Description and evaluation of a detailed gas-phase chemistry scheme in the TM5-MP global chemistry transport model (r112)

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Abstract. This work documents and evaluates the tropospheric gas-phase chemical mechanism MOGUNTIA in the three-dimensional chemistry transport model TM5-MP. Compared to the modified CB05 (mCB05) chemical mechanism previously used in the model, the MOGUNTIA includes a detailed representation of the light hydrocarbons (C1-C4) and isoprene, along with a simplified chemistry representation of terpenes and aromatics. Another feature implemented in TM5-MP for this work is the use of the Rosenbrock solver in the chemistry code, which can replace the classical Euler Backward Integration method of the model. Global budgets of ozone (O_3), carbon monoxide (CO), hydroxyl radicals (OH), nitrogen oxides (NO_X) and volatile organic compounds (VOCs) are analyzed and their mixing ratios are compared with a series of surface, aircraft and satellite observations for the year 2006. Both mechanisms appear to be able to satisfactorily represent observed mixing ratios of important trace gases, with the MOGUNTIA chemistry configuration yielding lower biases than mCB05 compared to measurements in most of the cases. However, the two chemical mechanisms fail to reproduce the observed mixing ratios of light VOCs, indicating insufficient primary emission source strengths, too fast oxidation, and/or a low bias in the secondary contribution to C2-C3 organics via VOC atmospheric oxidation. Relative computational memory and time requirements of the different model configurations are also compared and discussed. Overall, the MOGUNTIA scheme simulates a large suite of oxygenated VOCs that are observed in the atmosphere at significant levels. This significantly expands the possible applications of TM5-MP.

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1 Introduction

Chemistry transport models (CTMs) are tools to effectively study the temporal and spatial evolution of atmospheric species at regional and global scales, as well as to understand how the main physical and chemical processes in the troposphere (e.g., emissions, chemistry, transport, and deposition) influence air quality. Model investigations and analyses of the changes of important tropospheric pollutants, such as ozone (O_3) and carbon monoxide (CO), can further provide essential information about the oxidative capacity of the atmosphere and thus the lifetime of important climate gases like methane (CH_4). The oxidative capacity also controls the rate of formation and growth of aerosols by conversion of sulfur oxides into particulate sulfate (SO_4^{2-}) and volatile organic compounds (VOCs) into condensable organic matter that forms organic particles. Under certain tropospheric conditions (e.g., intense sunlight and high temperatures) the oxidation of VOCs in the presence of nitrogen oxides ($NO_X \equiv NO + NO_2$) enhances the formation of secondary pollutants, such as O_3 (Crutzen, 1974; Derwent et al., 1996; Monks et al., 2009). VOCs and NO_X arise from both natural and anthropogenic emission sources. NO_X can be further converted into other chemical species such as HNO_3 and particulate nitrate (NO_3^-), that together with SO_4^{2-} are key contributors to atmospheric acidity. The photochemical production of tropospheric O_3, a known toxic air pollutant that is transported over long distances, depends on the NO_X and VOC availability in a nonlinear manner (e.g., Seinfeld and Pandis, 2006). Under very high NO_X conditions, common in densely populated areas (i.e., VOC-limited regimes), the O_3 production is inhibited and reductions in NO_X emissions can locally increase O_3. In contrast, in rural areas, the O_3 production is more efficient, and NO_X emission reductions will decrease O_3 (i.e., NO_X-limited regimes). Thus, changes in emissions of NO_X and VOC may lead to nonlinear responses in ozone and the oxidation capacity of the troposphere. Overall, understanding the photochemical processes in the troposphere via robust model simulations is key to the development of effective abatement strategies on pollutants that affect both air quality and climate, as well as to the prediction of the future atmospheric composition.

The gas-phase photochemistry in the troposphere consists of numerous and complex reactions between odd oxygen ($O_X \equiv O + O_3$) and NO_X, coupled to the oxidation of various VOCs (e.g., Atkinson, 2000; Atkinson et al., 2004, 2006). Several chemical mechanisms of varying complexity in the representation of VOC oxidation are currently included in state-of-the-art CTMs. One of the most explicit mechanisms ever built for the simulation of the tropospheric VOC oxidation cycles, the Master Chemical Mechanism (MCM v3), comprises more than 12690 reactions, involving more than 4350 organic species, and about 46 associated inorganic reactions (Jenkin et al., 1997a; 2003). Note that recent updates further include detailed aromatic hydrocarbon (Bloss et al., 2005) and isoprene oxidation (Jenkin et al., 2015) mechanisms. Since this level of chemical complexity is far beyond the computational resources potentially available for three-dimensional (3-D) global tropospheric CTMs, simplifications are required that retain the essential features of the chemistry. To this end, various chemical mechanisms of tropospheric chemistry have been developed with different levels of complexity, involving mainly reductions of the number of VOCs considered by lumping organic species into representative surrogates. For example, the Statewide Air Pollution Research Center mechanism (SAPRC-99) is a well-documented gas-phase chemical mechanism used in many CTMs, including a rather detailed representation of tropospheric VOC oxidation based on an evaluation against over 1700 experiments

performed in different smog chambers (e.g., Carter, 1995, 2010). SAPRC-99 does not model the oxidation of each VOC individually as the MCM, but it uses a molecular lumping approach to assign VOCs to a smaller number of reactive species. Other well-documented mechanisms often used in CTMs are the Regional Atmospheric Chemistry Mechanism (RACM; e.g., Geiger et al., 2003; Goliff et al., 2013; Stockwell et al., 1997) and the Model of Ozone and Related Chemical Tracers

5   mechanism (MOZART; Emmons et al., 2010; Horowitz et al., 2003). A molecular lumping mechanism has been also developed and initially used in the Model of the Global Universal Tracer transport In the Atmosphere (MOGUNTIA) 3-D climatological CTM (e.g., Kanakidou and Crutzen, 1999; Poisson et al., 2000; Baboukas et al., 2000), as well as in box model applications for field data interpretation (e.g., Poisson et al., 2001; Vrekoussis et al., 2006); that latter chemical mechanism has been the starting point for the model development presented here.

10 A mechanism that has been extensively used in numerous chemistry and climate modeling studies is the Carbon Bond Mechanism (CBM). CBM has several different versions with different levels of complexity (e.g., reaction rate constants updates, additions of inorganic reactions, as well as additions of organic species to better represent the respective species and radicals in the atmosphere), such as the CB4 (e.g., Gery et al., 1988; Houweling et al., 1998; Luecken et al., 2008), the CBM 2005 (CB05; e.g., Yarwood et al., 2005; Williams et al., 2013, 2017; Flemming et al., 2015) and the CBM-Z (Zaveri and

15 Peters, 1999). The lumped-structure approach of the CBM has been extensively evaluated against chamber studies (e.g., Yarwood et al., 2005).

Several studies focused on the impact of the chemical complexity of the gas-phase mechanism on tropospheric simulations. These studies indicate an inevitable compromise between model accuracy and computational efficiency (e.g., Cai et al., 2011; Gross and Stockwell, 2003; Luecken et al., 2008; Sander et al., 2019). Indeed, for a given atmospheric condition, even different

20 versions of the same mechanism (e.g., the CBM family) may give significantly different results. For instance, the more explicit representation of VOCs in CB05 leads to a higher production of O_3 compared to the more lumped CB4 mainly due to a higher production of peroxy radicals, aldehydes and organic peroxides (Saylor and Stein, 2012). A comparison of CB05 with RACM (Kim et al., 2009) revealed that the most considerable differences appeared in areas with significant biogenic emissions, due to the more complex chemistry of aldehydes in the presence of anthropogenic alkenes and alkanes. Box-model comparisons

25 between the MCM and various state-of-the-art simplified tropospheric chemistry schemes also indicated that the differences between the chemistry schemes can be rather significant under high VOC loadings (Emmerson and Evans, 2009). Thus, the choice of a gas-phase mechanism for a model may introduce uncertainties in predictions of regulated gas-phase pollutants (e.g., Knote et al., 2015). Computational restrictions, such as memory and computing time savings, are always a critical point to consider for large-scale 3-D simulations, especially when higher spatial resolutions are applied. On the other hand, the

30 ability to validate the results of a particular chemical scheme in a global model can be significantly higher for the more extensive schemes that provide an explicit treatment of gases, such as in comparisons with satellite retrievals and *in situ* observations of a series of individual species.

In this work, a detailed and complete chemistry scheme is implemented in the global CTM TM5-MP, the massively parallel (MP) version of the Tracer Model version 5 (TM5), with the aim to investigate whether the consistent biases in important

tropospheric tracers, such as O_3, CO, OH, NO_X and light VOCs, found in previous work (e.g., Huijnen et al., 2010; van Noije et al., 2014; Williams et al., 2013, 2017) are sensitive to the chemistry scheme that is used. For this, we use the well-documented tropospheric gas-phase chemistry scheme MOGUNTIA (e.g., Myriokefalitakis et al., 2008 and refs. therein; along with recent updates), and benchmark its performance in TM5-MP. Section 2 provides a short description of the current model version, focusing in particular on the new features implemented in the gas-phase chemistry and the chemistry integration method. In particular, we describe here the implementation of the Kinetic PreProcessor (KPP) software (Damian et al., 2002; Sandu and Sander, 2006) in TM5-MP, which offers higher flexibility for testing, updating, and further developing the chemistry code in the model. Note that we are mostly focusing here on the performance of the new chemical scheme in comparison to the scheme previously included in the model, i.e., the modified CB05 (mCB05). This model was introduced by Huijnen et al. (2010) and Williams et al. (2013), and further updated by Williams et al. (2017). In Sect. 3, the model's performance is analyzed for the different chemical configurations used for this study and in Sect. 4 a detailed budget analysis of important gas-phase species is presented. Section 5 presents the evaluation of the different configurations of this work. The model's ability to reproduce the variability of important tropospheric species in both space and time is discussed, along with the associated uncertainties in atmospheric burdens and lifetimes. Finally, in Sect. 6 the main conclusions are presented, and some of the benefits and drawbacks of both chemical mechanisms are discussed, together with proposed directions for future model development.

2 Model description

2.1 General

The well-documented offline 3-D global CTM TM5 (Krol et al., 2005) is used for this study. Historically, the model has evolved from the original TM2 model (Heimann et al., 1988), via the TM3 model (Houweling et al., 1998; Tsigaridis and Kanakidou, 2003), to TM4 (van Noije et al., 2004; Myriokefalitakis et al., 2008) and TM5 (Krol et al., 2005; Huijnen et al., 2010; van Noije et al., 2014; Williams et al., 2017). In TM5-MP, the parallelization of the model has been redesigned, allowing for affordable global simulations at high resolution, i.e., 1°x1° globally (Williams et al., 2017). Moreover, in this new MP version, the two-way zoom capability of TM5 is no longer available. All applications of TM5 share the same methods for model discretization and operator splitting (Krol et al., 2005), the treatment of the meteorological fields, and the mass conserving tracer transport (Bregman et al., 2003). TM5-MP is driven by meteorological fields from the ECMWF ERA-Interim reanalysis (Dee et al., 2011) with an update frequency of 3 hours. The advection scheme used is based on the slopes scheme (Russell and Lerner, 1981) and deep and shallow cumulus convection is parameterized according to Tiedtke (1989). The performance of the transport in the model has been evaluated by Peters et al. (2004) using sulfur hexafluoride simulations and by analyzing the vertical and horizontal distribution of radon (^{222}Rn) (Koffi et al., 2016; Williams et al., 2017). More recently, global transport features, such as the transport times associated with inter-hemispheric transport, vertical mixing in the

troposphere, transport to and in the stratosphere, as well as, transport of air masses between land and ocean, were evaluated via an inter-comparison of six global transport models (Krol et al., 2018).

TM5-MP is primarily designed for simulation of the troposphere (i.e., no explicit stratospheric chemistry is considered in the model). To capture stratospheric ozone effects on actinic fluxes and to ensure realistic ozone stratosphere-troposphere exchange (STE), the overhead stratospheric profile is nudged to the ozone data set provided for the Coupled Model Intercomparison Project phase 6 (CMIP6; van Noije et al., manuscript in preparation). The boundary conditions for CH_4, both in the lower troposphere and the stratosphere, are also based on the respective global mean value from CMIP6 data set (see also Sect. 2.4) to scale the monthly 2-D climatological fields as derived from HALOE measurements (Grooß and Russell, 2005), with the same nudging heights and relaxation times as for the case of stratospheric O_3. This approach is justified due to the relatively long lifetime of CH_4. Additionally, for HNO_3 and CO in the stratosphere monthly mean latitudinal climatologies derived from ODIN space-based observations are applied by prescribing the ratio of HNO_3/O_3 (Jégou et al., 2008; Urban et al., 2009) and CO/O_3 (Dupuy et al., 2004), respectively. Note, however, that when we present the chemical budgets in the troposphere, a tropopause definition using the O_3 mixing ratio threshold of 150 ppb (e.g., Stevenson et al., 2006) is applied. Moreover, budget results using the 100 ppb O_3 mixing ratios (e.g., Lamarque et al., 2012) as a tropopause level in the model are also provided. For clarity, we note that, based on these threshold values, the different model configurations presented in this work (see Sect. 2.5) lead to identical tropopause heights.

The gas-phase chemistry of the TM5-MP model is supplemented with the in-cloud oxidation of SO_2 through aqueous-phase reactions with H_2O_2 and O_3, that depend on the acidity of the solution (Dentener and Crutzen, 1993). The heterogeneous conversion of N_2O_5 into HNO_3 on the available surface area of cloud droplets, cirrus particles, and hydrated sulfate aerosols is also accounted for. For cloud droplets, the number of droplets per unit volume is calculated using the liquid water content provided in the ECMWF meteorological data used by TM5-MP, assuming an effective droplet radius of 8 μm. For the heterogeneous conversion of N_2O_5 on hydrated sulfate particles, the approach of Dentener and Crutzen (1993) is employed, using a global mean reaction probability (γ value) of 0.02 and 0.01 on water and ice surfaces, respectively. Heterogeneous conversions also consider the total reactive surface area density of aerosols, with contributions to accumulation mode aerosol from sulfate, nitrate, and ammonium being calculated by the EQuilibrium Simplified Aerosol Model (EQSAM) approach (Metzger et al., 2002). The distribution of these aerosol species is calculated online and coupled to the gas-phase precursors NH_3, H_2SO_4, and HNO_3. Note that the aerosol microphysics module M7 (Vignati et al., 2004) is used in the model, as described in Aan de Brugh et al. (2011) and van Noije et al. (2014), along with recent updates on the inclusion of secondary organic aerosols (van Noije et al., manuscript in preparation). For N_2O_5, the uptake coefficient (γ) is considered as a function of temperature and relative humidity (Evans and Jacob, 2005), whilst for HO_2 and NO_3 radicals fixed γ values of 0.06 and 10^{-3}, respectively, are adopted across all aerosol types (Jacob, 2000).

The model considers the wet removal of atmospheric species by liquid and ice precipitation, by both in-cloud and below-cloud scavenging. The fraction of gases removed by precipitation depends on Henry's law (see Table S1 in the supplement), together with the dissociation constants, temperature, and liquid or ice water content. In-cloud scavenging in stratiform precipitation

considers an altitude dependent precipitation formation rate (also describing the conversion of cloud water into rainwater). For convective precipitation, highly soluble gases are assumed to be scavenged entirely in the vigorous convective updrafts producing rainfall rates of >1 mm/hour. Removal is exponentially scaled down for lower rainfall rates. For the dry deposition, the removal is calculated online in the model, based on a series of surface and atmospheric resistances on a $1°×1°$ spatial resolution (Wesely, 1989; Ganzeveld and Lelieveld, 1995; Ganzeveld et al., 1998). Overall, the calculated deposition velocities show both seasonal and diurnal cycles since they are calculated using 3-hourly meteorological and surface parameters, based on the uptake resistances for vegetation (in-canopy aerodynamic, soil, and leaf resistance), soil, water, snow, and ice (see Table S2). A more detailed description of dry and wet deposition schemes for the removal of gases can be found in de Bruine et al. (2017).

2.2 Gas-phase chemistry

2.2.1. The original MOGUNTIA chemical scheme

The new chemical mechanism that has been implemented in TM5-MP for this study was originally developed for box (Poisson et al., 2001) and global (Kanakidou and Crutzen, 1999; Poisson et al., 2000) modelling studies, and initially coupled to the global 3-D CTM MOGUNTIA (Zimmermann, 1988). Since then, the scheme has been continuously updated for box modelling, coupled to the global TM4 model, and applied in numerous studies (e.g., Tsigaridis and Kanakidou, 2002; Gros et al., 2002; Myriokefalitakis et al., 2008; Daskalakis et al., 2015).

The MOGUNTIA chemical scheme employs a rather detailed oxidation scheme of light alkanes (CH_4, C_2H_6, and C_3H_8), light alkenes (C_2H_4 and C_3H_6), acetylene (C_2H_2), and isoprene (C_5H_8). Acetaldehyde (CH_3CHO), glyoxal (GLY; CHOCHO), glycolaldehyde (GLYAL; HOCHCHO), methylglyoxal (MGLY; CH_3COCHO) and acetone (CH_3COCH_3) are also explicitly treated in the mechanism. The oxidation pathways of methacrolein (MACR; $CH_3(CH_2)CH=O$) and methylvinyl ketone (MVK; $CH_3C(O)CH=CH_2$) are also considered, together with the formation of formic (HCOOH) and acetic acid (CH_3COOH). Higher VOCs (i.e., $C_{n>4}$), besides isoprene, are represented in the mechanism by the surrogate species n-butane (n-C_4H_{10}), motivated by the similar O_X and hydrogen oxides (HO_X) yields per oxidized carbon atom (e.g., see Poisson et al., 2000; Stavrakou et al., 2009a). The second-generation oxidation products of higher hydrocarbons of biogenic origin (such as terpenes) and aromatics are also considered to follow the gas-phase oxidation pathways of the respective isoprene and surrogate n-C_4H_{10} oxidation species.

The reactions of peroxy radicals (RO_2) with hydrogen peroxide (HO_2), methyl peroxide (CH_3O_2) and NO lead to organic hydroperoxides (ROOH), carbonyls and organic nitrates, respectively. ROOH is removed by photolysis and reaction with OH. The addition of NO to the formed RO_2 radicals leads to alkyl nitrates ($RONO_2$), which are much longer lived than NO_X. $RONO_2$ can thus be transported over longer distances than NO_X and serve as a sink for NO_X in high-NO_X regimes and as a source for NO_X in low-NO_X regimes. The $RONO_2$ compounds explicitly considered in this study are identified by R=CH_3, C_2H_5, C_3H_7, C_4H_9, HOC_2H_4O, and $C_5H_8(OH)$, i.e., the first-generation product of isoprene oxidation. Additionally, the reactions of the acyl

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peroxy radicals (RC(O)O$_2$) with NO$_2$ produce peroxyacyl nitrates (RC(O)O$_2$NO$_2$), in particular PAN (R=CH$_3$), which is the most abundant organic nitrate observed in the troposphere and the only species of this group that is considered here. Thermal decomposition is dominant for peroxyacyl nitrates, while it is negligible for alkyl nitrates. NO$_3$ radical reactions with aldehydes, alcohols, n-C$_4$H$_{10}$, dimethylsulfide (DMS) and unsaturated hydrocarbons are also considered. A more detailed

5   description of the chemical scheme used for this study can be found in Poisson et al. (2000) and Myriokefalitakis et al. (2008).

2.2.2 Updates of the MOGUNTIA chemical mechanism

Several updates have been applied to the original MOGUNTIA chemical scheme with respect to the previous implementations (e.g., Poisson et al., 2000; Myriokefalitakis et al., 2008). These updates include reactions of major hydrocarbons, their rate constants and oxidation pathways. Concerning the terpene chemistry, we here consider one lumped monoterpene species

10   (C$_{10}$H$_{16}$) for all terpenes (assuming a 50:50 α-:β-pinene distribution), in contrast to the consideration of the explicit oxidation of α- and β-pinene as performed in the previous implementations of the MOGUNTIA scheme (e.g., Myriokefalitakis et al., 2008, 2010). Thus, monoterpenes represent here all terpenes and terpenoids species. Likewise, toluene is used to represent all aromatics replacing benzene, xylene, and toluene used previously (Myriokefalitakis et al., 2008, 2010). Besides these compounds, toluene is also used to represent trimethyl-benzenes and higher aromatics. Moreover, for this work the coupling

15   of the gas-phase chemistry with the aqueous-phase oxidation scheme of SO$_2$, as well as the gas-phase oxidation of dimethyl sulfide (DMS), methyl sulfonic acid (MSA) and ammonia (NH$_3$), follows the oxidation scheme outlined by Williams et al. (2013), which is slightly simpler compared to the MOGUNTIA scheme used in previous studies (e.g., Myriokefalitakis et al., 2010). Note that the lumping mentioned above, and the simplifications implemented here, aim at limiting the number of species without degrading the general performance of the chemical scheme for global-scale tropospheric chemistry.

20   Isoprene (2-methyl-1,3-butadiene; ISOP) oxidation has been extended with the production of isoprene epoxydiols (IEPOX) and hydroperoxyaldehydes (HPALD), and the HO$_X$-recycling mechanism under low-NO$_X$ conditions (Paulot et al., 2009; Peeters and Müller, 2010a; Crounse et al., 2011; Browne et al., 2014). The latter species replaces the lumped second-generation oxidation product considered in previous implementations of the MOGUNTIA mechanism (Poisson et al., 2000; Myriokefalitakis et al., 2008). The oxidation of isoprene by the OH radical leads to the formation of several isomers of an

25   unsaturated hydroxy hydroperoxide. In the presence of NO$_X$, this leads to the formation of carbonyl compounds. However, under low-NO$_X$ conditions, the major product from unsaturated hydroxy hydroperoxides oxidation is IEPOX (i.e., cis- and trans-isomers). The organic peroxy radicals formed from OH oxidation of isoprene, can react with either 1) HO$_2$ to form hydroperoxides, or 2) NO to form hydroxynitrates, formaldehyde (HCHO), MVK, MACR and HO$_2$ (e.g., Paulot et al., 2009), or hydroperoxyenals (HPALDs). The latter are produced by the isomerisation of the initial isoprene organic hydroperoxy

30   radicals followed by reaction with O$_2$ and other oxidized products (Peeters et al., 2009; Peeters and Müller, 2010). Under HO$_2$-dominated conditions, the main products are unsaturated hydroperoxides (all possible isomers referred to as ISOPOOH; see Table 2). The fate of isoprene peroxy radicals is highly dependent on the mixing ratios of HO$_2$, NO, organic peroxy radicals, and the local meteorological conditions that affect thermal and photochemical reaction rates and wet and dry removal.

Subsequent reactions of ISOPOOH with OH produce epoxydiols (cis- and trans- isomers referred to as IEPOX) and regenerate OH radicals (Paulot et al., 2009). Moreover, the isoprene peroxy radical 1,6-H-shift isomerizations (Peeters et al., 2014; Peeters and Müller, 2010) lead to the formation of photolabile C5-hydroperoxyaldehydes (i.e., all possible isomers referred to as HPALDs; see Table 1). Overall, these additions to the chemistry scheme is expected to provide a better representation of OH regeneration during isoprene oxidation (e.g., Browne et al., 2014), compared to the previous implementation of the MOGUNTIA mechanism.

