

Can a “state of the art” chemistry transport model simulate Amazonian tropospheric chemistry?

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[1] We present an evaluation of a nested high-resolution Goddard Earth Observing System (GEOS)-Chem chemistry transport model simulation of tropospheric chemistry over tropical South America. The model has been constrained with two isoprene emission inventories: (1) the canopy-scale Model of Emissions of Gases and Aerosols from Nature (MEGAN) and (2) a leaf-scale algorithm coupled to the Lund-Potsdam-Jena General Ecosystem Simulator (LPJ-GUESS) dynamic vegetation model, and the model has been run using two different chemical mechanisms that contain alternative treatments of isoprene photo-oxidation. Large differences of up to 100 Tg C yr⁻¹ exist between the isoprene emissions predicted by each inventory, with MEGAN emissions generally higher. Based on our simulations we estimate that tropical South America (30–85°W, 14°N–25°S) contributes about 15–35% of total global isoprene emissions. We have quantified the model sensitivity to changes in isoprene emissions, chemistry, boundary layer mixing, and soil NO_x emissions using ground-based and airborne observations. We find GEOS-Chem has difficulty reproducing several observed chemical species; typically hydroxyl concentrations are underestimated, whilst mixing ratios of isoprene and its oxidation products are overestimated. The magnitude of model formaldehyde (HCHO) columns are most sensitive to the choice of chemical mechanism and isoprene emission inventory. We find GEOS-Chem exhibits a significant positive bias (10–100%) when compared with HCHO columns from the Scanning Imaging Absorption Spectrometer for Atmospheric Chartography (SCIAMACHY) and Ozone Monitoring Instrument (OMI) for the study year 2006. Simulations that use the more detailed chemical mechanism and/or lowest isoprene emissions provide the best agreement to the satellite data, since they result in lower-HCHO columns.

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1. Introduction

[2] The Amazon Basin, covering nearly 6 million square kilometers and containing the world’s largest rainforest, is

the most productive and diverse ecosystem on Earth. One of the reasons why the Amazon rainforest is important to climate is because tropical vegetation emit a wide range of highly reactive nonmethane biogenic volatile organic compounds

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(BVOCs) into the atmosphere. These BVOCs play a critical role in global atmospheric chemistry and climate as their photochemical reactions influence the oxidation capacity of the atmosphere [Poisson *et al.*, 2000; Monson and Holland, 2001], and the lifetimes and distributions of other key trace gases, such as carbon monoxide (CO) and methane (CH₄) [Granier *et al.*, 2000; Collins *et al.*, 2002]. The most important BVOC emitted by vegetation is isoprene since it accounts for about half of the total global BVOC budget [Guenther, 2002], and because of its influence on tropospheric ozone [Jenkin and Clemitshaw, 2000; Sanderson *et al.*, 2003; Fiore *et al.*, 2005] and its precursor role in the formation of secondary organic aerosol [Claeys *et al.*, 2004; Kanakidou *et al.*, 2005]. Isoprene emissions are also relevant to carbon cycle studies as they represent a loss of fixed carbon from the terrestrial biosphere [Kesselmeier *et al.*, 2002] and a nonnegligible photochemical source of carbon dioxide [Folberth *et al.*, 2005].

[3] Despite the Amazon Basin being acknowledged as a significant isoprene source [Guenther *et al.*, 2006; Arneth *et al.*, 2008], there have only been a few measurement campaigns to date that have targeted this region (owing in part to inaccessibility), resulting in a limited number of in situ measurements. Leaf and branch level measurements [Kuhn *et al.*, 2002] offer insight to small-scale processes, whilst tower, balloon and airborne platforms provide information on localized emissions [e.g., Helmig *et al.*, 1998; Karl *et al.*, 2007; Kuhn *et al.*, 2007]. However, for the Amazon Basin as a whole, isoprene emissions are poorly quantified. Bottom-up emissions inventories, such as the widely used Model of Emissions of Gases and Aerosols from Nature (MEGAN) [Guenther *et al.*, 2006], are highly uncertain as they rely on upscaling sparse point measurements to landscape scales; where ecosystem diversity is high, as in tropical ecosystems, this is especially difficult. Moreover, the empirical algorithms that drive variations in the bottom-up emissions are mostly based on studies of extratropical plant species, which may not be applicable to tropical vegetation. Model studies conducted at relatively coarse spatial scales have generally been unable to accurately reproduce the sparse observations of Amazonian isoprene fluxes and surface concentrations without some form of adjustment or scaling of the bottom-up emissions [von Kuhlmann *et al.*, 2004; Jöckel *et al.*, 2006; Müller *et al.*, 2008; Barkley *et al.*, 2008].

[4] Formaldehyde (HCHO), a short-lived trace gas and high yield product of isoprene oxidation, can provide additional information on surface isoprene emissions on length scales of order 100 km [Palmer *et al.*, 2003]. However, since HCHO originates from a variety of atmospheric and surface sources, careful disaggregation is needed to accurately derive satellite-based (i.e., top-down) isoprene emission estimates. Globally, the largest source of HCHO is produced from the oxidation of methane [Stavrakou *et al.*, 2009a], though this only maintains ambient background concentrations, such as found in the remote marine atmosphere. Over land, BVOC oxidation over densely vegetated areas and anthropogenic VOC oxidation over urban areas create strong regional HCHO enhancements, easily observed from space. Biomass burning and wild fires are also significant localized HCHO sources, owing to directly released HCHO during incomplete combustion and from the oxidation of coemitted VOCs [Andreae and Merlet, 2001].

[5] During the last decade satellite observations of HCHO columns, retrieved using ultraviolet (UV) absorption spectroscopy [Chance *et al.*, 2000; De Smedt *et al.*, 2008], have been used in several studies to map top-down isoprene emissions on continental and global scales [see, e.g., Palmer *et al.*, 2003, 2006; Shim *et al.*, 2005; Fu *et al.*, 2007; Millet *et al.*, 2007; Barkley *et al.*, 2008; Stavrakou *et al.*, 2009b]. Recent work by Barkley *et al.* [2008], using HCHO column data from the Global Ozone Monitoring Experiment (GOME) [European Space Agency, 1995; Burrows *et al.*, 1999] and Scanning Imaging Absorption Spectrometer for Atmospheric Chartography (SCIAMACHY) [Bovensmann *et al.*, 1999] instruments, identified potentially large gaps in our quantitative understanding of Amazonian isoprene emissions, and unexplained observed seasonal variations [Barkley *et al.*, 2009]. In contrast, SCIAMACHY top-down estimates derived by Stavrakou *et al.* [2009b] gave better agreement with the MEGAN inventory for the Amazon region; in that study emissions were calculated using the Model for Hydrocarbon Emissions by the Canopy (MOHYCAN) [Müller *et al.*, 2008] forced with European Center for Medium-Range Weather Forecast (ECMWF) meteorological analyses. These differences, in part, reflect the difficulties and uncertainties associated with inferring isoprene emissions from satellite measurements of HCHO columns. Critically, the accuracy of the top-down emissions largely depends on the ability of the chosen chemistry transport model (CTM), the intermediary used to invert the retrieved HCHO columns, to accurately simulate the chemistry and dynamics of the rainforest atmosphere. Model spatial resolution is important in this respect, both to capture localized dynamical and chemical processes, and distinguish biogenic and pyrogenic contributions to the observed HCHO columns [Barkley *et al.*, 2008].

[6] However, modeling Amazonian tropospheric chemistry is difficult [Ganzeveld *et al.*, 2002; von Kuhlmann *et al.*, 2004; Butler *et al.*, 2008]. Besides the large uncertainties associated with the bottom-up BVOC emissions (typically >100%), the subsequent oxidation chemistry occurring in the prevalent low-NO_x conditions is poorly understood and often inadequately parameterized. In particular, there has been much emphasis placed on the underestimation of hydroxyl (OH) concentrations over dense tropical rainforests by most CTMs and chemistry-climate models [e.g., Butler *et al.*, 2008]. Various plausible mechanisms to recycle or regenerate OH through improved isoprene degradation schemes have been proposed and compared with observational data [Butler *et al.*, 2008; Lelieveld *et al.*, 2008; Paulot *et al.*, 2009a; Stavrakou *et al.*, 2010]. Given the large uncertainties of these chemical mechanisms much remains unresolved. Although significant effort has gone into improving our understanding of tropical oxidation chemistry, the influence of physical and micrometeorological processes on reactive gas exchange can be equally, if not more, important [Pike *et al.*, 2010; Pugh *et al.*, 2010]. Correct parameterizations of both in-canopy (e.g., soil NO_x emissions, deposition) and above-canopy (e.g., convection, turbulent mixing, segregation effects) processes are essential to properly model the tropical atmosphere [Ganzeveld and Lelieveld, 2004; Karl *et al.*, 2004; Ganzeveld *et al.*, 2008; Pugh *et al.*, 2010].

[7] In this manuscript we present the first comprehensive evaluation of a high-resolution simulation of tropospheric chemistry over tropical South America, performed by a

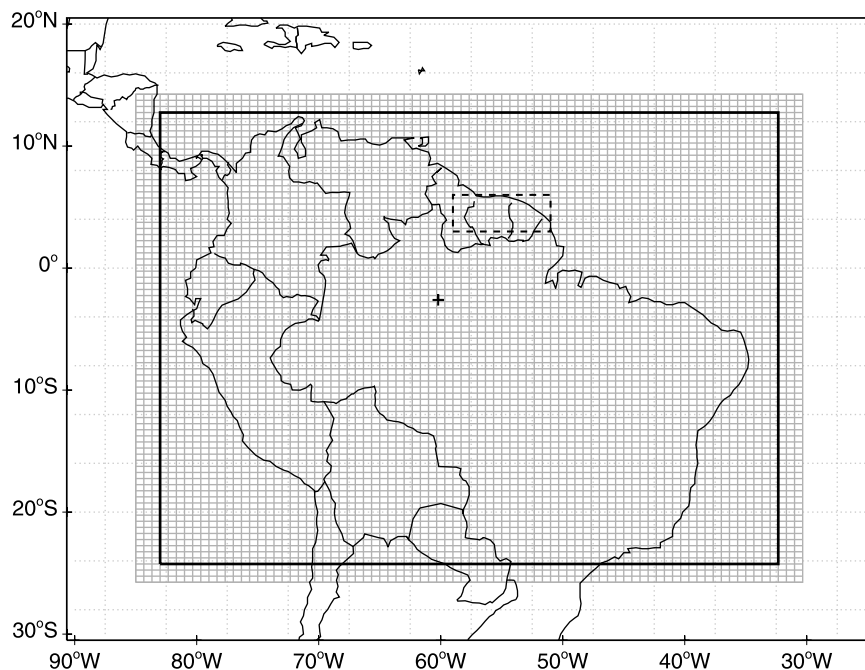


Figure 1. Schematic showing the domain of the Amazon nested grid. The thick black line is the actual boundary of the nested window; $0.5^\circ \times 0.667^\circ$ grid cells outside the black line represents the buffer zone for the boundary conditions. The GEOS-Chem $4^\circ \times 5^\circ$ horizontal grid is shown by the dotted light grey lines. The location of the TROFFEE campaign [Karl *et al.*, 2007] is shown as the black cross (see section 4.1). The domain of the GABRIEL campaign [Stickler *et al.*, 2007] is shown inset by the black dashed line (see section 4.2).

nested grid version of the widely used Goddard Earth Observing System (GEOS)-Chem CTM [Bey *et al.*, 2001]. Given that in future work we want to use GEOS-Chem to derive top-down isoprene emissions for this region, and knowing the large uncertainties associated with simulating Amazonian BVOC emissions and tropospheric chemistry, the objective of this paper is two fold. First, we assess the ability of GEOS-Chem to accurately simulate isoprene oxidation chemistry and other observed key atmospheric constituents over the Amazon rainforest. Secondly, we evaluate the model's usefulness to interpret satellite observations of HCHO, or in other words, its suitability for inferring top-down isoprene emission estimates. To achieve these objectives we limit our focus on the model sensitivity to four key processes: surface BVOC emissions, chemistry, boundary layer mixing, and soil NO_x emissions. To determine the relative importance of each process we use surface and airborne observations from previous Amazon field campaigns, along with satellite observations of HCHO vertical columns, to validate the model output. This work is novel in that for the first time GEOS-Chem will be forced with two contrasting BVOC emission inventories to assess their influence on Amazonian tropospheric chemistry.

[8] This manuscript is structured as follows. In section 2 we provide an overview of the GEOS-Chem model and outline the chemical mechanism. We discuss and compare the two BVOC emission inventories separately in section 3. In section 4 we evaluate the model using the surface/airborne field measurements and the satellite HCHO column data; details of the satellite HCHO retrievals are provided in

the auxiliary material.¹ We discuss the potential implications resulting from the model validation in section 5, and conclude the paper in section 6.

2. GEOS-Chem

2.1. Overview

[9] GEOS-Chem is a global 3-D chemistry transport model [Bey *et al.*, 2001] which we use here to simulate tropospheric chemistry over tropical South America. To reproduce the trace gas distributions over Amazon rainforest at relatively fine spatial scales we run GEOS-Chem (v8-03-01) in a one-way high-resolution nested grid mode. This nested grid capability of GEOS-Chem was first developed to study the east Asia region by Wang *et al.* [2004], and was more recently updated by Chen *et al.* [2009]. Here we adapt the model to be centered over the Amazon Basin, as shown in Figure 1. The model has a horizontal resolution of $0.667^\circ \times 0.5^\circ$ (longitude \times latitude) which is consistent with the forcing meteorology taken from the Goddard Earth Observing System (GEOS-5) of the NASA Global Modeling and Assimilation Office (GMAO) [Rienecker *et al.*, 2008]. The GEOS-5 meteorological data are updated every 6 h for 3-D variables and every 3 h for surface fields and mixing depths. In the vertical coordinate, we run the model with 47 hybrid eta levels extending from the surface to 0.01 hPa, with the boundary layer up to 2 km resolved by 14 layers (with

¹Auxiliary materials are available in the HTML. doi:10.1029/2011JD015893.

Table 1. GEOS-Chem Anthropogenic, Biofuels, and Selected Biomass Burning Emissions From the Amazon for 2006^a

Species ^b	Units	Anthropogenic	Biofuels	Biomass Burning
CO	Tg	15.70	15.53	50.85
NO _x	Tg N	1.01	0.16	0.49
Acetone	Tg C	0.04	0.03	0.22
Acetaldehyde	Tg C	–	0.06	0.22
ALK4	Tg C	1.52	0.05	0.09
C ₂ H ₆	Tg C	0.18	0.14	0.41
C ₃ H ₆	Tg C	0.74	0.39	0.49
C ₃ H ₈	Tg C	0.28	0.06	0.34
HCHO	Tg	–	0.13	0.52
MEK	Tg C	0.05	0.11	0.15
SO ₂	Tg S	–	–	0.27
NH ₃	Tg NH ₃	–	–	0.68
BC	Tg	–	–	0.32
OC	Tg	–	–	2.53

^aNO_x emissions from soils, fertilizers, and lightning are 0.91, 0.03, and 0.83 Tg N, respectively.

^bSpecies definition as follows: ALK4, C_{4,5} alkanes; C₂H₆, ethane; C₃H₈, propane; MEK, >C₃ ketones; C₃H₆, propene; BC, black carbon; OC, organic carbon.

midpoints at approximately 70, 200, 330, 470, 600, 740, 880, 1000, 1160, 1300, 1440 m altitude for a column based at sea level).

[10] Tracers are transported using a semi-Lagrangian approach for advection [Lin and Rood, 1996] and a relaxed Arakawa-Schubert (RAS) scheme for moist convection [Moorthi and Suarez, 1992]. Two options are available for the vertical mixing of chemical tracers within the planetary boundary layer (PBL): (1) a full-mixing scheme which ensures emissions and mixing ratios are constant with altitude at each chemistry time step (60 mins) and (2) a nonlocal scheme that can include “local” mixing between adjacent layers and, depending on the stability of the PBL, “nonlocal” mixing due to turbulent eddies. The nonlocal scheme is based on the formulation of Holtslag and Boville [1993] and was implemented into GEOS-Chem by Lin and McElroy [2010]; in their study it yielded a more realistic simulation of NO₂ and O₃ at midlatitudes.

