Simulating organic species with the global atmospheric chemistry general circulation model ECHAM5/MESSy1: a comparison of model results with observations


Air Chemistry Department, Max-Planck Institute of Chemistry, P.O. Box 3060, 55020 Mainz, Germany

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Correspondence to: A. Pozzer (pozzer@mpch-mainz.mpg.de)
Abstract

The atmospheric-chemistry general circulation model ECHAM5/MESSy1 is evaluated with observations of different organic ozone precursors. This study continues a prior analysis which focused primarily on the representation of atmospheric dynamics and ozone. We use the results of the same reference simulation and apply a statistical analysis using data from numerous field campaigns. The results serve as a basis for future improvements of the model system. ECHAM5/MESSy1 generally reproduces the spatial distribution and the seasonal cycle of carbon monoxide (CO) very well. However, for the background in the northern hemisphere we obtain a negative bias (mainly due to an underestimation of emissions from fossil fuel combustion), and in the high latitude southern hemisphere a yet unexplained positive bias. The model results agree well with observations of alkanes, whereas severe problems in the simulation of alkenes are present. For oxygenated compounds the results are ambiguous: The model results are in good agreement with observations of formaldehyde, but systematic biases are present for methanol and acetone. The discrepancies between the model results and the observations are explained (partly) by means of sensitivity studies.

1 Introduction

Ozone chemistry in the troposphere is highly dependent on precursor species like NO$_x$ (=NO+NO$_2$), CO, methane and non-methane hydrocarbons (NMHC). These trace gases not only play an important role in ozone formation but they also control hydroxyl radicals HO$_x$ (=OH+HO$_2$) through many complex reaction cycles (Atkinson, 2000; Logan, 1985; Houweling et al., 1998; Seinfeld and Pandis, 1997). Their reactions are strongly interconnected, and the lifetimes of the trace gases range from seconds to years. Three-dimensional (3-D) global models which calculate both transport and chemistry are required to study and/or predict the distribution and the temporal development of these species.
Here we evaluate ECHAM5/MESSy1 (further denoted as E5/M1) with data from numerous field campaigns and measurement stations, focusing on CO and NMHCs, notably alkanes, alkenes, and a selection of oxygenated compounds.

After a brief introduction of the model setup and the observational datasets (Sect. 2), we present an overview of the reference simulation and the ability of the model to reproduce the observations (Sect. 3). A more detailed analyses of specific species (Sects. 4 to 7) follows. In the course of our analyses we deduce several hypotheses to explain the discrepancies between our model results and the observations. These hypotheses are subject of sensitivity studies, which we discuss (Sect. 8) as a basis of our conclusions (Sect. 9).

2 Model and observations

2.1 Model description and setup

E5/M1 is a combination of the general circulation model ECHAM5 (Roeckner et al., 2006) (version 5.3.01) and the Modular Earth Submodel System (MESSy, version 1.1). The implementation follows the MESSy standard (Jöckel et al., 2005). A first description and evaluation of the model system has recently been published (Jöckel et al., 2006). More details about the model system can be found at http://www.messy-interface.org, where a comprehensive description of the model is provided.

The results evaluated here are from the reference simulation S1, as described by Jöckel et al. (2006). The simulation period covers almost 8 years from January 1998 to October 2005. For our analysis we are focusing on the year 2000, which is expected to be represented by the model with the highest consistency, mainly because the chosen emission setup of primarily emitted species was compiled for this year. We applied the anthropogenic emissions from the EDGAR database (version 3.2 “fast-track”, van Aardenne et al., 2005) for the year 2000 as described by Ganzefeld et al. (2006).
Dry and wet deposition processes have been extensively described by Kerkweg et al. (2006a) and Tost et al. (2006a), respectively, while the emission procedure has been explained by Kerkweg et al. (2006b). The chemistry is calculated with the MECCA submodel by Sander et al. (2005). The applied spectral resolution of the ECHAM5 base model is T42, corresponding to a horizontal resolution of the quadratic Gaussian grid of $\approx 2.8^\circ \times 2.8^\circ$. The applied vertical resolution is 90 layers of which about 25 are located in the troposphere. The model setup includes feedbacks between chemistry and dynamics via the radiation calculations.