The MOGUNTIA chemistry scheme is in line with the VOCs oxidation pathways as proposed by the Master Chemistry Mechanism (MCM v3.3.1) (e.g., Bloss et al., 2005; Saunders et al., 2003). The thermal and pressure-dependent reaction rate coefficients of the MOGUNTIA chemical mechanism are taken (when available) from the IUPAC kinetic data evaluation (Atkinson et al., 2004; Wallington et al., 2018) and supplemented with reaction rates based on recommendations given by JPL (Burkholder et al., 2015). Photolysis frequencies needed to drive MOGUNTIA are taken from the IUPAC database (Atkinson, 1997; Atkinson et al., 2004) along with the updates from MCM v3.3.1 (Bloss et al., 2005; Jenkin et al., 1997, 2003, 2015; Saunders et al., 2003). Note that the model calculates online the photolysis frequencies as described in Williams et al. (2012). The comprehensive lists of all photochemical and thermal kinetic reactions included in the current MOGUNTIA chemical scheme are presented in Tables 1 and 2, respectively.

2.3 The chemical solver

The KPP version 2.2.3 (Damian et al., 2002; Sandu and Sander, 2006) is here employed to generate Fortran 90 code for the numerical integration of the gas-phase chemical mechanisms. An important advantage of this approach is that the implementation of a KPP generated code in the model is less prone to errors than coding the mechanism manually. Upon the translation of the chemistry mechanisms (e.g., species, reactions, rate coefficients) from the KPP language into a Fortran 90 code, a model driver was developed to arrange the respective couplings to TM5-MP. Minor changes, however, were needed in the KPP code to deal with TM5-MP I/O requirements. The photolysis and the thermal reactions are not calculated in KPP, but explicitly calculated by the respective modules of TM5-MP and then directly provided to the aforementioned chemistry driver. To this end, only the integration method has been updated in the model, replacing the default hand-coded chemical solver set-up. Moreover, the NO emission rates (as well as the dry deposition terms of all deposited species) are imported to KPP through the application of appropriate production (and loss) rates, as previously done for the EBI solver, owing mainly to the numerical stiffness of the NO-NO_2-O_3 photo-stationary state and their fast interactions (e.g., see Huijnen et al., 2010). In this study, the Rosenbrock solver is used as the numerical integrator (Sander et al., 2019). Rosenbrock solver has been shown to be robust and capable of integrating very stiff sets of equations (Sander et al., 2011). For all previous versions of the model, the Euler Backward Iterative (EBI) solver (Hertel et al., 1993) was used. This holds for the modified CB4 (Houweling et al., 1998), the mCB05 (Williams et al., 2013) and the MOGUNTIA (Myriokefalitakis et al., 2008) mechanisms. Note, however, that EBI was originally designed for the CB4 mechanism (Gery et al., 1989) and it is a rather fast and robust solver suitable for the use in large-scale atmospheric models that incorporate operator splitting (Huang and Chang, 2001).

The favorable comparison of the Rosenbrock solver against other widely used methods, such as Facsimile (Curtis and Sweetenham, 1987), has already been described in the literature (e.g., Sander et al., 2005). Focusing specifically on the comparison of a series of Rosenbrock solvers to EBI, Sandu et al. (1997) concluded that, although EBI appears robust, especially when it is used with a relatively large timestep, the Rosenbrock methods with variable timesteps are significantly

5   more accurate and clearly superior for accuracies in the range of 1% compared to EBI, for a range of species examined. The main aim of this study is not to compare the two chemistry solvers (i.e., the Rosenbrock vs. the EBI). Instead, we present model simulations using the Rosenbrock solver as produced by KPP for the mCB05 scheme (see Sect. 2.5) to isolate the impact of the solver on various species mixing ratios of this work.

2.4 Emission set-up

10   For the present study, emissions from anthropogenic activities including aircraft emissions (Hoesly et al., 2018) and biomass burning (speciated for agricultural waste burning, deforestation fires, boreal forest fires, peat fires, savanna fires and temperate forest fires; van Marle et al. (2017)), are adopted from the sectoral and gridded historical inventories as developed for the CMIP6 (Eyring et al., 2016). In more details, anthropogenic and biomass burning emissions of CO, NO_X, black carbon aerosol (BC), particulate organic carbon (OC), sulfur dioxide and sulfates (SO_x), as well as speciated non-methane volatile organic

15   compounds (NMVOCs) are considered, such as emissions of ethane (C_2H_6), methanol (CH_3OH), ethanol (C_2H_5OH), propane (C_3H_8), acetylene (C_2H_2), ethane (C_2H_4), propene (C_3H_6), isoprene (C_5H_8), monoterpenes ($C_{10}H_{16}$), benzene (C_6H_6), toluene (C_7H_8), xylene (C_8H_{10}) and other aromatics, higher alkenes, higher alkanes, HCHO, acetaldehyde (CH_3CHO), acetone (CH_3COCH_3), dimethylsulfide (DMS; C_2H_6S), formic acid (HCOOH), acetic acid (CH_3COOH), methyl ethyl ketone (MEK; $CH_3CH_2COCH_3$), methylglyoxal (MGLY; CH_3COCHO), and hydroxyacetaldehyde ($HOCH_2CHO$). Note that all biomass

20   burning emissions (open forest and grassland fires) are vertically distributed in the model over latitude-dependent injection heights, i.e., for tropical (30° S–30° N), temperate (30–60° S/N) and high-latitude (60–90° S/N) forest fires (see Appendix in van Noije et al., 2014).

Biogenic emissions from vegetation include isoprene, terpenes and other volatile organic compounds, and CO. Emissions are based on the Model of Emissions of Gases and Aerosols from Nature (MEGAN) version 2.1 (Sindelarova et al., 2014). Isoprene

25   and terpenes emissions are distributed over the first ~50 m from the surface and a diurnal cycle is imposed. The biogenic emissions from soils include NO_X (Yienger and Levy, 1995), NH_3 and terrestrial DMS emissions from soils and vegetation (Spiro et al., 1992). Oceanic emissions of CO and NMVOCs come from the POET database (Granier et al., 2005), oceanic emissions of NH_3 from Bouwman et al. (1997), while the DMS oceanic emissions are calculated online (van Noije et al., manuscript in preparation) using the sea water concentration climatology from Lana et al. (2011). The NO_X production by

30   lightning is parameterized based on convective precipitation fields (Meijer et al., 2001) and the SO_x fluxes from continuously emitting volcanoes are taken from Andres and Kasgnoc (1998). Note that we focus below on the more detailed representation of emissions as used for the MOGUNTIA chemical scheme. Emissions of other tropospheric species in the gas and the particulate phase are described in detail in previous studies (e.g., van Noije et al., 2014).

The MOGUNTIA chemical scheme considers direct emissions of CO, CH$_4$, HCHO, HCOOH, CH$_3$OH, C$_2$H$_6$, C$_2$H$_4$, C$_2$H$_2$, CH$_3$CHO, CH$_3$COOH, C$_2$H$_5$OH, HOCH$_2$CHO, CHOCHO, C$_3$H$_8$, C$_3$H$_6$, n-C$_4$H$_{10}$, MEK, C$_5$H$_8$, C$_{10}$H$_{16}$, C$_7$H$_8$ as well as NO$_X$, NH$_3$, DMS, and SO$_X$. Butanes, pentanes, hexanes, and higher alkanes emissions are summed up into the lumped n-C$_4$H$_{10}$ species, which represents the alkanes containing four or more carbon atoms. For reactivity purposes, higher alkenes emissions containing four or more carbon atoms (butenes and higher alkenes) are accounted for as equivalent C$_3$H$_6$ emissions. Higher ketones (i.e., except for acetone) from open biomass burning emissions are represented as MEK. Emissions of benzene (C$_6$H$_6$), toluene (C$_7$H$_8$), xylene (C$_8$H$_{10}$), trimethyl-benzenes, and other higher aromatics and VOCs are represented by toluene as in the MOZART mechanism (Emmons et al., 2010a). Note that when VOC emissions are assigned to a lumped species, adjustments are made to preserve their atmospheric reactivity (see also notes in Tables 1 and 2).

The explicit parameterization of VOC species in the MOGUNTIA chemical scheme requires emissions that are not routinely included in available emission databases. Direct biofuel and biomass burning emissions of light carbonyls have been reported in several studies (e.g., Christian et al., 2003; Fu et al., 2008; Hays et al., 2002), and these represent a significant contribution to the VOC budget (e.g., Fu et al., 2008; Myriokefalitakis et al., 2008; Stavrakou et al., 2009b, 2009a; Vrekoussis et al., 2009). For this reason, emissions from biofuel use of 1.4 Tg yr^{-1}, 2.4 Tg yr^{-1}, and 1.6 Tg yr^{-1} are considered for GLYAL, GLY, and MGLY, respectively. For the biomass burning sector, we use global emissions of GLYAL and GLY of 4.3 Tg yr^{-1} and 5.2 Tg yr^{-1}, respectively. We base these emission rates on the HCHO emissions distribution, because mass emission rates of low molecular weight carbonyls, such as HCHO and GLY (e.g., Hays et al., 2002) are highly correlated. Global emissions of roughly 1.4 Tg yr^{-1} (Emmons et al., 2010) are also considered for MEK, accounting for anthropogenic emissions (Rodigast et al., 2016), such as domestic burning and solvent use (e.g., Ware, 1988). For all other carbonyls, primary anthropogenic emissions are considered negligible (e.g., Fu et al., 2008). A list of the global annual emission strengths considered for the MOGUNTIA chemical configuration is presented in Table 3. For completeness, we note that primary aerosol emissions of OC, BC, sea salt, and dust are also considered in the model with sea-salt and dust emissions calculated online. A more detailed description of the gas and aerosol emissions used in the model will be presented in van Noije et al. (manuscript in preparation).

2.5 Simulations

We will present the analysis of TM5-MP simulations with the mCB05 and MOGUNTIA chemical mechanisms for the year 2006, which has been the chosen year of previous benchmarking studies (Huijnen et al., 2010; Williams et al., 2013, 2017). All simulations have been performed at 1°x1° horizontal resolution (e.g., Williams et al., 2017) and 34 vertical layers, and use a 1-year spin-up (i.e., for the year 2005). The same emission datasets have been used in all simulations, albeit with higher speciation for the MOGUNTIA chemical scheme. Overall, two simulations have been performed for the mCB05 configuration: one employing the EBI solver (mCB05(EBI)) and one employing the KPP-generated Rosenbrock solver (mCB05(KPP)). This approach isolates differences that are caused solely by the applied chemistry solver. By comparing MOGUNTIA, generated by KPP, with mCB05(KPP), the differences due to the chemistry set-up in the model are isolated.

3 Model performance

Concerning the TM5-MP performance, simulations performed on the ECMWF CRAY XC40 high-performance computer facility using 360 cores, indicate that the coupling of KPP software alone, increases the time spent in chemistry by ~59% and overall slows down the code by ~18% compared to the (hand coded) EBI version for the mCB05 mechanism. As expected, the coupling of the MOGUNTIA atmospheric chemistry scheme further increases the model runtime. MOGUNTIA uses 100 transported and 28 non-transported tracers, numbers that are significantly larger than the mCB05 configuration (i.e., 69 transported and 21 non-transported tracers). As a result, time spent to transport the tracers increases by ~43% and the chemistry calculations slow down by ~55%. Altogether, the newly coupled MOGUNTIA chemistry scheme in TM5-MP is computationally ~27% more expensive than the mCB05(EBI) configuration. Overall, the mCB05(EBI), mCB05(KPP) and MOGUNTIA configurations simulate 0.73, 0.60 and 0.44 year per day simulation time, respectively (Table S3a). Note that an additional series of simulations with 450 cores leads only to marginal changes (Table S3b). Finally, the runtime values for the different model configurations presented here are highly hardware dependent, owing mainly to the large I/O component associated with reading the meteorological fields.

4 Comparison of budgets and tropospheric mixing ratios

4.1 Ozone (O_3)

Table 4 presents a detailed description of the chemical budget of tropospheric ozone as calculated by the TM5-MP model, for the three chemical configurations. Following Stevenson et al. (2006), chemical production of ozone is derived from all reactions that convert NO to NO_2, since NO_2 is rapidly photo-dissociated and forms O_3, i.e.,

$$NO + HO_2 \rightarrow NO_2 + OH \tag{1}$$

$$NO + RO_2 \rightarrow NO_2 + RO \tag{2}$$

where, RO_2 represents all the major organic peroxy radicals of the corresponding chemistry mechanism used in the model. For the MOGUNTIA scheme RO_2 includes CH_3O_2, $C_2H_5O_2$, $HYEO_2$, n-$C_3H_7O_2$, i-$C_3H_7O_2$, ACO_2, $HYPO_2$, n-C_4H_9O, $MEKO_2$, $ISOPO_2$, $IEPOXO_2$, $MVKO_2$, $MACRO_2$, $TERO_2$, and $AROO_2$ radicals. For mCB05, RO_2 includes the CH_3O_2 radical and XO_2 (i.e., the operator for the NO to NO_2 conversion which represents all lumped alkyl-peroxy radicals in mCB05; see Williams et al., 2017 and Yarwood et al., 2005).

The chemical O_3 loss is derived as the sum of the 1) O_3 photolysis to $O(^1D)$, i.e.,

$$O_3 + hv \rightarrow O(^1D) + O_2 \tag{3}$$

followed by reaction with H_2O to form OH, i.e.,

$$O(^1D) + H_2O \rightarrow 2OH \tag{4}$$

2) O_3 destruction by HO_2 and OH catalytic cycles, i.e.,

$$O_3 + HO_2 \rightarrow OH + 2\,O_2 \tag{5}$$

$$O_3 + OH \rightarrow HO_2 + O_2 \tag{6}$$

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and 3) reactions of O_3 with unsaturated VOCs. Chemical loss calculations exclude contribution from HNO_3, NO_3 and N_2O_5 and other fast cycles between ozone-related species, as proposed by Stevenson et al. (2006).

For the MOGUNTIA scheme, the tropospheric chemical production is calculated to be 5709 Tg yr^{-1}, which is only ~10 Tg yr^{-1} smaller compared to the mCB05(KPP) configuration. Chemical destruction in the troposphere is similar in the MOGUNTIA and mCB05(KPP) chemistry configurations (Table 4). The use of EBI compared to the Rosenbrock solver decreases the O_3 chemical production (5719 vs. 5589 Tg yr^{-1}) and destruction (5216 vs. 5192 Tg yr^{-1}) terms in the troposphere (Table 4). Besides some expected differences due to the behavior of the two solvers, the calculated differences may also be partly attributed to the mass fixer for NO_Y (i.e., the sum of NO, NO_2, NO_3, HNO_3, HNO_4, $2 \times N_2O_5$, PAN and the organic nitrate compounds) that is applied in the mCB05(EBI) configuration to ensure no artificial loss of nitrogen. NO_Y fixing occurs mainly over highly polluted regions with active NO_X photochemistry to improve the accuracy of the EBI solver.

Focusing on the impact of the stratosphere on the tropospheric O_3 budget, the net STE flux of O_3 for the MOGUNTIA configuration is somewhat lower (~1%) than for mCB05(KPP). Considering that all configurations use the same stratospheric ozone relaxation parameterization, this difference can only be attributed to the chemical schemes. Note that the global STE of O_3 is defined by simply considering the chemical production and loss budget terms, as proposed by Stevenson et al. (2006). Thus, differences in the O_3 stratospheric inflow budgets for the three chemistry configurations (Table 4) do not imply that the tropospheric chemistry impacts on O_3 transport from the stratosphere, but rather that the global budget is closed by an inferred stratospheric input term. Thus, the higher net chemical production of O_3 in the troposphere implies a lower contribution from the stratosphere to the troposphere for roughly the same deposition losses. The calculated net influx from the stratosphere for the MOGUNTIA configuration (~424 Tg yr^{-1}) remains within one standard deviation of a multi-model mean (552 ± 168 Tg yr^{-1}), as reported by both Stevenson et al. (2006) and Young et al. (2013). MOGUNTIA calculations are also in line with estimates (~400 Tg yr^{-1}) based on observations (Hsu, 2005; Olsen, 2004), although higher compared to the 306 Tg yr^{-1} calculated by an earlier version of the TM5 model driven by the same meteorological fields (van Noije et al., 2014). Overall, compared to the mCB05(EBI) simulation, the lower net stratosphere-troposphere exchange flux simulated in the MOGUNTIA configuration brings the model results closer to the current best estimates of the net STE.

The MOGUNTIA configuration also results in a reduction of roughly 2% in the tropospheric O_3 burden compared to both mCB05 configurations. No significant change in the O_3 lifetime in the troposphere (i.e., 22.3 - 22.8 days) is found and the calculated lifetimes remain close to other model estimates of ~22 days (Stevenson et al., 2006; Young et al., 2013). Compared to previous studies, the tropospheric O_3 burden calculated using the MOGUNTIA chemical configuration (~375 Tg) is ~12% higher compared to the multi-model mean estimate of Stevenson et al. (2006) (336 ± 27 Tg), the 335±10 Tg burden derived from O_3 climatology from pre-2000 data (Wild, 2007), and ~20% higher compared to the tropospheric burden of 309 Tg reported by van Noije et al. (2014). The calculated burden for the MOGUNTIA chemistry configuration is also ~11% higher compared to the burden derived from the ACCMIP models (337 ± 23 Tg; Young et al. 2013), roughly 17% higher than the burden reported by Schultz et al. (2018) and 8-15% higher than the Lamarque et al. (2012) estimations who used a tropopause level at 100 ppb of O_3 mixing ratios. Table 4 also presents the relative differences of the budget calculations when a tropopause

level of 100 ppb O_3 is adopted. Note that tropospheric burden estimates remain susceptible to the tropopause definition, leading potentially to significant differences between modelling studies. For this reason, the tropopause level(s) should always be reported when comparing modelling estimates. Overall, the use of the MOGUNTIA mechanism tends to bring the model closer to other published estimates, by lowering the O_3 burden compared to the mCB05 scheme in TM5-MP.

5   Ozone surface and zonal mean mixing ratios simulated by the MOGUNTIA configuration for the year 2006 are presented in Figs. 1a,b, respectively. Figures 1c,d show small differences in surface and zonal mean mixing ratios between MOGUNTIA and mCB05(KPP). Differences in surface simulated O_3 mixing ratios between the two mechanisms are evident mainly downwind of regions with biogenic and tropical fire emissions. The mCB05(KPP) simulation shows higher mixing ratios (~2-4 ppb) over the ITCZ, India and East Asia (up to ~10 ppb). This is mainly attributed to the different representation of VOCs,

10   with MOGUNTIA being significantly more explicit than mCB05. This behavior can also be observed in the zonal mean O_3 distribution presented in Fig. 1d, where the impact of the different representation of VOCs, originating mainly from the tropics, is reaching the mid- and upper troposphere lifted by convection following the upward branch of the tropical Hadley cell. The use of different solvers alone does not result in any critical difference in the O_3 mixing ratios for mCB05 (Fig. 1e,f), presenting only some small negative differences of ~1 ppb downwind of regions with high anthropogenic emissions (e.g., India) for

15   mCB05(EBI).

4.2 Hydroxyl radical (OH)

The hydroxyl radical (OH) is the primary oxidant in the atmosphere under sunlit conditions, initiating the oxidation of various VOCs, and thus the production of hydroperoxy (HO_2) and organic peroxy (RO_2) radicals. However, due to the high complexity of OH recycling pathways in atmospheric VOC degradation, the different representations of VOC oxidation pathways in

20   chemical mechanisms may lead to significant discrepancies between models. CH_4 is routinely used as a diagnostic for the calculated OH abundance in the troposphere since its background concentration is highly sensitive to the OH abundance in the tropics, where the water vapor and the biogenic emissions are high. Uncertainties in CH_4 global sources (e.g., a rapid rise in the CH_4 growth rates since 2007; Nisbet et al., 2019) together with uncertainties in anthropogenic emissions of the NO_X, CO, and NMVOC (e.g., Hoesly et al., 2018), may cause considerable divergence in model simulated CH_4 mixing ratios, for different

25   simulation years. For the present study, however, the surface mixing ratios of CH_4 are prescribed according to the CMIP6 recommendations for each simulation year (van Noije et al., manuscript in preparation).

Table 5 presents the global tropospheric OH production budgets for the various chemical configurations. The MOGUNTIA configuration yields a gas-phase OH formation via O_3 photolysis in the presence of water molecules (Reactions 3 and 4) of about 1878 Tg yr^{-1}. Additionally, the radical recycling terms (Reactions 1 and 5) contribute 1987 Tg yr^{-1}, the H_2O_2

30   photodissociation, i.e.,

$$H_2O_2 + h\nu \rightarrow 2\ OH \tag{7}$$

produces 303 Tg yr^{-1}, and all other reactions add another 120 Tg yr^{-1} to the global tropospheric OH production in the model. Overall, the total tropospheric OH production amounts to 4288 Tg yr^{-1}, which is in close agreement with the budget estimations

by Lelieveld et al. (2016), i.e., ~4270 Tg yr^{-1}. Some difference is however expected due to the definition of the troposphere in Lelieveld et al. (2016), where they define the tropopause in the tropics using temperature, and in the extratropics using potential vorticity gradients. We remind the reader that for the present study the chemical troposphere is defined using a threshold of 150 ppb O_3. It is striking that the OH chemical production calculated for the MOGUNTIA model set-up is much higher (28 -
5   35%) than for previous TM5 model configurations (i.e., 3355±30 and 3184±20 Tg yr^{-1}) as presented by van Noije et al. (2014) using a similar 150 ppb O_3 tropopause. This difference is mainly attributed to the various updates of the model compared to the version used in Noije et al. (2014), such as the emission database and the applied VOC representation (i.e., CMIP5; Lamarque et al. (2010) vs. CMIP6 for this study), the chemistry scheme (i.e., CB4 vs. MOGUNTIA), and the photolysis scheme (i.e., the previous implemented Landgraf et al. (1998) photolysis scheme vs. the Modified Band Approach scheme
10 implemented by Williams et al. (2012)).

Focusing on the differences between the MOGUNTIA and the CB05(KPP) mechanism, the MOGUNTIA OH production is very close to CB05(KPP) on a global scale (Table 5). Note that for mCB05, the comparison of the two solvers indicates that EBI calculates a ~1% lower chemical destruction of OH in the troposphere than Rosenbrock. The contribution of the CO and CH_4 oxidation terms to the global tropospheric OH losses are calculated as 41% and 15%, respectively, for the MOGUNTIA
15 scheme. This is slightly higher (by ~6% and ~3%, respectively) compared to mCB05(KPP).

Focusing further on the MOGUNTIA scheme, the calculated tropospheric CH_4 chemical lifetime is ~8.0 yr, as obtained through dividing the CH_4 global atmospheric mean burden (~4871 Tg) by the loss due to oxidation by OH radicals in the troposphere (~607 Tg yr^{-1}). Accounting, however, for additional CH_4 sinks due to oxidation in soils and the stratosphere with assumed lifetimes of 160 yr and 120 yr (Ehhalt et al., 2001), respectively, an atmospheric lifetime of about 7.18 yr is derived, which is
20 roughly 15% shorter than the ensemble model mean atmospheric lifetime reported by Stevenson et al. (2006) of 8.45±0.38 yr. The multi-model chemistry-climate simulations performed during the Atmospheric Chemistry and Climate Model Intercomparison Project (ACCMIP) (Naik et al., 2013; Voulgarakis et al., 2013), revealed vast diversities among models with a wide range of CH_4 chemical lifetime values (i.e., ~7-14 yr) and a mean value of 9.7±1.5 yr (i.e., 5–10% higher than observation-derived estimates). Lelieveld et al. (2016) derived a CH_4 chemical lifetime of 8.5 yr for the year 2010 and Schultz
25 et al. (2018) estimated a tropospheric CH_4 chemical lifetime of about 9.9 yr using also an O_3 threshold of 150 ppb to define the tropopause. Finally, Lamarque et al. (2012) reported a chemical lifetime of ~8.7 yr by taking a tropopause level at 100 ppb O_3.