[11] To provide appropriate boundary conditions to the nested Amazon grid, the GEOS-5 meteorological data is degraded to a 4° × 5° horizontal resolution and a global simulation performed, with the tracer mixing ratios saved every 3 h (consistent with the temporal resolution of the surface meteorology). The archived tracer mixing ratios in coarse grid cells adjacent to the nested domain, are then supplied to a delineated buffer zone (of three 0.667° × 0.5° grid cells width) in an area-weighting, grid-filled procedure (as described by Wang et al. [2004]) to give the dynamic boundary constraints.

2.2. Chemical Mechanism

[12] The standard GEOS-Chem chemical mechanism provides a relatively detailed treatment of coupled O₃-NO_x-VOC and aerosol chemistry [Horowitz et al., 1998; Bey et al., 2001; Fiore et al., 2002; Park et al., 2004] integrated using the Kinetic Preprocessor (KPP) solver of Sandu and Sander [2006]. Emissions from anthropogenic, biogenic and pyrogenic sources are provided, and the explicit photo-oxidation schemes of methane, ethane, propane, lumped >C₃ alkanes,

lumped >C₂ alkenes and isoprene are considered, with organic peroxides recycled [see Palmer et al., 2003; Fiore et al., 2005]. Details of non-BVOC emissions are described separately in the auxiliary material; their annual emissions for 2006 are summarized in Table 1. Photolysis rates are calculated using the Fast-J algorithm of Wild et al. [2000], which takes into account Rayleigh scattering as well as Mie scattering by aerosols and clouds. Dry deposition of aerosols and gases are based on a standard resistance-in-series model [Wesely, 1989] as described by Wang et al. [1998]. We update the model to include the deposition of six extra species (isoprene, methyl vinyl ketone (MVK), methacrolein (MACR), acetone, acetaldehyde and methyl hydroperoxide) based on their Henry’s Law constant and a reactivity factor, using values consistent with the study of von Kuhlmann et al. [2004]; Table 2 summarizes the dry deposition losses for 2006. Wet deposition (rainout and washout) of aerosols and gases are described by Liu et al. [2001] and Mari et al. [2000], respectively. In addition to the standard GEOS-Chem chemical scheme, an alternative mechanism (hereafter referred to as the Caltech mechanism) has been devised following the detailed work of Paulot et al. [2009a, 2009b]. This scheme includes a more explicit treatment of the production of organic nitrates, acids and epoxides from the photo-oxidation of isoprene, and contains an extra 13 transported species and approximately 50 more photochemical reactions. The formation of epoxides is of particular relevance to the Amazon region, given that extra OH is regenerated in low-NO_x conditions [Paulot et al., 2009a].

[13] We include a sensitivity test, discussed in section 4, to quantify the effect of artificial OH recycling as proposed by Lelieveld et al. [2008], by modifying within each scheme the reaction of first generation isoprene peroxy (ISOPO₂) and hydroperoxyl (HO₂) radicals, in the formation of isoprene hydroxy peroxides (ISOPOOH) as follows:



In this study we set $n = 2$ given the uncertainty reported in recent literature [Butler et al., 2008; Kubistin et al., 2010; Pugh et al., 2010; Stavrou et al., 2010]. Consistent with these previous studies, we also reduce the rate coefficient of the isoprene and OH reaction (here by 10%), owing to possible segregation effects due to incomplete mixing [Krol et al., 2000; Karl et al., 2007], acknowledging this rate reduction is highly uncertain and likely to vary considerably within the real atmosphere.

[14] Owing to the recent updates made to the chemical mechanism [e.g., Millet et al., 2010; Paulot et al., 2009a, 2009b], we reassess the GEOS-Chem HCHO yields from isoprene oxidation using the Master Chemical Mechanism (MCM) [Jenkin et al., 2003; Saunders et al., 2003] as our reference chemistry (see auxiliary material). The time-dependent HCHO yields from the oxidation of a pulse release of 1 ppbv of isoprene are given in Table 3. They show that under high-NO_x conditions (≥1.0 ppbv) the new GEOS-Chem yields are similar to those previously published by Palmer et al. [2003, 2006]. The short-term HCHO yield (i.e., that achieved by the end of the first day) calculated by GEOS-Chem’s standard and Caltech mechanisms are within 20% of the MCM value and comprise 98% and 91% of their ultimate yields, respectively.

Table 2. GEOS-Chem Dry Deposition Losses Over the Amazon Domain for 2006 Based on the S_{FM} Scenario^a

Species ^b	Amount ^c (Tg)	v_d^d (cm s^{-1})	Lifetime ^e (days)
NO ₂	0.43	0.16	27
O ₃	80.06	0.25	97
PAN	0.61	0.15	71
Isoprene	0.06	0.01	1054
HNO ₃	4.01	1.52	10
H ₂ O ₂	14.32	0.69	9
Acetone	1.08	0.17	103
Acetaldehyde	0.30	0.51	24
MVK	3.62	0.16	9
MACR	3.32	0.16	9
PMN	0.23	0.15	11
PPN	0.02	0.15	151
R ₄ N ₂	0.04	0.15	53
HCHO	6.10	0.51	9
N ₂ O ₅	0.02	1.52	39
MP	6.10	0.24	19

^aSee Table 5.^bSpecies definition as follows: MVK, methyl vinyl ketone; MACR, methacrolein; MP, methyl hydroperoxide; PMN, peroxy methacryloyl nitrate; PPN, peroxypropionyl nitrate; R₄N₂, C_{4,5} alkyl nitrates.^cNote that losses of isoprene, acetone, and acetaldehyde are in Tg C.^dAverage deposition velocity over land and ocean.^eAverage tropospheric lifetime against dry deposition.

[15] In low-NO_x conditions (≤ 0.1 ppbv), more relevant to our study domain, the short-term HCHO yields computed by the standard and Caltech mechanisms are within about 10% of the MCM yield and by the end of the day have reached 84% and 78% of their ultimate yield, respectively. This is a significant improvement on the studies of *Palmer et al.* [2003, 2006] since previously the GEOS-Chem yield was only within about 40% of the MCM after 1 day.

[16] Although the GEOS-Chem HCHO yields are broadly consistent with the MCM, the reader should be aware these model values only serve as an approximate guide, since they reflect significant uncertainties in BVOC oxidation chemistry. Furthermore, within the real atmosphere the yield depends on its local environment and is therefore likely to vary considerably [*Palmer et al.*, 2003].

3. BVOC Emissions

[17] Since BVOC emissions from terrestrial vegetation are the primary driver of tropical oxidation chemistry, it is essential to model their emissions as accurately as possible. Most CTMs employ only a single BVOC emission inventory, typically one of the *Guenther et al.* [1995, 1999, 2006] algorithms, owing to their ease of use. In this work we go a step further by implementing two different isoprene emission inventories into GEOS-Chem, to quantify their impact on the subsequent HCHO column distributions. The first inventory we use is MEGAN [*Guenther et al.*, 2006], which is canopy-scale model. The second inventory is a leaf-scale emission algorithm developed by *Arneth et al.* [2007a, 2010] which is coupled to the Lund-Potsdam-Jena General Ecosystem Simulator (LPJ-GUESS) dynamic vegetation model [*Smith et al.*, 2001]. The main difference between these models is that MEGAN is based on extrapolating available leaf/branch enclosure measurements and ecosystem observations (using a canopy model), to derive areal basal canopy emission

factors. Note that for the Amazon region the emission factors are solely based on above-canopy measurements. These standardized emission capacities are coupled to static vegetation maps, with emission variability modeled using empirical algorithms forced by external meteorology and seasonal estimates of leaf area. Alternatively, within LPJ-GUESS analogues to leaf level emission capacities are assigned per plant functional type (PFT), and short-term emission variability in response to temperature and light modeled using semi-mechanistic algorithms that link emissions to their chloroplastic production. The scaling to canopy, seasonal changes in leaf area index, and PFT distributions are explicitly calculated permitting ecosystem emissions to be estimated. In sections 3.1 and 3.2, we outline the details of each model and then provide a comparison of their respective emissions for the Amazon region in section 3.3.

3.1. MEGAN Emissions

[18] The latest MEGAN release (version 2.1) [*Guenther and Wiedinmyer*, 2007; *Sakulyanontvittaya et al.*, 2008] can calculate emission rates for 138 chemical species using 20 explicit and lumped chemical classifications. Here we focus only on the (explicit) species: isoprene, methylbutenol (MBO), and several monoterpene compounds (α pinene, β pinene, limonene, myrcene, sabinene, 3-carene, and ocimene). The emissions, E , of these compounds are parameterized by

$$E = E_0 \times \gamma_{CE} \times \gamma_{Age} \times \gamma_{SM} \times \rho, \quad (1)$$

where E_0 are the basal emissions (in μg of compound $\text{m}^{-2} \text{h}^{-1}$) normalized to standard conditions (current air temperature = 303 K, photosynthetic active radiation (PAR) = 1500 $\mu\text{mol m}^{-2} \text{s}^{-1}$, leaf area index = 5), which are multiplied by emission activity factors that simulate changes in the emission rate owing to the changes in the canopy environment γ_{CE} , leaf age γ_{Age} , and soil moisture γ_{SM} . In this study we neglect the effect of soil moisture ($\gamma_{SM} = 1$). Similarly, we assume standard conditions for any extra production or loss of the BVOC within the vegetation canopy by setting $\rho = 1$. For isoprene, this implies typical canopy losses of about 4% [*Guenther et al.*, 2006].

[19] In GEOS-Chem two different versions of MEGAN can be employed by switching on/off the relevant flags in the model control file. There are subtle but important differences between these model versions which are essentially based on the calculation of the canopy term γ_{CE} . The two models are as follows:

[20] 1. The PCEEA model is a simplified parameterized canopy environment emission activity (PCEEA) algorithm that is described in detail by *Guenther et al.* [2006] and *Guenther and Wiedinmyer* [2007]. Here γ_{CE} is calculated by

$$\gamma_{CE} = \gamma_T \times \gamma_{PAR} \times \gamma_{LAI}, \quad (2)$$

where γ_T , γ_{PAR} and γ_{LAI} are activity factors to account for the effect of current and past variations in temperature, light and leaf area on the emissions for the ‘whole’ canopy environment.

[21] 2. The hybrid algorithm model uses a combination of the new parameterizations of *Guenther et al.* [2006] and *Guenther and Wiedinmyer* [2007], and some older *Guenther et al.* [1995, 1999] algorithms. This hybrid model uses an

explicit canopy model to calculate variations in light and leaf area at five sublayers (denoted by l) within the canopy. Here γ_{CE} is calculated by

$$\gamma_{CE} = \gamma_T \times \sum_{l=1,5} \gamma_{PAR^l} \times LAI_l, \quad (3)$$

where LAI_l is the cumulative leaf area index at layer l , and γ_{PAR^l} is based on *Guenther et al.* [1999]. Note that (1) γ_T calculated here is equivalent to the γ_T in the PCEEA algorithm (equation (2)) and (2) in both approaches we also take into account the light dependency of monoterpenes based on the work of *Sakulyanontvittaya et al.* [2008].

[22] In standard GEOS-Chem simulations, the basal MEGAN emission factors are regridded from a default $0.5^\circ \times 0.5^\circ$ grid to the GEOS-Chem horizontal resolutions and driven by 3 hourly surface air temperatures (at 2 m height), and by diffuse and direct PAR from the GEOS-5 assimilation system. To simulate changes in vegetation we use gridded MODIS observations of monthly mean leaf area index (LAI) made over 2000–2008 (version C5; default resolution: $0.25^\circ \times 0.25^\circ$) [*Myneni et al.*, 2007]. The average leaf area index for vegetated areas within each grid cell, LAI_v , is estimated by dividing the LAI by the fraction of the cell covered by vegetation following the approach of *Guenther et al.* [2006].

3.2. LPJ-GUESS Emissions

[23] LPJ-GUESS is a dynamic global vegetation modeling framework [*Arneeth et al.*, 2010] that combines process descriptions for plant and soil carbon and water cycling of LPJ [*Sitch et al.*, 2003] with explicit formulation of vegetation resource competition for light and water and successional dynamics [*Smith et al.*, 2001]. For global simulations, the model represents vegetation by 10 plant functional types. Among these, tropical forests and woodlands are simulated via a dynamically changing mix of the tropical broadleaf rain green and tropical broadleaf evergreen plant functional types, as well as understorey vegetation that can be either of the C3 or C4 photosynthetic pathway.

[24] Emissions of isoprene and monoterpenes are calculated following *Arneeth et al.* [2007a] and *Schurgers et al.* [2009a]. Leaf production of these terpenoids is linked to photosynthetic electron flow, reflecting their metabolic pathway [*Niinemets et al.*, 1999]. For monoterpene emissions from broadleaf and herbaceous plant functional types, LPJ-GUESS assumes a production-driven emission pattern, with little or no contribution from leaf storage pools [*Kuhn et al.*, 2004; *Bäck et al.*, 2005; *Schurgers et al.*, 2009b]. The short-term variation of BVOC emissions is thus driven by variation in temperature and light, relatively similar to that described by more empirical algorithms such as MEGAN [*Arneeth et al.*, 2007b], whereas the overall emission totals and seasonal patterns are also greatly influenced by variation in leaf area index and overall vegetation productivity, and PFT composition. As LPJ-GUESS operates on a daily time step, estimates of sub-daily variation were created by applying empirical temperature and light algorithms [*Guenther et al.*, 1995] operating at 3 hourly time step to the emission daily totals.

[25] Since dynamic vegetation models need a spin-up period, the meteorological drivers light, precipitation and air

temperature from the Climate Research Unit of the University of East Anglia (CRU, <http://www.cru.uea.ac.uk/>) were corrected by equivalent variables from the GEOS-5 forecast [*Rienecker et al.*, 2008] to match the chemistry simulations. First, the grid points in the GEOS-5 data set, at a longitude spatial resolution of 0.667° , were matched with the corresponding closest grid point of the 0.5° CRU data set. Corrections at each grid point were based on monthly average values for the two data sets for the period January 2005 to December 2006. Based on the observed differences between the two data sets, a least squares fitting procedure was adopted, applying a general sinus function. The GEOS-5 (years 2005–2009) and GEOS-5 adjusted CRU values (years 1901–2004) were then used to follow the standard simulation protocol of 500 year spin-up period with detrended data and constant atmospheric CO_2 concentration to compute equilibrium soil and vegetation carbon pools, followed by a 20th and early 21st century simulation of vegetation dynamics and BVOC emissions [*Sitch et al.*, 2003; *Arneeth et al.*, 2007a]. We call this scenario *LPJ(GC)*. In addition, to provide a reference point for the *LPJ(GC)* scenario, we also conducted a simulation in which vegetation dynamics and BVOC emissions are based on the default CRU meteorology alone, denoted *LPJ(CRU)*.

3.3. Comparison of MEGAN and LPJ-GUESS

[26] Figures 2 and 3, which show the monthly mean emissions maps for 2006 and the time series of the monthly totals over 2005–2008, respectively, illustrate there are substantial spatial and temporal differences in the BVOC emissions predicted by the MEGAN and LPJ-GUESS inventories. Given the large uncertainties in modeling BVOC emissions [*Arneeth et al.*, 2008], we have to assume that all these estimates are plausible. The annual totals given in Table 4, show that for isoprene, the estimates lie almost within a factor of 2 of one another, consistent with known uncertainties of tropical emissions [*Guenther et al.*, 2006]. Furthermore, the maximum monthly emissions, which range from 2.4 to $9.7 \text{ mg m}^{-2} \text{ h}^{-1}$ during 2005–2006, are also consistent with previously reported measurements [see, e.g., *Kuhn et al.*, 2007, and references therein]. The discrepancy between the monoterpene annual totals is more considerable; uncertainties in the basal emission rates and algorithm differences are the likely origin. For example, owing to the lack of storage, monoterpene emissions from broadleaf trees only occur during the day in the LPJ-GUESS model, whereas in MEGAN the emissions are continuous.

