF from AM3 to AM4.0 has been attributed by Zhao et al. (2018b) to a decrease in the
515	strength of the aerosol indirect effect, resulting from the increase in horizontal resolution and
516	improvements to the representations of aerosol convective wet deposition (Paulot et al., 2016)
517	and aerosol activation.

### 518 **6. Summary**

519	AM4.1 includes considerable advances in resolution and physics as in AM4.0 (Zhao et al,
520	2018ab) as well as a comprehensively revised suite of chemistry parameterizations to improve
521	consistency in treatment across species and with advances in the underlying science over the last
522	decade. AM4.1 is able to maintain the fidelity of AM4.0 while substantially increasing in
523	comprehensiveness and associated climate-chemistry interactions and feedbacks. This
524	development effort has also led to considerable improvement in model fidelity compared to
525	GFDL's previous-generation coupled chemistry-climate model (AM3) with respect to observed
526	atmospheric composition for aerosol, CO, ozone, as well as climate phenomena such as sudden
527	stratospheric warmings.

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800

## 801 Figure captions

- 802 Figure 1. Schematic description of forcing, dynamics, physics, aerosol, and chemistry
- 803 interactions in AM4.1. Terms depicted in gray (left) are prescribed as inputs to the model, while
- 804 chemical processes included in the orange box are calculated interactively within the
- atmospheric model. The light blue box (top) includes physical processes calculated in AM4.1.
- The green box (bottom left) represents the land component (LM4.1), which is coupled to AM4.1.
- 807 The dark blue box (bottom right) includes specified ocean surface boundary conditions.
- **Figure 2.** Global annual totals (in Tg a<sup>-1</sup>, using mass as indicated on y-axis label) for
- 809 anthropogenic (fossil fuel + biomass burning + ship + aircraft) emissions of NO, CO, SO<sub>2</sub>, NH<sub>3</sub>,
- 810 BC, and primary OM in AM3 (CMIP5 emissions) and AM4.1 (CMIP6 emissions) AMIP
- 811 simulations.
- Figure 3. Annual mean surface air temperatures (°C) in AM4.1 AMIP simulation (1980-2014)
- and CRU-TS-3.22 observations (1979-2013). Differences between simulated and observed
- 814 surface air temperatures in AM4.1, AM4.0, and AM3 AMIP simulations.
- 815 **Figure 4.** Annual mean precipitation (mm day<sup>-1</sup>) for 1980-2014 in AM4.1 AMIP simulation and
- 816 GPCP v2.3 observations. Differences between simulated and observed precipitation in AM4.1,
- 817 AM4.0, and AM3 AMIP simulations.
- 818 **Figure 5.** Annual mean zonal mean zonal wind (m s<sup>-1</sup>) in AM4.1 AMIP simulation (1980-2014)
- and ERA40 reanalysis (1981-2000). Differences between simulated and observed zonal winds in
  AM4.1, AM4.0, and AM3 AMIP simulations.
- 821 Figure 6. Monthly and annual (ANN) stratospheric sudden warming (SSW) frequency for 1870–
- 822 2014 from AM4.0 and AM4.1, and 1957–2002 from ERA40. SSW is defined as in Charlton and
- 823 Polvani (2007). Error bars indicate the 95% confidence interval (the statistical test of the SSW
- frequency is calculated as in Charlton et al., 2007).
- 825 **Figure 7.** Annual mean net radiation flux at top of atmosphere (W m<sup>-2</sup>) in AM4.1 AMIP
- simulation (1980-2014) and CERES EBAF v2.8 observations (2000-2015). Differences between
- simulated and observed net radiation flux in AM4.1, AM4.0, and AM3 AMIP simulations.
- 828 Figure 8. MAM mean surface MDA8 ozone mixing ratios (ppbv) for 2000-2008 from: (a) TOAR
- 829 observations regridded to the same 1° x 1° grid as AM4.1, (b) AM3 AMIP simulation, (c) AM4.1
- 830 AMIP simulation. Here, mn is the mean and rmsd is the root-mean-square deviation between
- 831 observations and simulations.
- 832 **Figure 9.** Same as Figure 8, but for JJA.

- 833 Figure 10. Climatological mean tropospheric ozone column in AM3 (upper left; Dobson Units,
- 834 DU), AM4.1 (lower left; DU), and the % bias compared to the OMI/MLS satellite estimate of the
- Tropospheric Ozone Column (Ziemke et al., 2019) for AM3 (upper right; %) and AM4.1 (lower 835
- 836 right; %). RMSE is provided in DU.
- 837 Figure 11. Comparison of time series of total ozone column (DU) for the annual mean (a) global
- 838 mean (90°S-90°N), (b) tropics (25°S-25°N), (c) northern mid-latitudes (35°N-60°N), (d)
- 839 southern mid-latitudes (35°S-35°N), and for the (e) March mean in the Arctic (60°N-90°N) and
- 840 (f) October mean in the Antarctic (60°S-90°S) from AM3 (red) and AM4.1 (blue) against NASA
- 841 and NOAA observations from the multi-satellite merged ozone total column [Frith et al., 2013]
- 842 (SBUV; open triangles) and version 3.4 of the NIWA-BS total column ozone database [Bodeker
- 843 et al., 2005] (NIWA; closed circles). The numbers in each panel indicate linear correlation 844 coefficient (R) for model against each of the measurement datasets (top for NIWA and bottom
- 845 for SBUV).
- 846 Figure 12. Absolute difference in tropospheric CO column between AM3 and MOPITT (left
- panel) and AM4.1 and MOPITT (right panel) for winter (December-January-February, DJF, top) 847 848 and summer (June-July-August, JJA, bottom).
- 849 Figure 13. Comparison of surface CO mixing ratios (ppbv) from AM4.1 (red) and AM3 (blue)
- 850 against NOAA Global Monitoring Division (GMD) flask observations (Pétron et al., 2019, for
- 851 1988-2014). Left panels show model bias (top) and correlation coefficient (bottom) versus
- 852 observations, plotted by station latitude. Right panels show monthly time series comparisons at
- 853 selection stations. The root mean square error (RMSE) and correlation coefficient (R) are
- 854 indicated on plots.
- 855 Figure 14. Comparison of AM4.1 (2000-2014) against IMPROVE (a,b) and NADP (c,d)
- 856 observations of concentrations in surface air (top) and in precipitation (bottom) of sulfate (left)
- 857 and nitrate (right).
- Figure 15. Comparison of simulated (AM3, 1979-2008; AM4.0, 1980-2014; and AM4.1, 1980-858
- 859 2014) and observed (University of Miami) annual mean surface concentrations (ug  $m^{-3}$ ) of (first
- 860 row) sulfate, (second row) dust, and (third row) sea salt sodium at 28 locations and (bottom) their
- 861 ratios (simulated/observed) at each location (for AM4.1 only). Shaded contours indicate
- 862 simulated surface concentrations (top colorbar) and symbols indicate the ratio of
- 863 simulated/observed concentrations (bottom colorbar, symbol points upwards if ratio greater than
- 864 one, downwards if less than one).
- 865 Figure 16. Comparison of simulated aerosol optical depths (550 nm) with AERONET
- observations over the 2000-2014 period for (top) AM4.1, (middle) AM4.0, and (bottom) AM3 866
- 867 AMIP simulation. Dashed lines in left panels denote slopes of 0.5 and 2. Color in right panels