4.3 Carbon monoxide (CO)

Table 6 presents the chemical CO budget calculated by TM5-MP for the three chemical configurations. The different model
30 configurations show that approximately 62±1% of the CO global production in the troposphere is due to the oxidation of CH_4 and NMVOC, with the remaining owing to direct emissions. Overall, the global CO budget is significantly affected by the interactions between OH and CO. Thus, changes in OH tropospheric chemical production (i.e., ~ -0.2% from mCB05(KPP) to MOGUNTIA) modulate the tropospheric secondary formation of CO from the oxidation of CH_4 and NMVOC (~ -10% change)

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and the CO chemical loss (~ -3% change) in the model. The global chemical production (i.e., the sum of chemical production terms in troposphere and stratosphere; Table 6) of CO for both the MOGUNTIA and the mCB05(KPP) chemical configurations, i.e., 2018 and 1844 Tg yr^{-1}, respectively, is however higher than the multi-model mean estimate (1505 ± 236 Tg yr^{-1}) reported by Shindell et al. (2006), which can be partially attributed to the different year of NMVOC emissions used (i.e., 2000 vs. 2006 for this work).

The dominant chemical reaction responsible for the increase in the tropospheric CO chemical production for the MOGUNTIA compared to mCB05(KPP) chemical configuration is the HCHO oxidation by OH radicals (i.e., ~15% increase compared to mCB05(KPP)). Indeed, although the lumped nature of the mCB05(KPP) mechanism leads to a higher tropospheric HCHO chemical production (~1896 Tg yr^{-1}) compared to MOGUNTIA configuration (~1843 Tg yr^{-1}), the HCHO tropospheric chemical destruction is calculated roughly 2% higher for the MOGUNTIA scheme. HCHO is mainly formed via the oxidation of CH_4, isoprene, and other NMVOC in the model. However, for both mCB05 configurations, the HCHO production via CH_3O_2H photolysis is calculated to be ~1.65 times higher compared to MOGUNTIA. The latter scheme seems to recycle the methyl-peroxy radical (CH_3O_2) more efficiently via the CH_3O_2 gas-phase reactions with organic peroxy radicals (RO_2) produced by higher-order NMVOC oxidation. In contrast, other higher aldehydes that represent the second most important producer of CO contribute more significantly in MOGUNTIA than in mCB05. This could be due to the more detailed representation of the higher aldehydes in the MOGUNTIA mechanism (e.g., considering the production and destruction reaction of GLY, GLYAL, and C_2H_5CHO) compared to the single lumped species (i.e., the ALD2) that represents all higher aldehydes in mCB05.

The global annual mean burden of CO for the MOGUNTIA chemical scheme is 361 Tg, almost the same as in the mCB05(KPP) configuration, but ~2 % lower compared to mCB05(EBI). Higher CO losses by OH oxidation and deposition in MOGUNTIA lead to a CO atmospheric lifetime of ~44 days, i.e., about 6% shorter compared to the mCB05(KPP) chemical mechanism. Note that the reduction in the atmospheric lifetime of CO is in line with the reduction in the atmospheric lifetime of CH_4 (~3%), reflecting overall an increase in tropospheric OH mixing ratios for the MOGUNTIA configuration compared to mCB05(KPP); i.e., higher OH levels in the atmosphere lead to a proportionally larger CO and CH_4 sinks.

Focusing further on the impact of the solver alone, we calculate roughly a 3% reduction in the CO atmospheric burden when the EBI solver is applied on the mCB05 mechanism in the model. This is directly connected to the ~1% increase in OH mixing ratios that is calculated when the Rosenbrock solver is used in the model. Furthermore, the CO tropospheric production is increased by ~0.5% in mCB05(KPP) compared to mCB05(EBI). Overall, the presented differences between the EBI and Rosenbrock solvers confirm that the choice of solver may impact on the simulated mixing ratios, owing mainly to the use of a constant versus a variable timestep in the chemistry integration (e.g., see Sandu et al., 1997).

Zonal mean CO mixing ratios at the surface for the year 2006 using the MOGUNTIA scheme are presented in Fig. 2 (a,b). Compared to mCB05(KPP), the results from MOGUNTIA show slightly higher surface CO mixing ratios (up to ~2 ppb) over highly populated regions, such as India. This regional increase is due to the differences in surface OH mixing ratios, owing mainly to the differences in NO_X chemistry between the two simulations (see also Sect. 5.2). In contrast, in South America

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negative differences of ~5-15 ppb are calculated at the surface (Fig. 2c). The effective HO_x regeneration together with the detailed VOC representation and oxidation pathways considered in MOGUNTIA result in an increase of the surface OH mixing ratios in locations with high biogenic VOC emissions. This subsequently leads to a regional decrease in the tropospheric CO mixing ratios compared to the mCB05(KPP) configuration. Similar results are found for the zonal mean CO distribution. Free

5   tropospheric CO mixing ratios in the tropics are also affected due to effective tropical convection. Finally, the use of different solvers for the mCB05 mechanism does not lead to any notable differences in the annual mean CO mixing ratios (Fig. 2e,f).

5 Model evaluation

Model simulations are evaluated with a series of surface, flask, aircraft, and sonde measurements, as well as with satellite retrievals and climatological data. The simulated NO_2 tropospheric columns are compared with satellite retrievals from the European project Quality Assurance for Essential Climate Variables (QA4ECV) project (Boersma et al., 2017), provided by the Ozone Monitoring Instrument (OMI) and the SCanning Imaging Absorption SpectroMeter for Atmospheric CHartographY (SCIAMACHY) instruments. The simulated OH mixing ratios are evaluated against calculations of global mean tropospheric values from other modelling studies, as well as against climatological data compiled by Spivakovsky et al. (2000). Modeled O_3 mixing ratios are evaluated against surface observations and ozonesonde data for the year 2006, as compiled by the World Ozone and Ultraviolet Radiation Data Centre (WOUDC; http://www.woudc.org; last access 20/08/2019); surface observations from the European Monitoring Evaluation Program network (EMEP; http://www.emep.int; last access 20/08/2019) have been also used. For the CO model evaluation, flask observations for the year 2006 are used, as compiled by National Oceanic and Atmospheric Administration Earth System Research Laboratory, Global Monitoring Division (NOAA, https://www.esrl.noaa.gov/gmd; last access 20/08/2019). O_3 and CO mixing ratios in the upper troposphere/lower stratosphere (UTLS) are compared to *in-situ* measurements from the MOZAIC (Measurement of Ozone and Water Vapour by Airbus In-Service Aircraft) data record (Thouret et al., 1998). The modelled CO total columns are compared with satellite retrievals from Measurement of Pollution in the Troposphere (MOPITT) instrument, version MOP02J_V008 (Deeter et al., 2013, 2019; Ziskin, 2019), i.e., the combined thermal/near-infrared data product. Finally, light VOCs (i.e., C_2H_4, C_2H_6, C_3H_6, C_3H_8) as simulated for the year 2006 are evaluated against flask measurements from the NOAA database, and against climatological data from aircraft campaigns, as produced by Emmons et al. (2000). Overall, to quantify and discuss the model performance, commonly used statistical parameters are calculated, such as the correlation coefficient (R), which reflects the strength of the linear relationship between model results and observations (the ability of the model to simulate the observed variability), the absolute bias (BIAS), the normalized mean bias (NMB), and the root mean square error (RMSE) as a measure of the mean deviation of the model from the measurement due to random and systematic errors. All equations used for the statistical analysis of model results are provided in the supplementary material (Eq. S1–S5).

5.1 Nitrogen dioxide (NO_2)

NO_X is a rate-limiting precursor of O_3 formation and thus an essential species for other tropospheric oxidants, such as OH. NO_X is emitted by both natural (lightning, soils, and fires) and anthropogenic combustion sources, with lightning mainly impacting NO_X mixing ratios at the top of convective up-drafts and anthropogenic fuel emissions being the principal source of NO at the surface. Tropospheric NO_2 vertical column densities retrieved from OMI (Boersma et al., 2017) are compared against the MOGUNTIA and mCB05(KPP) simulations (Fig. 3). Note that since the differences between mCB05(EBI) and mCB05(KPP) are small for tropospheric NO_2 columns, mCB05(EBI) is not shown. NO_2 column densities are retrieved using a consistent set of retrieval parameters and validated against ground-based MAX-DOAS measurements (Boersma et al., 2018).

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To consider the vertical sensitivity of the satellite measurements to NO_2 molecules at different altitudes, the tropospheric column averaging kernels, provided in the QA4ECV data product, are applied separately to both sets of modelled NO_2 vertical profiles, extracted from the hourly 3-D model output by linear and nearest-neighbor interpolation in space and time. The resulting NO_2 tropospheric column density is what would have been retrieved by the satellite if the actual vertical profile of

5   NO_2 mixing ratios were identical to the modeled profile. The tropospheric NO_2 columns retrieved from the satellite are averaged per model grid cell and day, resulting in a comparison dataset consisting of one NO_2 vertical column density per model grid cell and day.

For the MOGUNTIA configuration, the model shows a mean overestimation of 1.78×10^{14} (R=0.71) and 1.96×10^{14} molecules cm^{-2} (R=0.95) against OMI measurements for daily and annual values, respectively, performing slightly better than the

10   correlation of mCB05(KPP) configuration (R=0.71 and R=0.94 for daily and annual values). An overview of the statistical comparison of the three model simulations against OMI measurements is given in Fig. S1a. Some discrepancies, especially in the Northern Hemisphere (NH) may be attributed to the absence of a significant seasonal cycle in monthly anthropogenic emissions. Over the biomass burning source regions in Africa, the model overestimates the satellite retrievals. When the model is compared against NO_2 tropospheric columns from the SCIAMACHY instrument using the QA4ECV retrieval (not shown),

15   the MOGUNTIA configuration shows a similar improvement over mCB05(KPP), as with the OMI data.

Williams et al. (2017) showed that the TM5-MP model significantly underestimates the NO and NO_2 mixing ratios, both at the surface and in vertical profiles. The model satisfactorily reproduces the NO_2 mixing ratios in the boundary layer but overestimates mixing ratios at higher altitudes and in pristine environments. The MOGUNTIA scheme shows generally a better agreement with satellite retrievals compared to the mCB05(KPP) configuration, as expressed by a higher correlation

20   coefficient and a generally lower bias (Fig. S1a). The differences between the two chemistry schemes can be mainly attributed to the representation of organic NO_X reservoir species (i.e., the organic nitrates; ORGNTRs) in the two mechanisms (Fig. S2). Overall, since deep convection may efficiently transport ORGNTRs to the upper troposphere, the more explicit representation of VOC chemistry in the MOGUNTIA chemistry scheme alters the distribution of ORGNTR compared to the more lumped chemistry of the mCB05. Although production of ORGNTR is about 10% larger in the MOGUNTIA scheme, the ORGNTR

25   burden is dominated by the loss term (Table S4). Due to the more detailed ORGNTR representation in the MOGUNTIA scheme, the destruction becomes significantly more efficient compared to the mCB05 configuration. As a result, the global ORGNTR burden calculated using the MOGUNTIA scheme in the model is about 60% smaller.

Several modelling studies have compared the simulated NO_2 columns with *in situ* and satellite observations (e.g., Travis et al., 2016; Williams et al., 2017). These studies demonstrated an overestimate of the observed NO/NO_2 ratios compared to

30   observations in higher altitudes, possibly due to a respective underestimate of peroxy radicals in the upper troposphere that contribute to the NO to NO_2 conversion. A deviation in the NO/NO_2 ratio has also been reported for the GEOS-Chem model (Silvern et al., 2018; Travis et al., 2016). This model significantly underestimated the observed upper tropospheric NO_2 observations from the SEAC[4]RS aircraft campaign over the southeast United States. Silvern et al. (2018) calculated that the reaction with ozone accounts for roughly 75% of the NO to NO_2 conversion in the upper troposphere; thus, this deviation from

the photochemical equilibrium could be due to an error in kinetic data. Overall, the authors indicated that reducing the NO_2 photolysis by 20% and increasing the low-temperature $NO + O_3$ reaction rate constant by 40%, improves the model simulation of the NO/NO_2 ratio in the upper tropospheric data significantly compared to the aircraft data. Another source of uncertainty could be the strength of the direct soil emissions that, according to Miyazaki et al. (2017), are lower in our model (i.e., ~5 Tg-N yr^{-1}; Yienger and Levy, 1995) compared to the emissions of 7.9 Tg-N yr^{-1} derived using a multi-constituent satellite data assimilation.

5.2 Hydroxyl radical (OH)

Figures 4a and 4b illustrate the zonal mean tropospheric distributions of OH for two seasons (i.e., boreal winter and boreal summer) for 2006, as simulated with the MOGUNTIA chemistry scheme. The highest atmospheric mixing ratios of OH in the model are calculated in the tropics from close to surface up to roughly the tropopause, as a result of intense solar radiation and high humidity in the region, with the main OH maximum being roughly below 400 hPa (and a secondary maximum at ~300 hPa). The differences in OH zonal mean mixing ratios compared to the mCB05(KPP) configuration are presented in Figs. 4c,d. During the boreal winter, the mCB05(KPP) configuration results on average in lower OH mixing ratios in the northern subtropical lower troposphere (~3-6%) than the MOGUNTIA simulation (Fig. 4c), with the largest differences (~20-30%) around 20°-40° N. In the subtropical Southern Hemisphere (SH) during boreal summer, OH mixing ratios are on average lower (~2-3%) in the MOGUNTIA configuration than in mCB05(KPP) (Fig. 4d) almost everywhere, except for a small increase (up to 10%) at around 30° S. These small differences in OH mixing ratios are mainly related to the HOx regeneration and differences of NO_X and ORGNTR species that influence the distribution of OH in the troposphere. The more detailed representation of ORGNTR in the MOGUNTIA chemistry scheme results in more efficient NOx release upon the ORGNTR destruction (Table S4), leading overall to O_3 formation in remote locations, and thus to the stimulation of HOx recycling at higher altitudes. Note that globally the $NO + HO_2$ reaction is roughly 9% higher in the MOGUNTIA configuration on an annual basis compared to mCB05(KPP) (see Table 5).

Focusing on global means, a global mean tropospheric OH concentration of 10.1×10^5 molecules cm^{-3} is obtained from the MOGUNTIA chemistry configuration for the year 2006, which is roughly 4% higher than in the mCB05(KPP) configuration, but closer to the low end of the multi-model mean of $11 \pm 1.6 \times 10^5$ molecules cm^{-3} as derived by Naik et al. (2013) for the year 2000, and the mean tropospheric mixing ratios of 11.3×10^5 molecules cm^{-3} as calculated by Lelieveld et al. (2016) for the year 2013. In the tropical troposphere (30°S - 30°N), the mean OH level in the MOGUNTIA configuration of 16.74×10^5 molecules cm^{-3} is ~6% higher than in mCB05(KPP). In all model configurations, higher OH mixing ratios are calculated in the NH compared to the SH, directly related to the asymmetry in the hemispheric O_3 and NO_X burdens. Figures 4e,f show the climatological mean OH mixing ratios from the surface up to ~200hPa from Spivakovsky et al. (2000), reduced by 8% based on the observed decay of methyl-chloroform mixing ratios (see Huijnen et al., 2010; van Noije et al., 2014). The mean tropospheric OH concentration for the MOGUNTIA configuration is calculated to be roughly 25% and 30% higher compared to the optimized climatology from Spivakovsky et al. (2000) for boreal winter and summer, respectively. Moreover, a ~28%

higher NH/SH ratio of annual mean hemispheric OH mixing ratios in the troposphere is derived for the MOGUNTIA configuration compared to Spivakovsky et al. (2000). The NH/SH ratios are calculated ~1.37 and ~1.35 for the MOGUNTIA and the mCB05(KPP) configuration, respectively, being on the high end of other modeling estimates, such as the multi-model estimate of an NH/SH ratio of 1.28 ± 0.10 by Naik et al. (2013) and the 1.20 ratio as reported by Lelieveld et al. (2016).

5   **5.3 Ozone (O_3)**

The evaluation of modeled O_3 mixing ratios against surface observations for the three simulations for the year 2006 is presented in Fig. 5. The seasonal cycle across surface stations is generally well captured by all model configurations for most of the cases. TM5-MP, however, generally overestimates O_3 mixing ratios at most NH sites and for all model configurations, as, for example, can be seen at the Barrow (Fig. 5a) and Mace Head (Fig. 5b) stations, especially during the summer (June-July-
10   August, JJA) season, when O_3 is overestimated by about 8 and 3 ppb, respectively. However, at Viznar (Spain) and Mauna Loa (USA) (Figs. 5c and 5d, respectively), model results are closer to the observed O_3 mixing ratios, showing overall lower biases (i.e., ~1-3 ppb). In the SH (except for the polar circle), the model simulates the seasonal cycle of the O_3 surface mixing ratios well, however, with average positive biases of ~6-10 ppb in Cape Point (South Africa) and Baring Head (New Zealand) (Figs. 5e,f). At the South Pole (USA) and Sayowa (Japan) stations in Antarctica (Figs. 5g,h), the model also captures the
15   observed seasonality well (R= ~0.9), except for a negative bias of ~3 ppb during the local winter season. Focusing further on the chemistry mechanisms applied in the model, a slightly better consistency is achieved for the MOGUNTIA chemistry scheme in most of the cases. For the mCB05 chemistry scheme, the choice of the solver does not result in any notable difference in simulated surface O_3 mixing ratios. Considering all surface O_3 observations available for the year 2006 (Fig. S3), the MOGUNTIA chemistry configuration tends to overestimate the available observations with a mean bias of ~6.5 ppb. Note that
20   although the differences between the different chemistry configurations for surface O_3 are small, the mCB05(KPP) configuration shows the lowest bias (~5.2 ppb) whereas the mCB05(EBI) bias is closer to that of the MOGUNTIA configuration (~6.1 ppb).

Ozonesonde observations are used to evaluate the models' ability to reproduce the O_3 vertical profiles. Indicatively, Fig. 6 presents the comparison of model results with ozonesonde observations in 2006 at the Hohenpeissenberg in Germany and at
25   the Macquarie Island in the Southwestern Pacific Ocean, at five pressure levels (900 hPa, 800 hPa, 500 hPa, 400 hPa, and 200 hPa) covering the boundary layer and the low and high free troposphere. For this evaluation, all ozonesonde data have been binned to the 34 model pressure levels (see Sect. 3). The seasonal cycle at the two stations is well captured by each model configuration. For the highest model levels, above 200 hPa, all simulations are very close to the measurements, since O_3 mixing ratios are mainly determined by the upper boundary condition that is used (see Sect. 2.1). Comparisons for other WOUDC
30   stations around the globe for the year 2006 are presented in the supplementary material (Fig. S4). Overall, all model simulations capture the O_3 distribution quite well at almost all sites in the lower troposphere. The MOGUNTIA scheme shows a slightly better agreement with observations than the mCB05 configurations with smaller biases in most of the cases, especially at lower levels (i.e. from ~900hPa and up to ~500hPa). Concerning the impact of the chemistry solver, the vertical O_3 concentration

simulated using the mCB05 mechanism shows no notable differences between the use of KPP and EBI in most of the cases. Overall, considering all available ozonesonde data for the year 2006 (Fig. S4), the MOGUNTIA chemistry in TM5-MP results in an overestimation of the ozonesondes observations by roughly 16% (R = 0.96, BIAS = 4.7 ppb, NME=15.6%), which is slightly smaller compared to the mCB05 chemistry configurations.

5 Figure S5 presents a comparison of O_3 mixing ratios in the upper troposphere/lower stratosphere (UTLS) simulated by TM5-MP for the two chemistry configurations (i.e., mCB05(KPP) and MOGUNTIA) with *in-situ* observations from the MOZAIC airborne program (see Sect. 3.1), as a function of latitude. The accuracy of the MOZAIC O_3 measurements is ±2 ppb (Marenco et al., 1998). For this comparison, the MOZAIC measurements are binned on the vertical grid of TM5-MP. The model evaluation at pressure levels < 300 hPa indicates there is good agreement of both configurations with the observed mixing
10 ratios. A positive bias in April in the order of ~20 ppb is calculated for the model, but smaller biases are found around the tropics and in the latitudes north of 40ºN (Fig. S5a). In October (Fig. S5b), a constant positive bias of roughly 20 ppb is calculated for both configurations. This could be caused by the limited vertical resolution of this model version in the UTLS region. Note that 34 vertical levels were employed for this study with a higher resolution in the upper troposphere–lower stratosphere region. Part of the model overestimation could also be attributed to systematic errors, as also reported in previous
15 studies (e.g., Huijnen et al., 2010). Possible causes include cumulative effects such as a lack of a diurnal or weekly variation in the NO_X emissions from the road transport sector, an underestimation of surface deposition during summer, or errors in the representation of nocturnal boundary layer dynamics (e.g., see Williams et al., 2012).

5.4 Carbon monoxide (CO)

Figure 7 presents the model performance concerning surface CO mixing ratios, by comparing a series of flask observations
20 for the year 2006. CO is underestimated at most sites in the NH for all TM5-MP configurations, e.g., at the Barrow Observatory and Mace Head station (Figs. 7a,b), especially during boreal spring (March-April-May, MAM), by about 30 ppb on average. In the tropics, negative biases (~16-20 ppb) are observed at Mauna Loa and Mahe Island (Figs. 7c,d). At other stations in the SH, the model simulates the CO surface mixing ratios well with both positive and negative biases depending on the season (Figs. 7e,f). In Antarctica, at the South Pole and Sayowa stations (Figs. 7g,h), the model also shows a small positive bias up to
25 ~3 ppb during the local winter season. The seasonal cycle across stations is generally well captured by all model's chemistry configurations (i.e., R = 0.7- 0.9). The full set of CO comparisons with flask data is further presented in the supplement (Fig. S6). Overall, the MOGUNTIA and the mCB05(KPP) configurations underestimate the flask observations for the year 2006 with a negative bias of around 30 ppb, and with a correlation coefficient for both configurations of R=0.45. Notably, the mCB05(EBI) model configuration tends to produce lower biases in the SH, where the emission strengths are in general low,
30 compared to the other two configurations (i.e., approximately -3 vs. -4 and -5 ppb for mCB05(KPP) and MOGUNTIA, respectively). In contrast, the MOGUNTIA chemistry configuration results in lower biases in the NH where the majority of anthropogenic emissions occur (i.e., approximately -30 vs. -31 and -33 ppb for mCB05(EBI) and mCB05(KPP), respectively).

Total CO columns from the MOGUNTIA and the mCB05(KPP) model configurations are compared to the total column densities retrieved from the MOPITT satellite instrument (Deeter et al., 2013, 2019; Ziskin, 2019) for the year 2006 (Fig. 8). Co-sampling with averaging kernel has been applied to the modelled CO concentration profiles (i.e., in the same manner as for NO_2; see Sect. 5.1). Note that when the absolute difference in surface pressure between the MOPITT retrieval and the TM5-MP simulation is larger than 5 hPa, the measurements were excluded from the comparison. For the MOGUNTIA configuration, the model shows a mean underestimation of -8.54×10^{16} (R=0.82) and -1.18×10^{17} molecules cm^{-2} (R=0.91) compared to daily and annual averages of MOPITT data, respectively. However, the correlation is slightly improved compared to the mCB05(KPP) configuration (R=0.78 and R=0.88 for daily and annual values, respectively). As in the comparison with surface data, the biases in total column CO in the MOGUNTIA and mCB05(KPP) configurations deteriorated compared to the mCB05(EBI) configuration, albeit biases are still small (~-5% and ~-7% for daily and annual values, respectively). As this pattern can be seen in both KPP configurations, this difference seems to be caused by the implementation of the more accurate Rosenbrock solver. An overview of the statistical comparison of the three model configurations against MOPITT CO measurements is given in Fig. S1b.