[27] Focusing on isoprene, we find the MEGAN hybrid and PCEEA algorithms (not unexpectedly) produce very similar emissions throughout the year, with very high emissions in the dry season (broadly August–November) owing to higher light levels and slightly warmer temperatures [*Barkley et al.*, 2008]. Typically, MEGAN predicts the highest emissions along the Brazilian border with Peru and Bolivia, owing to a large percentage of high emitting species (e.g., bamboo forest) [*Barkley et al.*, 2008]. Isoprene emissions from the MOHYCAN model [*Müller et al.*, 2008], which also uses MEGAN, generally have a similar spatial distribution (a consequence of using same basal emission factors) but much lower annual totals, most likely owing to different choice of meteorology and canopy model, and through the

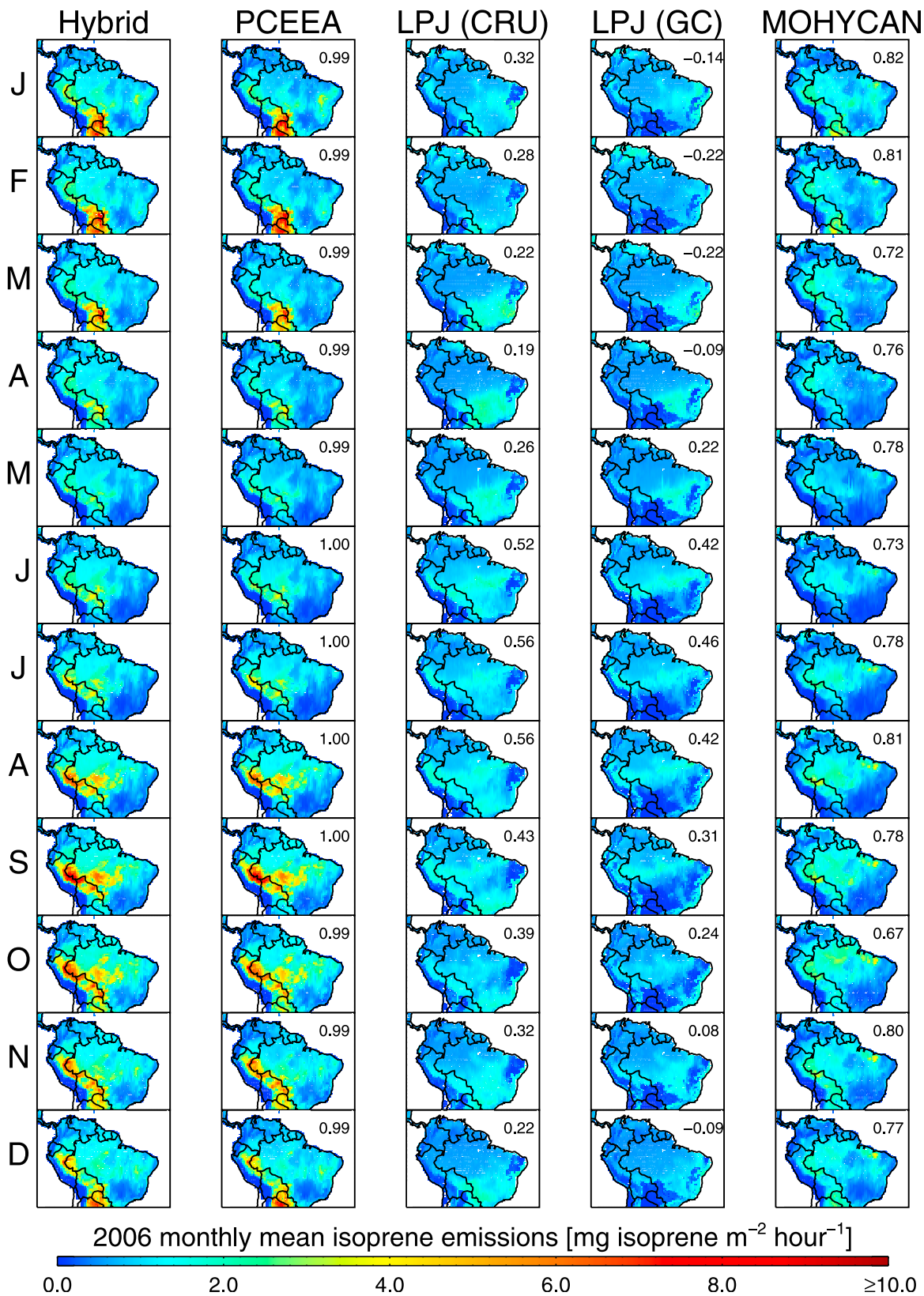


Figure 2. Monthly mean isoprene emissions for 2006 (in $\text{mg isoprene m}^{-2} \text{h}^{-1}$) calculated by the MEGAN hybrid and PCEEA algorithms and the LPJ-GUESS model forced by the default CRU meteorology, LPJ (CRU), and the GEOS-5 meteorology, LPJ(GC). For comparison, the monthly mean emissions from the MOHYCAN model [Müller *et al.*, 2008] remapped to the GEOS-Chem nested grid are also shown. The correlation of the emissions, relative to the MEGAN hybrid model, are shown inset.

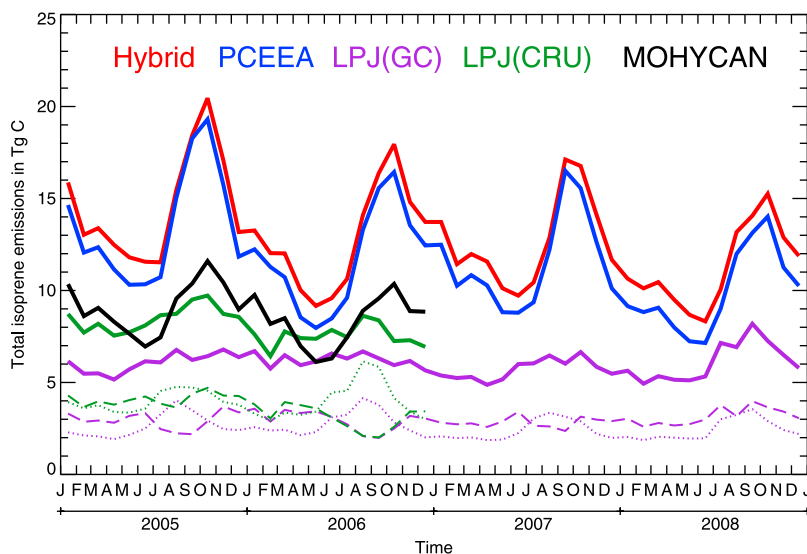


Figure 3. Time series of the total monthly isoprene emissions (in Tg C) calculated by the MEGAN hybrid (solid red line) and PCEEA (solid blue line) algorithms and the LPJ-GUESS model forced by the default CRU meteorology (solid green line) and the GEOS-5 meteorology (solid purple line). The purple and green dashed and dotted lines represent the LPJ emissions from tropical broadleaf evergreen and rain green trees, respectively. For comparison, the monthly totals calculated by the MOHYCAN model [Müller *et al.*, 2008] are also shown (solid black line).

inclusion of the soil moisture activity factor (which we set to unity in equation (1)). Therefore, as Figure 2 clearly illustrates, the way MEGAN is implemented can produce sizeable differences between emission estimates, even on a regional basis.

[28] The LPJ-GUESS model produces lower emissions everywhere compared with the hybrid/PCEEA estimates, though when forced with the CRU meteorology the model yields monthly and annual totals that agree well with MOHYCAN (Figure 3). In the LPJ-GUESS simulations, the largest sources of isoprene are from broadleaf evergreen and rain green trees; emissions from temperate tree species and grasses are much lower (about 2–5% of emission total). The highest emissions tend to occur toward southeastern Brazil, where emissions from broadleaf rain green trees dominate during November–June. Outside this time period emissions from broadleaf evergreens are typically higher than emission from rain green trees, owing to a decrease in the latter’s leaf area. Although on a continental scale the total emissions from either LPJ-GUESS simulation show little seasonal variation, indicating the Amazon is a constant emission source, a clear seasonal signal is evident for tropical broadleaf evergreens

with higher emissions in the dry season and fairly constant emissions otherwise.

4. Model Evaluation

[29] Our validation strategy is based on using a combination of surface, aircraft and satellite observations to assess model performance. Each of these observing platforms are representative of differing spatial and temporal scales and through combined use we can effectively relate the surface concentrations of isoprene and its oxidation products to the satellite retrieval of the HCHO vertical column. In this work we focus on the observations made from two campaigns: (1) the Tropical Forest and Fire Emissions Experiment (TROFFEE) [Yokelson *et al.*, 2007] undertaken in August–September 2004 and (2) the Guyanas Atmosphere–Biosphere Exchange and Radicals Intensive Experiment with the Learjet (GABRIEL) [Stickler *et al.*, 2007] performed in October 2005.

[30] In addition to these in situ observations we use satellite data from SCIAMACHY and the Ozone Monitoring Instrument (OMI) [Levelt *et al.*, 2006] to evaluate GEOS-Chem’s ability to model the HCHO column distributions, from which

Table 3. Time-Dependent HCHO Yields per Carbon From the Oxidation of Isoprene^a

Chemical Scheme	HCHO Yield in High NO _x Conditions				HCHO Yield in Low NO _x Conditions			
	Midday	Afternoon	Midnight	5 Days	Midday	Afternoon	Midnight	5 Days
MCM	0.45	0.48	0.49	0.50	0.22	0.30	0.35	0.47
GEOS-Chem (standard)	0.37	0.40	0.41	0.42	0.22	0.29	0.32	0.38
GEOS-Chem (Caltech)	0.33	0.38	0.39	0.43	0.23	0.29	0.31	0.40

^aHCHO yields are calculated using the Master Chemical Mechanism (MCM) [Jenkin *et al.*, 2003; Saunders *et al.*, 2003] and the GEOS-Chem standard and Caltech schemes (see section 2.2) and correspond to the oxidation of 1 ppbv of isoprene released at 07:00 within a tropical environment. High- and low-NO_x regimes are simulated by holding NO_x constant at 1 and 0.1 ppbv, respectively. The afternoon yield is computed at 14:00; the yield after 5 days is taken to be the ultimate yield [see, e.g., Palmer *et al.*, 2006].

Table 4. Total Annual Isoprene and Monoterpene Emissions From the Amazon Nested Grid for 2005–2008 as Calculated by the MEGAN Hybrid and PCEEA Algorithms and the LPJ-GUESS Model Using the GEOS-5 Meteorology and the Original CRU Meteorology^a

Algorithm	Annual Emissions (Tg C)							
	Isoprene				Monoterpenes			
	2005	2006	2007	2008	2005	2006	2007	2008
Hybrid	174	154	152	135	32	30	30	28
PCEEA	162	140	138	119	31	29	29	27
LPJ(GC)	73	75	69	73	5	5	5	5
LPJ(CRU)	102	90	–	–	7	6	–	–
MOHYCAN	102	100	–	–	–	–	–	–

^aLPJ(GC), LPJ-GUESS model using the GEOS-5 meteorology; LPJ (CRU), LPJ-GUESS model using the original CRU meteorology. For comparison the total isoprene emissions calculated by the MOHYCAN model [Miller *et al.*, 2008] are also included. Methylbutenol (MBO) emissions are negligible (<0.1 Tg C).

a transfer function is derived to infer the top-down isoprene emissions [Barkley *et al.*, 2008]. Details of these instruments and their respective retrieval algorithms are provided in the auxiliary material. The three most important distinctions between the sensors (relevant here) is the their coverage, overpass time, and ground pixel size. SCIAMACHY has a ground pixel size of $60 \times 30 \text{ km}^2$ and a 10:00 local overpass time, with global coverage achieved at the equator every 6 days. OMI has a pixel size which ranges $14 \times 26 \text{ km}^2$ to $28 \times 160 \text{ km}^2$ (depending on viewing geometry), a 13:30 local overpass time and can achieve global coverage in a single day. Since OMI has a higher spatial resolution and better temporal coverage than SCIAMACHY, it has a lower measurement uncertainty owing to improved sampling statistics and reduced cloud contamination. In this work we

restrict our model-satellite analysis to the year 2006 to save computational time; 2006 was chosen in preference to 2005, owing to the strong drought that occurred in 2005 [Zeng *et al.*, 2008].

[31] To determine the relative importance of model processes on the surface and column concentrations we performed a comprehensive set of sensitivity simulations with various key parameters changed or adjusted, as outlined in Table 5. Boundary conditions for each simulation were provided separately for the two GEOS-Chem chemical mechanisms, owing to the different number of reactive species and transported tracers within each scheme. Scenarios using the same chemical mechanism, but with different parameters changed (e.g., emissions) used the same boundary conditions, to ensure identical edge constraints and to save computational time. The spin-up period for each simulation, irrespective of the chemical scheme, originated from the same initialization time relevant to each study period. In this analysis we define our arbitrary baseline simulation, S_{FM} , to which other scenarios are compared, as that using the standard chemical mechanism, full boundary layer mixing and with isoprene emissions based on the MEGAN hybrid algorithm.

4.1. Comparison With TROFFEE Observations

[32] The ground-based component of the TROFFEE campaign took place approximately 60 km NNW of Manaus in Central Amazonia. Measurements were performed on an instrumented flux tower (Z14: 2.6°S, 60.2°W, 55 m height) between 14 and 29 September 2004. The tower was surrounded by a forest of canopy height 30 m and with an average leaf area index of about 5–6. A proton transfer reaction mass spectrometry (PTR-MS) instrument situated on the tower was used in conjunction with the eddy covariance technique to infer the surface fluxes of isoprene and monoterpenes, and concentrations of isoprene and the sum of its

Table 5. Description of the GEOS-Chem Sensitivity Simulations

Scenario	Chemical Mechanism	Isoprene Emissions		Boundary Layer Mixing Scheme	OH Recycling ^a	Soil NO _x Scaling ^b
		Model	Scaling			
S_{FM}	standard	MEGAN (hybrid)	1.000	full mixing	no	1.00
S_{NL}	standard	MEGAN (hybrid)	1.000	nonlocal	no	1.00
$S_{FM(NO_x)}$	standard	MEGAN (hybrid)	1.000	full mixing	no	1.25
$S_{FM(IE)}$	standard	MEGAN (hybrid)	0.635	full mixing	no	1.00
$S_{FM(PCEEA)}$	standard	MEGAN (PCEEA)	1.000	full mixing	no	1.00
$S_{FM(OH)}$	standard	MEGAN (hybrid)	1.000	full mixing	yes	1.00
$S_{NL(OH)}$	standard	MEGAN (hybrid)	1.000	nonlocal	yes	1.00
$S_{FM(LPJ(CRU))}$	standard	LPJ (CRU)	1.000	full mixing	no	1.00
$S_{FM(LPJ(GC))}$	standard	LPJ (GC)	1.000	full mixing	no	1.00
$S_{FM(IE, OH)}$	standard	MEGAN (hybrid)	0.635	full mixing	yes	1.00
C_{FM}	Caltech	MEGAN (hybrid)	1.000	full mixing	no	1.00
C_{NL}	Caltech	MEGAN (hybrid)	1.000	nonlocal	no	1.00
$C_{FM(NO_x)}$	Caltech	MEGAN (hybrid)	1.000	full mixing	no	1.25
$C_{FM(IE)}$	Caltech	MEGAN (hybrid)	0.635	full mixing	no	1.00
$C_{FM(PCEEA)}$	Caltech	MEGAN (PCEEA)	1.000	full mixing	no	1.00
$C_{FM(OH)}$	Caltech	MEGAN (hybrid)	1.000	full mixing	yes	1.00
$C_{NL(OH)}$	Caltech	MEGAN (hybrid)	1.000	nonlocal	yes	1.00
$C_{FM(LPJ(CRU))}$	Caltech	LPJ (CRU)	1.000	full mixing	no	1.00
$C_{FM(LPJ(GC))}$	Caltech	LPJ (GC)	1.000	full mixing	no	1.00
$C_{FM(IE, OH)}$	Caltech	MEGAN (hybrid)	0.635	full mixing	yes	1.00

^aSee reaction (R1).

^bScaling factor is only applied to soil NO emissions; fertilizer emissions are unscaled.