- 868 shows the percentage difference between model and AERONET (i.e., 100%× [model-
- 869 AERONET]/AERONET).
- Figure 17. Monthly climatology (2003–2014) of aerosol optical depth simulated by AM3
- 871 (purple line), AM4.0 (green line) and AM4.1 (orange line) and measured by MODIS (TERRA:
- star, AQUA: cross) and MISR (filled circles) satellite instruments. Each panel represents a
- spatial average over the corresponding region on the background map. The numbers in each box
- 874 show the correlation coefficients (left) and normalized root mean square error (right) compared
- to MODIS-TERRA (purple: AM3, green: AM4.0, orange: AM4.1).
- 876 Figure 18. Climatological (1980-2014) annual mean airmass-weighted tropospheric OH
- 877 concentration averaged globally (top-most row) and regionally for individual atmospheric
- subdomains from AM4.1 (black) compared with those from AM3 (1980-2008, red), ACCMIP
- 879 ensemble mean (orange), and climatological mean values from Spivakovsky et al. (2000)
- 880 (purple). Values for AM4.1 and AM3 also show +/- standard deviation about the mean.
## 881 Tables

Species	Units	anthro	biomass burning	biogenic / natural	ocean	animals	soil	ship	aircraft	Total
ACET	TgC/yr	1.47	0.98	15.09	0	0	0	0	0	17.53
BC	TgC/yr	7.83	1.77	0	0	0	0	0.17	0	9.76
C2H4	TgC/yr	4.88	3.82	0	0	0	0	0.14	0	8.83
C2H5OH	TgC/yr	2.40	0.07	4.82	0	0	0	0	0	7.29
C2H6	TgC/yr	5.22	2.71	0.80	0.78	0	0	0.17	0	9.67
C3H6	TgC/yr	9.50	5.85	0.85	1.29	0	0	0.16	0	17.66
C3H8	TgC/yr	5.05	0.53	1.63	1.05	0	0	0.49	0	8.76
C4H10	TgC/yr	52.93	2.34	0	0	0	0	1.10	0	56.38
C10H16	TgC/yr	0	1.24	57.37	0	0	0	0	0	58.61
CH2O	TgC/yr	1.00	1.94	0	0	0	0	0	0	2.94
СНЗОН	TgC/yr	0.30	3.24	85.61	0	0	0	0	0	89.14
СО	Tg/yr	612.40	356.68	159.24	19.80	0	0	0.69	0.57	1149.37
DMS	Tg/yr	0	0	0	42.72	0	0	0	0	42.72
DUST	Tg/yr	0	0	0	0	0	2507.67	0	0	2507.67
H2	Tg/yr	24.50	9.01	0	2.98	0	2.98	0.03	0	39.48
ISOP	Tg/yr	0.00	0.57	499.78	0	0	0	0.00	0	500.36
NH3	Tg/yr	60.82	4.30	0	3.89	0.15	2.95	0.02	0	72.13
NO	TgN/yr	35.52	6.23	3.29	0	0	3.59	6.89	0.93	56.45
ОМ	Tg/yr	31.26	26.81	0	16.21	0	0	0.20	0	74.48
SSALT	Tg/yr	0	0	0	6254.24	0	0	0	0	6254.24
SO2	TgS/yr	51.26	1.14	3.59	0	0	0	4.44	0.14	60.56

# **Table 1.** Annual total emissions for year 2014 in AM4.1.

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- **Table 2.** Effective radiative forcings and feedbacks (in W m<sup>-2</sup>) to greenhouse gases, aerosols,
- anthropogenic forcings, land use, quadrupling  $CO_2$ , and SST perturbations. The AM3 (1990) and
- AM4.0 (1990) experiments are identical to those described by Zhao et al. (2018a). The AM4.0
- 887 (2014) and AM4.1 (2014) experiments use 30-year climatological simulations, i.e., the piClim
- 888 experiments from the Radiative Forcing Model Intercomparison Project (RFMIP; Pincus et al.,
- 889 2016).

	AM3 (1990)	AM4.0 (1990)	AM4.0 (2014)	AM4.1 (2014)
GHG ERF	2.63	2.61	3.14	3.22
Aerosol ERF	-1.69	-0.96	-0.73	-0.70
Anthro ERF	N/A	N/A	2.33	2.37
LU ERF	N/A	N/A	-0.33	-0.28
4xCO2 ERF	7.19	N/A	8.23	7.72
Cess feedback (SST +2K)	-2.86	-3.59	-3.64	-4.14

890

















60N

30N

EQ

30S

60S-

60

45

32

21

12

5

-5

-12

-21

-32

-45

-60

















ESM4\_amip\_D1 1980-2014



min=0.027 \_\_\_\_\_ max=1695











Journal of Advances in Modeling Earth Systems

Supporting Information for

## The GFDL Global Atmospheric Chemistry-Climate Model AM4.1: Model Description and Simulation Characteristics

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Table S1 Table S2 Table S3

### Introduction

The tables contained in this electronic supplement describe the chemical tracers, photolysis reactions, and kinetic reactions included in AM4.1.

Figure S1 shows an evaluation of lightning flash frequency in AM3 and AM4.1 against spaceborne Optical Transient Detector (OTD) and Lightning Imaging Sensor (LIS) retrievals. Figures S2–S5 show evaluations of mean surface MDA8 O3 concentrations for 2005–2014 from model simulations and observations for MAM, JJA, SON, and DJF seasons. Figure S6 shows an evaluation of the monthly regional climatology of aerosol optical depth simulated by AM3, AM4.0, and AM4.1 with retrievals from MODIS and MISR satellite instruments. Bars denote the speciated aerosol optical depth in AM4.1.



**Figure S1.** Lightning flash frequency (flash km<sup>-2</sup> a<sup>-1</sup>) from (a) spaceborne Optical Transient Detector (OTD) and Lightning Imaging Sensor (LIS) retrievals (Cecil et al., 2014) and parameterized in (b) AM<sub>3</sub> and (c) AM<sub>4.1</sub>.