Figure S5 further presents the comparison of CO mixing ratios in the upper troposphere/lower stratosphere (UTLS) simulated by TM5-MP with *in-situ* measurements from the MOZAIC airborne program (see Sect. 3.1). Model evaluation at pressure levels < 300 hPa shows a good correlation for both configurations in the SH, with a small positive bias (up to ~20 ppb) for the mCB05(KPP) configuration in April around the equator and a small negative bias (~10 ppb) for the MOGUNTIA configuration for latitudes below 10°N. Both configurations present a strong negative bias (~30 ppb) for latitudes above 20°N (Fig. S5c). In October (Fig. S5d), both the mCB05(KPP) and MOGUNTIA configurations tend to underestimate the observations with a negative bias of ~20 ppb, except for a small positive bias between 0-20°N. This positive model bias in the UTLS could point to a stronger convective uplift (e.g., Krol et al., 2018) in tropical Africa in April, or to possible misrepresentations of biomass burning emission strengths and horizontal and vertical distributions (e.g., Daskalakis et al., 2015; Nechita-Banda et al., 2018). Indeed, MOZAIC data show an increase in CO mixing ratios from the NH (April) to the SH (October), mainly due to the impact of biomass burning processes. Overall, the model configurations of this work present both positive and negative biases compared to the MOZAIC observations, with observations indicating larger latitudinal CO variability than simulated.

5.5 Volatile organic compounds (VOCs)

5.5.1 Ethane and propane

Ethane (C_2H_6) is the lightest alkane with emissions primarily of anthropogenic origin, associated mainly with fossil fuel extraction and use. In the model, the global ethane emission is 11 Tg yr^{-1} (Table 3) with an atmospheric lifetime of about 56 days for all chemistry configurations, in close agreement with other studies (e.g., Hodnebrog et al., 2018). Flask measurements indicate that C_2H_6 surface mixing ratios are strongly underestimated by all configurations at Mace Head (Fig. 9a) by ~80%, mainly during the winter, indicating also an opposite annual cycle. The latter can be attributed to the misinterpretation of

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seasonal variation of anthropogenic emission and/or to the C_2H_6 oxidation by OH radicals in the model. Significant underestimations are also observed in the tropics at Mauna Loa, Hawaii (Fig. 9c), of roughly 98% (R ≈ -0.5). In contrast, at Cape Grim, Australia (Fig. 9e), the model is better reproducing the measured C_2H_6 mixing ratios for all configurations, with a higher correlation coefficient (R = 0.5) and an NME of around 63%.

5   The underestimation of the C_2H_6 mixing ratio likely indicates that the model lacks primary emissions of C_2H_6 and can thus better reproduce atmospheric observations in the SH where the anthropogenic emissions are not as strong as in the NH. Dalsøren et al. (2018) showed recently that an increase of natural and anthropogenic fossil fuel emissions by a factor of two to three may significantly improve the simulated C_2H_6 and C_3H_8 mixing ratios compared to observations. Note that this increase in emissions would result in source estimates close to those calculated by the first global 2-D modeling study of these two

10   hydrocarbons by Kanakidou et al. (1991). To investigate, here, how the model responds to an increase of ethane emissions, sensitivity simulations with the MOGUNTIA configuration are performed by 1) doubling and 2) quadrupling the anthropogenic C_2H_6 fossil fuel emissions, resulting in total C_2H_6 emissions of ~17.1 Tg yr^{-1} and ~29.5 Tg yr^{-1}, respectively. The global tropospheric burdens have been also increased by a factor of ~1.4 and 2.2, respectively. The comparison, however, with flask data (Fig. S7) indicates that the increase of C_2H_6 anthropogenic emissions does not significantly affect the simulated

15   mixing ratios in the model at these specific stations. Overall, this means that even a more aggressive increase of emissions (at least over specific regions) is required, other missing sources are needed to be considered in the model, or that the oxidation of C_2H_6 is too fast in the model. The full set of C_2H_6 comparisons with flask data is presented in the supplement (Fig. S8). Propane (C_3H_8) is also emitted mainly from anthropogenic sources, and in the current simulations the total emission is 8.5 Tg yr^{-1} (Table 3), lower compared to other reported emission estimates of ~15 Tg yr^{-1} (Jacob et al., 2002). Model comparison with

20   flask observations (Fig. 9) shows that the model tends to underestimate the measured mixing ratios for all simulations, however, with higher correlation coefficients compared to C_2H_6 in most of the cases. C_3H_8 is underestimated in the NH at Mace Head (Fig. 9b) during the winter and autumn seasons by 72-74%. In the tropics, strong negative biases of ~100 ppt are observed at Mauna Loa (Fig. 9d). However, the model simulates the C_3H_8 surface mixing ratios better in the SH at Cape Grim compared to stations in the NH (Figs. 9b,d,f) due to the weaker impact of anthropogenic emissions. In contrast to the C_2H_6 evaluation

25   however, the model satisfactorily simulates the observed C_3H_8 mixing ratios at the South Pole (Fig. 9h), with a small overestimation during the local summer season. The full set of C_3H_8 comparisons with flask data is presented in Fig. S9. As for the case of C_2H_6, to further investigate the impact of emissions on the simulated C_3H_8 mixing ratios, additional simulations are performed by 1) doubling and 2) quadrupling the anthropogenic fossil fuel emissions, resulting overall in total C_3H_8 emissions of ~14.9 Tg yr^{-1} and ~27.9 Tg yr^{-1}, respectively. The global C_3H_8 tropospheric burdens have been increased by a

30   factor of ~1.7 and 3.2, respectively. Figure S7 indicates that an increase of C_3H_8 emissions by two times tends to significantly improve the model simulations, whereas a respective increase by four times tends to overestimate the observed mixing ratios. Comparison with C_2H_6 and C_3H_8 aircraft climatological data (Fig. 10) further indicates that all chemistry configurations tend to underestimate the observed mixing ratios (~20-60%) in most of the cases, especially in the upper troposphere. In more detail, at Boulder and East Brazil, the model significantly underestimates the observed mixing ratios for both compounds,

while at Hawaii C_2H_6 is underestimated, but C_3H_8 is well simulated by all three configurations. In contrast, at Easter Island, all schemes overestimate the observed mixing ratios for both compounds, although the MOGUNTIA overestimate is larger for C_2H_6 and lower for C_3H_8 compared to the two mCB05 configurations. The full sets of C_2H_6 and C_3H_8 comparisons with aircraft climatological data are presented in the supplement (Fig. S10 and Fig. S11, respectively). Overall, considering that the model

5   reasonably simulates the oxidative capacity of the atmosphere, direct emissions are the likely reason for these differences, since both alkanes are oxidized in the troposphere by OH radicals and no secondary production terms of these alkanes are known. Note, however, that alkane emission fluxes are on the low side, as also reported by other studies (e.g., Aydin et al., 2011; Huijnen et al., 2019; Monks et al., 2018).

5.5.2 Ethene and propene

10   Ethene is mainly emitted from biogenic sources, as well as, by the incomplete combustion from biomass burning, power plants, and combustion engines. C_2H_4 emissions in the model are roughly 30 Tg yr^{-1} (Table 3), close to the estimate of Huijnen et al. (2019), but on the high side compared to the 21 Tg yr^{-1} reported by Toon et al. (2018). The three chemistry configurations produce similar mixing ratios of C_2H_4 in most of the cases. Nevertheless, the comparison with aircraft observations (Fig. 11) indicates underestimated mixing ratios in the upper troposphere. In more detail, the model reproduces well (R=0.97) the

15   vertical distribution of C_2H_4 at Boulder (USA). However, observed mixing ratios close to the surface (up to ~ 2 km) are overestimated by the model, while observations at the higher levels (up to ~ 6 km) are underestimated. In the tropics, the observed mixing ratios in the lower and upper troposphere (e.g., at Hawaii) are slightly overestimated by the model for all configurations, although for the MOGUNTIA configuration this overestimate is the lowest. In remote regions, where the impact of direct emissions is negligible (e.g., at the Easter Island), the model overestimates C_2H_4 close to the surface (~1 km),

20   but some negative biases appear aloft. At higher altitudes, however, all configurations overestimate the observed C_2H_4 mixing ratios (Fig. 11g), but again the MOGUNTIA model configuration better reproduces the observations. Overall, these deviations from the observations could be attributed to 1) the not well-resolved background concentrations by the model, 2) the severe uncertainties in emission fluxes, and 3) a not well-understood chemistry (e.g., Huijnen et al., 2019; Pozzer et al., 2007), such as the C_2H_4 production during the VOC decomposition in the atmosphere.

25   Propene (C_3H_6) emissions in the model are ~32 Tg yr^{-1} (Table 3). The two mCB05 configurations produce similar C_3H_6 mixing ratios, but the MOGUNTIA tends to simulate higher values, especially in the tropics, at Hawaii (Fig. 11d) and at East Brazil (Fig. 11f). Close to the surface, where the impact of the emissions is stronger, the model severely overestimates observations (Figs. d,f), except for Japan (Fig 11b). For the MOGUNTIA configuration, this overestimation is more substantial in the tropics compared to the mCB05 chemistry scheme. An overestimation of the observed mixing ratio close to the surface is also found

30   in other regions, especially in the SH, such as in Eastern Brazil (Fig. 11f) or in remote regions, where the direct impact of emissions is negligible, such as in the Easter Island (Fig. 11h). However, at Easter Island (Fig. 11h), the model fails to reproduce the observed C_3H_6 vertical profile, resulting in a significant underestimation of the observed mixing ratios. Overall, even though the evaluation of vertical profiles should be considered here only as a climatological comparison, the reason for

the model underestimation of C_3H_6 mixing ratios at higher altitudes, is likely a combination of the emission strengths, the simulated vertical distribution, and the potential but still unaccounted secondary production from higher VOC oxidation. All comparisons for C_2H_4 and C_3H_6 with aircraft climatological data are presented in Figs. S12 and S13, respectively.

6. Summary and conclusions

This study documents and evaluates the implementation of the tropospheric chemistry scheme MOGUNTIA in the global chemistry and transport model TM5-MP. The MOGUNTIA scheme is a comprehensive gas-phase chemistry mechanism that explicitly accounts for the oxidation of light hydrocarbons, coupled with an updated representation of isoprene oxidation, along with a simplified representation of terpenes and aromatics chemistry. The newly coupled chemistry scheme in TM5-MP is compared to the existing chemistry scheme of the model, the mCB05. Another feature implemented in the TM5-MP chemistry code is the Rosenbrock solver, that replaces the classical EBI method. For this, a simple preprocessor directive has been implemented in the model to choose between the two solvers during model compilation. In the case of the Rosenbrock solver, the KPP software has been used to generate the chemistry code coupled with the TM5-MP. To further examine the impact of the solver on the TM5-MP atmospheric simulations and performance, the mCB05 scheme is also tested using the Rosenbrock solver.

Global budgets of O_3, CO, and OH, for all simulations performed for this work, are calculated and compared with estimates published in the literature. In more detail, the O_3 budget calculated with the MOGUNTIA chemistry scheme falls within one standard-deviation of mean estimates from other modelling studies. However, the new MOGUNTIA scheme reduces the tropospheric O_3 burden by ~3% compared to the mCB05 configurations. For tropospheric CO, a respective reduction in the atmospheric lifetime (~6%) provides evidence that the implementation of the MOGUNTIA chemistry leads to an increase in the oxidative capacity of the troposphere in TM5-MP. This also holds for the atmospheric CH_4 chemical lifetime that is calculated here to be about 8.0 yr for the MOGUNTIA chemistry scheme, which is roughly 3-5% shorter compared to mCB05(KPP) and mCB05(EBI) configurations.

The large-scale variability in space and time of modeled tropospheric NO_2, OH, O_3, CO, and light VOCs (i.e., C_2H_6, C_2H_4, C_3H_8, C_3H_6) has been evaluated for the year 2006 and compared to several sets of in-situ observations, satellite retrievals, and climatological data. Overall, both the lumped-structure (i.e., the mCB05) and the lumped-molecule (i.e., the MOGUNTIA) mechanisms appear to be able to satisfactorily represent the tropospheric chemistry. In most of the cases, lower biases compared to measurements are calculated when the MOGUNTIA chemistry configuration is used. The model simulates well the major observed features of the spatial and temporal variability in surface observations for O_3 and CO. The observed background surface O_3 mixing ratios are captured with a bias of ~6.5 ppb for the MOGUNTIA configuration, very close to the mCB05 configurations. Ozone in the vertical matches on average within ~5 ppb for all configurations, and the model is able to capture well the variability observed by ozone sondes. In contrast, the model underestimates the available CO flask observations by roughly 30% for all configurations, most likely linked to uncertainties in the seasonal cycle of anthropogenic emissions and the representation of biomass burning CO emissions. For the model comparison with observed light VOC mixing ratios, all chemistry configurations clearly show that significant uncertainties still exist regarding their emission strength or poorly understood chemistry, such as the secondary chemical production during the decomposition of higher VOC in the atmosphere. Sensitivity simulations performed indicate that increases of emissions may improve the simulation of the

atmospheric mixing ratios of some light VOCs, such as the C_3H_8. However, our results suggest that changes in emissions should not just be based on fixing the model's emissions using a specific (constant) value, but that scientifically accepted methods should be used. Future studies should therefore aim at improving source estimates and a better understanding of the processes that govern the budgets of light VOCs. From a chemistry point of view, it would be interesting to study the chemical formation pathways from higher VOCs. Inverse modelling or data-assimilation studies might be also used to "optimize" the emissions in order to minimize the differences between observations and model simulations.

The presented model configurations result in a benchmark of the TM5-MP tropospheric chemistry version upon which future model improvements may take place. Inherent uncertainties need to be reduced and further work is required, focusing mainly on the most poorly understood chemistry-related processes. For example, further attention concerns the uncertainties in NO-NO_2-O_3 cycling along with the atmospheric fate of ORGNTRs and their impacts on the oxidative capacity of the troposphere. Attention is also needed for the treatment of aerosols and clouds, in particular ice clouds and their impact on photolysis frequencies. Other issues that need to be resolved are related to the significant uncertainties in light hydrocarbons mixing ratios – as clearly noticed by the model comparison to surface and aircraft observations – and their potential impact on the oxidative capacity of the troposphere. Considering that both chemistry schemes underestimate light VOCs mixing ratios in most of the cases, the use of a more detailed scheme such as the MOGUNTIA will allow us to better understand the causes of this deviation compared to the lumped representation of VOC chemistry in the mCB05 mechanism. This is especially relevant over tropical regions with high biogenic VOC emissions under low-NO_X conditions. For this, a more dedicated comparison of the model with *in-situ* observations and satellite retrievals is needed. MOGUNTIA contains also an ample number of oxygenated VOCs that are observed in the atmosphere at significant levels and further involved in aerosol formation, making the scheme appropriate for detailed studies. On top of this, the implementation of the KPP software in the model makes the code a lot more flexible for chemistry updates compared to the previous EBI-based chemistry versions. The use of the KPP in TM5-MP reduces the uncertainties in solving stiff chemistry equations and opens up new possibilities on model development, such as the construction of an adjoint of the chemistry mechanism that can be used in 4D-VAR data assimilation systems (e.g., Henze et al., 2007). Another possible application is to more accurately explore atmospheric chemistry-climate interactions, since TM5-MP is also coupled to the Earth System Model EC-Earth (e.g., Van Noije et al., 2014; Van Noije et al., manuscript in preparation). Note, however, that despite the clear benefits regarding code development and management, the use of a more sophisticated solver such as the Rosenbrock, and the implementation of a detailed chemistry scheme such as the MOGUNTIA, make the code computationally more expensive. Overall, this work shows that the newly coupled chemistry version of TM5-MP works as good – or better in some of the cases – as the previous chemistry versions of the model, opening opportunities for further chemistry developments and more detailed tropospheric investigations by the TM and EC-Earth communities.

Author contributions. This paper resulted from the deliberations of 27th International TM5 Meeting, 28-29 June 2018, Utrecht, the Netherlands (SM, MCK, TvN, PLS, SH, ND, MK). SM and MCK developed the chemistry code coupled to the model. SM and MK provided the original chemistry scheme equations. JEW developed both the photolysis code and mCB05 chemical mechanism, including the implementation of updated photolysis frequencies for the additional organics included in the MOGUNTIA chemistry scheme. AG contributed to reaction data updates and coupling. AH developed and provided model evaluation tools with satellite retrievals. VH provided model evaluation tools and a collection of observation data. SM, ND, AH, and PLS performed the model evaluation. SM wrote the manuscript and all authors contributed to the preparation of this paper.

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Tables

Table 1. Photolysis reactions (J) in the MOGUNTIA chemistry scheme.

#	Reactants		Products[#]	References
J1	$O_3 + h\nu$	→	$O(^1D)$	1
J2	$H_2O_2 + h\nu$	→	2 OH	1
J3	$NO_2 + h\nu$	→	NO + O	1
J4	$NO_3 + h\nu$	→	$NO_2 + O$	1
J5	$NO_3 + h\nu$	→	NO	1
J6	$N_2O_5 + h\nu$	→	$NO_2 + NO_3$	1
J7	$N_2O_5 + h\nu$	→	$NO + NO_3$	1
J8	$HONO + h\nu$	→	OH + NO	1
J9	$HNO_3 + h\nu$	→	$NO_2 + OH$	1
J10	$HNO_4 + h\nu$	→	$NO_2 + HO_2$	1
J11	$HCHO + h\nu$	→	CO	1
J12	$HCHO + h\nu$	→	$CO + 2 HO_2$	1
J13	$CH_3OOH + h\nu$	→	$HCHO + HO_2 + OH$	1
J14	$CH_3ONO_2 + h\nu$	→	$HCHO + HO_2 + NO_2$	1
J15	$CH_3OONO_2 + h\nu$	→	$CH_3OO + NO_2$	1
J16	$CH_3OONO_2 + h\nu$	→	$HCHO + HO_2 + NO_3$	1
J17	$CH_3C(O)OONO_2 + h\nu$	→	$CH_3C(O)OO + NO_2$	J10
J18	$CH_3C(O)OONO_2 + h\nu$	→	$CH_3OO + NO_3 + CO_2$	J10
J19	$CH_3C(O)OOH + h\nu$	→	$CH_3C(O)OO + OH$	J13
J20	$C_2H_5OOH + h\nu$	→	$CH_3CHO + HO_2 + OH$	J13
J21	$C_2H_5ONO2 + h\nu$	→	$HCHO + CO + HO_2 + NO_2$	1
J22	$HOCH_2CH_2OOH + h\nu$	→	$2 HCHO + HO_2 + OH$	f 0.5 * J13
J23	$HOCH_2CH_2OOH + h\nu$	→	$HOCH_2CHO + HO_2 + OH$	(1 – f) 0.5 *J13
J24	$HOCH_2CH_2ONO_2 + h\nu$	→	$2 HCHO + HO_2 + NO_2$	f 0.5 * JORGN
J25	$HOCH_2CH_2ONO_2 + h\nu$	→	$HOCH_2CHO + HO_2 + NO_2$	(1 – f) 0.5 * JORGN
J26	$CH_3CHO + h\nu$	→	$CH_3OO + CO + HO_2$	1
J27	$HOCH_2CHO + h\nu$	→	$CH_3OH + CO$	1
J28	$CHOCHO + h\nu$	→	$2 CO + 2 HO_2$	1
J29	$CHOCHO + h\nu$	→	HCHO + CO	1
J30	$CHOCHO + h\nu$	→	2 CO	1
J31	$CH_3C(O)CH_3 + h\nu$	→	$2 CH_3OO + CO$	1
J32	$CH_3C(O)CH_3 + h\nu$	→	$CH_3C(O)OO + CH_3OO$	1
J33	$HOCH_2C(O)CH_3 + h\nu$	→	$CH_3C(O)OO + HCHO + HO_2$	1
J34	$CH_3C(O)CH_2OOH + h\nu$	→	$0.3 CH_3C(O)CHO\ 0.7(CH_3C(O)OO + HCHO) + OH$	J13
J35	$n\text{-}C_3H_7OOH + h\nu$	→	$C_2H_5CHO + HO_2 + OH$	0.5 * J13
J36	$n\text{-}C_3H_7ONO_2 + h\nu$	→	$C_2H_5CHO + HO_2 + NO_2$	1
J37	$i\text{-}C_3H_7OOH + h\nu$	→	$CH_3C(O)CH_3 + HO_2 + OH$	0.5 * J13
J38	$i\text{-}C_3H_7ONO_2 + h\nu$	→	$CH_3C(O)CH_3 + HO_2 + NO_2$	1
J39	$C_2H_5CHO + h\nu$	→	$C_2H_5OO + CO + HO_2$	1

J40	$HOC_3H_6OOH + hv$	→	$CH_3CHO + HCHO + HO_2$	J13
J41	$CH_3COCHO + hv$	→	$CH_3C(O)OO + CO + HO_2$	1
J42	$C_4H_9OOH + hv$	→	$0.67(CH_3CH_2COCH_3 + HO_2) + 0.33(C_2H_5OO + CH_3CHO) + OH$	J13
J43	$C_4H_9ONO_2 + hv$	→	$0.67(CH_3CH_2COCH_3 + HO_2) + 0.33(C_2H_5OO + CH_3CHO) + NO_2$	J_{ORGN}
J44	$CH_3CH_2C(O)CH_3 + hv$	→	$CH_3C(O)OO + C_2H_5OO$	1
J45	$CH_3CH(OOH)COCH_3 + hv$	→	$CH_3CHO + CH_3C(O)OO + OH$	J13
J46	$CH_3CH(ONO_2)COCH_3 + hv$	→	$CH_3CHO + CH_3C(O)OO + NO_2$	J_{ORGN}
J47	$ISOPOOH + hv$	→	$HCHO + 0.64\ MVK + 0.36\ MACR + HO_2 + OH$	13
J48	$ISOPONO_2 + hv$	→	$HCHO + 0.64\ MVK + 0.36\ MACR + HO_2 + NO_2$	J_{ORGN}
J49	$MACR + hv$	→	$0.5\ MACROO + 0.5\ HCHO + 0.175\ CH_3C(O)OO + 0.325\ CH_3OO + 0.825\ CO + HO_2$	1
J50	$MACROOH + hv$	→	$CH_3COCH_2OH + CO + HO_2 + OH$	J13
J51	$MACRONO_2 + hv$	→	$CH_3COCH_2OH + CO + HO_2 + NO2$	J_{ORGN}
J52	$MVK + hv$	→	$0.6\ (C_3H_6 + CO) + 0.4\ (CH_3C(O)OO + CH_3OO + HCHO)$	1
J53	$MVKOOH + hv$	→	$0.7(CH_3C(O)OO + HOCH_2CHO) + 0.3(CH_3COCHO + HCHO + HO_2) + OH$	J13
J54	$MVKONO_2 + hv$	→	$0.7(CH_3C(O)OO + HOCH_2CHO) + 0.3(CH_3COCHO + HCHO + HO_2) + NO_2$	J_{ORGN}
J55	$CH_3C(O)C(O)CH_3 + hv$	→	$2\ CH_3C(O)OO$	1
J56	$CH_3C(O)COOH + hv$	→	$CH_3C(O)OO + HO_2 + CO_2$	1
J57	$HPALD + hv$	→	$0.5\ HOCH_2C(O)CH_3 + 0.5\ CH_3COCHO + 0.25\ HOCH_2CHO + 0.25\ CHOCHO + HCHO + HO_2 + OH$	4, 5
J58	$O_2 + hv$	→	O_3	1

[#] The reaction products O_2, H_2, and H_2O are not shown.