Table 6. Comparison of Modeled Surface Emissions and Mixing Ratios With Ground-Based Observations From the TROFEE Campaign^a

Scenario	Isoprene Emissions				Isoprene VMR				MVK+MACR VMR				Mean OH (10:00–14:00 LT) ($\times 10^6 \text{ cm}^{-3}$)	HCHO VCD (10:00 LT) ($\times 10^{16} \text{ cm}^{-2}$)
	Mean \pm 1 SD ($\text{mg m}^{-2} \text{ h}^{-1}$)	B (%)	PPA (%)	r	Mean \pm 1 SD (ppbv)	B (%)	PPA (%)	r	Mean \pm 1 SD (ppbv)	B (%)	PPA (%)	r		
Observations	2.12 ± 2.82	–	–	–	4.60 ± 2.73	–	–	–	2.43 ± 1.08	–	–	–	–	2.06^b
S_{FM}	2.80 ± 2.91	32	–31	0.81	7.86 ± 1.76	71	–19	0.14	4.24 ± 0.59	75	–14	0.14	0.48	2.57
S_{NL}	2.80 ± 2.91	32	–31	0.81	15.12 ± 3.89	228	68	0.13	4.44 ± 0.57	83	–10	0.32	0.37	2.57
$S_{FM(OH)}$	2.80 ± 2.91	32	–31	0.81	3.93 ± 1.29	–15	–55	–0.01	2.57 ± 0.32	6	–46	–0.06	1.70	2.53
$S_{NL(OH)}$	2.80 ± 2.91	32	–31	0.81	9.33 ± 2.98	103	16	0.11	2.98 ± 0.29	23	–36	0.10	1.43	2.59
$S_{FM(NOx)}$	2.80 ± 2.91	32	–31	0.81	7.59 ± 1.73	65	–21	0.15	4.23 ± 0.58	75	–14	0.13	0.51	2.59
$S_{FM(PCEE4)}$	2.22 ± 2.46	5	–45	0.81	5.95 ± 1.20	29	–38	0.24	3.71 ± 0.48	53	–23	0.12	0.56	2.45
$S_{FM(rIE)}$	1.78 ± 1.85	–16	–56	0.81	3.99 ± 1.03	–13	–57	0.13	2.49 ± 0.33	3	–49	0.09	0.73	1.83
$S_{FM(rIE, OH)}$	1.78 ± 1.85	–16	–56	0.81	2.25 ± 0.82	–51	–73	0.01	1.53 ± 0.21	–37	–68	–0.12	1.99	1.78
C_{FM}	2.80 ± 2.91	32	–31	0.81	5.97 ± 1.65	30	–32	0.13	3.28 ± 0.50	35	–35	0.03	0.74	2.14
C_{NL}	2.80 ± 2.91	32	–31	0.81	13.05 ± 3.87	184	56	0.14	3.63 ± 0.50	50	–29	0.13	0.50	2.16
$C_{FM(OH)}$	2.80 ± 2.91	32	–31	0.81	2.31 ± 1.00	–50	–70	0.06	1.65 ± 0.34	–32	–64	–0.07	3.59	2.12
$C_{NL(OH)}$	2.80 ± 2.91	32	–31	0.81	6.53 ± 2.49	42	–8	0.30	2.22 ± 0.44	–8	–52	–0.15	2.75	2.12
$C_{FM(NOx)}$	2.80 ± 2.91	32	–31	0.81	5.65 ± 1.62	23	–35	0.14	3.28 ± 0.49	35	–34	0.02	0.81	2.18
$C_{FM(PCEE4)}$	2.22 ± 2.46	5	–45	0.81	4.16 ± 1.14	–10	–51	0.22	2.75 ± 0.37	13	–43	0.00	0.96	2.06
$C_{FM(rIE)}$	1.78 ± 1.85	–16	–56	0.81	2.66 ± 0.94	–42	–67	0.09	1.83 ± 0.27	–25	–62	0.02	1.41	1.59
$C_{FM(rIE, OH)}$	1.78 ± 1.85	–16	–56	0.81	1.37 ± 0.65	–70	–82	0.07	1.03 ± 0.23	–56	–77	–0.09	4.07	1.59

^aKarl *et al.* [2007]. VMR is the volume mixing ratio, B is the normalized mean bias (equation (4)), PPA is the paired peak accuracy (equation (5)), r is the Pearson correlation coefficient, and VCD stands for vertical column densities.

^bThe mean HCHO vertical column density retrieved by SCIAMACHY [De Smedt *et al.*, 2008] averaged over 17–24 September 2004 calculated using only columns with cloud fractional cover $\leq 40\%$; only four observations matched this selection criteria.

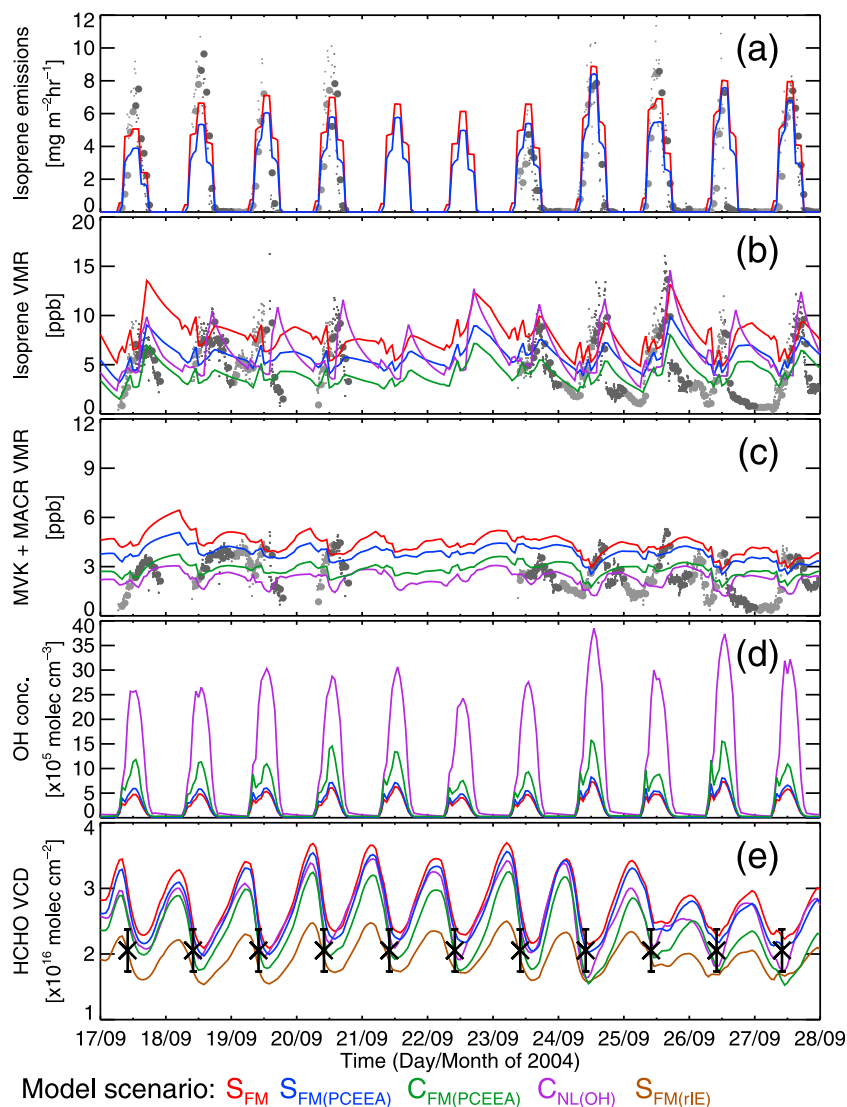


Figure 4. Comparison of model (a) isoprene emissions and surface mixing ratios of (b) isoprene and (c) MVK+MACR, with in situ tower measurements made in a primary forest reserve (tower Z14: 2.6°S, 60.2°W) as part of the TROFFEE campaign during 17–28 September 2004 [Karl *et al.*, 2007]. For clarity we only show the S_{FM} (red), $S_{FM(PCEEAA)}$ (blue), $C_{FM(PCEEAA)}$ (green), $C_{NL(OH)}$ (purple), and $S_{FM(rIE)}$ (brown) simulations (see Table 5). Light and dark grey dots represent morning and afternoon observations, respectively; large dots are hourly averages, and small dots are the raw data. We also show the model's (instantaneous) hourly (d) surface OH concentrations and (e) HCHO vertical column densities (VCD), along with the mean SCIAMACHY observation (retrieved by De Smedt *et al.* [2008]) averaged at this model grid cell (over 17–27 September), shown as the black cross with ± 1 standard deviation error bars.

oxidation products MVK and MACR. The instrumental setup and measurement procedure are discussed in detail by Karl *et al.* [2007]. For temporal consistency the observations were assembled into hourly means and compared to the hourly model output (adjusted to the local time) at the grid cell containing the tower location. All model runs were spun-up from 1 August 2004 (from a previously initialized 6 month simulation) to remove transient influences.

[33] Table 6 summarizes the main results of this comparison and Figure 4 shows the model and observed time series of surface isoprene fluxes, isoprene, MVK+MACR and OH

surface concentrations, and HCHO vertical columns. Here, and elsewhere, we use the normalized mean bias (B)

$$B = 100\% \times \frac{\sum(M_i - O_i)}{\sum(O_i)} \quad (4)$$

and the paired peak accuracy (PPA)

$$PPA = 100\% \times \frac{M(x', t') - O(x', t')}{O(x', t')}, \quad (5)$$

along with the Pearson correlation coefficient, r , to quantify the agreement between the modeled (M) and observed (O) quantities; x' and t' correspond to the location and time of the maximum observed value. The main results of this comparison can be summarized as follows:

4.1.1. Isoprene Emissions

[34] Figure 4a shows that at this location the MEGAN model reproduces the measured isoprene emissions fairly well (when taking into account the different spatial scales of the point measurements and the averaged grid cell emissions). Of the two MEGAN variants, the PCEEA model gives the best agreement to the observations, with a bias of less than 5%, although it does not capture the peak values as well ($PPA = -45\%$). In contrast, the hybrid model better reproduces the peak emissions ($PPA = -31\%$), but has a much larger bias overall ($B = 32\%$) as it tends to overestimate emissions during the early morning and latter part of the afternoon. The correlation is the same for both algorithms ($r = 0.81$). Scaling the hybrid emissions to those reported by Müller *et al.* [2008] reduces the bias by about half but increases the PPA. Since the LPJ-GUESS emissions were unavailable for the TROFFEE period, as an alternative, the hourly model emissions over 2005–2008 at this site were compared instead. We find that LPJ-GUESS estimates very rarely exceed the corresponding PCEEA or hybrid emissions (even when scaled to match the Müller *et al.* [2008] totals). A previous comparison by Arneeth *et al.* [2007a] of the LPJ-GUESS emissions to the TROFFEE data also found lower model isoprene emissions than observed.

4.1.2. Isoprene Surface Mixing Ratios

[35] Figure 4b shows that unlike the biogenic emissions, the isoprene mixing ratios are in general not reproduced well by GEOS-Chem, irrespective of the scenario employed. The observed isoprene mixing ratios demonstrate a pronounced diurnal cycle with midday peak values and a nocturnal minima. The absence of isoprene during the night indicates that most of the daytime emissions have been oxidized or deposited within the canopy. All scenarios fail to reproduce this diurnal pattern, as nighttime mixing ratios never approach zero, indicating that either (1) the model emissions are too high (clearly not the case), (2) the emitted isoprene has not been completely oxidized or has been poorly mixed out of the bottom model layer, or (3) we are missing a substantial nocturnal sink, e.g., surface deposition [Karl *et al.*, 2010]. The range of simulated isoprene mixing ratios is large (1–15 ppbv) suggesting that even the supposedly simple transition from emissions to surface mixing ratios can be seriously distorted by individual or any potential combination of model processes. Of the scenarios tested, the $C_{FM(PCEEA)}$ has the smallest bias ($B = -10\%$) but doesn't capture the peak isoprene mixing ratios ($PPA = -51\%$); the correlation between the model simulations and measurements is always poor ($r < 0.3$). In addition, as Figure 4b shows, the timing of the daytime peak emissions is not well captured, regardless of the choice of chemical or boundary layer mixing scheme.

4.1.3. MVK+MACR Surface Mixing Ratios

[36] The measured MVK+MACR mixing ratios also demonstrate a clear diurnal cycle which GEOS-Chem fails to reproduce (Figure 4c). Although the spread of the simulated mean mixing ratio is smaller than found for isoprene, the correlations are worse and even negative in some cases. Typically the modeled MVK+MACR mixing ratios are

overestimated during the night and underestimated throughout the day, and show very little diel variation.

4.1.4. Mean Surface OH Concentrations

[37] The mean OH concentrations between 10:00 and 14:00 local time (LT) differ greatly, as shown in Figure 4d, and are primarily dependent on the chemical mechanism used and the inclusion of artificial OH recycling [Lelieveld *et al.*, 2008]. Although the Caltech chemical mechanism generates slightly more OH than the standard GEOS-Chem chemistry scheme, it is only through the OH recycling, via reaction R1, that the model really stimulates high OH levels. For example, the mean OH concentrations for the S_{FM} and C_{FM} simulations are 0.48 and 0.74×10^6 molecules cm^{-3} , respectively, but when OH recycling is included as in the $S_{FM(OH)}$ and $C_{FM(OH)}$ scenarios, levels reach 1.70 and 3.59×10^6 molecules cm^{-3} . Despite this extra OH production, the model values are still relatively low compared with the observed range of values reported in the literature, typically $3\text{--}8 \times 10^6$ molecules cm^{-3} [see, e.g., Kuhn *et al.*, 2007; Lelieveld *et al.*, 2008]. Hydroxyl concentrations can also be enhanced by a factor of 1.5–1.9 through simply reducing the isoprene emissions, as in the $S_{FM(rIE)}$ and $C_{FM(rIE)}$ simulations, owing to isoprene's significant OH reactivity.

4.1.5. HCHO Vertical Column

[38] Figure 4e reveals the model HCHO columns show a diurnal variation in response to both the surface emissions and oxidation chemistry, and to also the variation in the PBL depth. The HCHO columns have a maximum during the night when the PBL height is at its shallowest and a minimum in the day when the PBL has grown to its maximum extent. The choice of chemical scheme and baseline isoprene emissions have the greatest influence on the magnitude of the model HCHO columns. The standard chemical mechanism always produces larger HCHO columns than both the Caltech scheme and the mean SCIAMACHY HCHO column retrieved at this location (see Table 6). The Caltech mechanism yields HCHO columns much closer in agreement to the SCIAMACHY HCHO column, and in the case of the $C_{FM(PCEEA)}$ scenario produces an exact match at the time of the satellite overpass. Down scaling the hybrid isoprene emissions to the Müller *et al.* [2008] estimates results in lower-HCHO columns at the time of the SCIAMACHY overpass, but which fall closer to the ± 1 standard deviation limits of the observed SCIAMACHY column.