The model dynamics has been weakly nudged (Jeuken et al., 1996; Jöckel et al., 2006; Lelieveld et al., 2006) towards the analysis data of the ECMWF operational model (up to 100 hPa) in order to represent the realistic meteorology in the troposphere. This allows a direct comparison with observations.

2.2 Observations

For our comparison we applied two types of data sets: aircraft and surface measurements. Although the aircraft measurements (Emmons et al., 2000) cover only limited periods, they provide valuable information about the vertical distribution of the analysed trace gases. The surface measurements consist of a large number of multi-year surface observations collected from the literature (Solberg et al., 1996). They, in contrast to the aircraft observations, have limited spatial extension, though they cover an entire (climatological) year and are well suited for the analysis of the seasonal cycle. An additional important global dataset of surface measurements is the NOAA/CMDL flask network (Novelli et al., 1998) which encompasses multiple years of CO measurements.

Both dataset types are important for a meaningful evaluation of the model. The aircraft measurements are compared only with the year 2000 of the model simulation, while the surface measurements are compared with climatological monthly averages of the model results. The NOAA/CMDL flask measurements are compared with model calculated monthly averages.

For a quantitative statistical analysis, correlations between the model results and the
aircraft observations are calculated with respect to the altitude, while the correlations between the model results and the surface measurements are calculated with respect to time.

3 Overview of the results

To first provide an overview of the overall model performance regarding important O$_3$ precursors, we statistically compare model results and observations of the following species: alkanes and alkenes with up to three carbon atoms, oxygenated compounds (methanol, acetone, formaldehyde, acetaldehyde), hydroperoxide and isoprene. Aircraft observations are additionally compared to model results for methane, ozone and nitric acid.

3.1 Aircraft measurements

Table 1 summarises the comparison of E5/M1 model results with aircraft measurements. Figure 1 depicts the corresponding Taylor diagram, visualising the quality of the simulation for several tracers in a single diagram. It shows the correlation coefficient between model results and observations ($R$) by the angle to the ordinate. The standard deviation of the model normalised to the standard deviation of the observations ($\sigma_{\text{model}}/\sigma_{\text{obs}}$) is the distance from the origin. The observations are therefore located at a correlation of 1 and a normalised standard deviation of 1. The better a model reproduces the observations, the closer are the resulting points located to this “ideal” point. Detailed explanation of this diagram has been present by Taylor (2001).

Correlations and biases have been also calculated using an “uncertainties” weight, maintaining the relationship between the three statistical quantities visualised in the Taylor diagram. The uncertainties weight is calculated with the geometric sum of “model error” and “instrument error”. The model error is the standard deviation from the averaged output values, and the measurement error is a combination of instrumen-
tal errors and variance. We refer to Jöckel et al. (2006, Appendix D), for a detailed explanation of this approach. The results of this recalculation are shown in Fig. 2 and listed in Table 2.

According to this analysis, the discrepancies between model results and measurements are smaller than the uncertainties if the absolute value of the weighted bias (i.e., in units of the normalised standard deviation, Fig. 2 and Table 2) for a specific tracer is less than one. This implies that the observations are generally well reproduced by the model with the exception of \( \text{C}_2\text{H}_4 \), \( \text{C}_3\text{H}_6 \), \( \text{CH}_3\text{COCH}_3 \), \( \text{CH}_3\text{OH} \) and PAN. \( \text{C}_3\text{H}_6 \) and \( \text{CH}_3\text{OH} \) are not even shown in Fig. 2; due to their very high normalised standard deviations they are outside the shown range. The inability of the model to reproduce the vertical distribution of these compounds, as indicated by the relatively low correlation with all aircraft measurements included in the database, requires a more detailed analysis. This will be conducted in Sect. 5–7.

3.2 Surface measurements

Figures 3–4 and Tables 3–4 summarise the comparison between the model simulation (climatological monthly averages of the 7 years 1998–2004) and the climatology derived from the station measurements.