[1] http://iupac.pole-ether.fr

[2] Atkinson, (1997):

$R_1 = 2.7 \times 10^{14} \exp(-6350/T)$
$R_2 = 6.3 \times 10^{-14} \exp(-550/T)$

$f = R_1/(R_1 + R_2 \times [O_2])$

[3] J_{ORGN} is calculated based on average of σ-values for 1-$C_4H_9ONO_2$ and 2-$C_4H_9ONO_2$ as described in Williams et al. (2012)

[4] Browne et al. (2014)

[5] Peeters and Müller (2010)

Table 2. Thermal reactions (K) in MOGUNTIA chemistry scheme.

#	Reactants		Products[#]	Rate expression[$]	References
K0a	O(^{1}D) (+ M)		O	$3.3\times10^{-11}\exp(55/T)[O_2]$ + $2.5\times10^{-11}\exp(110/T)$ [N$_2$]	1
K0b	O(^{1}D) +H$_2$O		OH + OH	$1.63\times10^{-10}\exp(60/T)$	1
K1	O$_3$ + OH	→	HO$_2$	$1.7 \times 10^{-12}\exp(-940/T)$	1
K2	HO$_2$ + O$_3$	→	OH	2.03×10^{-16} (T/300)$^{4.57}$ exp(693/T)	1
K3	HO$_2$ + OH	→	H$_2$O	$4.8 \times 10^{-11}\exp(250/T)$	1
K4	HO$_2$ + HO$_2$	→	H$_2$O$_2$	$2.2\times10^{-13}\exp(600/T)$ 1.9×10^{-33} [N$_2$] exp(980/T) 1.4×10^{-21} [H$_2$O] exp(2200/T)	1
K5	H$_2$O$_2$ + OH	→	HO$_2$	$2.9 \times 10^{-12}\exp(-160/T)$	1
K6	HO$_2$ + NO	→	NO$_2$ + HO	$3.45 \times 10^{-12}\exp(270/T)$	1
K7	NO + O$_3$	→	NO$_2$	$2.07 \times 10^{-12}\exp(-1400/T)$	1
K8	NO + NO$_3$	→	2NO$_2$	$1.8 \times 10^{-11}\exp(110/T)$	1
K9	NO$_2$ + O$_3$	→	NO$_3$	$1.4 \times 10^{-13}\exp(-2470/T)$	1
K10	OH + NO {+ M}	→	HONO	$7.4\times10^{-31} \times(T/300)^{-2.4}$ [N$_2$] $3.3\times10^{-11}(T/300)^{-0.3}$ Fc = 0.81	1
K11	OH + NO$_2$ {+ M}	→	HONO$_2$	$3.2 \times 10^{-30}(T/300)^{-4.5}$[N$_2$] 3.0×10^{-11} Fc = 0.41	1
K12	NO$_2$ + NO$_3$ {+ M}	→	N$_2$O$_5$	$3.6 \times 10^{-30}(T/300)^{-4.1}$[N$_2$] $1.9 \times 10^{-12}(T/300)^{0.2}$ Fc = 0.35	1
K13	NO$_2$ + HO$_2$	→	HO$_2$NO$_2$	$1.4 \times 10^{-31}(T/300)^{-3.1}$[N$_2$] 4.0×10^{-12} Fc = 0.40	1
K14	HO$_2$ + NO$_3$	→	OH + NO$_2$	4.0×10^{-12}	1
K15	HONO + OH	→	NO$_2$	$2.5\times10^{-12}\exp(260/T)$	1
K16	HNO$_3$ + OH	→	NO$_3$	$2.4\times10^{-14}\exp(460/T)$ $6.5\times10^{-34}\exp(1335/T)$ $2.7\times10^{-17}\exp(2199/T)$	1
K17	HO$_2$NO$_2$ + OH	→	NO$_2$	$1.9 \times 10^{-12}\exp(270/T)$	1
K18	HO$_2$NO$_2$	→	HO$_2$ + NO$_2$	$4.1 \times 10^{-5}\exp(-10650/T)[N_2]$ $6.0 \times 10^{15}\exp(-11170/T)$ Fc = 0.40	1
K19	N$_2$O$_5$	→	NO$_2$ + NO$_3$	$1.3 \times 10^{-3}(T/300)^{-3.5}\exp(-11000/T)[N_2]$ $9.7 \times 10^{14}(T/300)^{0.1}\exp(-11080/T)$ Fc = 0.35	1
K20	OH + H$_2$	→	HO$_2$	$7.7\times10^{-12}\exp(-2100/T)$	1
K21	CH$_4$ + OH	→	CH$_3$OO	$2.45 \times 10^{-12}\exp(-1775/T)$	2
K22	CH$_3$OO + HO$_2$	→	CH$_3$OOH	$3.8 \times 10^{-13}\exp(780/T)$* $(1-1/(1+498.0\exp(-1160/T)))$	1, 3

K23	$CH_3OO + HO_2$	$\rightarrow$	HCHO	$3.8 \times 10^{-13}\exp(780/T)*$ $(1/(1+498.0\exp(-1160/T)))$	1, 3
K24	$CH_3OO + NO$	$\rightarrow$	$0.999\ (HCHO + HO_2 + NO_2) +$ $0.001\ CH_3ONO_2$	$2.3 \times 10^{-12}\exp(360/T)$	1, 3
K25	$CH_3OO + NO_2$	$\rightarrow$	$CH_3O_2NO_2$	$2.5 \times 10^{-30}(T/300)^{-5.5}[N_2]$ 1.8×10^{-11} $F_c = 0.36$	1
K26	$CH_3OO + NO_3$	$\rightarrow$	$HCHO + NO_2$	1.2×10^{-12}	1
K27	$CH_3OO + CH_3OO$	$\rightarrow$	$2HCHO + 2HO_2$	$7.18\exp(-885/T)\ x$ $1.03 \times 10^{-13}\exp(365/T)$	3
K28	$CH_3OO + CH_3OO$	$\rightarrow$	$CH_3OH + HCHO$	$(1 - 7.18\exp(-885/T)\ x$ $1.03 \times 10^{-13}\exp(365/T)$	3
K29	$CH_3OOH + OH$	$\rightarrow$	$HCHO + OH$	$0.4 \times 5.3 \times 10^{-12}\exp(190/T)$	1
K30	$CH_3OOH + OH$	$\rightarrow$	CH_3OO	$0.6 \times 5.3 \times 10^{-12}\exp(190/T)$	1
K31	$CH_3ONO_2 + OH$	$\rightarrow$	$HCHO + NO_2$	$4.0 \times 10^{-13}\exp(-845/T)$	1
K32	CH_3OONO_2	$\rightarrow$	$CH_3O_2 + NO_2$	$9.0 \times 10^{-5}\exp(-9690/T\ )\ [N_2]$ $1.1 \times 10^{16}\exp(-10560/T\ )$ $F_c = 0.36$	1
K33	$HCHO + OH$	$\rightarrow$	$CO + HO_2$	$5.4 \times 10^{-12}\exp(135/T)$	1
K34	$HCHO + NO_3$	$\rightarrow$	$CO + HO_2 + HNO_3$	$2.0 \times 10^{-12}\exp(-2440/T)$	1
K35	$CH_3OH + OH$	$\rightarrow$	$HCHO + HO_2$	$2.85 \times 10^{-12}\exp(-345/T)$	1
K36	$CH_3OH + NO_3$	$\rightarrow$	$HCHO + HO_2 + HNO_3$	$9.4 \times 10^{-13}\exp(-2650/T)$	1
K37	$HCOOH + OH$	$\rightarrow$	$CO_2 + HO_2$	4.5×10^{-13}	1
K38	$CO + OH$	$\rightarrow$	$CO_2 + HO_2$	$5.9 \times 10^{-33}(300/T)^{1.4}$ $1.1 \times 10^{-12}(300/T)^{-1.3}$ $1.5 \times 10^{-13}(300/T)^{-0.6}$ $2.9 \times 10^{9}(300/T)^{-6.1}$	2
K39	$C_2H_6 + OH$	$\rightarrow$	C_2H_5OO	$6.9 \times 10^{-12}\exp(-1000/T)$	1
K40	$C_2H_5OO + HO_2$	$\rightarrow$	C_2H_5OOH	$6.4 \times 10^{-13}\exp(710/T)$	1
K41	$C_2H_5OO + NO$	$\rightarrow$	$CH_3CHO + HO_2 + NO_2$	$(1 - RTC2P) \times 2.55 \times 10^{-12}\exp(380/T)$	1, 4
K42	$C_2H_5OO + NO$	$\rightarrow$	$C_2H_5ONO_2$	$RTC2P \times 2.55 \times 10^{-12}\exp(380/T)$	1, 4
K43	$C_2H_5OO + CH_3OO$	$\rightarrow$	$CH_3CHO + HCHO + 2HO_2$	$0.8 \times (6.4 \times 10^{-14} \times 1.03 \times 10^{-13}\exp(365/T))^{0.5}$	3
K44	$C_2H_5OO + CH_3OO$	$\rightarrow$	$0.5\ CH_3CHO + 0.5\ CH_3CH_2OH + CH_3OH$	$0.2 \times (6.4 \times 10^{-14} \times 1.03 \times 10^{-13}\exp(365/T))^{0.5}$	3
K45	$C_2H_5OOH + OH$	$\rightarrow$	C_2H_5OO	$1.90 \times 10^{-12}\exp(190/T)$	3
K46	$C_2H_5OOH + OH$	$\rightarrow$	$CH_3CHO + OH$	8.01×10^{-12}	3
K47	$C_2H_5ONO_2 + OH$	$\rightarrow$	$CH_3CHO + NO_2$	$6.7 \times 10^{-13}\exp(-395/T)$	1
K48	$CH_3CHO + OH$	$\rightarrow$	$CH_3C(O)OO$	$4.7 \times 10^{-12}\exp(345/T)$	1
K49	$CH_3CHO + NO_3$	$\rightarrow$	$CH_3C(O)OO + HNO_3$	$1.4 \times 10^{-12}\exp(-1860/T)$	1
K50	$CH_3C(O)OO + HO_2$	$\rightarrow$	$CH_3C(O)OOH$	$0.41 * 5.2 \times 10^{-13}\exp(980/T)$	3
K51	$CH_3C(O)OO + HO_2$	$\rightarrow$	$CH_3COOH + O_3$	$0.15 * 5.2 \times 10^{-13}\exp(980/T)$	3
K52	$CH_3C(O)OO + HO_2$	$\rightarrow$	$CH_3O_2 + CO_2 + OH$	$0.44 * 5.2 \times 10^{-13}\exp(980/T)$	3
K53	$CH_3C(O)OO + NO$	$\rightarrow$	$CH_3OO + CO_2 + NO_2$	$7.5 \times 10^{-12}\exp(290/T)$	1
K54	$CH_3C(O)OO + NO_2$	$\rightarrow$	$CH_3C(O)OONO_2$	$3.28 \times 10^{-28}(T/300)^{-6.87}[N_2]$	1

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			$1.125 \times 10^{-11}(T/300)^{-1.105}$	
			$Fc = 0.3$	
K55	$CH_3C(O)OO + NO_3$	$\rightarrow$ $CH_3OO + NO_2$	4.0×10^{-12}	2
K56	$CH_3C(O)OO + CH_3OO$	$\rightarrow$ $CH_3C(O)OOH + HCHO$	$0.9 * 2.0 \times 10^{-12}exp(500/T)$	2
K57	$CH_3C(O)OO + CH_3OO$	$\rightarrow$ $CH_3COOH + HCHO$	$0.1 * 2.0 \times 10^{-12}exp(500/T)$	2
K58	$CH_3C(O)OO + CH_3C(O)OO$	$\rightarrow$ $2 (CH_3OO + CO_2)$	$2.9 \times 10^{-12}exp(500/T)$	2
K59	$CH_3C(O)OO + CH_3COCH_2O_2$	$\rightarrow$ $CH_3COOH + CH_3COCHO$	2.5×10^{-12}	2
K60	$CH_3C(O)OO + CH_3COCH_2O_2$	$\rightarrow$ $CH_3OO + CH_3COCH_2OH + CO_2$	2.5×10^{-12}	2
K61	$CH_3C(O)OO + C_2H_5OO$	$\rightarrow$ $CH_3CHO + 2 CH_3OO$	$0.7 * 4.4 \times 10^{-13}exp(1070/T)$	1, 3
K62	$CH_3C(O)OO + C_2H_5OO$	$\rightarrow$ $CH_3CHO + CH_3COOH$	$0.3 * 4.4 \times 10^{-13}exp(1070/T)$	1, 3
K63	$CH_3C(O)OONO_2 + OH$	$\rightarrow$ $HCHO + CO + NO_2$	3.0×10^{-14}	1
K64	$CH_3C(O)OONO_2$	$\rightarrow$ $CH_3C(O)OO + NO_2$	$1.1 \times 10^{-5}exp(-10100/T)[N_2]$ $1.9 \times 10^{17}exp(-14100/T)$ $Fc = 0.3$	1
K65	$CH_3C(O)OONO_2$	$\rightarrow$ $CH_3ONO_2 + CO_2$	$2.1 \times 10^{12} exp(-12525/T)$	5
K66	$CH_3C(O)OOH + OH$	$\rightarrow$ $CH_3C(O)OO$	1.1×10^{-11}	3
K67	$C_2H_4 + OH$	$\rightarrow$ $HOCH_2CH_2OO$	$8.6 \times 10^{-29}(T/300)^{-3.1}[N_2]$ $9.0 \times 10^{-12}(T/300)^{-0.85}$ $Fc = 0.48$	1
K68	$C_2H_4 + NO_3$	$\rightarrow$ $HOCH_2CH_2ONO_2$	$3.3 \times 10^{-12}exp(-2880/T)$	1
K69	$C_2H_4 + O_3$	$\rightarrow$ $1.37 HCHO + 0.63 CO + 0.13 HO_2 + 0.13 OH$	$6.82 \times 10^{-15}exp(-2500/T)$	1
K70	$HOCH_2CH_2OO + HO_2$	$\rightarrow$ $HOCH_2CH_2OOH$	1.3×10^{-11}	1
K71	$HOCH_2CH_2OO + NO$	$\rightarrow$ $NO_2 + 2HCHO + HO_2$	$(1-RTC2P) \times f \times 2.7 \times 10^{-12} exp(360/T)$	3
K72	$HOCH_2CH_2OO + NO$	$\rightarrow$ $NO_2 + HOCH_2CHO + HO_2$	$(1-RTC2P) \times (1-f) \times 2.7 \times 10^{-12} exp(360/T)$	3
K73	$HOCH_2CH_2OO + NO$	$\rightarrow$ $HOCH_2CH_2ONO_2$	$RTC2P \times 2.7 \times 10^{-12} exp(360/T)$	1
K74	$HOCH_2CH_2OO + CH_3OO$	$\rightarrow$ $HOCH_2CHO + HCHO + 2HO_2$	$0.8 * (7.8 \times 10^{14}exp(1000/T) * 1.03 \times 10^{-13}exp(365/T))^{0.5}$	3
K75	$HOCH_2CH_2OO + CH_3OO$	$\rightarrow$ $HOCH_2CHO + CH_3OH$	$0.2 * (7.8 \times 10^{14}exp(1000/T) * 1.03 \times 10^{-13}exp(365/T))^{0.5}$	3
K76	$HOCH_2CH_2OOH + OH$	$\rightarrow$ $HOCH_2CH_2OO$	K45	
K77	$HOCH_2CH_2OOH + OH$	$\rightarrow$ $HOCH_2CHO + OH$	1.38×10^{-11}	3
K78	$HOCH_2CH_2ONO_2 + OH$	$\rightarrow$ $HOCH_2CHO + NO_2$	K47	
K79	$C_2H_2 + OH$	$\rightarrow$ $0.636(CHOCHO + OH) + 0.364(HCOOH + CO + HO_2)$	$5.0 \times 10^{-30}(T/300)^{-1.5}[N_2]$ 1.0×10^{-12} $Fc = 0.37$	1
K80	$C_2H_2 + NO_3$	$\rightarrow$ $0.635 CHOCHO + 0.365(HCOOH + CO) + HNO_3$	1.0×10^{-16}	1
K81	$C_2H_2 + O_3$	$\rightarrow$ $0.635 CHOCHO + 0.365(HCOOH + CO)$	1.0×10^{-20}	1

K82	$HOCH_2CHO + OH$	→	$HCHO + CO_2$	6.4×10^{-12}	1
K83	$HOCH_2CHO + OH$	→	$CHOCHO + HO_2$	1.6×10^{-12}	1
K84	$CHOCHO + OH$	→	$2CO + HO_2$	$3.1 \times 10^{-12} \exp(340/T)$	1
K85	$CHOCHO + NO_3$	→	$2CO + HO_2 + HNO_3$	4.0×10^{-16}	1
K86	$CH_3COOH + OH$	→	$CH_3OO + CO_2$	$4.0 \times 10^{-14} \exp(850/T)$	1
K87	$CH_3CH_2OH + OH$	→	$0.95\ (CH_3CHO + HO_2) + 0.05\ HOCH_2CH_2OO$	$3.0 \times 10^{-12} \exp(20/T)$	1
K88	$C_3H_8 + OH$	→	$0.264\ n\text{-}C_3H_7O_2 + 0.736\ i\text{-}C_3H_7O_2$	$7.6 \times 10^{-12} \exp(-585/T)$	1, 3
K89	$n\text{-}C_3H_7O_2 + HO_2$	→	$n\text{-}C_3H_7OOH$	$0.52 \times 2.91 \times 10^{-13} \exp(1300/T)$	3
K90	$n\text{-}C_3H_7O_2 + NO$	→	$C_2H_5CHO + HO_2 + NO_2$	$(1 - RTC3P) \times 2.9 \times 10^{-12} \exp(350/T)$	1, 4
K91	$n\text{-}C_3H_7O_2 + NO$	→	$n\text{-}C_3H_7ONO_2$	$RTC3P \times 2.9 \times 10^{-12} \exp(350/T)$	1, 4
K92	$n\text{-}C_3H_7O_2 + CH_3OO$	→	$C_2H_5CHO + CH_3OH$	$0.8 \times (3.5 \times 10^{-13} \times 3.0 \times 10^{13})^{0.5}$	3
K93	$n\text{-}C_3H_7O_2 + CH_3OO$	→	$C_2H_5CHO + HCHO + 2HO_2$	$0.2 \times (3.5 \times 10^{-13} \times 3.0 \times 10^{13})^{0.5}$	3
K94	$n\text{-}C_3H_7OOH + OH$	→	$n\text{-}C_3H_7O_2$	K76	
K95	$n\text{-}C_3H_7OOH + OH$	→	$C_2H_5CHO + OH$	1.66×10^{-11}	3
K96	$n\text{-}C_3H_7ONO_2 + OH$	→	$C_2H_5CHO + NO_2$	5.8×10^{-13}	1
K97	$i\text{-}C_3H_7O_2 + HO_2$	→	$i\text{-}C_3H_7OOH$	K89	
K98	$i\text{-}C_3H_7O_2 + NO$	→	$CH_3COCH_3 + HO_2 + NO_2$	$(1 - RTC3S) * 2.7 \times 10^{-12} \exp(360/T)$	1, 4
K99	$i\text{-}C_3H_7O_2 + NO$	→	$i\text{-}C_3H_7ONO_2$	$RTC3S * 2.7 \times 10^{-12} \exp(360/T)$	1, 4
K100	$i\text{-}C_3H_7O_2 + CH_3OO$	→	$CH_3COCH_3 + HCHO + 2HO_2$	$0.8 * (1.03 \times 10^{-13} \exp(365/T) * 1.6 \times 10^{-12} \exp(-2200/T))^{0.5}$	3
K101	$i\text{-}C_3H_7O_2 + CH_3OO$	→	$CH_3COCH_3 + CH_3OH$	$0.2 * (1.03 \times 10^{-13} \exp(365/T) \times 1.6 \times 10^{-12} \exp(-2200/T))^{0.5}$	3
K102	$i\text{-}C_3H_7OOH + OH$	→	$i\text{-}C_3H_7O_2$	$1.9 \times 10^{-12} \exp(190/T)$	3
K103	$i\text{-}C_3H_7OOH + OH$	→	$CH_3COCH_3 + OH$	1.66×10^{-11}	3
K104	$i\text{-}C_3H_7ONO_2 + OH$	→	$CH_3COCH_3 + NO_2$	$6.2 \times 10^{-13} \exp(-230/T)$	1
K105	$C_2H_5CHO + OH$	→	$CH_3C(O)OO + CO$	$4.9 \times 10^{-12} \exp(405/T)$	1
K106	$C_2H_5CHO + NO_3$	→	$CH_3C(O)OO + CO + HNO_3$	6.3×10^{-15}	1
K107	$CH_3COCH_3 + OH$	→	CH_3COCH_2OO	$8.8 \times 10^{-12} \exp(-1320/T) + 1.7 \times 10^{-14} \exp(423/T)$	1
K108	$CH_3COCH_2OO + NO$	→	$CH_3COCHO + NO_2 + HO_2$	$2.7 \times 10^{-13} \exp(360/T)$	3
K109	$CH_3COCH_2OO + HO_2$	→	CH_3COCH_2OOH	$1.36 \times 10^{-13} \exp(1250/T)$	3
K110	$CH_3COCH_2OOH + OH$	→	$0.7\ CH_3COCHO + 0.3\ CH_3COCH_2OO + OH$	$1.90 \times 10^{-12} \exp(190/T)$	3
K111	$C_3H_6 + OH$	→	HOC_3H_6OO	$8 \times 10^{-27}(T/300)^{-3.5}[N_2]$ $3.0 \times 10^{-11}(T/300)^{-1.0}$ $Fc = 0.5$	1
K112	$C_3H_6 + NO_3$	→	$0.35\ n\text{-}C_3H_7ONO_2 + 0.65\ i\text{-}C_3H_7ONO_2$	$4.6 \times 10^{-13} \exp(-1155/T)$	1, 3
K113	$C_3H_6 + O_3$	→	$0.62\ HCHO + 0.62\ CH_3CHO + 0.38\ CH_3OO + 0.56\ CO + 0.36\ HO_2 + 0.36\ OH + 0.2\ CO_2$	$5.77 \times 10^{-15} \exp(-1880/T)$	1, 3
K114	$HOC_3H_6OOH + OH$	→	$0.928\ CH_3COCH_2OH + 0.072\ HOC_3H_6OO + 0.928\ OH$	$2.44 \times 10^{-11} + 1.9 \times 10^{-12} \exp(190/T)$	3
K115	$HOC_3H_6OO + HO_2$	→	HOC_3H_6OOH	K89	3
K116	$HOC_3H_6OO + NO$	→	$CH_3CHO + HCHO + HO_2 + NO_2$	$(1 - 0.35RTC3P - 0.65RTC3S) * 2.55 \times 10^{-12} \exp(380/T)$	1, 3
K117	$HOC_3H_6OO + NO$	→	$0.35\ n\text{-}C_3H_7ONO_2 + 0.65\ i\text{-}C_3H_7ONO_2$	$(0.35RTC3P + 0.65RTC3S) *$	1, 3