4.1.6. Discussion of TROFFEE Analysis

[39] To better understand the processes leading to the modeled diurnal cycle of HCHO columns, we show in Figure 5 time versus altitude plots for the model output in the lowermost 3 km of the atmosphere for the days 18–20 September. In the early morning when sunlight initiates the emissions, isoprene is continually emitted and mixed throughout the PBL and isoprene oxidation by OH dominates production of HCHO. Isoprene mixing ratios are usually quite low at midday when OH production is a maximum. However, in the late afternoon and early evening when the PBL decreases in height, isoprene and its oxidation products that are not destroyed earlier in the day are lofted above the falling boundary layer (which at night dips beneath the depth of the first model layer adjacent to the surface). Here they can accumulate at night and be further oxidized by ozone and the nitrate radical. In particular, there is a very large buildup of MVK, MACR, and HCHO during the night

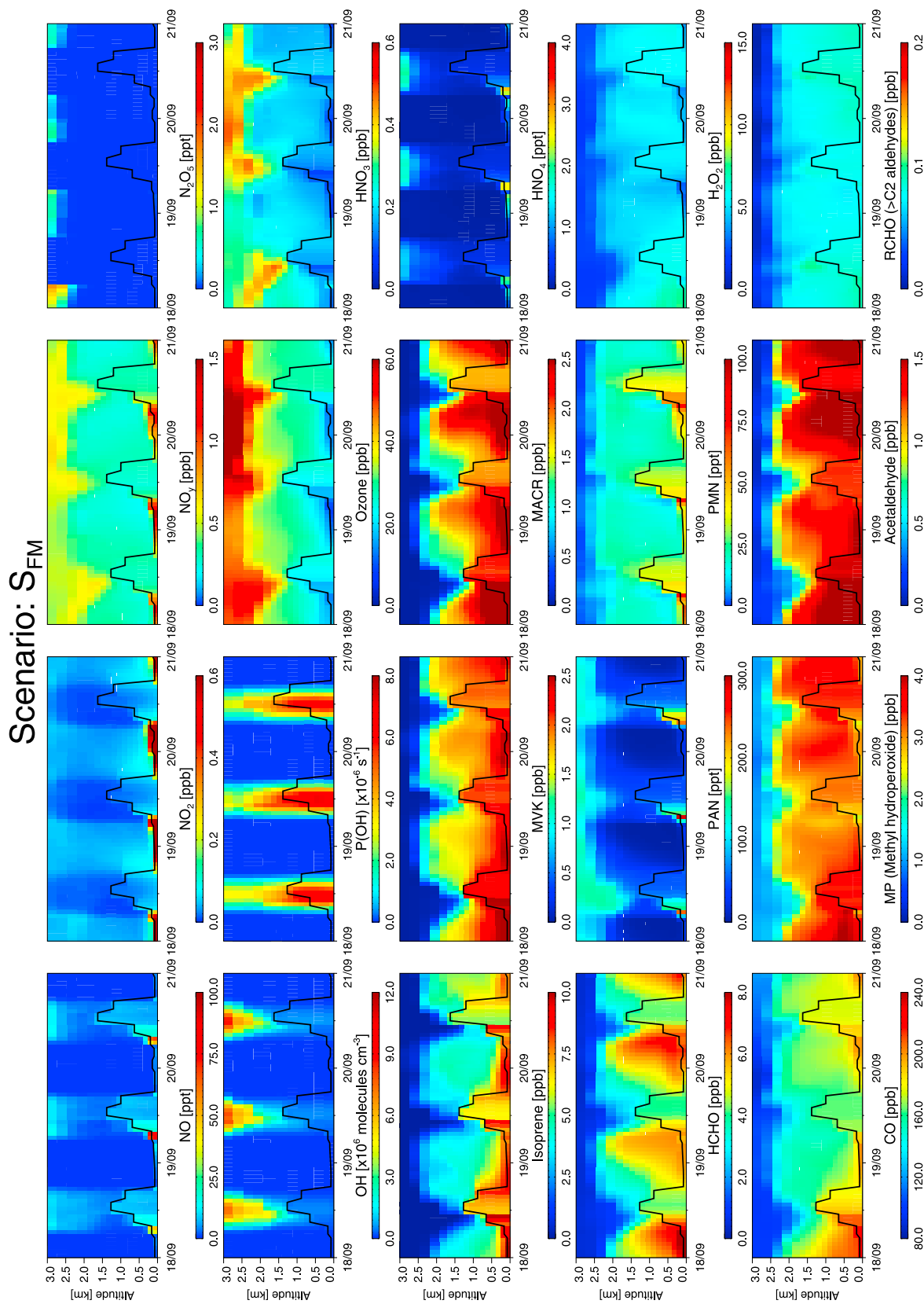


Figure 5. Time versus altitude plot of key species from the baseline model simulation S_{FM}, over 18–20 October 2004 at the location of the TROFEEE ground-based site (60.2°W 2.6°S) [Karl *et al.*, 2007]. The variation of the modeled planetary boundary layer is shown in each plot as the thick black line. Note that PMN is peroxy methacryloyl nitrate and PPN is peroxypropionyl nitrate; P(OH) is the primary production rate of OH from the photolysis of ozone.

even at altitudes as high as 2 km. In the early morning when the inversion breaks down, these accumulated species are then entrained into the rising boundary layer during the onset of daytime photochemistry. Hence, during the early morning, there is an extra ‘nocturnal’ source of HCHO within the boundary layer in addition to that produced from the photo-oxidation of the newly activated isoprene emissions. During the day, photolysis and oxidation by OH partly removes the MVK, MACR, and HCHO that has accumulated at higher altitudes during the night. Thus in our baseline simulation, high levels of isoprene and HCHO are not just constrained to the boundary layer but can exist up to about 2 km throughout the day. The night time accumulation of isoprene oxidation products in the residual layer and the failure of GEOS-Chem to vertically resolve the nocturnal PBL most likely results in the overestimation of the daytime HCHO columns. The entrained oxidation products also affect daytime isoprene oxidation, as they represent an additional sink for OH [Ganzeveld *et al.*, 2008]. High levels of MVK, MACR, and HCHO within the tropical nocturnal residual layer have also been reported in other studies [see, e.g., Ganzeveld *et al.*, 2008; Pugh *et al.*, 2010]; we discuss this feature more in section 5.

[40] Whilst changing a model parameterization may not appear to have a great effect on the HCHO column, and could be considered unimportant, it can greatly affect surface mixing ratios of isoprene and other key trace species (see, e.g., Figures S1–S3 and Text S1 within the auxiliary material). However, based on this set of simulations, we find the magnitude of the HCHO columns at this location are most sensitive to the surface isoprene emissions and chemical mechanism.

4.2. Comparison With GABRIEL Observations

[41] The GABRIEL aircraft campaign took place over Suriname, French Guiana and Guyana (3°–6°N, 51°–59°W) during October 2005, coinciding with the region’s dry season. During this month southeasterly trade winds transported clean Atlantic air masses westward where they were then subjected to significant reactive exchange over pristine tropical forests. Flights were performed over both land and ocean, sampling in the boundary layer and free troposphere (approximately 300 m to 10 km). The GABRIEL mission was unique, in that both OH and HO₂ were measured for the first time, along with a suite of VOC and oxygenated compounds. Overviews of the mission are given by Stickler *et al.* [2007] and Lelieveld *et al.* [2008]; more detailed descriptions of the airborne instrumentation and measurement procedures are described elsewhere [Williams *et al.*, 2007; Gebhardt *et al.*, 2008; Eerdeken *et al.*, 2009; Martinez *et al.*, 2010].

[42] To facilitate a meaningful comparison, we sample the model at the measurement times and locations along each flight track. Data from all flights were then averaged into 500 m bins (between 0 and 8 km) and the resulting profiles compared. We show the model and observed vertical profiles in Figure 6 and provide a statistical summary for the lowest 0–2 km of the atmosphere in Table S1 (see auxiliary material). The main results are as follows:

4.2.1. Isoprene Emissions

[43] Eerdeken *et al.* [2009] derived a mean surface isoprene emission flux from the measured concentrations of isoprene and its oxidation products of 6.9 mg m⁻² h⁻¹ for

the Guyana rainforest, after compensating for chemistry and detrainment. Model emissions spatially averaged over the GABRIEL domain and temporally averaged over 11:00–13:00 LT, were slightly lower than this inferred value being 4.2, 4.0, 2.5, 4.0, and 2.0 mg m⁻² h⁻¹ for the hybrid, PCEEA, hybrid-scaled, LPJ(CRU), and LPJ(GC) algorithms, respectively.

4.2.2. Isoprene

[44] There is a large spread in the isoprene mixing ratios simulated by GEOS-Chem (0.3–2.2 ppbv), though they mostly fall within the ±1 standard deviation of the mean observation. Generally, the model captures the shape of the observed isoprene observations ($r > 0.9$), although the measurements demonstrate a slower falloff with altitude. Several simulations (e.g., S_{FM} , $S_{FM(NOx)}$, $C_{FM(LPJCRU)}$) show very close agreement to the observations, whilst other scenarios, particularly those with OH recycling or with the lowest isoprene emissions, exhibit significant model bias.

4.2.3. MVK+MACR

[45] As with isoprene, the modeled MVK+MACR mixing ratios decrease faster with altitude than the observations; typically the profile differences are of order 0.5 ppbv. For simulations using the standard chemical scheme, there is a large variation in the simulated mixing ratios. Some scenarios have low overall bias but cannot capture the peak value well (e.g., S_{FM} or S_{NL}), other scenarios capture the peak but have a significant bias (e.g., $S_{FM(OH)}$ or $S_{NL(OH)}$). In contrast, the Caltech mechanism always produces an underestimation of the MCK+MACR mixing ratios and has a greater bias (approximately –20% to –70%).

4.2.4. HCHO

[46] Although the model simulations capture the observed HCHO vertical distribution, there is almost always a positive model bias. Below 1.5 km the modeled mixing ratios are of order 1–1.5 ppbv greater than aircraft measurements; above 1.5 km, the difference between the model and observations is small. Simulations using the Caltech mechanism consistently yield better agreement to the observations (unless OH recycling is included) than standard chemistry simulations. The $C_{FM(PCEEA)}$ scenario has the lowest overall bias ($B = 3\%$, $PPA = 8\%$).

4.2.5. OH and HO₂

[47] GEOS-Chem consistently underestimates the observed OH and HO₂ concentrations by about –20% to –60%. Only the Caltech mechanism with additional OH recycling, produces OH levels anywhere near those measured (e.g., scenario $C_{FM(PIE,OH)}$ yields $B = -17\%$ and $PPA = -17\%$). The general disagreement with the OH and HO₂ observations, suggests HO_x is inadequately generated and/or recycled in either model mechanism. It should be noted however, that above 3 km the differences between the modeled and observed HO₂ concentrations are fairly small.

4.2.6. Ozone

[48] The model reproduces the observed ozone vertical distribution reasonably well. The standard mechanism produces a closer match to observations below 2 km, whereas the Caltech mechanism overestimates mixing ratios near the surface but better captures the ozone profile at higher altitudes. As shown in Figure 6, the wide variety of different model scenarios tested have almost negligible impact on the simulated ozone mixing ratios.

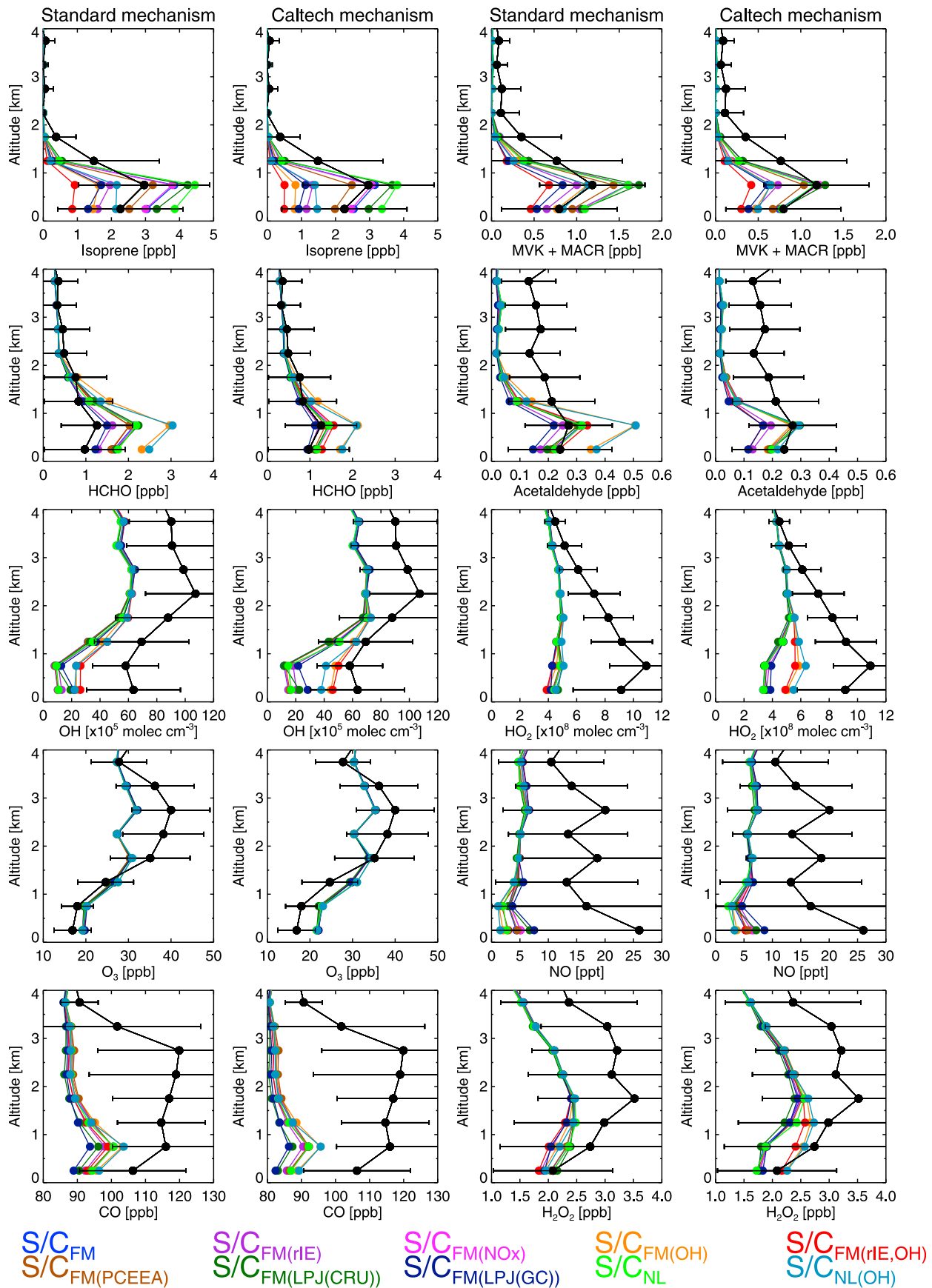


Figure 6

4.2.7. NO and CO

[49] NO is significantly underestimated by GEOS-Chem. Increasing soil NO_x emissions or using the Caltech chemical mechanism does little to alleviate this discrepancy. This negative bias may partially explain why model OH concentrations are also so low since the reaction of NO with HO₂ is a pathway for OH regeneration (if there is sufficient NO_x available). GEOS-Chem also fails to reproduce the observed CO distribution and typically underestimates the mixing ratios by 20 ppbv or more (i.e., $B = -15\%$ to -30%).

4.2.8. Acetaldehyde and H₂O₂

[50] Formation of hydrogen peroxide (H₂O₂) is an important sink for HO₂, especially in low-NO_x environments. Near the surface GEOS-Chem captures the observed H₂O₂ mixing ratios well, when using the standard chemical scheme, but simulates lower values than the measurements at higher altitudes. The Caltech chemistry scheme generates lower levels of H₂O₂ near the surface on average, unless OH recycling is included. Observed acetaldehyde mixing ratios are also captured reasonably by GEOS-Chem near the surface, but at higher altitudes are also underestimated.

4.2.9. Discussion of GABRIEL Analysis

[51] From this evidence it is clear that GEOS-Chem has difficulty in simulating several key chemical species measured during the GABRIEL campaign. In particular, HO_x, H₂O₂, NO and CO are all underestimated by the model, even with the more comprehensive Caltech chemical mechanism, whilst isoprene, MVK+MACR, and HCHO are generally overestimated. For isoprene and its oxidation products there is a large variation in the modeled values, indicating a high degree of sensitivity to the choice of model chemistry, emissions and boundary layer mixing. The observed slower decrease in the mixing ratios with altitude, suggests that either the vertical mixing within the boundary layer is underestimated or that the height of PBL is incorrect. Observed boundary layer heights, derived from potential temperature and water vapor measurements, were estimated to be approximately 500 m and 1450 m over land and 460 m and 500 m over ocean at 09:00 and 15:00 LT, respectively [Eerdeken et al., 2009]. The corresponding model PBL heights were 571 m and 1188 m over land, and 723 m and 718 m over ocean, averaged over 08:00–10:00 and 14:00–15:00 LT, respectively. Therefore the underestimation of the model PBL height overland potentially yields higher tracer mixing ratios than observed, given that the dilution volume for reactive emissions and chemical species is smaller. As noted by Eerdeken et al. [2009] and Ganzeveld et al. [2008], the true altitude to which species were mixed, referred to as the mixed layer (ML) depth, was actually at least 1900 m which further decreases surface emissions and boundary layer mixing ratios. As the full- or nonlocal mixing schemes are constrained to the dynamic PBL height, neither mix tracers to the higher ML depth which may account for the higher than observed model values. Above

the PBL, both MVK+MACR and HCHO are higher than the simulated values which may suggest convective transport is also underestimated within this region.