**Figure S2.** MAM mean surface MDA8 O<sub>3</sub> concentrations for 2005-2014 from: (a) TOAR (inset map for CNMEP 2013-2017) observations on the same 1° x 1° grid used for AM4.1, (b) AM4.1 simulation with AM<sub>3</sub>-like chemistry, (c) standard AM4.1 simulation. Here, mn is the mean and rmsd is the root-mean-square deviation between observations and simulations.



Figure S<sub>3</sub>. Same as Figure S<sub>2</sub>, but for JJA.



Figure S4. Same as Figure S2, but for SON.



Figure S5. Same as Figure S2, but for DJF.



**Figure S6.** Monthly climatology (2003–2014) of aerosol optical depth simulated by AM3 (purple line), AM4.0 (green line) and AM4.1 (orange line) and measured by MODIS (TERRA: star, AQUA: cross) and MISR (filled circles) satellite instruments. Each panel represents a spatial average over the corresponding region on the background map. The numbers in each box show the correlation coefficients (left) and normalized root mean square error(right) compared to MODIS-TERRA (purple: AM3, green: AM4.0, orange: AM4.1). Bars denote the speciated aerosol optical depth in AM4.1 (species indicated by shading, with colors indicated below).

 Table S1. Chemical species represented in AM4.1.

 Table S2. Photolysis reactions represented in AM4.1.

 Table S3. Kinetic reactions represented in AM4.1.

Name	Description		
Aerosol tracers			
SO4	Sulfate aerosol		
ВСРНОВ	Hydrophobic black carbon aerosol		
BCPHIL	Hydrophilic black carbon aerosol		
OMPHOB	Hydrophobic organic aerosol		
OMPHIL	Hydrophilic organic aerosol		
SOA	Secondary organic aerosol		
DUST1	Mineral dust aerosol (r=0.1-1.0µm)		
DUST2	Mineral dust aerosol (r=1.0-2.0µm)		
DUST3	Mineral dust aerosol (r=2.0-3.0µm)		
DUST4	Mineral dust aerosol (r=3.0-6.0µm)		
DUST5	Mineral dust aerosol (r=6.0-10.0µm)		
SSALT1	Sea salt aerosol (r=0.1-0.5µm)		
SSALT2	Sea salt aerosol (r=0.5-1.0µm)		
SSALT3	Sea salt aerosol (r=1.0-2.5µm)		
SSALT4	Sea salt aerosol (r=2.5-5.0µm)		
SSALT5	Sea salt aerosol (r=5.0-10.0µm)		
Gas-phase tracers			
O3	Ozone		
O (d)	Atomic oxygen		
O1D (d)	O( <sup>1</sup> D)		
N2O	Nitrous oxide		
N (d)	Atomic nitrogen		
NO	Nitric oxide		
NO2	Nitrogen dioxide		
NO3	Nitrate radical		
HNO3	Nitric acid		
HO2NO2	Pernitric acid		
N2O5	Dinitrogen pentoxide		

CH4	Methane
CH3O2 (d)	Methylperoxy radical
СНЗООН	Methylhydroperoxide
CH2O	Formaldehyde
СО	Carbon monoxide
OH (d)	Hydroxyl radical
HO2 (d)	Hydroperoxy radical
H2O2	Hydrogen peroxide
С3Н6	Propane
ISOP	Isoprene (C5H8)
PO2 (d)	C <sub>3</sub> H <sub>6</sub> OHO <sub>2</sub>
СН3СНО	Acetaldehyde (CH <sub>3</sub> CHO)
POOH (d)	C <sub>3</sub> H <sub>6</sub> OHOOH
CH3CO3 (d)	Acetylperoxy radical
CH3COOOH (d)	Peroxyacetic acid
PAN	Peroxyacetyl nitrate (CH <sub>3</sub> CO <sub>3</sub> NO <sub>2</sub> )
C2H6	Ethane
C2H4	Ethene
C4H10	Butane
MPAN	CH2CCH3CO3NO2
ISOPO2 (d)	HOCH <sub>2</sub> COOCH <sub>3</sub> CHCH <sub>2</sub>
MVK	CH <sub>2</sub> CHCOCH <sub>3</sub>
MACR	CH <sub>2</sub> CCH <sub>3</sub> CHO
MACRO2 (d)	CH <sub>3</sub> COCHO <sub>2</sub> CH <sub>2</sub> OH
MACROOH (d)	CH <sub>3</sub> COCHOOHCH <sub>2</sub> OH
C2H5O2 (d)	Ethylperoxy radical
C2H5OOH (d)	Ethylhydroperoxide
C10H16	Terpene
С3Н8	Propane
C3H7O2 (d)	Propylperoxy radical

CH3COCH3	Acetone
СНЗОН	Methanol
С2Н5ОН	Ethanol
GLYALD	HOCH <sub>2</sub> CHO
HYAC	CH <sub>3</sub> COCH <sub>2</sub> OH
EO2 (d)	HOCH <sub>2</sub> CH <sub>2</sub> O <sub>2</sub>
EO (d)	HOCH <sub>2</sub> CH2O
ISOPOOH	HOCH <sub>2</sub> COOHCH <sub>3</sub> CHCH <sub>2</sub>
H2	Molecular hydrogen
SO2	Sulfur dioxide
DMS	Dimethyl sulfide (CH <sub>3</sub> SCH <sub>3</sub> )
NH3	Ammonia
NH4NO3	Nitrate aerosol
NH4	Ammonium aeorsol
HC1	Hydrochloric acid
HOC1	Hypochlorous acid
ClONO2	Chlorine nitrate
Cl	Atomic chlorine
ClO	Chlorine monoxide
C12O2	Dichlorine dioxide
C12	Molecular chlorine
HOBr	Hypobromous acid
HBr	Hydrobromic acid
BrONO2	Bromine nitrate
Br	Atomic bromine
BrO	Bromine monoxide
BrC1	Bromine chloride
H (d)	Atomic hydrogen
H2O	Water vapor
ROH (d)	С3Н7ОН
RCHO (d)	C <sub>2</sub> H <sub>5</sub> CHO