As can be seen in Fig. 3 and Table 3, the model generally overestimates the mixing ratios of these trace gases at the surface, with the exception of acetone (\( \text{CH}_3\text{COCH}_3 \)) and formaldehyde (HCHO). As further shown in Fig. 4 and Table 4, only the biases of ethene (\( \text{C}_2\text{H}_4 \)), propene (\( \text{C}_3\text{H}_6 \)) and PAN exceed one normalised standard deviation, and consequently the discrepancy between the model results and the observations cannot be explained by the model variability and/or uncertainties of the observations. Furthermore, Fig. 3 shows that the amplitude of the seasonal cycle is underestimated for many NMHCs, since the absolute value of the normalised standard deviation is lower than 1. We hence infer that the model is able to reproduce (with the exception of the aforementioned trace gases) the observed magnitude of the tracer mixing ratios and the phase of the seasonal cycle, though with a generally underestimated
Nevertheless, the model underestimates NMHCs in comparison to the aircraft measurements (Table 1), mainly in the upper troposphere. Jöckel et al. (2006) showed that the global air-mass weighted average OH abundance in the middle and upper troposphere in the simulation is lower than that calculated by Spivakovsky et al. (2000). This implies that the oxidation reactions are probably not responsible for the underestimation of NMHCs. Rather, there are indications that the convection scheme applied in the model does not sufficiently transport these species to the upper troposphere (Tost, 2006; Tost et al., 2006b).

4 Carbon monoxide, CO

Carbon monoxide provides the most important sink for OH (Lelieveld et al., 2002; Logan et al., 1981; Thompson, 1992). A correct simulation of this tracer is very important for studies of atmospheric oxidants. The emissions of CO applied in the present simulation have been described by Ganzeveld et al. (2006), and references therein). There are large uncertainties with respect to the amount of CO globally lost by dry deposition (e.g., 115–230 Tg/yr, Sanhueza et al., 1998, 540±430 Tg/yr, Moxley and Cape, 1997, 150 Tg/yr, von Kuhlmann et al., 2003b). In a recent study Horowitz et al. (2003) estimated the global dry deposition of CO to be only around 2 Tg/yr. Following this study, in our model simulation the dry deposition of CO was switched off. Although the simulated CO will be influenced by taking into account the process of dry deposition, the effect is potentially significant for the budget only in very remote regions. We will investigate this hypothesis further in Sect. 8.

Thanks to the large dataset of CO observations available from the NOAA/CMDL network (Novelli et al., 1998), a more detailed analysis is possible for CO than for the other studied trace gases. This allows the direct comparison of 7 years (1998–2004) of monthly averages model results with the corresponding observations. It cannot be expected that the model simulation fully reproduces the inter-annual variability due to
the prescribed climatological emissions for the year 2000. Nevertheless, the meteorological inter-annual variability is included through the applied nudging procedure. The correlation between the model results and the observations is generally good (see Sect. 3.2) with $R^2 = 0.67$.

As evident in Fig. 5, where data from selected locations are shown, the phase of the seasonal cycle of CO is well reproduced (e.g., Alert, Canada (ALT), or Mace Head, Ireland (MID)). The Taylor diagram in Fig. 6 completes the picture and confirms the high correlation between the model results and the measurements. The correlation is high at remote locations (south of 60° S) with $R \approx 0.9$, indicating that the main processes controlling the CO abundance are well represented by the model. However, at locations between 20° N and 40° N the correlation is lower, because these regions are strongly influenced by local emissions, including industry and biomass burning. Even though the amplitude of the seasonal cycle is well reproduced by the model at some locations (e.g., Mauna Loa, Hawaii, (MLO)), it overall tends to be underestimated (see Fig. 6, normalised standard deviation $\leq 1$). In the northern hemisphere (e.g., Zeppelin, (ZEP) or Alert, Canada, (ALT)) the maximum during winter seems to be systematically underestimated.

Looking further at the overall biases of the model simulation compared to the observations, some of the discrepancies between model results and observations can be resolved. Fig. 7 depicts the relative biases at all NOAA/CMDL stations. Strong positive biases are present mainly in polluted regions where the model resolution is not sufficient to resolve the details of the source distributions and the meteorology. As pointed out by Haas-Laursea and Hartley (1997), the flask samples have been collected under non-polluted conditions, i.e., for stations close to local sources only certain wind directions have been selected to avoid local contamination. The model results are not filtered in the same way, and at the rather low model grid resolution, local sources are sometimes located in the same grid box as the measurement station. Therefore, the simulated mixing ratios are potentially higher than the observed. Sampling the simulated data one grid-box upwind of the polluted locations, the correlation between model
results and observations increases drastically and the overestimation by the model almost disappears (Fig. 5).