[52] To understand the subsequent impact of this poor model agreement on the simulated HCHO columns, we relate the aircraft measurements to the satellite observations by additionally sampling the model along a longitudinal transect (45–60°W; 4.5°N) directly across the flight paths, averaging the HCHO columns over 09:00–11:00 and 12:00–15:00 LT to coincide with the overpass times of the respective SCIAMACHY (10:00 LT) and OMI (13:30 LT) instruments. We show the variation of the satellite and model HCHO columns along the transect in Figure 7. Both GEOS-Chem and the satellite observations agree on the transect gradient, with larger HCHO columns over land than ocean. Compared with the SCIAMACHY observations, simulations using the standard chemical scheme underestimate the observed HCHO columns over the ocean but overestimate them over land; model runs performed using the Caltech chemistry option are consistently lower than the retrieved HCHO column measurements. The spread of the relative difference, i.e., $(M - O)/O$, of the model columns (M) compared to the SCIAMACHY data (O) is considerable: -50% to 50% . When compared with the OMI HCHO column data, the Caltech chemical scheme produces a better match to the observations as the magnitude of the relative difference is mostly $<25\%$. The standard chemical scheme performs worse with the relative difference typically $>25\%$ (over land).

[53] In this context, it is not obvious why the observed differences in the HCHO profiles shown in Figure 6 of only 1–1.5 ppbv can translate to such a large deviations in the model HCHO columns. To illustrate why this is the case, we sequentially perturb two typical synthetic oceanic and continental HCHO profiles (below 1 km) to assess the relative change in the HCHO vertical column. As shown in Figure 8, over the ocean where HCHO columns approach background levels, even a small increase of HCHO of 0.5 ppbv within the lowest 1 km can create a 17% relative increase in the HCHO column; a 1 or 2 ppbv change results in increases of 33% or 66%. Similarly, over land where HCHO columns are higher owing to VOC oxidation, a 1 or 2 ppbv perturbation creates a 20% and 40% relative change, respectively. In other words, a 20–30% model bias relative to a SCIAMACHY or OMI measurement will occur if GEOS-Chem fails to reproduce ‘true’ HCHO near surface mixing ratios within 1 ppbv (irrespective of any retrieval uncertainties). Therefore, the model should therefore at least attempt to simulate HCHO within boundary layer at the sub 1 ppbv level, in order to ensure meaningful comparisons to satellite HCHO column observations.

4.3. Comparison With Satellite HCHO Columns

[54] To determine top-down isoprene emission with any confidence, it is a necessary prerequisite that the model

Figure 6. Observed and model vertical profiles over Suriname, French Guiana, and Guyana (3°–6°N 51°–59°W) during the GABRIEL campaign in October 2005 [Stickler et al., 2007]. The model has been sampled at the same times and locations as the measurements and the data sets averaged into 500 m altitude bins. The observations are shown by the solid black lines with ± 1 standard deviation error bars; the model simulations are represented by the solid color lines (see Table 5). Here S/C_{FM} represents either S_{FM} or C_{FM} scenario (see Table 5).

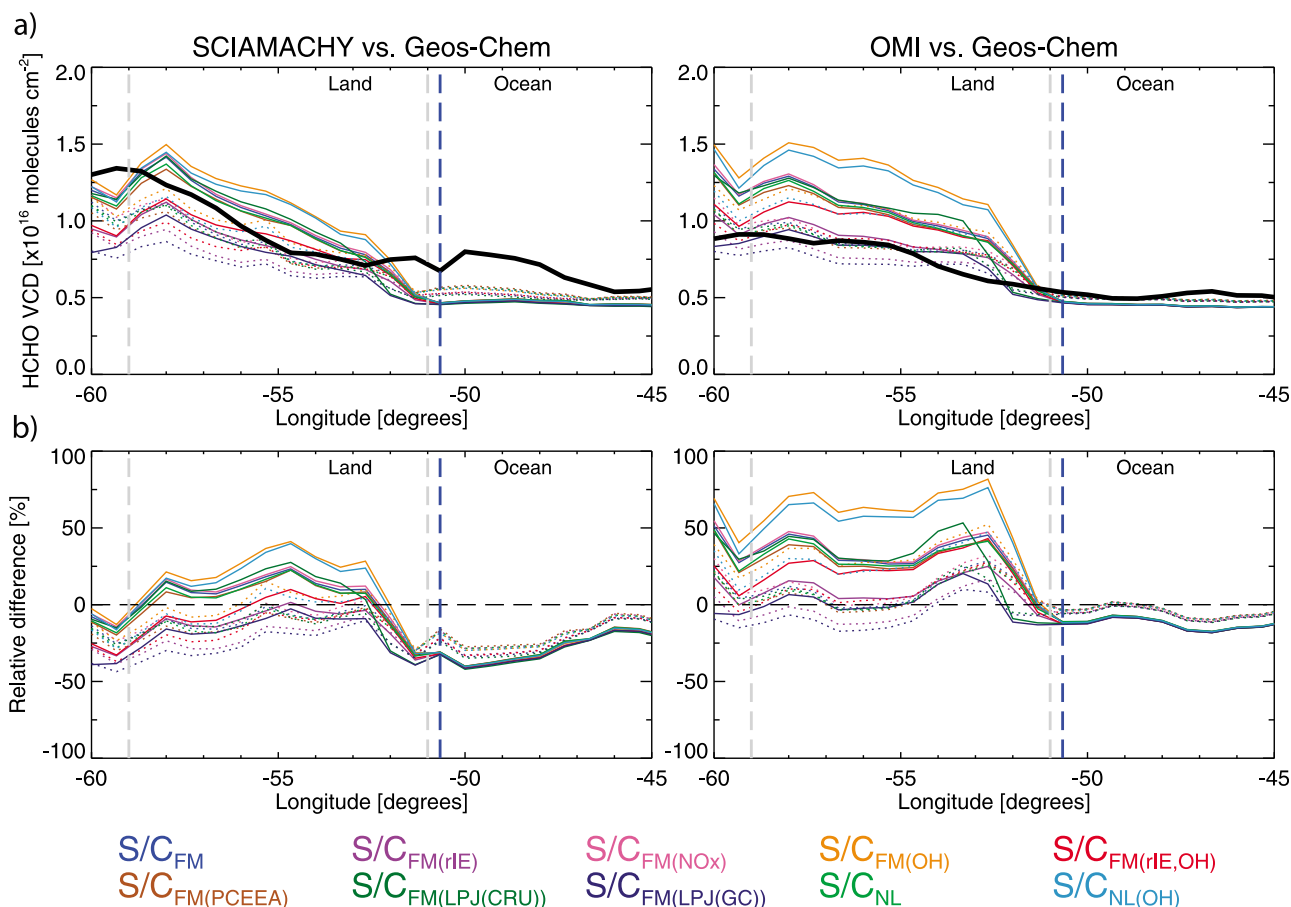


Figure 7. Observed and modeled HCHO vertical columns along a longitudinal transect at 4.5°N during the GABRIEL campaign October 2005 [Stickler et al., 2007]. (left) HCHO columns averaged over 09:00–11:00 LT corresponding with the overpass of SCIAMACHY (10:00 LT) (solid black line). (right) HCHO columns averaged over 12:00–15:00 LT corresponding with the overpass of OMI (13:30 LT) (solid black line). The model simulations using the standard and Caltech chemical mechanisms are represented by the solid and dashed color lines, respectively (see Table 5). (a) The mean HCHO columns (averaged over 3–17 October) and (b) the relative difference between the model and observations are shown. The SCIAMACHY and OMI HCHO data has been smoothed along the transect with a box filter (of nine grid cells width) to remove noise. The vertical grey lines represent the longitudinal limits of the Gabriel aircraft flights; the blue vertical line denotes the GEOS-Chem land-ocean crossover.

provides a reasonable simulation of the observed HCHO column spatial distributions. To establish whether this is the case, we show in Figures 9 and 10 the monthly mean SCIAMACHY and OMI HCHO distributions for 2006, along with the model output from the C_{FM} , $C_{FM(PCEEA)}$, $C_{FM(rIE)}$, $C_{FM(LPJ(CRU))}$, and $C_{FM(LPJ(GC))}$ scenarios. Here we sample the model at the same times and locations as the respective SCIAMACHY and OMI measurements and average the data in the same way (as described in the auxiliary material). We also show the time series of the monthly mean HCHO column statistics in Figure 11, and provide a summary of the annual mean statistics in Table 7.

[55] The SCIAMACHY observations show a pronounced seasonal cycle with high-HCHO columns in the dry season and low values in the wet-to-dry transitional period [Barkley et al., 2009]. When compared to the observations, the model demonstrates a large positive bias (often >50%), with simulations using the standard chemical mechanism performing

much worse than those using Caltech scheme. Despite the model's positive bias, the various scenarios do not capture the observed maximum values well, often underestimating them except during March and September. Of the Caltech simulations, four scenarios yield much better results than the others: $C_{FM(rIE)}$, $C_{FM(LPJ(CRU))}$, and $C_{FM(LPJ(GC))}$ and $C_{FM(rIE,OH)}$. These scenarios have notably lower isoprene emissions than MEGAN as discussed in section 3, which may suggest the unscaled MEGAN emissions are too high. For example, the C_{FM} and $C_{FM(PCEEA)}$ model runs produce very high-HCHO columns for much of the continental interior. The $C_{FM(rIE)}$ simulated distributions shown in Figure 9 are more similar to those observed by SCIAMACHY, except that they are much higher in both the wet-to-dry transitional period and in the dry season, and contain very high HCHO columns over Brazilian border with Peru and Bolivia (i.e., where the highest emissions occur as shown in Figure 2). The $C_{FM(LPJ(CRU))}$, and $C_{FM(LPJ(GC))}$ simulations have a low

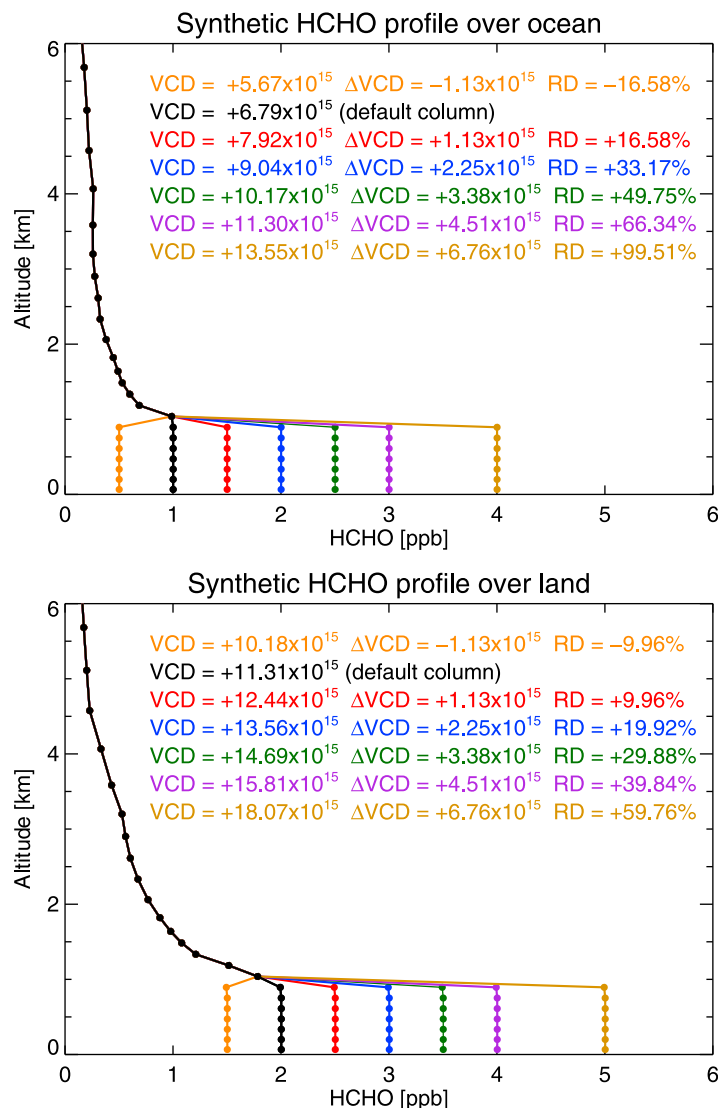


Figure 8. Influence of simulated HCHO near-surface mixing ratios on the corresponding VCD (in molecule cm^{-2}). The default model profile (black line) is representative of the HCHO vertical distribution over the (top) ocean and (bottom) land. The colored lines represent successive perturbations to the default profile of -0.5 (orange), $+0.5$ (red), $+1.0$ (blue), $+1.5$ (green), $+2.0$ (purple), and $+3.0$ ppbv throughout the lowest 1 km; Δ VCD is the corresponding change in the vertical column due to each perturbation and RD is the relative difference.

annual mean bias overall of 15% and 10%, respectively. However, as shown in Figure 11 most of this bias originates from during the wet-to-dry transitional period when isoprene emissions are particularly uncertain [Barkley *et al.*, 2009]. The spatial distributions from these two scenarios are different to the MEGAN-based simulations, but have better correlations in earlier and latter parts of the year as shown in Figure 11. Therefore despite some significant differences, e.g., the evident disagreement in southeastern Brazil during May, the HCHO column distributions based on the LPJ-GUESS emissions generally give better agreement to the SCIAMACHY data than those based on MEGAN.

[56] Unlike SCIAMACHY, OMI only observes very high HCHO columns deep within the central regions during August–September and has smaller annual mean (about 15% less relative to SCIAMACHY). Subsequently the

positive model bias relative to the OMI observations, is notably higher than with the SCIAMACHY data. Although the simulations using the Caltech chemistry once more have a lower model bias than simulations using the standard mechanism, the bias is still substantial ranging from about 20–100%. Again, the $C_{FM(VIE)}$, $C_{FM(LPJ(CRU))}$, and $C_{FM(LPJ(GC))}$ and $C_{FM(VIE,OH)}$ simulations perform best and have the lowest bias. That said, the spatial correlation is particularly poor in some months, e.g., during May and June for the $C_{FM(LPJ(CRU))}$ simulation. Furthermore, all model runs demonstrate a strong seasonality regarding spatial agreement to the OMI observations, as the correlation is highest during July–October and lowest in April. Overall, the $C_{FM(LPJ(GC))}$ simulation performs best. However, an extra simulation in which the isoprene emission within the C_{FM} scenario were reduced by 50%

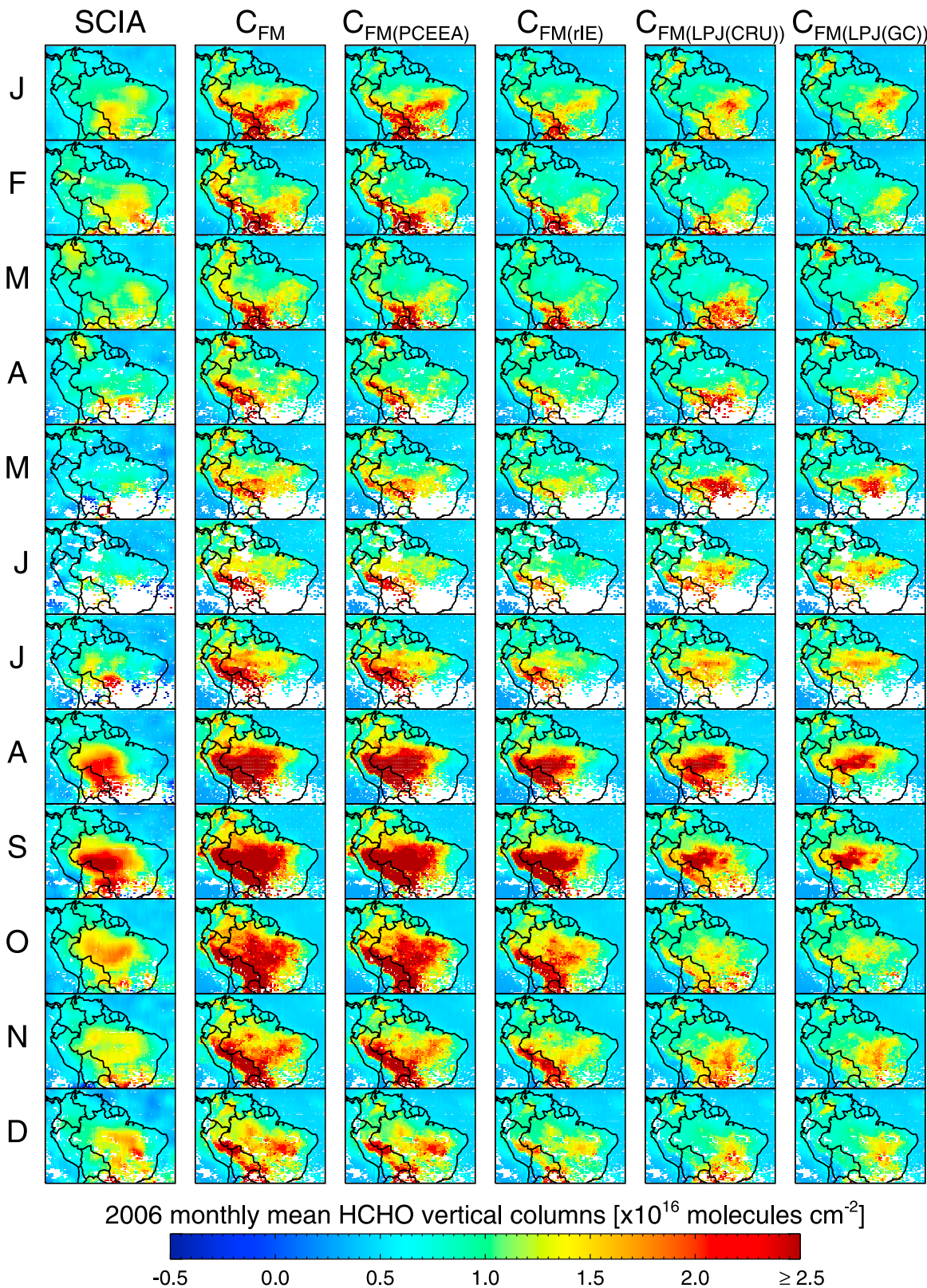


Figure 9. Monthly mean HCHO columns for 2006, corresponding to 09:00–11:00 LT, as observed by SCIAMACHY and simulated by GEOS-Chem using the Caltech chemical mechanism (see Table 5). The SCIAMACHY data have been averaged onto the nested $0.667^\circ \times 0.5^\circ$ grid using observations with cloud cover $\leq 40\%$ and smoothed with a 9×9 box filter to remove noise. The model data have been sampled at the same times and locations as the SCIAMACHY observations but are not smoothed. White-colored regions correspond to areas with no usable SCIAMACHY data.