				$2.55 \times 10^{-12} \exp(380/T)$	
K118	$HOC_3H_6OO + CH_3OO$	→	$CH_3CHO + 2HCHO + 2HO_2$	$0.8 * 6.0 \times 10^{-13}$	3
K119	$HOC_3H_6OO + CH_3OO$	→	$CH_3COCH_2OH + CH_3OH$	$0.2 * 6.0 \times 10^{-13}$	3
K120	$CH_3COCH_2OH + OH$	→	$CH_3COCHO + HO_2$	$1.6 \times 10^{-12} \exp(305/T)$	1
K121	$CH_3COCHO + OH$	→	$CH_3C(O)OO + CO$	$1.9 \times 10^{-12} \exp(575/T)$	1
K122	$CH_3COCHO + NO_3$	→	$CH_3C(O)OO + CO + HNO_3$	5.0×10^{-16}	1
K123	$CH_3C(O)COOH + OH$	→	$CH_3C(O)OO + CO_2$	8.0×10^{-13}	3
K124	$C_4H_{10} + OH$	→	C_4H_9OO	$9.8 \times 10^{-12} \exp(-425/T)$	3
K125	$C_4H_{10} + NO_3$	→	$C_4H_9OO + HNO_3$	$2.8 \times 10^{-12} \exp(-3280/T)$	1
K126	$C_4H_9OO + HO_2$	→	C_4H_9OOH	$0.625 * 2.91 \times 10^{-13} \exp(1300/T)$	3
K127	$C_4H_9OO + NO$	→	$NO_2 + 0.67(CH_3CH_2COCH_3 + HO_2) + 0.33(C_2H_5OO + CH_3CHO)$	$(1 - RTC4P) \times 8.3 \times 10^{-12}$	1, 4
K128	$C_4H_9OO + NO$	→	$C_4H_9ONO_2$	$RTC4P \times 8.3 \times 10^{-12}$	1, 4
K129	$C_4H_9OO + CH_3OO$	→	$HCHO + HO_2 + 0.67(CH_3CH_2C(O)CH_3 + HO_2) + 0.33(CH_3CHO + CH_3CH_2OO)$	$0.8 * 1.3 \times 10^{-12}$	3
K130	$C_4H_9OO + CH_3OO$	→	$CH_3CH_2COCH_3 + CH_3OH$	$0.2 * 1.3 \times 10^{-12}$	3
K131	$C_4H_9OOH + OH$	→	C_4H_9OO	$1.90 \times 10^{-12} \exp(190/T)$	3
K132	$C_4H_9OOH + OH$	→	$CH_3CH_2COCH_3 + OH$	2.15×10^{-11}	3
K133	$C_4H_9ONO_2 + OH$	→	$CH_3CH_2COCH_3 + NO_2$	8.6×10^{-13}	1
K134	$CH_3CH_2COCH_3 + OH$	→	$CH_3CH(OO)COCH_3$	$1.5 \times 10^{-12} \exp(-90/T)$	1
K135	$CH_3CH(OO)COCH_3 + HO_2$	→	$CH_3CH(OOH)COCH_3$	K126	
K136	$CH_3CH(OO)COCH_3 + NO$	→	$CH_3CHO + CH_3C(O)OO + NO_2$	$(1 - RTC4S) \times 2.55 \times 10^{-12} \exp(380/T)$	1, 4
K137	$CH_3CH(OO)COCH_3 + NO$	→	$CH_3CH(ONO_2)COCH_3$	$RTC4S \times 2.55 \times 10^{-12} \exp(380/T)$	1, 4
K138	$CH_3CH(OOH)COCH_3 + OH$	→	$CH_3CH(OO)COCH_3$	K131	
K139	$CH_3CH(OOH)COCH_3 + OH$	→	$CH_3C(O)C(O)CH_3 + OH$	1.88×10^{-11}	3
K140	$CH_3CH(ONO_2)COCH_3 + OH$	→	$CH_3C(O)C(O)CH_3 + NO_2$	1.2×10^{-12}	1
K141	$ISOP + OH$	→	$0.98 \ ISOPOO + 0.0003 \ ELVOC + 0.007 \ SVOC$	$2.7 \times 10^{-11} \exp(390/T)$	1, 3
K142	$ISOP + NO_3$	→	$ISOPONO_2$	$2.95 \times 10^{-12} \exp(-450/T)$	1, 3
K143	$ISOP + O_3$	→	$0.98 * (0.3 \ MACR + 0.3 \ MACROO + 0.2 \ MVK + 0.2 \ MVKOO + 0.78 \ HCHO + 0.22 CO + 0.125 \ HO_2 + 0.125 OH) + 0.0001 \ ELVOC + 0.009 \ SVOC$	$1.05 \times 10^{-14} \exp(-2000/T)$	1, 3
K144	$ISOPOO + HO_2$	→	$ISOPOOH$	$2.06 \times 10^{-13} \exp(1300/T)$	3, 7
K145	$ISOPOO + NO$	→	$HCHO + 0.64 \ MVK + 0.36 \ MACR + HO_2 + NO_2$	$(1-RTC5S) * 2.7 \times 10^{-12} \exp(360/T)$	3
K146	$ISOPOO + NO$	→	$ISOPONO_2$	$RTC5S * 2.7 \times 10^{-12} \exp(360/T)$	3
K147	$ISOPOO + NO_3$	→	$HCHO + 0.64 \ MVK + 0.36 \ MACR + HO_2 + NO_2$	2.3×10^{-12}	3
K148	$ISOPOO + CH_3OO$	→	$0.64 \ MVK + 0.36 \ MACR + 2HCHO + 2HO_2$	$0.8 * 2.65 \times 10^{-12}$	3

K149	ISOPOO + CH$_3$OO	→	0.64 MVK + 0.36 MACR + HCHO + CH$_3$OH	0.2 * 2.65 x 10^{-12}	3
K150	ISOPOO	→	HPALD + HO$_2$	4.12×10^{8}exp(-7700/T)	6, 7
K151	ISOPOOH + OH	→	IEPOX + OH	1.9×10^{-11}exp(-390/T)	8
K152	ISOPOOH + OH	→	ISOPOO	0.7 * 3.8×10^{-12}exp(-200/T)	8
K153	ISOPOOH + OH	→	0.64 CH$_3$COCHO + 0.64 HOCH$_2$CHO + 0.36 HOCH$_2$C(O)CH$_3$ + 0.36 CHOCHO + OH	0.3 * 3.8×10^{-12}exp(-200/T)	8, 9
K154	ISOPONO$_2$ + OH	→	0.64 CH$_3$COCHO + 0.64 HOCH$_2$CHO + 0.36 HOCH$_2$C(O)CH$_3$ + 0.36 CHOCHO + NO$_2$	1.77×10^{-11}exp(-500/T)	8
K155	HPALD + OH	→	0.5 HOCH$_2$C(O)CH$_3$ + 0.5 CH$_3$C(O)CHO + 0.25 HOCH$_2$CHO + 0.25 CHOCHO + HCHO + HO$_2$ + OH	4.6×10^{-11}	6
K156	IEPOX + OH	→	IEPOXOO	5.78×10^{-11}exp(-400/T)	8
K157	IEPOXOO + HO$_2$	→	0.725 HOCH$_2$C(O)CH$_3$+ 0.275 HOCH$_2$CHO + 0.275 HOCH$_2$CHO + 0.275 CH$_3$C(O)CHO + 1.125 OH + 0.825 HO$_2$ + 0.2 CO$_2$ + 0.375 HCHO + 0.074 HCOOH + 0.251 CO	7.4×10^{-13}exp(700/T)	8
K158	IEPOXOO + NO	→	0.725 HOCH$_2$C(O)CH$_3$+ 0.275 HOCH$_2$CHO + 0.275 HOCH$_2$CHO + 0.275 CH$_3$C(O)CHO + 1.125 OH + 0.825 HO$_2$ + 0.2 CO$_2$ + 0.375 HCHO + 0.074 HCOOH + 0.251 CO + NO$_2$	2.7×10^{-12}exp(360/T)	3
K159	IEPOXOO + NO$_3$	→	0.725 HOCH$_2$C(O)CH$_3$+ 0.275 HOCH$_2$CHO + 0.275 HOCH$_2$CHO + 0.275 CH$_3$C(O)CHO + 1.125 OH + 0.825 HO$_2$ + 0.2 CO$_2$ + 0.375 HCHO + 0.074 HCOOH + 0.251 CO + NO$_2$	1.74 * 2.3×10^{-12}	3
K160	MVK + OH	→	MVKOO	2.6 x 10^{-12}exp(610/T)	1
K161	MVK + NO$_3$	→	0.65 HCOOH + 0.65 CH$_3$COCHO + 0.35 HCHO + 0.35 CH$_3$C(O)OOH + HNO3	6.0 x 10^{-16}	1
K162	MVK + O$_3$	→	0.38 CH$_3$COCHO + 0.2088 CH$_3$C(O)OO + 0.26 CH$_3$COCOOH + 0.26 CO + 0.0432 CH$_3$COOH + 0.108 CH$_3$CHO + 0.62 HCHO + 048 CO$_2$ + 0.54 HO$_2$ + 0.1008 OH	8.5 x 10^{-16}exp(-1520/T)	1, 3
K163	MVKOO + HO$_2$	→	MVKOOH	K144	
K164	MVKOO + NO	→	0.295 CH$_3$C(O)CHO + 0.295 HCHO + 0.670 CH$_3$CHO + 0.670 HOCHCHO + 0.295 HO$_2$ + 0.965 NO$_2$ + 0.0352 MVKONO$_2$	2.7 x 10^{-12}exp(360/T)	3
K165	MVKOOH + OH	→	CH$_3$C(O)CHO + CO + 2HO$_2$ + OH	2.55 x 10^{-11}	3
K166	MVKOOH + OH	→	MVKOO	1.9 x 10^{-12}exp(190/T)	3
K167	MVKONO$_2$ + OH	→	CH$_3$C(O)CHO + CO + HO$_2$ + NO$_2$	1.33 x 10^{-12}	3
K168	MACR + OH	→	MACROO	8.0 x 10^{-12}exp(380/T)	1
K169	MACR + NO$_3$	→	MACROO + HNO$_3$	3.4 x 10^{-15}	1
K170	MACR + O$_3$	→	0.90 CH$_3$COCHO + 0.5 HCHO + 0.5 CO + 0.14 HO$_2$ + 0.24 OH	1.4 x 10^{-15}exp(-2100/T)	1, 3
K171	MACROO + HO$_2$	→	MACROOH	0.625 * 2.91 x 10^{-13}exp(1300/T)	3
K172	MACROO + NO	→	0.987 (CH$_3$COCH$_2$OH + CO + NO$_2$ + HO$_2$) + 0.013 MACRONO$_2$	K164	1, 3
K173	MACROOH + OH	→	CH$_3$COCH$_2$OH + CO + OH	3.77 x 10^{-11}	

K174	MACROOH + OH	→	MACROO	K166
K175	MACRONO$_2$ + OH	→	CH$_3$COCHO + CO + HO$_2$ + NO$_2$	4.34 x 10^{-12}	3
K176	TERP + OH	→	0.81 TERPOO + 0.05 ELVOC + 0.14 SVOC	0.5 * 1.34 x 10^{-11} exp(410/T) + 0.5 * 1.62 x 10^{-11} exp(460/T)	1, 10
K177	TERP + NO$_3$	→	TERPOO + HNO$_3$	0.5 * 1.2 x 10^{-12} exp(490/T) + 0.5 * 2.5 x 10^{-12}	1, 10
K178	TERP + O$_3$	→	0.915 MACR + 0.36 MVK + 0.24 PRV + 1.68 HCHO + 0.16 CO + 0.6 HCOOH + 0.08 C3H6 + 0.68 OH + 0.05 ELVOC + 0.14 SVOC	0.5 * 8.22 x 10^{-16} exp(-640/T) + 0.5 * 1.39 x 10^{-15} exp(-1280/T)	1, 10
K179	TERPOO + HO$_2$	→	2 ISOPOOH	K144
K180	TERPOO + NO	→	2 (HCHO + 0.64MVK + 0.36MACR + HO$_2$) + NO$_2$	K145
K181	TERPOO + NO	→	2 ISOPONO$_2$	K146
K182	TERPOO + NO$_3$	→	2 (HCHO + 0.64MVK + 0.36MACR + HO$_2$) + NO$_2$	K147
K183	TERPOO + CH$_3$OO	→	2 (0.64MVK + 0.36MACR + 2HCHO + 2HO$_2$)	K148
K184	TERPOO + CH$_3$OO	→	2 (0.64MVK + 0.36MACR + HCHO + CH$_3$OH)	K149
K185	AROM + OH	→	AROMOO + HO$_2$	A1 * 1.8 x 10^{-12}exp(340/T) + A2 * 1.72 x 10^{-11} + A3 * 2.3 x 10^{-12}exp(-190/T)	1, 11
K186	AROM + NO$_3$	→	AROMOO + HNO$_3$	A1 * 7.8 x 10^{-17} + A2 * 3.54 x 10^{-16}	1, 11
K187	AROM + O$_3$	→	AROMOO	A1 * 1.0x 10^{-21} + A2 * (2.4 x 10^{-13}exp(-5586/T) + 5.37 x 10^{-13}exp(-6039/T) + 1.91 x 10^{-13}exp(-5586/T))/3	1, 11, 12
K188	AROMOO + HO$_2$	→	C$_4$H$_9$OOH + CHOCHO + HCHO	K126
K189	AROMOO + NO	→	NO$_2$ + 0.67CH$_3$CH$_2$COCH$_3$ + 0.67 HO$_2$ + 0.33C$_2$H$_5$OO + 0.33CH$_3$CHO + CHOCHO + HCHO	K127
K190	AROMOO + NO	→	C$_4$H$_9$ONO$_2$ + CHOCHO + HCHO	K128
K191	AROMOO + CH$_3$OO	→	HCHO + HO$_2$ + 0.67(CH$_3$CH$_2$C(O)CH$_3$ + HO$_2$) + 0.33(CH$_3$CHO + CH$_3$CH$_2$OO) + CHOCHO + HCHO	K129
K192	AROMOO + CH$_3$OO	→	CH$_3$CH$_2$COCH$_3$ + CH$_3$OH + CHOCHO + HCHO	K130
K193	SO$_2$ + OH	→	HO$_2$ + H$_2$SO$_4$	3.3 x 10^{-31}(T/300)$^{-4.3}$[N$_2$] 1.6 x 10^{-12} (T/300)$^{-0.7}$ Fc = 0.6	2
K194	DMS + OH	→	CH$_3$OO + HCHO + SO$_2$	1.1 x 10^{-11}exp(-240/T)	2
K195	DMS + OH	→	0.75 CH$_3$OO + 0.75 HCHO + 0.75 SO$_2$ + 0.25 MSA	1.0 x 10^{-39}[O$_2$] exp(5820/T) / (1 + 5.0 x 10^{-30}[O$_2$] exp(6280/T))	2
K196	DMS + NO$_3$	→	CH$_3$OO + HCHO + SO$_2$ + HNO$_3$	1.9 x 10^{-13}exp(520/T)	2
K197	NH$_3$ + OH	→	NH$_2$ + HO$_2$	1.7 x 10^{-12}exp(-710/T)	2
K198	NH$_2$ + O$_2$	→	NH$_2$O$_2$	6.0 x 10^{-21}	2

K199	$NH_2 + O_3$	→	NH_2O_2	$4.3 \times 10^{-12}\exp(-930/T)$	2
K200	$NH_2 + OH$	→	NH_2O_2	3.4×10^{-11}	2
K201	$NH_2 + HO_2$	→	NH_3	3.4×10^{-11}	2
K202	$NH_2 + NO$	→	$NH_2O_2 + NO_2$	$4.0 \times 10^{-12}\exp(450/T)$	2
K203	$NH_2 + NO_2$	→	$NH_2O_2 + NO$	$2.1 \times 10^{-12}\exp(650/T)$	2
K204	$NH_2O_2 + O_3$	→	NH_2	K199	
K205	$NH_2O_2 + HO_2$	→	NH_2	K201	
K206	$NH_2O_2 + NO$	→	$NH_2 + NO_2$	K202	

[#] The reaction products O_2, H_2, and H_2O are not shown.

[1] The chemical kinetic data and mechanistic information was taken from the website of the IUPAC Task Group on Atmospheric Chemical Kinetic Data Evaluation: www.iupac-kinetic.ch.cam.ac.uk

[2] The chemical kinetic data and mechanistic information was taken from the website of the NASA Panel for Data Evaluation (Evaluation No. 18, JPL Publication 15-10) http://jpldataeval.jpl.nasa.gov

[3] The chemistry mechanistic information was taken from the Master Chemical Mechanism (MCM v3.3.1):
- for non-aromatic schemes: Jenkin et al. (1997); Saunders et al. (2003)
- for the isoprene scheme: Jenkin et al. (2015)
- for aromatic schemes: Jenkin et al. (2003); Bloss et al. (2005)
- and via the website: http://mcm.leeds.ac.uk/MCM

[4] Atkinson (1997):

$R_1 = 2.7 \times 10^{14}\exp(-6350/T)$
$R_2 = 6.3 \times 10^{-14}\exp(-550/T)$

$f = R_1/(R_1 + R_2 \times [O_2])$

$R_1 = 1.94 \times 10^{-22}$ [AIR] $\exp(0.972 \times N_c)$
$R_2 = 0.826 \times (T/300)^{-8.1}$

$A = 1/(1+\log_{10}(R_1/R_2)^2)$

$RTC(N_c)P = 0.4 \times R_1/(1+R_1/R_2)\ 0.411^A$

$RTC(N_c)S = R1/(1+R_1/R_2)\ 0.411^A$

where, N_c is the number of carbons (i.e., 1-5)

[5] Orlando et al. (1992); Poisson et al. (2000)

[6] Peeters and Müller (2010)

[7] Crounse et al. (2011)

[8] Paulot et al. (2009)

[9] Browne et al. (2014)

[10] Average of α- and β-pinene

[11] A1, A2, A3 represents the relative contributions of *ortho-, meta-,* and *para*-xylene, toluene and benzene (roughly 0.4, 0.6 and 0.4, respectively, for the year 2006)

[12] Average of ortho-, meta- and para-isomers of xylene

Table 3. Global annual emissions of trace gases used for the MOGUNTIA chemistry scheme in TM5-MP for the year 2006, in Tg yr^{-1} unless specified otherwise.

Species	Long name	Emissions						
		Anthropogenic[&]	Biomass Burning	Biogenic	Soil	Oceanic	Other	Total
CO	carbon monoxide	600.5	386.4	90.2		19.9		1097
HCHO	formaldehyde	2.4	5.2	4.7				12.3
HCOOH	formic acid	4.6	1.8	3.5				9.8
CH$_3$OH	methanol	4.7	9.8	131.9				146.4
C$_2$H$_6$	ethane	6.2	3.4	0.3		1.0		10.9
C$_2$H$_4$	ethene	5.3	4.8	18.3		1.4		29.8
C$_2$H$_2$	acetylene	3.3						3.3
CH$_3$CHO	acetaldehyde	1.2	4.4	21.9				27.5
CH$_3$COOH	acetic acid	4.6	18.0	3.5				26.1
CH$_3$CH$_2$OH	ethanol	0.5	0.1	18.6				19.3
HOCH$_2$CHO	glycol-aldehyde	1.4	4.3					5.7
CHOCHO	glyoxal	2.4	5.2					7.6
C$_3$H$_8$	propane	6.5	0.7	0.03		1.3		8.5
C$_3$H$_6$	propene and higher alkenes	8.3	4.8	17.5		1.5		32.1
CH$_3$COCH$_3$	acetone	2.7	1.7	37.7				42.1
CH$_3$COCHO	methylglyoxal	1.6	3.4					5.0
C$_4$H$_{10}$	butane and higher alkanes (including butane, pentane, hexane, higher alkanes, and other vocs)	52.8	0.5	0.1				53.4
CH$_3$CH$_2$COCH$_3$	methyl-ethyl-ketone (including higher ketones except for acetone)	1.4	1.4	0.9				3.7
C$_5$H$_8$	isoprene			579.4				579.4
C$_{10}$H$_{16}$	monoterpenes			97.9				97.9
C$_7$H$_8$	toluene and aromatics (including toluene, xylene benzene, trimethylbenzene and higher aromatics)	25.3	4.0	1.5				30.8

		A	B	C	D	E	F	Total
NO_X [#]	nitrogen oxides	42.3	6.6		5.0		6.0 [*]	59.9
NH_3	ammonia	56.1	4.4		2.3	8.1		70.9
SO_2	sulfur dioxide	120.5	2.3				9.3 [$]	132.1
CH_3SCH_3	dimethylsulphide			1.7		95.8		97.5

[&] including aircraft emissions

[#] in Tg-N yr^{-1}

[*] NO_X production from lightning

[$] SO_2 from volcanoes

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Table 4. Tropospheric budgets of O_3 for the year 2006 in Tg(O_3) yr^{-1} and burden in Tg(O_3), using the 150 ppb O_3 mixing ratio to define tropopause level. In parentheses, the relative differences using the 100 ppb O_3 mixing ratios are also presented, calculated by reference to the 150 ppb O_3 tropopause level definition.

Production terms	mCB05 (EBI)		mCB05 (KPP)		MOGUNTIA		Loss terms	mCB05 (EBI)		mCB05 (KPP)		MOGUNTIA	
Stratospheric inflow*	632	(10%)	429	(32%)	424	(30%)	Deposition	955	(0%)	932	(0%)	913	(0%)
Trop. chem. production	5589	(-3%)	5719	(-3%)	5709	(-3%)	Trop. chem. loss	5192	(-1%)	5216	(-1)	5219	(-1%)
Trop. burden	385	(-8%)	384	(-8%)	375	(-8%)	Trop. lifetime (days)	22.8	(-8%)	22.8	(-8%)	22.3	(-6%)

*sum of the deposition and the tropospheric chemical loss minus the production

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Table 5. Tropospheric chemical budget of OH for the year 2006 in Tg(OH) yr^{-1}, using the 150 ppb O_3 mixing ratio to define tropopause level. In parentheses, the relative differences using the 100 ppb O_3 mixing ratios are also presented, calculated by reference to the 150 ppb O_3 tropopause level definition.

Production terms	mCB05 (EBI)		mCB05 (KPP)		MOGUNTIA		Loss terms	mCB05 (EBI)		mCB05 (KPP)		MOGUNTIA	
$O(^1D) + H_2O$	1960	(0%)	1953	(0%)	1878	(0%)	$OH + CO$	1665	(-2%)	1671	(-2%)	1775	(-2%)
$NO + HO_2$	1268	(-4%)	1312	(-4%)	1426	(-4%)	$OH + CH_4$	613	(0%)	626	(0%)	644	(-1%)
$O_3 + HO_2$	560	(-1%)	566	(-1%)	561	(-1%)	$OH + O_3$	254	(-2%)	260	(-2%)	262	(-3%)
$H_2O_2 + hv$	262	(-1%)	265	(-1%)	303	(-1%)	$OH + ISOP$	114	(-1%)	115	(-1%)	120	(0%)
Other	203	(-2%)	201	(-2%)	120	(-1%)	Other	1606	(-1%)	1626	(-1%)	1487	(-1%)

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Table 6. Global budgets of CO for the year 2006 in Tg(CO) yr^{-1} and burden in Tg(CO), using the 150 ppb O_3 mixing ratio to define tropopause level. In parentheses, the relative differences using the 100 ppb O_3 mixing ratios are also presented, calculated by reference to the 150 ppb O_3 tropopause level definition.