ISOPNB	HOCH <sub>2</sub> C(CH <sub>3</sub> )=CHCH <sub>2</sub> ONO <sub>2</sub>
ISOPNBO2 (d)	HOC5H7(OH)(O2)ONO2
MACRN	HC(O)C(CH <sub>3</sub> )=CHCOOH
MVKN	HC(O)C(CH <sub>3</sub> )=CHCOOH
R4N2	RONO <sub>2</sub>
MEK (d)	C <sub>2</sub> H <sub>5</sub> C(O)CH <sub>3</sub>
R4N1	RO <sub>2</sub> from R4N2
IEPOX	C5H10O3
IEPOXOO (d)	RO <sub>2</sub> from IEPOX
GLYX	СНОСНО
MGLY	CH <sub>3</sub> COCHO
MVKO2 (d)	HOCH <sub>2</sub> CH(O <sub>2</sub> )C(O)CH <sub>3</sub>
MVKOOH (d)	HOCH <sub>2</sub> CH(OOH)C(O)CH <sub>3</sub>
MACRNO2 (d)	RO2 from MACRN
MAO3 (d)	$CH_2=C(CH_3)C(O)OO$
MAOP (d)	CH2=C(CH3)C(O)OOH
MAOPO2 (d)	RO <sub>2</sub> from MAOP
ATO2 (d)	CH <sub>3</sub> C(O)CH <sub>2</sub> O <sub>2</sub>
АТООН	CH <sub>3</sub> C(O)CH <sub>2</sub> OOH
INO2 (d)	O2NOCH2C(OO)(CH3)CH=CH2
INPN (d)	O2NOCH2C(OOH)(CH3)CH=CH2
ISNOOA (d)	Peroxy radical from ISN1
ISN1 (d)	CH2CHCCH3OOCH2ONO2
Idealized tracers	
LCH4 (d)	Loss of CH <sub>4</sub> by reaction with OH
O3S	Stratospheric O <sub>3</sub>
O3S_E90	Stratospheric O <sub>3</sub> (tagged by E90)
E90	Tropopause tracer
AOANH	Northern hemisphere tracer lifetime
NH50	Northern hemisphere transport tracer
NOy (d)	Total reactive nitrogen

Cly (d)	Total reactive chlorine
Bry (d)	Total reactive bromine

Tracers marked with (d) are diagnostic (i.e., not transported).

Reaction
$O2 + hv \rightarrow 2*O$
$O3 + hv \rightarrow O1D + O2$
$O3 + hv \rightarrow O + O2$
$N2O + hv \rightarrow O1D + N2$
$NO + hv \rightarrow N + O$
$NO2 + hv \rightarrow NO + O$
$N2O5 + hv \rightarrow NO2 + NO3$
$HNO3 + hv \rightarrow NO2 + OH$
$NO3 + hv \rightarrow .89*NO2 + .11*NO + .89*O$
$HO2NO2 + hv \rightarrow NO2 + HO2$
$CH3OOH + hv \rightarrow CH2O + HO2 + OH$
$CH2O + hv \rightarrow CO + HO2 + H$
$CH2O + hv \rightarrow CO + H2$
$H2O + hv \rightarrow OH + H$
$H2O2 + hv \rightarrow 2*OH$
$CH3CHO + hv \rightarrow CH3O2 + CO + HO2$
$POOH + hv \rightarrow CH3CHO + CH2O + HO2 + OH$
$CH3COOOH + hv \rightarrow CH3O2 + OH + \{CO2\}$
$PAN + hv \rightarrow .6*CH3CO3 + .6*NO2 + .4*CH3O2 + .4*NO3$
$MPAN + hv \rightarrow MAO3 + NO2$
$MACR \rightarrow 1.34*HO2 + .66*MAO3 + 1.34*CH2O + 1.34*CH3CO3$
$MACR \rightarrow .66^{*}OH + 1.34^{*}CO$
$MVK + hv \rightarrow .7*C3H6 + .7*CO + .3*CH3O2 + .3*CH3CO3$
$C2H5OOH + hv \rightarrow CH3CHO + HO2 + OH$
$C3H7OOH + hv \rightarrow 0.82*CH3COCH3 + OH + HO2$
$CH3COCH3 + hv \rightarrow CH3CO3 + CH3O2$
$MGLY + hv \rightarrow CH3CO3 + CO + HO2$
$GLYX + hv \rightarrow 2.0*CO + 2.0*HO2$
$ISOPOOH + hv \rightarrow .402*MVK + .288*MACR + .69*CH2O + HO2$

 $HYAC + hv \rightarrow CH3CO3 + HO2 + CH2O$ 

 $GLYALD + hv \rightarrow 2*HO2 + CO + CH2O$ 

 $ISOPNB + hv \rightarrow RCHO + NO2 + HO2$ 

 $MACRN + hv \rightarrow NO2 + HYAC + MGLY + .5*CH2O + HO2 + .5*CO$ 

 $MVKN + hv \rightarrow GLYALD + NO2 + CH3CO3$ 

 $R4N2 + hv \rightarrow NO2 + .32*CH3COCH3 + .19*MEK + .18*CH3O2 + .27*HO2 +$ 

.32\*CH3CHO + .13\*RCHO

 $ClONO2 + hv \rightarrow Cl + NO3$ 

 $HOCl + hv \rightarrow OH + Cl$ 

 $Cl2O2 + hv \rightarrow 2*Cl + O2$ 

 $BrONO2 + hv \rightarrow Br + NO3$ 

 $HOBr + hv \rightarrow OH + Br$ 

 $BrCl + hv \rightarrow Br + Cl$ 

 $BrO + hv \rightarrow Br + O$ 

 $Cl2 + hv \rightarrow 2*Cl$
Reaction	Rate Constant
$O + O2 + M \rightarrow O3 + M$	6.00e-34*(300/T)**2.4
$O + O3 \rightarrow 2*O2$	8.00E-12*exp(-2060/T)
$N + O2 \rightarrow NO + O$	1.50E-11*exp(-3600/T)
$N + NO \rightarrow N2 + O$	2.10E-11*exp(100/T)
$CO + OH + M \rightarrow \{CO2\} + HO2$	k0=5.90E-33*(300/T)**1.4 ki=1.10E-12*(300/T)**-1.3 f=0.6
$CO + OH \rightarrow \{CO2\} + H$	k0=1.50E-13*(300/T)**-0.6 ki=2.10E+09*(300/T)**-6.1 f=0.6
$H2 + O1D \rightarrow HO2 + OH$	1.20E-10
$O + OH \rightarrow H + O2$	1.80E-11*exp(180/T)
$HO2 + O \rightarrow OH + O2$	3.00E-11*exp(200/T)
$OH + O3 \rightarrow HO2 + O2$	1.70E-12*exp(-940/T)
$HO2 + O3 \rightarrow OH + 2*O2$	1.00E-14*exp(-490/T)
$HO2 + HO2 \rightarrow H2O2$	k1 = 3.0E-13*exp(460/T) k2 = 2.1E-33*exp(920/T) k3 = 1.4E-21*exp(2200/T) k = (k1 + k2*[M]) * (1 + k3*[H2O])
$H2O2 + OH \rightarrow H2O + HO2$	1.80E-12
$OH + HO2 \rightarrow H2O + O2$	4.8E-11*exp(250/T)
$OH + OH \rightarrow H2O + O$	1.80E-12
$H2 + OH \rightarrow H2O + HO2$	2.8E-12*exp(-1800/T)
$O1D + N2 \rightarrow O + N2$	2.15E-11*exp(110/T)
$O1D + O2 \rightarrow O + O2$	3.3E-11*exp(55/T)
$O1D + H2O \rightarrow 2*OH$	1.63E-10*exp(60/T)
$N2O + O1D \rightarrow 2*NO$	7.25E-11*exp(20/T)
$N2O + O1D \rightarrow N2 + O2$	4.63E-11*exp(20/T)
$NO + HO2 \rightarrow NO2 + OH$	3.3E-12*exp(270/T)
$NO + O3 \rightarrow NO2 + O2$	3E-12*exp(-1500/T)
$NO2 + O \rightarrow NO + O2$	5.1E-12*exp(210/T)
$NO2 + O3 \rightarrow NO3 + O2$	1.2E-13*exp(-2450/T)