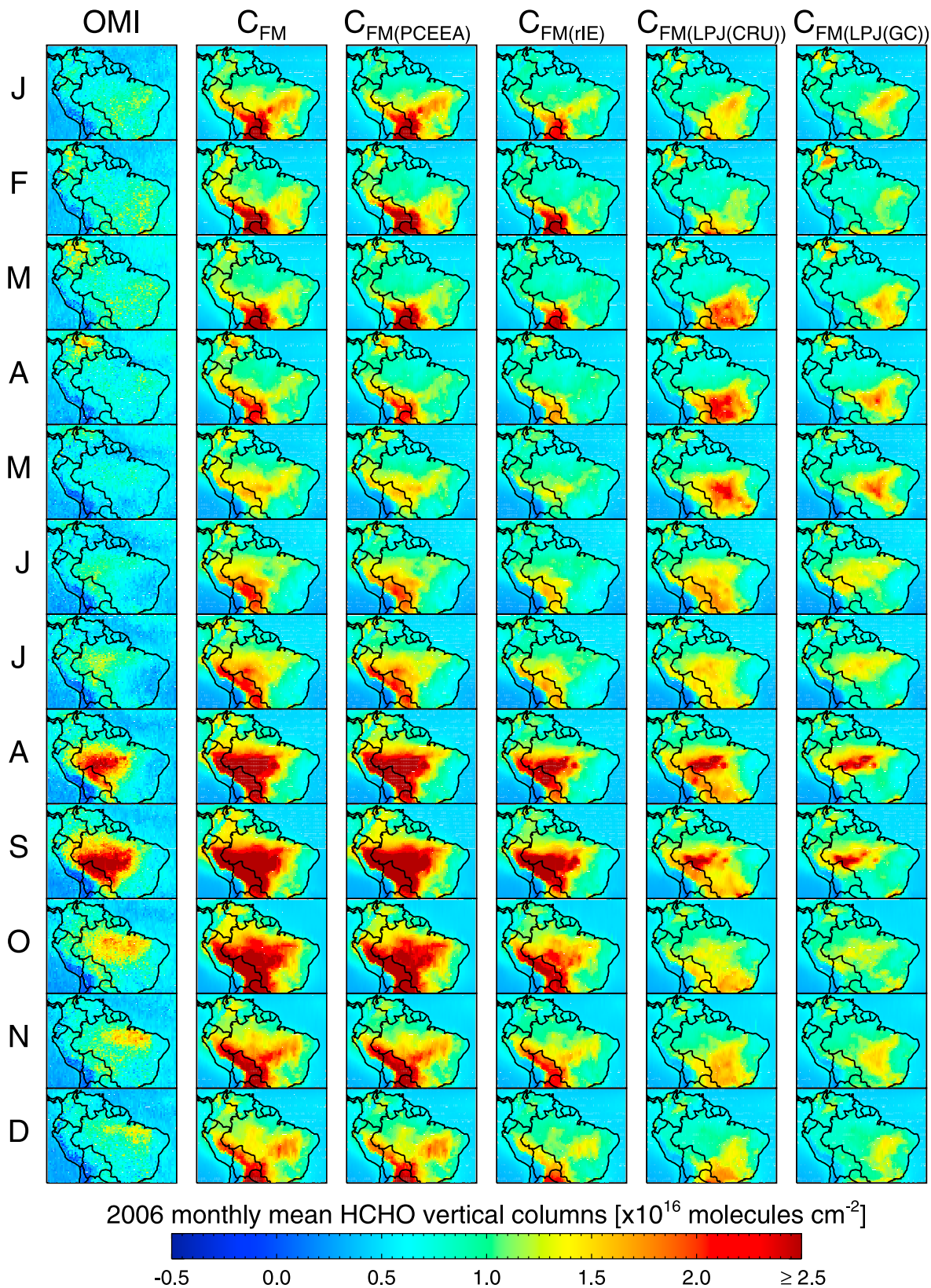


Figure 10. Monthly mean HCHO columns for 2006, corresponding to 12:00–15:00 LT, as observed by OMI and simulated by GEOS-Chem using the Caltech chemical mechanism (see Table 5). The OMI data have been averaged onto the nested $0.667^\circ \times 0.5^\circ$ grid using observations with cloud cover $\leq 40\%$. The model data have been sampled at the same times and locations as the OMI observations. Neither the OMI nor the GEOS-Chem data are smoothed.

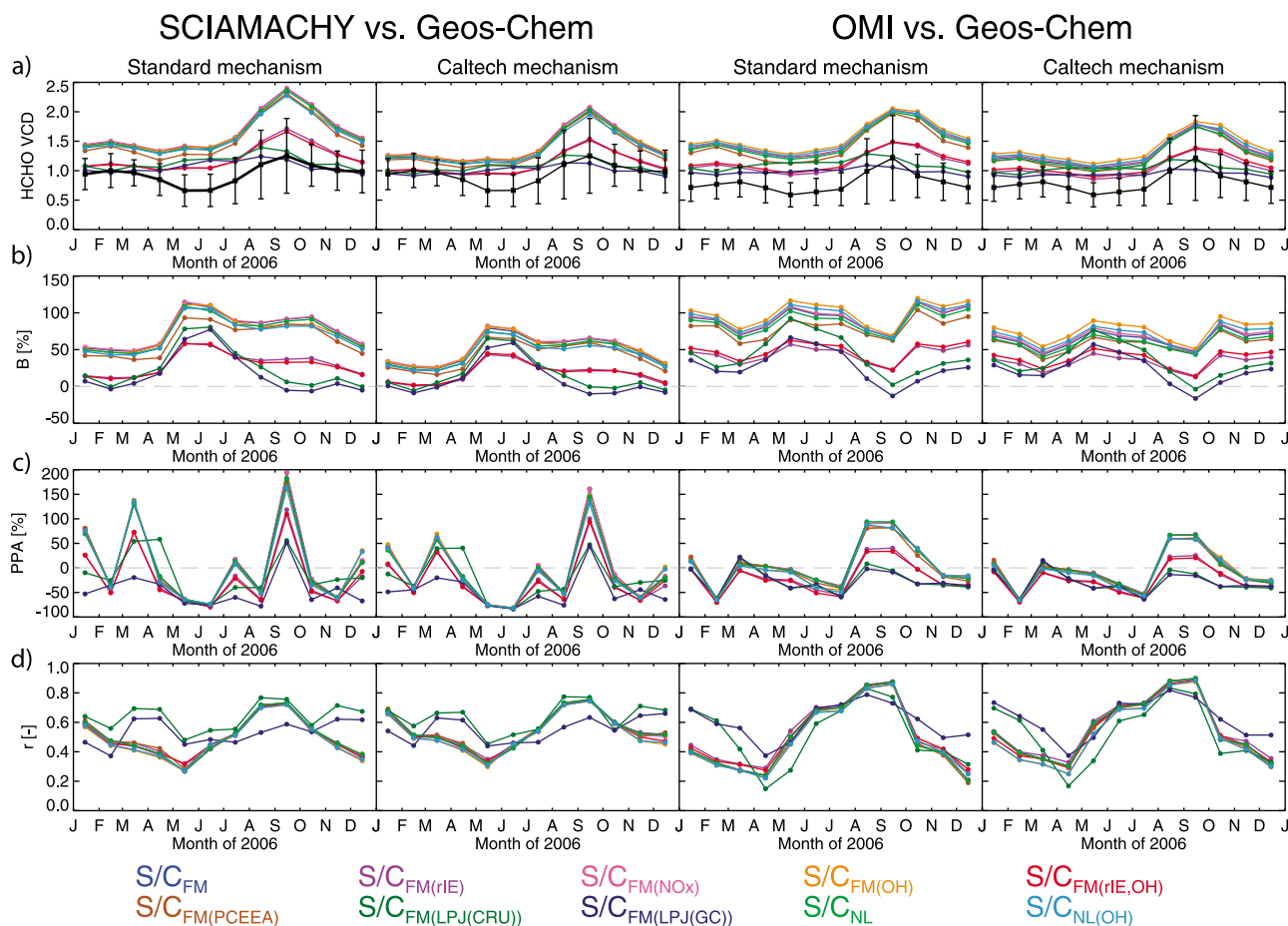


Figure 11. Time series of the monthly mean HCHO column statistics for 2006. (left) Comparison of SCIAMACHY and GEOS-Chem HCHO columns computed with standard and Caltech chemical mechanism. (right) Comparison of OMI HCHO with the model data. Both the SCIAMACHY and OMI data have been averaged onto the nested $0.667^\circ \times 0.5^\circ$ grid using observations with cloud cover $\leq 40\%$; the SCIAMACHY data are smoothed with a 9×9 box filter to remove noise. The model data have been sampled at the same times and locations as the respective satellite observations but are not smoothed. (a) The monthly mean HCHO column (in molecule cm^{-2}) with the observations and their ± 1 standard deviation error bars (plotted in black), (b) the normalized mean bias (%), (c) the paired peak accuracy (%), and (d) the correlation are shown. The model simulations are represented by the solid color lines (see Table 5).

improves the agreement to the satellite observations reducing the annual mean bias with respect to SCIAMACHY and OMI to 6% and 20%, respectively.

[57] We find the influence of boundary layer mixing, soil NO_x emissions, and OH recycling on the simulated HCHO columns to be minor (Figure 11). Use of nonlocal boundary layer mixing reduces the annual bias by about 4%, whilst increased soil NO_x emissions only slightly increase the bias. Use of OH recycling worsens the bias by about 10% when the model is compared with OMI data, and has a minimal effect when it is compared with SCIAMACHY data. This reinforces our earlier findings that the model HCHO columns are predominantly controlled by isoprene emissions and the choice of chemical mechanism. However, in the real atmosphere all these processes occur simultaneously and cannot be decoupled as done in the model (by operator splitting). For example, the influence of boundary layer mixing on the HCHO columns in reality may actually be substantial.

[58] Finally, the burning of tropical vegetation can release significant amounts of HCHO into the atmosphere [Stavrakou *et al.*, 2009a; Yokelson *et al.*, 2008]. To determine the influence biomass burning on the model HCHO columns, we conducted a simulation in which the biomass burning emissions within the C_{FM} scenario were switched off. We find only a small improvement (a reduction of about 5%) in the annual biases with respect to both the SCIAMACHY and OMI observations. Despite this relatively small impact on the HCHO columns for the region as a whole, emissions from biomass burning still represent a large source of uncertainty and may partially explain some of the discrepancy between the model and satellite data.

5. Discussion

[59] Having evaluated GEOS-Chem against the in situ and satellite observations we find the model has difficulty in

Table 7. Annual Model HCHO Column Statistics Over Land for 2006

Scenario	Mean HCHO VCD ^a		Mean B (%)		Mean PPA (%)		Mean <i>r</i>	
	Standard	Caltech	Standard	Caltech	Standard	Caltech	Standard	Caltech
SCIAMACHY	0.94 ^b							
<i>FM</i>	1.66	1.40	76	49	9	-8	0.49	0.55
<i>NL</i>	1.62	1.37	73	46	7	-11	0.50	0.55
<i>FM(OH)</i>	1.63	1.40	74	50	7	-8	0.48	0.53
<i>NL(OH)</i>	1.59	1.35	70	44	5	-12	0.48	0.53
<i>FM(NO_x)</i>	1.67	1.42	78	51	9	-7	0.49	0.54
<i>FM(PCEEA)</i>	1.55	1.33	64	41	6	-12	0.51	0.55
<i>FM(<i>r</i>IE)</i>	1.22	1.10	31	18	-19	-28	0.50	0.55
<i>FM(<i>r</i>IE,OH)</i>	1.21	1.11	29	18	-20	-28	0.49	0.54
<i>FM(LPJ(CRU))</i>	1.14	1.06	25	16	-14	-22	0.64	0.63
<i>FM(LPJ(GC))</i>	1.07	1.00	17	10	-46	-47	0.53	0.55
OMI	0.80 ^b							
<i>FM</i>	1.51	1.30	92	64	7	-4	0.49	0.55
<i>NL</i>	1.48	1.26	88	60	6	-6	0.49	0.55
<i>FM(OH)</i>	1.57	1.38	100	75	3	-6	0.48	0.52
<i>NL(OH)</i>	1.54	1.33	95	69	2	-9	0.48	0.52
<i>FM(NO_x)</i>	1.52	1.31	93	66	7	-4	0.49	0.55
<i>FM(PCEEA)</i>	1.43	1.25	80	57	3	-7	0.49	0.55
<i>FM(<i>r</i>IE)</i>	1.13	1.04	44	32	-18	-24	0.52	0.57
<i>FM(<i>r</i>IE,OH)</i>	1.16	1.08	48	38	-21	-25	0.51	0.55
<i>FM(LPJ(CRU))</i>	1.10	1.04	43	34	-25	-30	0.51	0.52
<i>FM(LPJ(GC))</i>	0.99	0.94	28	22	-25	-30	0.60	0.62

^aHCHO vertical column density in units of $\times 10^{16}$ molecules cm^{-2} .

^bMean observation retrieved by each instrument.

reproducing observed isoprene-related chemistry and HCHO columns over the Amazon Basin. In this section we discuss the potential reasons for this.

[60] We acknowledge that the bottom-up isoprene emissions are a large source of uncertainty. Both the MEGAN and LPJ-GUESS inventories suffer from a lack of constraining measurements, and only extensive field campaigns to catalogue the emissions from tropical plant species will address this problem. Despite the challenges of making measurements within the remote rainforest, campaigns need to target regions of high uncertainty, e.g., Brazilian-Peruvian border, and not locations simply because of logistical ease or existing infrastructure. Moreover, continuous measurements are required to quantify and understand the seasonal variability of isoprene emissions and their response to environmental and phenological changes [Barkley *et al.*, 2009]. However, through using different variants of MEGAN and LPJ-GUESS models we have at least provided a range of estimates, i.e., about 70–170 Tg a^{-1} over the time period 2005–2008. Of the two emission inventories which it is difficult to say which is better, the LPJ-GUESS estimates give slightly better agreement to the satellite data, but when the basal emission rates are known or measured, as for example at the TROFFEE site (Figure 4), then MEGAN reproduces the observed emissions with a modest bias (see Table 6). On the other hand, aircraft measurements of the above-canopy isoprene fluxes surrounding the TROFFEE site indicated that MEGAN underestimated the emissions from this region [Karl *et al.*, 2007], thus implying that LPJ-GUESS estimates would potentially be much lower than the observations. Similarly, down scaling the MEGAN emissions slightly improves the agreement to satellite HCHO data, if not the in situ observations; i.e., lower emissions potentially compensate for errors in the chemical mechanism. Despite the different approaches of MEGAN and LPJ-GUESS, in this work they are both forced with meteorological fields from

the same GEOS-5 assimilation system, except for the LPJ (CRU) based estimates. Therefore, differences between the two models are entirely due to the driving algorithms and assignment of isoprene emission capacities; both inventories have advantages and disadvantages [see, e.g., Ninemets *et al.*, 2010a, 2010b, and related online discussion].