Production terms	mCB05 (EBI)		mCB05 (KPP)		MOGUNTIA		Loss terms	mCB05 (EBI)		mCB05 (KPP)		MOGUNTIA	
Emissions	1097	(0%)	1097	(0%)	1097	(0%)	Deposition	98	(0%)	97	(0%)	99	(0%)
Trop. chem. production	1809	(-1%)	1818	(-1%)	1992	(-1%)	Trop. chem. loss	2840	(-6%)	2849	(-6%)	2924	(-2%)
Strat. chem. production	26	(69%)	26	(73%)	26	(65%)	Strat. chem. loss	87	(68%)	89	(69%)	90	(68%)
Atmos. burden	370	(0%)	360	(0%)	361	(0%)	Lifetime (days)	47.5	(2%)	46.2	(2%)	43.6	(3%)

Figures
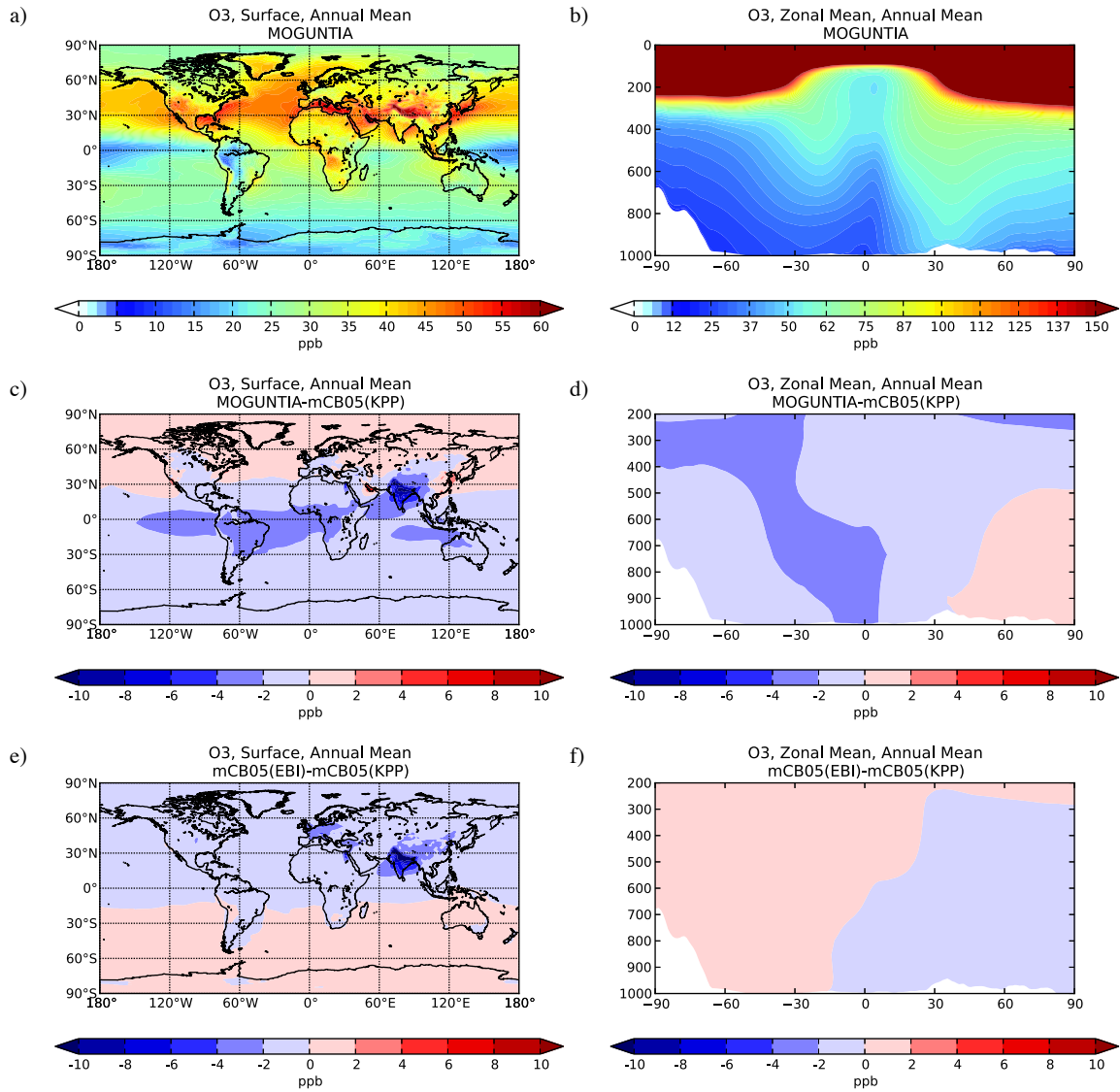
Figure 1: Simulated annual mean surface (left columns) and zonal mean (right columns) O₃ mixing ratios (ppb) for the MOGUNTIA chemistry scheme for the year 2006 (a,b), and the respective differences compared to mCB05(KPP) (c,d); the surface and zonal mean absolute differences between mCB05(KPP) and mCB05(EBI) are also presented (e,f).
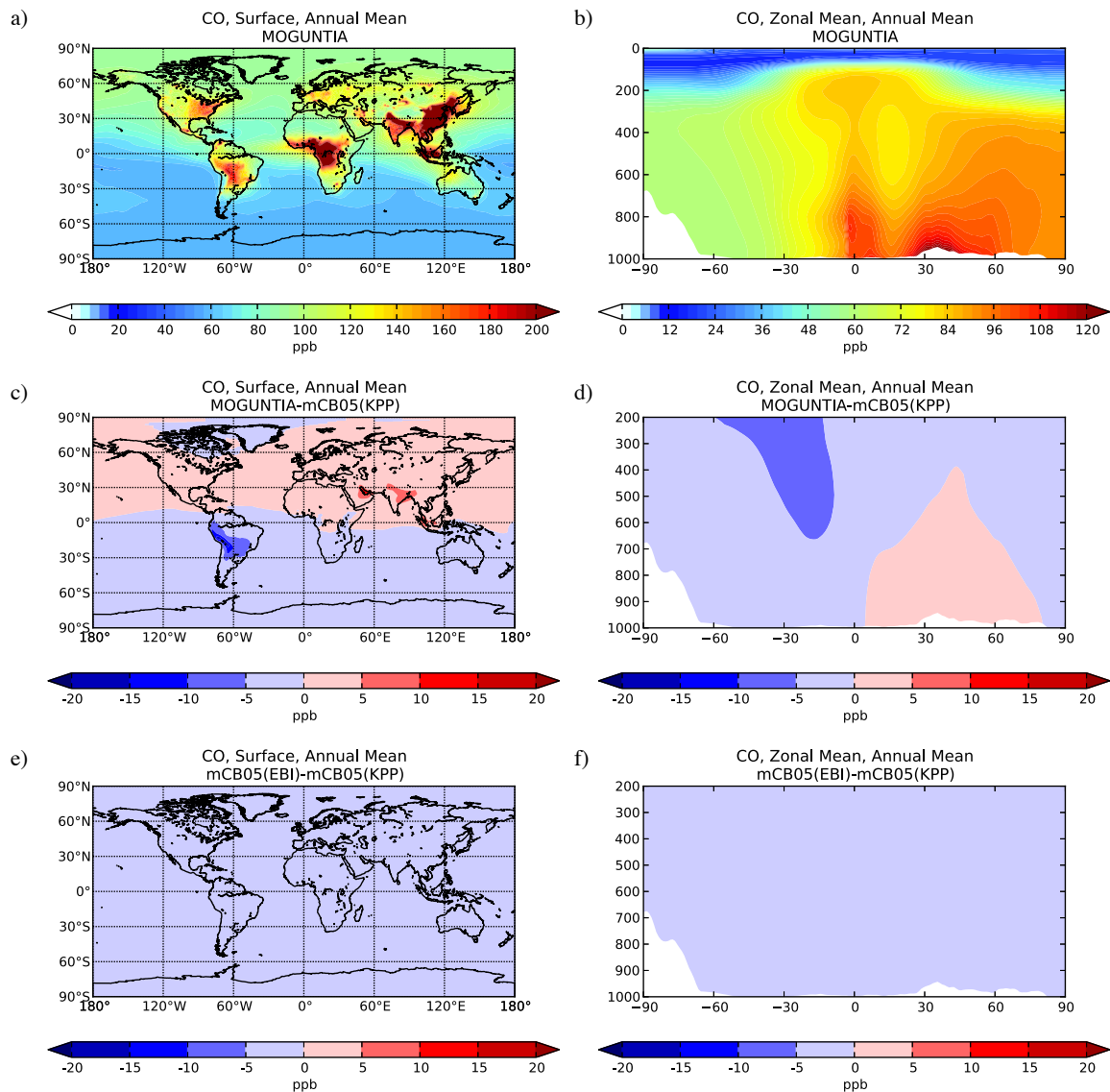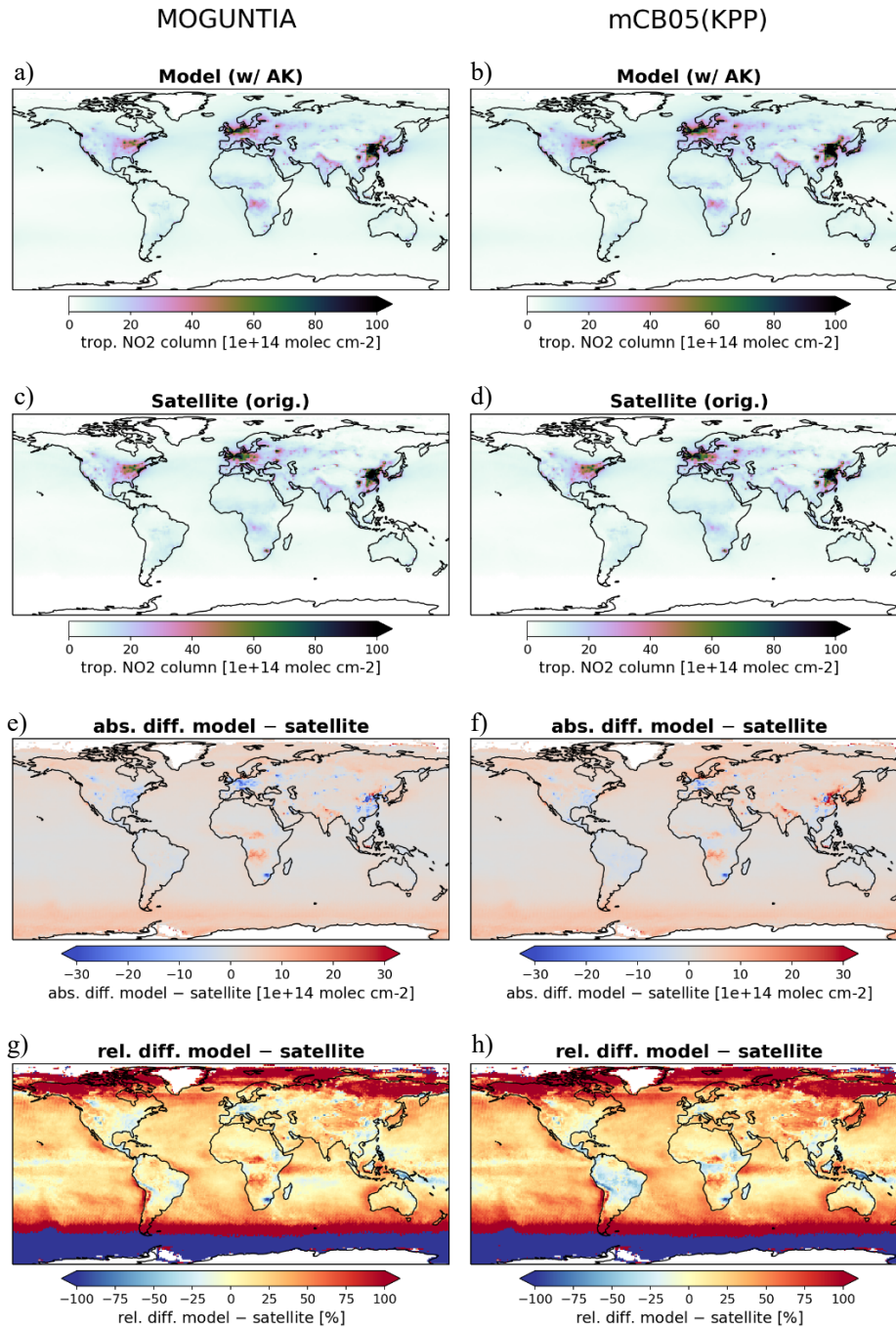
Figure 2: Simulated annual mean surface (left columns) and zonal mean (right columns) CO mixing ratios (ppb) for the MOGUNTIA chemistry scheme for the year 2006 (a,b), and the respective differences compared to mCB05(KPP) (c,d); the surface and zonal mean absolute differences between mCB05(KPP) and mCB05(EBI) are also presented (e,f).

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Trop. NO2 columns / OMI / 2006

MOGUNTIA

mCB05(KPP)

Figure 3: Annual mean comparison of tropospheric NO_2 vertical columns (molecules cm^{-2}) for the two chemistry schemes MOGUNTIA and mCB05(KPP) (a,b), against the Ozone Monitoring Instrument (OMI) satellite data (c,d), using the respective averaging kernel information for 2006. The absolute (e,f) and relative (g,h) differences are also presented.

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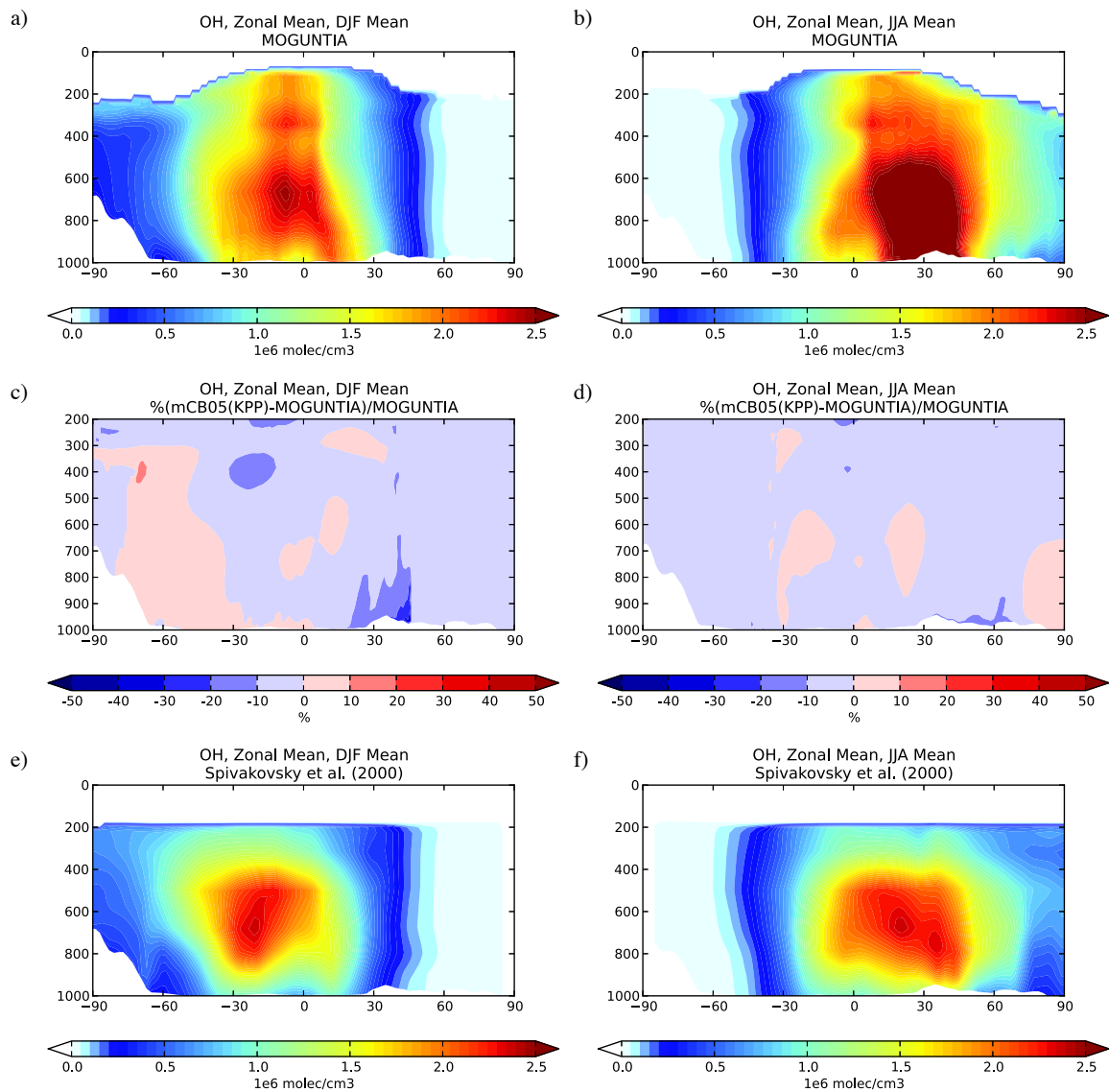
Figure 4: Zonal mean OH mixing ratios for December-January-February (DJF; left) and June-July-August (JJA; right) 2006, as simulated by the TM5-MP model with the MOGUNTIA chemistry scheme (top), the differences (%) between the mCB05(KPP) and the MOGUNTIA chemical configuration (middle), and the optimized climatological average from Spivakovsky et al. (2000), up to 200 hPa (bottom).
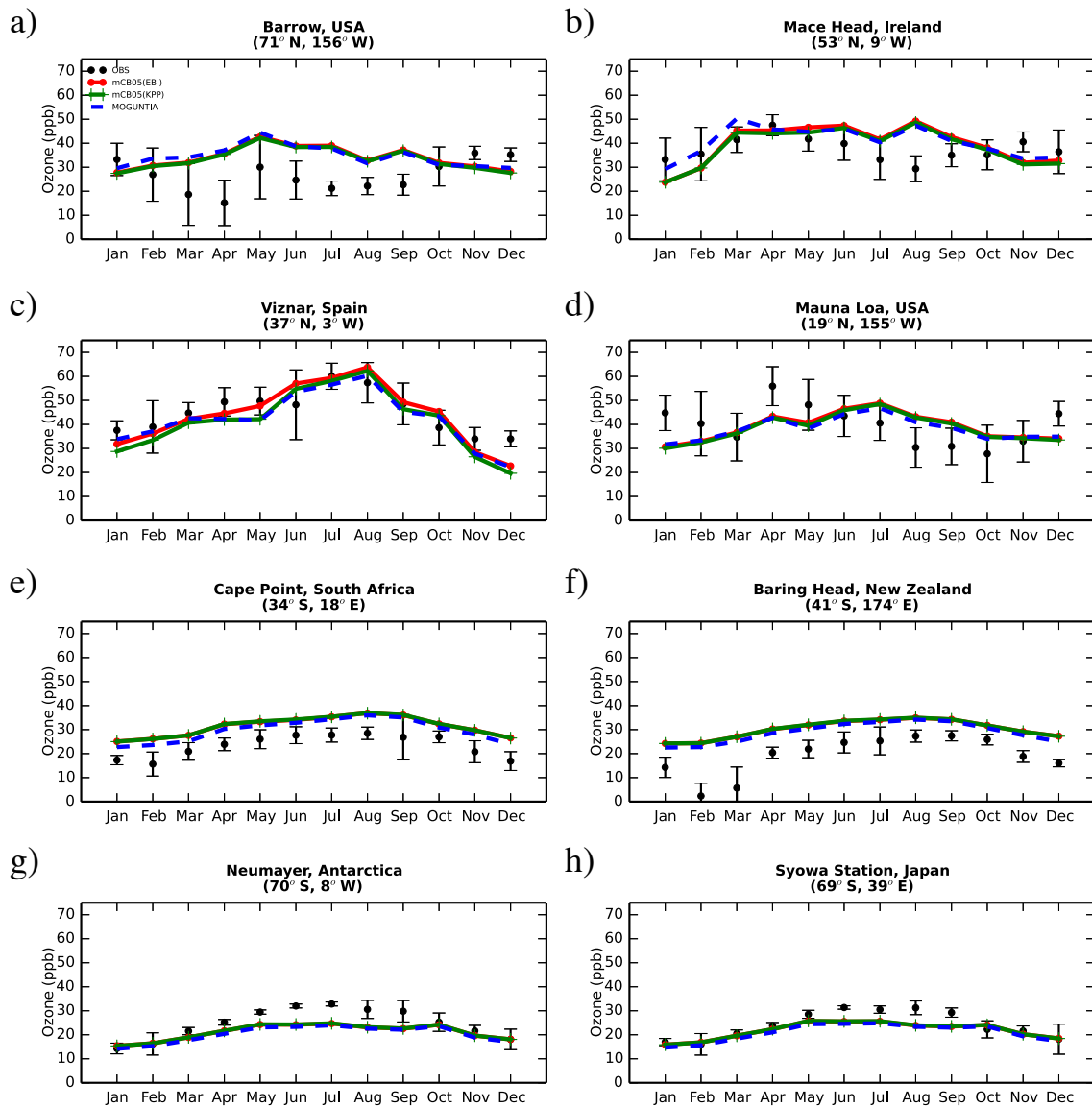
Figure 5: Monthly mean comparison of TM5-MP surface O$_3$ (ppb) against surface observations (black line) from EMEP and WOUDC databases for the two chemistry schemes, mCB05(KPP) (green line) and MOGUNTIA (blue line), using co-located model output for 2006 sampled at the measurement times; error bars indicate the standard deviation in the monthly means. For comparison, model results of the mCB05 with the EBI solver (red line) are also presented.