$NO3 + HO2 \rightarrow OH + NO2$	3.50E-12
$NO2 + NO3 + M \rightarrow N2O5 + M$	k0=2.00E-30*(300/T)**4.0 ki=1.40E-12*(300/T)**0.7 f=0.6
$\frac{1}{1} \frac{1}{1} \frac{1}$	$K_{eq} = 2.70 \text{F}_{-}27 \text{*exp}(11000/\text{T})$
$N2O5 + H2O \rightarrow 2*HNO3$	0
$NO2 + OH + M \rightarrow HNO3 + M$	k0=1.80E-30*(300/T)**3.0 ki=2.80E-11 f=0.6
$HNO3 + OH \rightarrow NO3 + H2O$	k1=2.4E-14*exp(460/T) k2=2.7E-17*exp(2199/T) k3=6.5E-34*exp(1335/T) k = k1 + k3*[M] / (1 + k3*[M] / k2)
$NO3 + NO \rightarrow 2*NO2$	1.5E-11*exp(170/T)
$NO2 + HO2 + M \rightarrow HO2NO2 + M$	k0=2.00E-31*(300/T)**3.4 ki=2.90E-12*(300/T)**1.1 f=0.6
$HO2NO2 + OH \rightarrow H2O + NO2 + O2$	1.3E-12*exp(380/T)
$HO2NO2 + M \rightarrow HO2 + NO2 + M$	$K_{eq} = 2.10E-27*exp(10900/T)$
$CH4 + OH \rightarrow CH3O2 + H2O + LCH4$	2.45E-12*exp(-1775/T)
CH4 + O1D $\rightarrow$ .75*CH3O2 + .75*OH + .25*CH2O + .2*HO2 + .2*H + .05*H2	1.50E-10
$CH3O2 + NO \rightarrow CH2O + NO2 + HO2$	2.8E-12*exp(300/T)
$CH3O2 + CH3O2 \rightarrow 2*CH2O + 2*HO2$	6.03E-13*exp(-453/T)
$CH3O2 + CH3O2 \rightarrow CH2O + CH3OH$	2.3E-14*exp(677/T)
$CH3O2 + HO2 \rightarrow CH3OOH + O2$	4.1E-13*exp(750/T)
CH3OOH + OH → .7*CH3O2 + .3*OH + .3*CH2O + H2O	3.8E-12*exp(200/T)
$CH2O + NO3 \rightarrow CO + HO2 + HNO3$	3.4E-13*exp(-1900/T)
$CH2O + OH \rightarrow CO + H2O + HO2$	5.5E-12*exp(125/T)
$OH + C2H4 + M \rightarrow .75*EO2 + .5*CH2O + .25*HO2 + M$	k0=1.00E-28*(300/T)**4.50 ki=7.50E-12*(300/T)**0.85 f=0.6
$EO2 + NO \rightarrow EO + NO2$	4.2E-12*exp(180/T)

$EO + O2 \rightarrow GLYALD + HO2$	1.00E-14
$EO \rightarrow 2*CH2O + HO2$	1.60E+11*exp(-4150/T)
$C2H4 + O3 \rightarrow CH2O + .12*HO2 + .5*CO + .12*OH$	1.2E-14*exp(-2630/T)
	k0=8.00E-27*(300/T)**3.5 ki=3.00E-11
$C3H6 + OH + M \rightarrow PO2 + M$	f=0.5
$C3H6 + O3 \rightarrow .4*CH3CHO + .244*OH +$ .244*HO2 + .42*CO + .58*CH2O + .036*CH3OH	5.5E-15*exp(-1880/T)
$C3H6 + NO3 \rightarrow R4N2$	4.6E-13*exp(-1156/T)
$PO2 + NO \rightarrow CH3CHO + CH2O + HO2 + NO2$	2.7E-12*exp(350/T)
$PO2 + HO2 \rightarrow POOH + O2$	7.5E-13*exp(700/T)
POOH + OH → .791*OH + .209*PO2 + .791*RCHO	8.78E-12*exp(200/T)
$CH3CHO + OH \rightarrow H2O + .95*CH3CO3 + .05*CH2O + .05*CO + .05*HO2$	4.63E-12*exp(350/T)
$CH3CHO + NO3 \rightarrow CH3CO3 + HNO3$	1.4E-12*exp(-1900/T)
$CH3CO3 + NO \rightarrow CH3O2 + \{CO2\} + NO2$	8.1E-12*exp(270/T)
$CH3CO3 + NO2 + M \rightarrow PAN + M$	k0=9.70E-29*(300/T)**5.6 ki=9.30E-12*(300/T)**1.5 f=0.6
$CH3CO3 + HO2 \rightarrow .15*O3 + .44*OH + .44*CH3O2 + .41*CH3COOOH$	5.2E-13*exp(980/T)
$CH3CO3 + CH3O2 \rightarrow .9*CH3O2 + CH2O + .9*HO2 + .9*{CO2}$	2E-12*exp(500/T)
$CH3COOOH + OH \rightarrow .5*CH3CO3 + .5*CH2O + .5*OH$	3.8E-12*exp(200/T)
$PAN + M \rightarrow CH3CO3 + NO2 + M$	$K_{eq} = 9.00E-29*exp(14000/T)$
$CH3CO3 + CH3CO3 \rightarrow 2*CH3O2 + 2*\{CO2\}$	2.5E-12*exp(500/T)
$C2H6 + OH \rightarrow C2H5O2 + H2O$	7.66E-12*exp(-1020/T)
$C2H5O2 + NO \rightarrow CH3CHO + HO2 + NO2$	2.6E-12*exp(365/T)
$C2H5O2 + HO2 \rightarrow C2H5OOH + O2$	7.5E-13*exp(700/T)
C2H5O2 + CH3O2 → .75*CH2O + .75*CH3CHO + HO2 + .25*CH3OH + .25*C2H5OH	3.00E-13