[61] However, the choice of driving variables is clearly important. This is apparent from the fact that MOHYCAN estimates, which use ECMWF meteorological fields, are about 70 Tg C yr^{-1} less than GEOS-Chem's MEGAN estimates, even though they are based on the same emission factors. Although the MOHYCAN model includes the effects of soil moisture, this extra activity term is unlikely to account for all the observed differences; over the Amazon soil moisture term is about 0.7–0.9 [see Müller *et al.*, 2008, Figure 2], i.e., a reduction of only 10–30% in the isoprene emissions at most. A key question therefore is accuracy of the GEOS-5 meteorological data. To assess the impact of meteorological data on the HCHO columns, we performed two extra sensitivity simulations based on the S_{FM} scenario, during which the 2 m air temperature (T2m) and PAR (diffuse and direct) were separately increased by 2 K and 10%, respectively. Owing to these perturbations the isoprene emissions increased from 154 Tg C to 198 Tg C and 176 Tg C, respectively. Whilst the increase in PAR increases the HCHO column bias by about 6–7%, the effect of temperature is more substantial increasing the bias by about an extra 25%. The impacts of these changes are nonlinear and difficult to extrapolate, but they at least demonstrate the sensitivity of MEGAN to the input meteorological data, and more importantly that any small bias in temperature will have a notable effect on the simulated HCHO columns.

[62] The failure of GEOS-Chem to simulate observed key species, coupled with its likely excessive production of HCHO, indicate potential inadequacies within the model's chemical mechanism. Given that extensive lumping of

oxidation intermediates occurs with the mechanism, some degree of uncertainty is expected; intercomparisons of other lumped CTM chemical mechanisms have shown considerable differences in their simulation of isoprene chemistry [e.g., *Emmerson and Evans*, 2009; *Archibald et al.*, 2010a]. That said, by midafternoon prior to the overpass of OMI, the GEOS-Chem yields are only within 3% of MCM yields. As GEOS-Chem consistently underestimates OH concentrations, it is therefore more likely that either OH is not being generated or recycled fast enough, coupled with the fact that the model isoprene emissions are simply much too high. If observed OH levels were accurately reproduced by the model it might help reconcile some of the differences to the in situ measurements and satellite HCHO data. Recently, three pathways for OH regeneration coupled to the photo-oxidation of isoprene have been proposed: (1) artificial OH recycling [*Lelieveld et al.*, 2008], (2) OH regeneration from the production of epoxides [*Paulot et al.*, 2009a], and (3) a radically new theoretical scheme called the Leuven Isoprene Mechanism (LIM0) [*Peeters et al.*, 2009; *Peeters and Müller*, 2010]. We have tested (1) and (2) within this work and found neither to sustain observed OH levels. We acknowledge that for OH recycling, an alternative value of n within reaction R1, or including OH formation via second generation peroxy radicals, with equivalent reduction in the ISOP+OH reaction might satisfy the Amazon conditions [e.g., *Butler et al.*, 2008]. However, as we have only one observational campaign to evaluate n , setting such values is arbitrary. In any case, the impact of OH recycling on the simulated HCHO columns is small, as also found in the study of *Stavrakou et al.* [2009a]. Similarly, epoxide formation as implemented in the Caltech scheme does also not simulate observed OH levels, and compared with OH recycling yields only a modest increase in OH. Studies by *Archibald et al.* [2010b] and *Stavrakou et al.* [2010] also found only moderate increases in OH levels produced via OH recycling and epoxide formation. Nevertheless, the modifications to GEOS-Chem's chemical scheme by *Paulot et al.* [2009a, 2009b] result in a much improved simulation of the HCHO columns, probably from improved characterization of organic nitrates [*Paulot et al.*, 2009b], despite remaining uncertainties in their yields and atmospheric fates [*Perring et al.*, 2009; *Lockwood et al.*, 2010].

[63] The LIM0 scheme has demonstrated that it has the potential to generate the very high OH levels over tropical rainforests [*Stavrakou et al.*, 2010]. The basis of this mechanism is that the isomerization reactions of specific isoprene peroxy radicals, outcompete traditional reactions with NO and HO₂, resulting in the formation of HO_x radicals and hydroperoxy aldehydes (HPALDs). These HPALDs can easily undergo photolysis and lead to additional production of OH and also HCHO. Despite some minor refinement to the scheme [*Peeters and Müller*, 2010], owing to issues raised by the studies of *Karl et al.* [2009], *Paulot et al.* [2009a], and *Archibald et al.* [2010b], the LIM0 mechanisms potentially represents the best option for realistic simulation of the Amazonian atmosphere, though it still needs extensive evaluation. For example, the low-NO_x HCHO yields are largely unknown owing to uncertainties in the isomerization reaction rates and HPALD photolysis products; the formation of HPALDs also partially suppress MVK and MACR production.

[64] The simulations performed at the TROFFEE location (section 4.1) show a pronounced nighttime buildup of isoprene and other oxygenated compounds. Clearly, this phenomena originates from the close diurnal chemical and dynamical coupling that occurs within the lowest part of the troposphere. Furthermore, the nocturnal accumulation of HCHO most likely contributes to the very high HCHO columns simulated by GEOS-Chem. This behavior likely arises because the model cannot generate sufficient OH concentrations to adequately remove the majority of the isoprene emitted during the day. By the late afternoon/early evening as the boundary layer collapses, a buildup of isoprene (and other compounds) occurs within model layers nearest the surface. During the night isoprene is further oxidized to form HCHO. However, observations have shown that near the surface, isoprene decays away very rapidly during the night with lifetime of ~2 h [*Hurst et al.*, 2001; *Holzinger et al.*, 2002; *Karl et al.*, 2004]. Likely causes of this loss mechanism include dry deposition [*Holzinger et al.*, 2002], or as postulated by *Hurst et al.* [2001] OH oxidation and vertical mixing with isoprene-depleted air from above. *Karl et al.* [2004] showed that substantial loss of isoprene could occur within the rainforest canopy, and that deposition velocities for isoprene and other oxygenated compounds were much greater than current model predictions. In that study, VOC mixing ratios within the surface layer just above the canopy exhibited a minimum in the early morning owing to venting with VOC depleted canopy air and growth of the morning boundary layer. *Ganzeveld et al.* [2008] suggested that intermittent nocturnal transport could couple the residual layer with the surface, allowing deposition to remove VOCs and oxygenated products. Such events might reduce the elevated mixing ratios that are simulated in the higher model layers. However, since GEOS-Chem does not explicitly model reactive exchange within the canopy, or vertically resolve the shallow nighttime PBL, it cannot mimic these nocturnal processes. In a simple attempt to mix tracers to the ML depth and allow some nocturnal vertical mixing, we performed a sensitivity simulation in which we adjusted the S_{FM} scenario to allow full mixing to occur to two extra model layers beyond the PBL. This alteration dramatically reduced the biases of isoprene and MVK+MACR with respect to the TROFFEE observations to about 4% and 35%, respectively; the HCHO column decreased by about 3%, reducing the bias by 4%. Clearly, in future work we must examine the role of diurnal boundary layer mixing much more closely.

[65] Tracers that escape the PBL are vertically mixed by moist convection and advection. A sensitivity study by *Stavrakou et al.* [2009a] using the Intermediate Model of Global Evolution of Species (IMAGES) CTM showed modeled HCHO columns were strongly influenced by the convection. Over the Amazon those authors found that a doubling of convective fluxes produced relative changes in the vertical column of up to -10% [see, e.g., *Stavrakou et al.*, 2009c, Figure 2]. Similarly, aircraft measurements over Suriname have shown that shallow convective mixing could transport air masses from the boundary layer up to about 4 km [*Warneke et al.*, 2001]. Although we have not examined the impact of convection in this work, a recent analysis by *Liu et al.* [2010] using GEOS-Chem and CO data, from the Microwave Limb Sounder (MLS) [*Waters et al.*, 2006] and the Tropospheric Emission Spectrometer (TES) [*Beer*,

2006], showed that GEOS-5 convection decays at too low an altitude over the Amazon, and that vertical mixing is potentially not strong enough. If convection is too weak, then not enough HCHO will be transported from the boundary layer to the free troposphere, where photolysis rates are faster and OH levels higher, and subsequently its photochemical destruction will be less than expected.

[66] Various studies have highlighted the importance of dry deposition as a major factor that influences isoprene and its subsequent oxidation products [Karl *et al.*, 2004; von Kuhlmann *et al.*, 2004; Pugh *et al.*, 2010]. Unfortunately, few studies exist by which to evaluate dry deposition scheme within the model. Karl *et al.* [2010] found average daytime deposition velocities (v_d) for MVK+MACR to be 1.6 and 2.4 cm s^{-1} within the tropical rainforests of Costa Rica and the Amazon, respectively. For the GABRIEL campaign, Ganzeveld *et al.* [2008] computed deposition velocities for MVK and MACR of about 1.0 cm s^{-1} during the day, and 0.1 cm s^{-1} at night. From the S_{FM} simulation performed for the TROFFEE comparison, we find similar average day and nighttime values for MVK (and MACR) of 0.8 and 0.1 cm s^{-1} , respectively; whilst for PAN we calculate a $v_{d(\text{Day})} = 0.7 \text{ cm s}^{-1}$, which is consistent with the study of Turnipseed *et al.* [2006]. However, recent work by Karl *et al.* [2010], based on plant exposure and wounding experiments, implies plants can adjust their metabolism according to their environmental conditions, and decrease their mesophyll resistance to deposition. Their incorporation of these new findings within a global CTM resulted in significant increases in global BVOC deposition fluxes, but more relevantly, also in dramatic changes of MVK and MACR concentrations over the Amazon (relative decreases of about 30–35%, see Figures S12 and S13 of their supporting online material). Such large increases in deposition fluxes have the potential to reduce mixing ratios of isoprene oxidation products and limit HCHO production.

[67] When the model is compared with the satellite HCHO data, we also have to consider the uncertainties associated with HCHO retrievals themselves. Here we briefly draw attention to two issues that are of particular relevance: the satellite overpass times, and the ancillary information used within the air mass factor (AMF) calculation (see auxiliary material). The time difference between the satellite overpasses is important since it affects each instrument's respective sampling, e.g., owing to relative cloud coverage, and because it also reflects potential temporal errors within the model, i.e., the model simulation better at 10:00 LT than at 13:30 LT? The better agreement between GEOS-Chem and SCIAMACHY may partially originate from this time difference. Computation of AMFs requires a priori knowledge of aerosols and cloud properties, surface albedo, and HCHO vertical distribution. This last term is provided by a CTM driven by a specific isoprene emission inventory, in this case MEGAN for both instruments (albeit implemented differently); a somewhat circular process. By implementing the MEGAN and LPJ-GUESS inventories within GEOS-Chem, we have shown they produce very different HCHO distributions over the Amazon; AMF calculations based these respective emissions will therefore yield different HCHO vertical columns. Based on aircraft observations over the United States [Millett *et al.*, 2006; Palmer *et al.*, 2003], the estimated AMF error due to uncertainties in the HCHO vertical profile is about 10%.

However, to our knowledge no one has yet assessed the spatial and temporal components of this error using model HCHO columns based on two different isoprene emission models. Recomputation of AMFs using our model simulations might help reconcile the differences to the satellite data. We intend to investigate this topic in future work.

6. Summary and Future Implications

[68] Top-down isoprene emissions derived from satellite observations of HCHO columns potentially offer information on the seasonal and interannual variability of the emissions from largely inaccessible tropical rainforests. However, their accuracy relies critically on the chosen chemistry transport model used to derive the emission estimates. We have presented an evaluation of a high-resolution GEOS-Chem simulation of tropospheric chemistry over tropical South America, in which we have driven model with two different isoprene emission inventories, MEGAN and LPJ-GUESS. The isoprene emissions simulated by the two models vary considerably and have large differences in their annual totals (up to 100 Tg C). Based on our simulations and assuming a global emission strength of 500 Tg C a^{-1} [Arnth *et al.*, 2008; Guenther *et al.*, 1995], we find the Amazon contributes anywhere between about 15–35% of the total global emissions.

[69] Knowing that in the future we want to use GEOS-Chem to determine top-down isoprene emissions, in this first evaluation of the model we have assessed the model's ability to reproduce isoprene and HCHO chemistry over this region. To achieve this we have performed a comprehensive set of sensitivity simulations to quantify the impact of BVOC emissions, chemistry, boundary layer mixing, and soil NO_x emissions on the simulated HCHO columns. Two different chemical mechanisms have been evaluated within the model, and the effects of OH recycling also examined.

[70] Validation of the model with in situ surface and aircraft measurements reveals GEOS-Chem has difficulty reproducing several observed chemical species. Simulated OH concentrations are generally underestimated, either because the isoprene emissions are too high and/or that too little OH is being generated and recycled. We find that use of the Caltech chemical scheme, which includes OH production via epoxide formation, or the inclusion of OH recycling does not fully resolve this problem. In addition, we find the model simulates high levels of isoprene and excessive of production of MVK+MACR and HCHO, which is potentially linked to their nighttime accumulation within the residual layer. Of the two chemical schemes tested, the Caltech scheme usually results in a more accurate simulation, based on the available in situ observations.

[71] In our analysis we find that magnitude of the model HCHO columns are dominated by the choice of chemical mechanism and bottom-up isoprene emission inventory; boundary layer mixing and soil NO_x emissions have a minor influence. Generally, GEOS-Chem exhibits a positive bias when the model HCHO columns are compared with observations from SCIAMACHY and OMI. Typically we find better agreement with SCIAMACHY than with OMI, although this may reflect differences in the respective retrieval schemes or be related to each instrument's overpass time. Simulations that use the Caltech chemical mechanism

and/or lowest isoprene emissions provide the best agreement to the satellite data, since they yield lower-HCHO columns. We find the LPJ-GUESS emissions produce HCHO columns closer in magnitude to the satellite observations, compared to simulations based on MEGAN. However, the lower LPJ-GUESS emissions may be partially offsetting other model errors.

[72] Despite these issues, GEOS-Chem can still be considered a valid tool for the inversion of HCHO columns measurements if used with care. As we have shown, the HCHO column is very sensitive to its vertical distribution within the boundary layer: a 20–30% model bias can result only from 1 ppbv uncertainty within the lowest 1 km. Given the uncertainty of tropical emissions and chemistry, such errors are easily feasible. Furthermore, inaccuracies of the forcing meteorology can have a significant impact on the agreement between the model the satellite data; a small 2 K temperature error can worsen the bias by about 25%. Owing to the differences between the two satellite products, and the large variation of the simulated HCHO columns, we suggest that top-down emission for the Amazon region cannot be derived using only one choice of model setup and instrument. Instead we believe an ensemble of top-down estimates is required that utilizes data from both emission inventories and both satellite data sets. Furthermore, the uncertainties of top-down emissions with respect to individual model processes need to be properly quantified.

[73] Ongoing examination of GEOS-Chem is necessary to resolve problematic issues, such as the low levels of OH, or the unusual nighttime chemistry within the residual layer. In a follow up study we will examine the diurnal evolution of chemistry, boundary layer mixing and dry deposition within the Amazon rainforest in much greater detail. However, we stress the need for observational data sets that span the lower troposphere, i.e., that provide day and night measurements of emission fluxes and key species both within and above the canopy, and out into the boundary layer and free troposphere. Whilst we acknowledge the many difficulties and expense of making such measurements, without such data it will be difficult to adequately constrain model simulations of the Amazon. Since CTMs and satellite observations are the only tools to study the reactive exchange of the Amazon Basin as a whole, reducing uncertainties in both approaches by using observations from detailed field campaigns undoubtedly represents a worthwhile investment.

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