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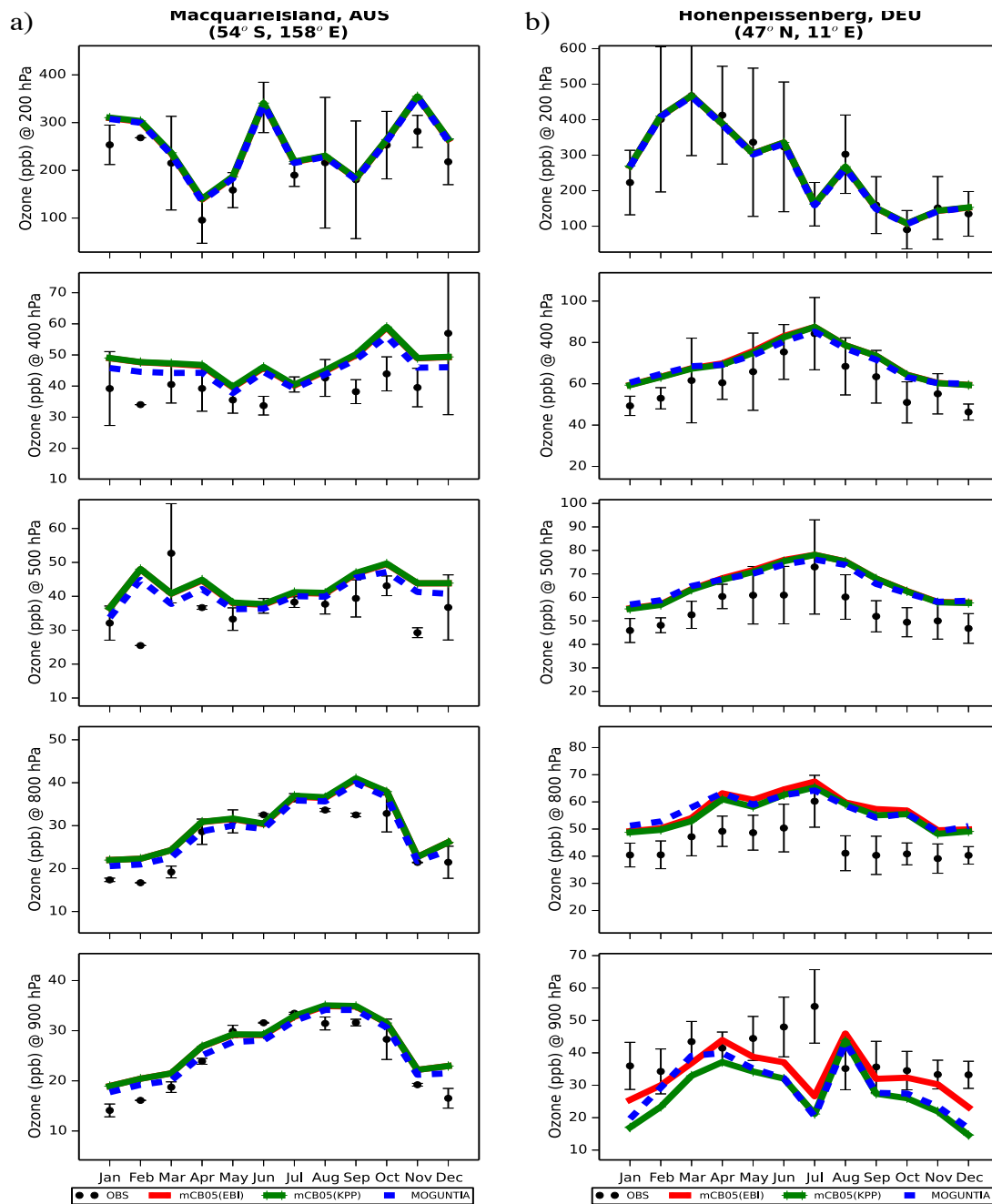
Figure 6: Monthly mean comparison of TM5-MP O$_3$ (ppb) against sonde observations (black dots, mean and standard deviation) at a) Hohenpeissenberg and b) Macquarie Island, for different pressure levels (900; 800; 500; 400; 200 hPa) for the two chemistry schemes, mCB05(KPP) (green line) and MOGUNTIA (blue line), using co-located model output for 2006 sampled at the measurement times; error bars indicate the standard deviation in the monthly means. For comparison, the results of mCB05 with the EBI solver (red line) are also presented.

60
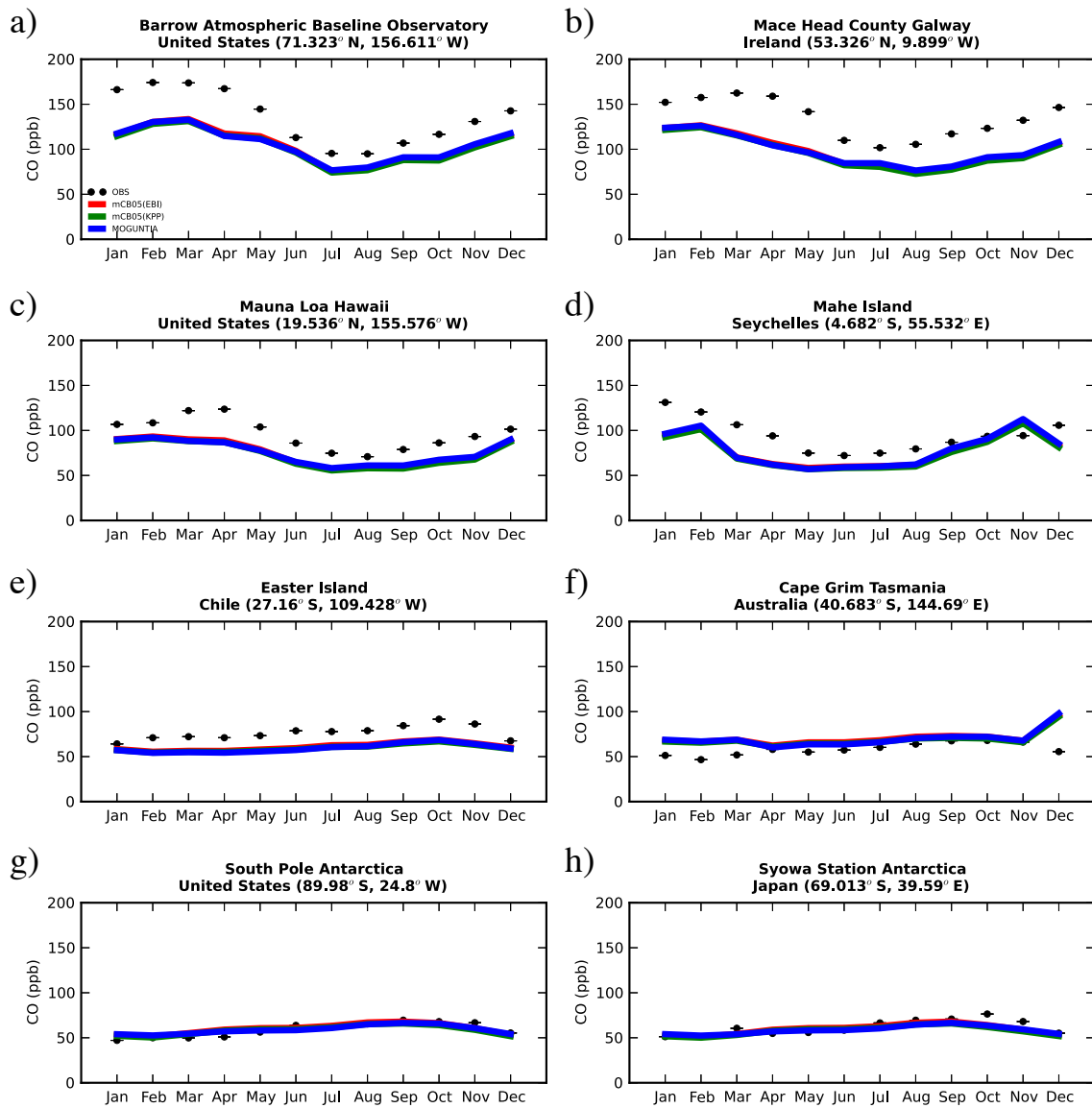
Figure 7: Monthly mean comparison of TM5-MP surface CO (ppb) against flask measurements (black line) for the two chemistry schemes, mCB05(KPP) (green line) and MOGUNTIA (blue line), using co-located model output for 2006 sampled at the measurement times; error bars indicate the standard deviation in the monthly means. For comparison, model results of the mCB05 with the EBI solver (red line) are also presented.
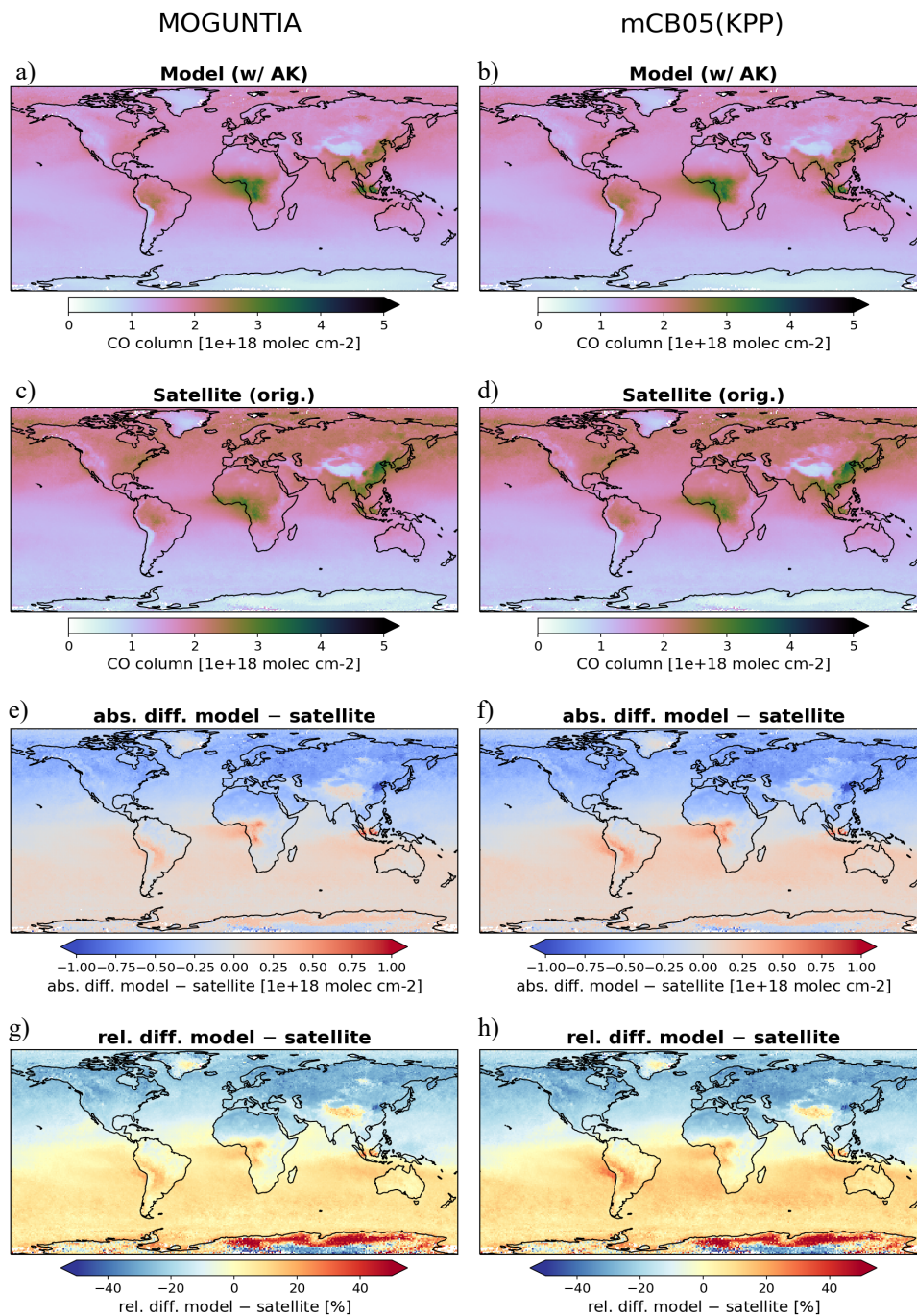
CO columns / MOPITT / 2006

MOGUNTIA

mCB05(KPP)

Figure 8: Annual mean comparison of total CO vertical columns (molecules cm^{-2}) for the two chemistry schemes of TM5-MP, MOGUNTIA and mCB05(KPP) (a,b), against MOPITT satellite data (c,d), using the respective averaging kernel information for 2006. The absolute (e,f) and relative (g, h) differences are also presented.
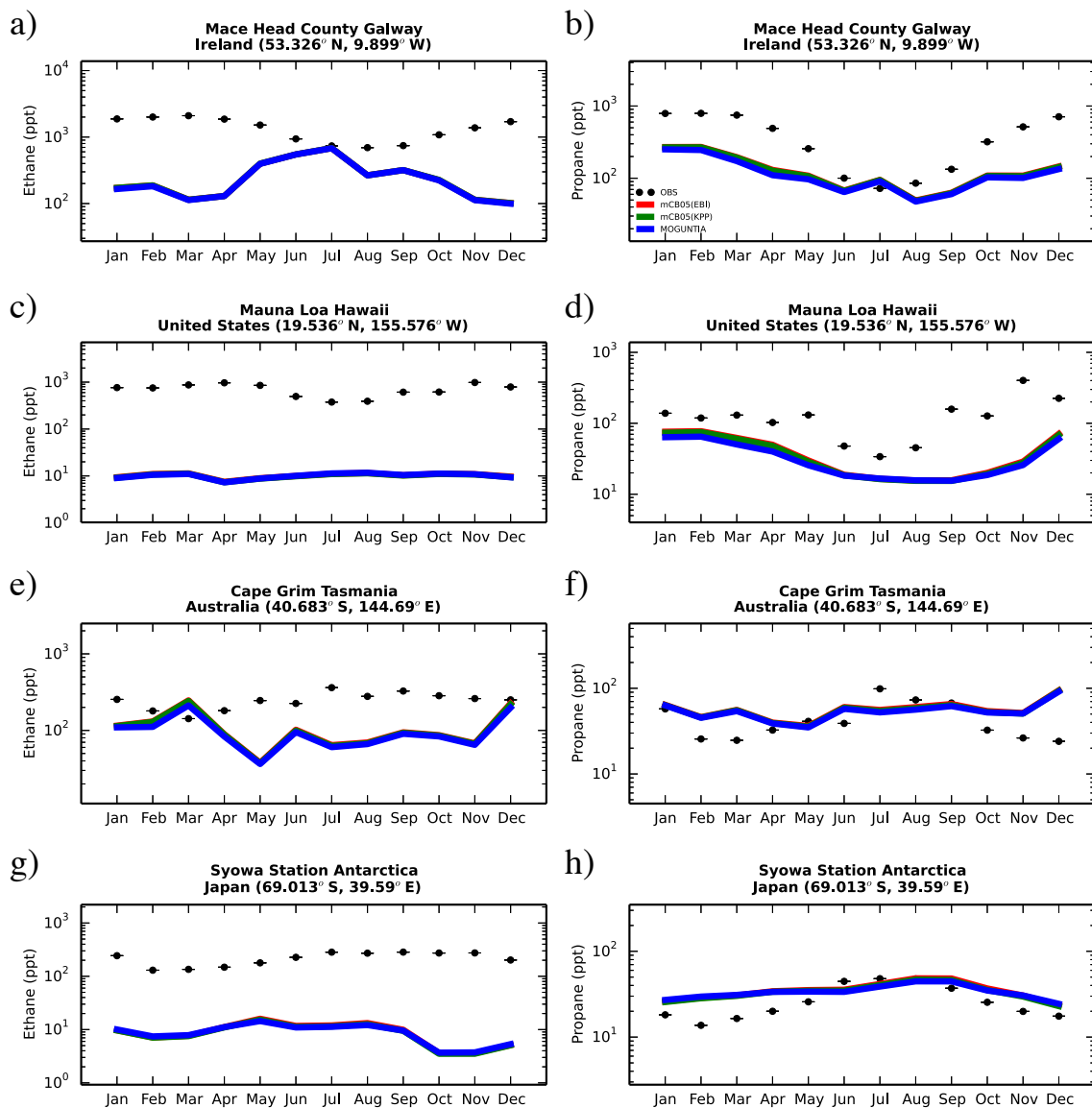
Figure 9: Monthly mean comparison of TM5-MP surface C_2H_6 (left column) and C_3H_8 (right column) against flask measurements (black dots) in ppt for the two chemistry schemes, mCB05(KPP) (green line) and MOGUNTIA (blue line), using co-located model output for 2006 sampled at the measurement times; error bars indicate the standard deviation in the monthly means. For comparison, model results of the mCB05 with the EBI solver (red line) are also presented.
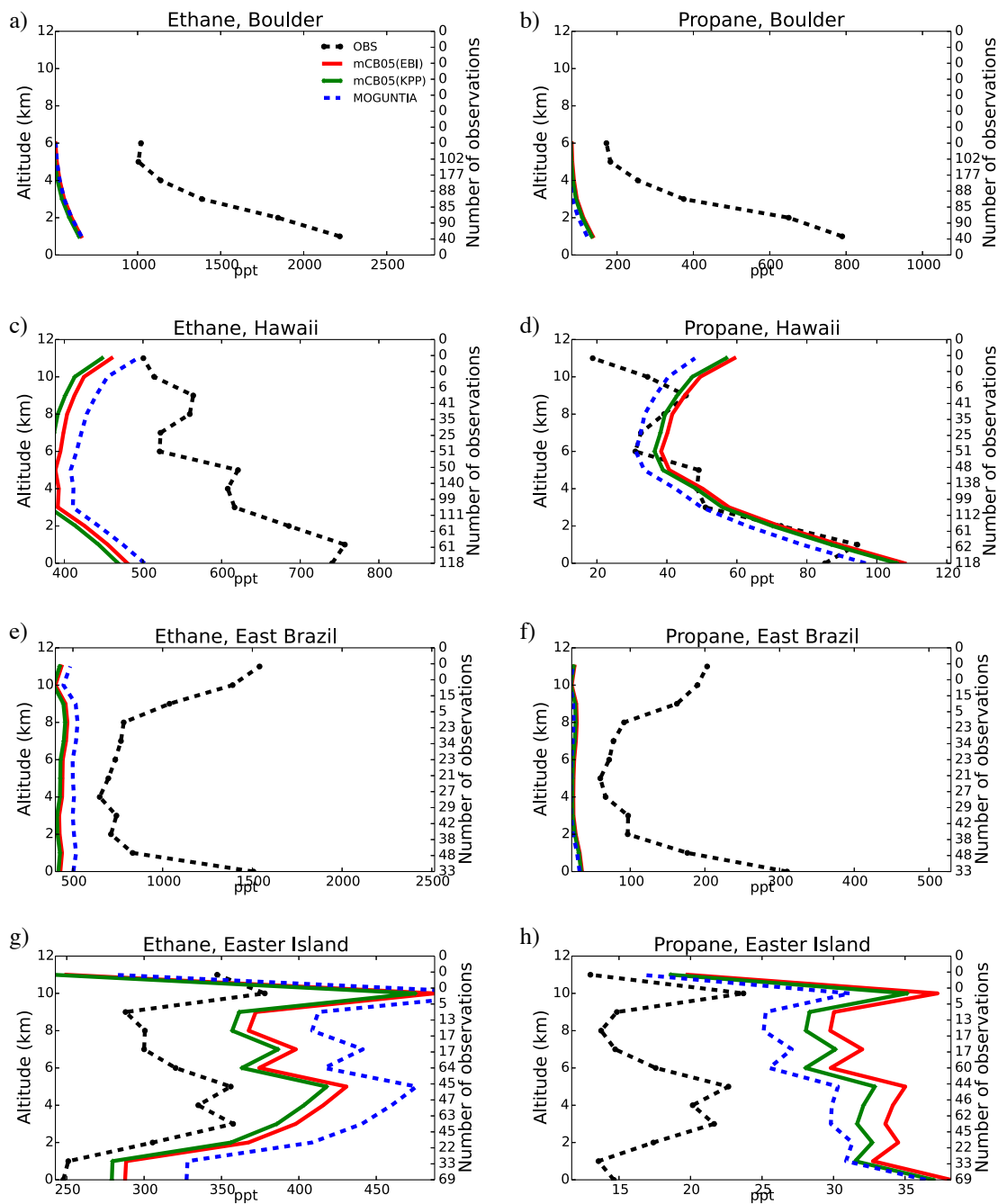
Figure 10: Comparison of TM5-MP vertical profiles (in km) of C₂H₆ (left column) and C₃H₈ (right column) against aircraft observations (black line) in ppt, for the two chemistry schemes, mCB05(KPP) (green line) and MOGUNTIA (blue line), using co-located model output for 2006 sampled at the measurement times; error bars indicate the standard deviation. For comparison, model results of the mCB05 with the EBI solver (red line) are also presented. The numbers on the right vertical axis indicate the number of available measurements.
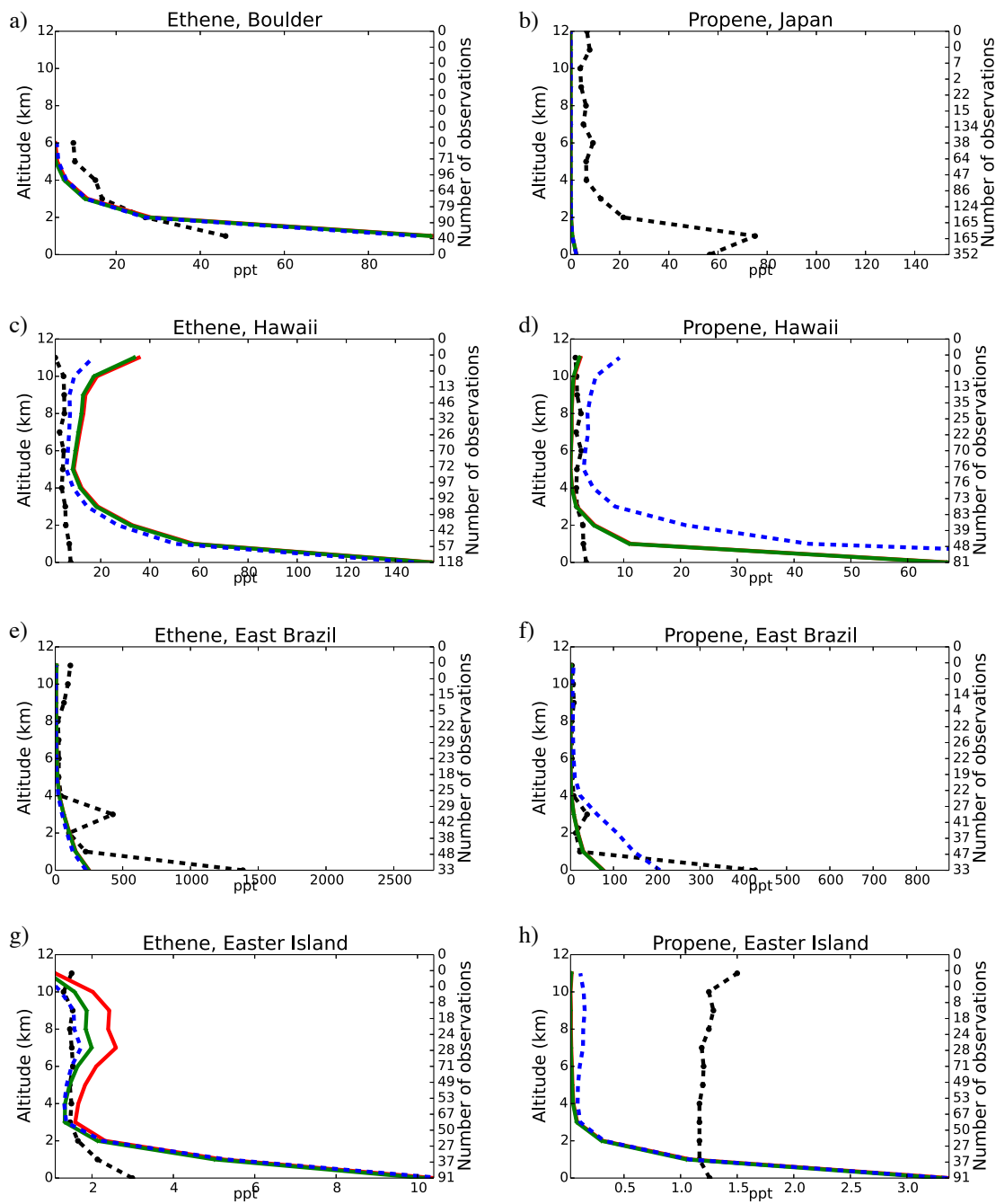
Figure 11: Comparison of TM5-MP vertical profiles (in km) of C_2H_4 (left column) and C_3H_6 (right column) against aircraft observations (black line) in ppt, for the two chemistry schemes, mCB05(KPP) (green line) and MOGUNTIA (blue line), using co-located model output for 2006 sampled at the measurement times; error bars indicate the standard deviation. For comparison, model results of the mCB05 with the EBI solver (red line) are also presented. The numbers on the right vertical axis indicate the number of available measurements.