$C2H5O2 + C2H5O2 \rightarrow 2*CH3CHO + 2*HO2$	4.10E-14
C2H5OOH + OH → .36*C2H5O2 + .64*CH3CHO + .64*OH	5.18E-12*exp(200/T)
$C4H10 + OH \rightarrow C3H7O2$	1.55E-11*exp(-540/T)
$C3H8 + OH \rightarrow C3H7O2 + H2O$	8.7E-12*exp(-615/T)
$C3H7O2 + NO \rightarrow .82*CH3COCH3 + NO2 + HO2$ $+ .27*CH3CHO$	4.2E-12*exp(180/T)
$C3H7O2 + HO2 \rightarrow C3H7OOH + O2$	7.5E-13*exp(700/T)
$C3H7O2 + CH3O2 \rightarrow CH2O + HO2 + .82*CH3COCH3$	3.75E-13*exp(-40/T)
$C3H7OOH + OH \rightarrow H2O + C3H7O2$	3.8E-12*exp(200/T)
$CH3COCH3 + OH \rightarrow ATO2 + H2O$	3.82e-11*exp(-2000/T) + 1.33e-13
$OH + CH3OH \rightarrow HO2 + CH2O$	2.9E-12*exp(-345/T)
$OH + C2H5OH \rightarrow HO2 + CH3CHO$	6.9E-12*exp(-230/T)
$ISOP + OH \rightarrow ISOPO2$	3.1E-11*exp(350/T)
$ISOPO2 \rightarrow HO2 + CH2O + .25*MGLY + .5*GLYALD + 0.25*GLYX + .5*HYAC + OH$	4.07E+08*exp(-7694/T)
$ISOP + O3 \rightarrow .325*MACR + .244*MVK + .845*CH2O + .11*H2O2 + .522*CO + .199*CH3CO3 + .026*HO2 + .27*OH + .128*C3H6 + .051*CH3O2$	1E-14*exp(-1970/T)
$ISOPO2 + NO \rightarrow .90*NO2 + .90*HO2 + .90*CH2O + .55*MVK + .35*MACR + .10*ISOPNB$	2.7E-12*exp(350/T)
ISOPO2 + HO2 → .937*ISOPOOH + .063*OH + .025*MACR + .038*MVK + .063*HO2 + .063*CH2O	2.06E-13*exp(1300/T)
ISOPO2 + CH3O2 → 1.1*HO2 + 1.22*CH2O + .28*MVK + .18*MACR + .3*RCHO + .24*CH3OH + .24*ROH	8.37E-14
$ISOPO2 + ISOPO2 \rightarrow 1.28*HO2 + .92*CH2O + .56*MVK + .36*MACR + .48*ROH + .5*RCHO$	1.54E-13
$ISOPNB + OH \rightarrow ISOPNBO2$	3.61E-12*exp(380/T)

ISOPNBO2 + NO → .09*GLYALD + .09*HYAC + .69*CH2O + 0.88*NO2 + .44*MACRN + .69*HO2 + .26*MVKN + 0.42*HNO3	2.4E-12*exp(360/T)
ISOPNBO2 + HO2 → .06*GLYALD + .06*HYAC + .44*CH2O + .28*MACRN + .16*MVKN + .06*NO2 + .44*HO2 + .5*OH + .5*HNO3	8.7E-14*exp(1650/T)
$ISOPNB + O3 \rightarrow .61*MVKN + .39*MACRN + .27*OH + CH2O$	3.70E-19
$ISOPOOH + OH \rightarrow .387*ISOPO2 + .613*OH + .613*RCHO$	4.75E-12*exp(200/T)
$ISOPOOH + OH \rightarrow OH + IEPOX$	1.9E-11*exp(390/T)
$IEPOX + OH \rightarrow IEPOXOO$	5.78E-11*exp(-400/T)
IEPOXOO + HO2 → .725*HYAC + .275*GLYALD + .275*GLYX + .275*MGLY + 1.125*OH + .825*HO2 + .375*CH2O + .251*CO	2.06E-13*exp(1300/T)
IEPOXOO + NO → .725*HYAC + .275*GLYALD + .275*GLYX + .275*MGLY + .125*OH + .825*HO2 + .375*CH2O + .251*CO + NO2	2.7E-12*exp(350/T)
$MVK + OH \rightarrow MVKO2$	2.6E-12*exp(610/T)
$MVK + O3 \rightarrow .202*OH + .202*HO2 + .535*CO + .05*CH3CHO + .95*MGLY + .05*CH2O$	8.5E-16*exp(-1520/T)
MVKO2 + NO → .965*NO2 + .249*HO2 + .249*CH2O + .716*CH3CO3 + .716*GLYALD + .249*MGLY + .035*MVKN	2.7E-12*exp(350/T)
MVKO2 + HO2 → .38*MVKOOH + .62*OH + .37*GLYALD + .37*CH3CO3 + .13*MEK + .25*HO2 + .12*CH2O + .12*MGLY + .033*RCHO	1.82E-13*exp(1300/T)
MVKO2 + CH3O2 → .14*HO2 + .14*CH2O + .36*CH3CO3 + .36*GLYALD + .14*MGLY + .25*MEK + .75*CH2O + .25*CH3OH + .25*ROH + .5*HO2	8.37E-14
$MVKOOH + OH \rightarrow .791*OH + .791*MEK + .209*MVKO2$	8.78E-12*exp(200/T)
$MVKN + OH \rightarrow NO3 + .65*MGLY + .35*CH2O$	1.60E-12
$MACR + OH \rightarrow .45*MAO3 + .55*MACRO2$	8E-12*exp(380/T)

$MACRO2 \rightarrow CO + HYAC + OH$	2.90E+07*exp(-5297/T)
$MACR + O3 \rightarrow .261*OH + .202*HO2 + .569*CO + .88*MGLY + 0.12*CH2O$	1.4E-15*exp(-2100/T)
$MACR + NO3 \rightarrow MAO3 + HNO3$	3.40E-15
$MACRO2 + NO \rightarrow .97*NO2 + .97*HO2 + .97*HYAC + .97*CO + .03*MACRN$	2.7E-12*exp(350/T)
MACRO2 + HO2 → .42*MACROOH + .58*OH + .58*HYAC + .58*CO + .58*HO2	1.82E-13*exp(1300/T)
MACRO2 + CH3O2 → .595*HYAC + .255*MGLY + .595*CO + 1.255*CH2O + 1.7*HO2 + .15*ROH	8.37E-14
$MACROOH + OH \rightarrow MACRO2$	1.84E-12*exp(200/T)
$MACROOH + OH \rightarrow HYAC + OH$	4.4E-12*exp(380/T)
$MACRN + OH \rightarrow MACRNO2$	3.20E-12
$MACRNO2 + NO \rightarrow .08*CH2O + .15*NO3 + .07*MGLY + .85*HYAC + 1.85*NO2$	2.7E-12*exp(350/T)
$MACRNO2 + HO2 \rightarrow .08*CH2O + .15*NO3 + .07*MGLY + .85*HYAC + .85*NO2 + OH$	1.82E-13*exp(1300/T)
$MAO3 + NO \rightarrow NO2 + .5*CH2O + .5*CO + .5*CH3O2 + .5*CH3CO3$	6.7E-12*exp(340/T)
$MAO3 + HO2 \rightarrow .44*OH + .15*O3 + .59*CH2O + .39*CH3O2 + .41*MAOP + .39*CO$	4.3E-13*exp(1040/T)
$MAO3 + CH3O2 \rightarrow CH2O + HO2 + CH2O + CH3CO3$	1.68E-12*exp(500/T)
$MAO3 + CH3O2 \rightarrow CH2O$	1.87E-13*exp(500/T)
$MAO3 + NO2 + M \rightarrow MPAN + M$	k0=9.00E-28*(300/T)**8.90 ki=7.70E-12*(300/T)**0.20 f=0.6
$MPAN \rightarrow MAO3 + NO2$	$K_{eq} = 9.00E-29*exp(14000/T)$
$MAOP + OH \rightarrow MAO3$	6.13E-13*exp(200/T)
$MAOP + OH \rightarrow MAOPO2$	3.6E-12*exp(380/T)
$MAOPO2 + HO2 \rightarrow HYAC + 2*OH$	1.82E-13*exp(1300/T)
MAOPO2 + CH3O2 → .7*HYAC + .7*OH + CH2O + .7*HO2 + .3*C2H5OH	8.37E-14

$MAOPO2 + NO \rightarrow HYAC + OH + NO2$	2.35E-12*exp(350/T)
MAOPO2 + NO $\rightarrow$ HNO3	3.5E-13*exp(350/T)
$ATO2 + NO \rightarrow .96*NO2 + .96*CH2O + .96*CH3CO3 + .04*R4N2$	2.8E-12*exp(300/T)
$ATO2 + HO2 \rightarrow .15*CH3CO3 + .15*OH + .15*CH2O + .85*ATOOH$	8.6E-13*exp(700/T)
$ATOOH + OH \rightarrow ATO2 + H2O$	2.66E-12*exp(200/T)
$ATOOH + OH \rightarrow MGLY + OH + H2O$	1.14E-12*exp(200/T)
$MPAN + OH \rightarrow HYAC + CO + NO2$	2.90E-11
$GLYALD + OH \rightarrow .8*CH2O + .8*CO + HO2 + .2*GLYX$	1.00E-11
$GLYX + OH \rightarrow HO2 + 2*CO$	3.1E-12*exp(340/T)
$GLYX + NO3 \rightarrow HNO3 + HO2 + 2*CO$	4.00E-16
$MGLY + OH \rightarrow CH3CO3 + CO$	1.50E-11
$MGLY + NO3 \rightarrow HNO3 + CO + CH3CO3$	1.4E-12*exp(-1860/T)
$HYAC + OH \rightarrow MGLY + HO2$	1.6E-12*exp(305/T)
$ISOP + NO3 \rightarrow INO2$	3.3E-12*exp(-450/T)
$INO2 + NO \rightarrow .7*ISN1 + .035*MVK +$ .035*MACR + .07*CH2O + .8*HO2 + 1.3*NO2 + .23*RCHO	2.7E-12*exp(350/T)
$INO2 + NO3 \rightarrow .7*ISN1 + .035*MVK +$ .035*MACR + .07*CH2O + .8*HO2 + 1.3*NO2 + .23*RCHO	2.30E-12
$INO2 + HO2 \rightarrow .22*MVK + .015*MACR +$ .235*OH + .235*NO2 + .235*CH2O + .765*INPN	2.06E-13*exp(1300/T)
$INPN + OH \rightarrow OH + NO2 + MEK$	1.9E-11*exp(390/T)
$INPN + OH \rightarrow .36*INO2 + .64*R4N2 + .64*OH$	5.18E-12*exp(200/T)
$ISN1 + NO3 \rightarrow ISNOOA + HNO3$	3.15E-13*exp(-448/T)
$ISNOOA + NO3 \rightarrow NO2 + R4N2 + CO + HO2$	4.00E-12
$ISNOOA + NO \rightarrow NO2 + R4N2 + CO + HO2$	6.7E-12*exp(340/T)

ISNOOA + HO2 $\rightarrow$ .25* {RCOOH} + .25*O3 + HNO3	5.2E-13*exp(980/T)
$ISN1 + O3 \rightarrow .3*R4N2 + .45*CO + .15*OH + .45*HO2 + .7*GLYX + .7*OH + .7*NO2 + .7*NO2 + .7*OH + .7*NO2 + .7*NO2 + .7*OH + .7*NO2 $	
.7*MGLY	4.15E-15*exp(-1520/T)
$ISN1 + OH \rightarrow ISNOOA$	7.48E-12*exp(410/T)
$R4N2 + OH \rightarrow R4N1 + H2O$	1.60E-12
$C10H16 + OH \rightarrow 1.64*ISOPO2 + 0.1*CH3COCH3$	1.2E-11*exp(440/T)
$C10H16 + O3 \rightarrow 1.122*MACR + .442*MVK + .765*O + 1.156*OH$	5.3E-16*exp(-530/T)
$C10H16 + NO3 \rightarrow 1.7*ISOPO2 + NO2$	1.2E-12*exp(490/T)
$N2O5 \rightarrow 2*HNO3$	aerosol (γ=0.02)
$NO3 \rightarrow HNO3$	aerosol (γ=0.02)
$HO2 \rightarrow H2O$	aerosol (γ=0.2)
$NO2 \rightarrow 0.5*HNO3 + 0.5*OH + 0.5*NO$	aerosol (γ=1e-5)
$SO2 \rightarrow SO4$	aerosol (γ from Zheng et al. 2015)
	k0=3.30E-31*(300/T)**4.30
	ki=1.60E-12
$SO2 + OH + M \rightarrow SO4 + M$	1=0.6
$SO2 + H2O2 \rightarrow SO4$	in-cloud (Paulot et al. 2017)
$SO2 + O3 \rightarrow SO4$	in-cloud (Paulot et al. 2017)
$DMS + OH \rightarrow SO2 + CH2O$	1.2E-11*exp(-280/T)
	k1=8.2E-39*exp(5376/T) k2=1.05E-5*exp(3644/T)
$DMS + OH \rightarrow 0.75 * SO2 + CH2O$	k=k1*[O2]/(1+k2*[O2]/[M])
$DMS + NO3 \rightarrow SO2 + HNO3 + CH2O$	1.9E-13*exp(530/T)
$NH3 \rightarrow NH4$	aerosol (γ=0)
$NH3 + OH \rightarrow H2O + HNO3$	1.7E-12*exp(-710/T)
$H + O3 \rightarrow OH + O2$	1.40E-10*exp(-470/T)
	k0=4.40E-32*(300/T)**1.30 ki=4.70E-11*(300/T)**0.20
$H + O2 + M \rightarrow HO2 + M$	f=0.6

$C1 + O3 \rightarrow C1O + O2$	2.3E-11*exp(-200/T)
$O + ClO \rightarrow Cl + O2$	2.8E-11*exp(85/T)
$C1O + NO \rightarrow NO2 + C1$	6.4E-12*exp(290/T)
$ClO + NO2 + M \rightarrow ClONO2 + M$	k0=1.80E-31*(300/T)**3.40 ki=1.50E-11*(300/T)**1.90 f=0.6
$O + CIONO2 \rightarrow CIO + NO3$	2.9E-12*exp(-800/T)
$Cl + CH4 \rightarrow HCl + CH3O2$	7.3E-12*exp(-1280/T)
$OH + HC1 \rightarrow H2O + C1$	2.6E-12*exp(-350/T)
$Cl + HO2 \rightarrow HCl + O2$	1.8E-11*exp(170/T)
$ClO + HO2 \rightarrow HOCl + O2$	2.7E-12*exp(220/T)
$ClO + OH \rightarrow HO2 + Cl$	7.4E-12*exp(270/T)
$CH2O + Cl \rightarrow HCl + HO2 + CO$	8.1E-11*exp(-30/T)
$OH + OH + M \rightarrow H2O2 + M$	k0=6.90E-31*(300/T)**1.00 ki=2.60E-11 f=0.6
$ClO + ClO + M \rightarrow Cl2O2 + M$	k0=1.60E-32*(300/T)**4.50 ki=2.00E-12*(300/T)**2.40 f=0.6
$C12O2 + M \rightarrow 2*C1O + M$	$K_{eq} = 9.30E-28*exp(8835/T)$
$Br + O3 \rightarrow BrO + O2$	1.7E-11*exp(-800/T)
$BrO + NO2 + M \rightarrow BrONO2 + M$	k0=5.20E-31*(300/T)**3.20 ki=6.90E-12*(300/T)**2.90 f=0.6
$BrO + ClO \rightarrow Br + Cl + O2$	2.3E-12*exp(260/T)
$BrO + HO2 \rightarrow HOBr + O2$	4.5E-12*exp(460/T)
$BrO + NO \rightarrow Br + NO2$	8.8E-12*exp(260/T)
$HOBr + O \rightarrow BrO + OH$	1.20E-10*exp(-430/T)
$Br + HO2 \rightarrow HBr + O2$	4.8E-12*exp(-310/T)
$Br + CH2O \rightarrow HBr + HO2 + CO$	1.7E-11*exp(-800/T)
$HBr + OH \rightarrow Br + H2O$	5.5E-12*exp(200/T)
$BrO + ClO \rightarrow BrCl + O2$	4.1E-13*exp(290/T)
$ClO + OH \rightarrow HCl + O2$	6E-13*exp(230/T)

$NO2 + NO3 \rightarrow NO + NO2 + O2$	4.5E-14*exp(-1260/T)
$NO3 + NO3 \rightarrow 2*NO2 + O2$	8.5E-13*exp(-2450/T)
	k0=9.00E-32*(300/T)**1.50
	ki=3.00E-11
$NO + O + M \rightarrow NO2 + M$	f=0.6
$N + NO2 \rightarrow N2O + O$	5.8E-12*exp(220/T)
$HOC1 + HC1 \rightarrow H2O + C12$	heterogeneous (Austin and Wilson, 2006)
$N2O5 + HC1 \rightarrow HNO3 + C1 + NO2$	heterogeneous (Austin and Wilson, 2006)
$N2O5 + H2O \rightarrow 2*HNO3$	heterogeneous (Austin and Wilson, 2006)
$CIONO2 + H2O \rightarrow HOC1 + HNO3$	heterogeneous (Austin and Wilson, 2006)
$C1ONO2 + HC1 \rightarrow C12 + HNO3$	heterogeneous (Austin and Wilson, 2006)
$HOBr + HCl \rightarrow BrCl + H2O$	heterogeneous (Austin and Wilson, 2006)
$HOC1 + HBr \rightarrow BrC1 + H2O$	heterogeneous (Austin and Wilson, 2006)
$HOBr + HBr \rightarrow 2*Br + H2O$	heterogeneous (Austin and Wilson, 2006)
$BrONO2 + H2O \rightarrow HOBr + HNO3$	heterogeneous (Austin and Wilson, 2006)

## Notes:

Read 6.00E-34 as 6.00 x 10<sup>-34</sup>.

T = temperature (K);

[M] = atmospheric density (molecules cm<sup>-3</sup>);

[O2] = molecular oxygen density (molecules cm<sup>-3</sup>);

[H2O] = water vapor density (molecules cm<sup>-3</sup>).

a) Rate constants are given in units of  $s^{-1}$  for first-order reactions,  $cm^3$  molec<sup>-1</sup>  $s^{-1}$  for second order reactions, and  $cm^6$  molec<sup>-2</sup>  $s^{-1}$  for third-order reactions.

b) Three-body reaction rate constants are defined by:  $a = 1 + [\log_{10}(k0 * [M] / ki)]^2$ ,  $k = [(k0 * [M]) / (1 + k0 * [M] / ki)] * f^{(1/a)}$ .

c) Rate constants for dissociation reactions are calculated based on the rate constant ( $k_f$ ) for the corresponding association ("forward") reaction and the equilibrium constant ( $K_{eq}$ ) using:  $k = k_f / K_{eq}$ .