In contrast to polluted regions, the model tends to somewhat underestimate CO compared to observations in remote regions of the northern hemisphere (Fig. 7). This can probably be attributed to underestimated anthropogenic emissions, which relatively strongly influence the background mixing ratio in the northern hemisphere. In fact, in the present simulation, as shown in Table 6, CO emissions from fossil fuel usage are at the lower end of the range of estimates in the literature.

In the southern hemisphere the model simulation produces higher mixing ratios of CO than observed (Fig. 7). This is particularly evident for locations south of 50° S. This significant bias is especially visible in Fig. 5, for Palmer station, Antarctic (PSA), and Halley Bay, Antarctic (HBA). This discrepancy is present in many other models (Hauglustaine et al., 1998; Wang et al., 1998; Bey et al., 2001; von Kuhlmann et al., 2003b; Horowitz et al., 2003; Park et al., 2004b; Folberth et al., 2006) and unexplained so far.

The hypothesis of underestimated emissions in polluted regions (i.e., primarily from fossil fuel usage) is supported by the analysis of the vertical profiles from the aircraft observations. Figure 8 (TRACE-P, China or PEM-WEST-B, China) shows that CO is clearly underestimated near China, especially in the planetary boundary layer, where the influence of the emissions is largest. This underestimation is also present further downwind (PEM-WEST-B, Philippine Sea, TRACE-P, Guam), however, it almost disappears in the central Pacific region (TRACE-P, Hawaii). Interestingly, the correct representation of carbon monoxide in East Asia is a problem for many atmospheric chemistry models. Kiley et al. (2003) demonstrated that many models are underestimating CO in the western Pacific region. Wang et al. (2004, and references therein) performed an inverse modelling analysis and calculated that an increase of the CO emissions in East Asia of around 45% from the a priori estimate (Streets et al., 2003) is required to match the observations. More recently, Streets et al. (2006) estimated 116 Tg/yr for the year 2000 and 157 Tg/yr for the year 2001 of CO emissions from
China with an uncertainty of 68%.

5 Non-methane Hydrocarbons

Comparison of the simulated non-methane hydrocarbon mixing ratios with observations yields a dual picture (see Sect. 3). Some of the simulated trace gases are in particularly good agreement with the measurements (e.g., propane \(\text{C}_3\text{H}_8\)), while others are largely uncorrelated with measurements (e.g., ethene \(\text{C}_2\text{H}_4\) and propene \(\text{C}_3\text{H}_6\)). In the case of \(\text{C}_3\text{H}_6\), the simulation does not reproduce the observed profiles. The simulated values are completely out of range of the observed values, even including measurement uncertainties and variability (see bias in Sect. 3). Moreover, the simulated and observed vertical profiles are uncorrelated (see Fig. 2), i.e., the model is unable to reproduce the shape of the profiles.

5.1 Alkanes (Ethane \(\text{C}_2\text{H}_6\) and Propane \(\text{C}_3\text{H}_8\))

Among all considered NMHCs, the alkanes are best reproduced by the model. The overall correlation \((R^2 > 0.75, \text{see Table 1})\) indicates a very good agreement between these simulated trace gases and the respective observations. Table 2 shows that the model results are well within the range of the measurements.

In the case of ethane, the improvement in the results compared to other models is mainly due to different spatial distribution patterns of the emissions. For example, the total emission of \(\text{C}_2\text{H}_6\) due to biomass burning in E5/M1 (Ganzeveld et al., 2006) is about 0.87 Tg/yr lower than in von Kuhlmann et al. (2003a), and the anthropogenic emissions are about 0.75 Tg/yr higher. Although the total is essentially unchanged, the different distribution improves the quality of the simulation. A good agreement with aircraft observations from the TOPSE campaign is achieved (Fig. 9), even though the model is at the lower end of the measurement range. The anthropogenic emissions in the model, in fact, are not sufficient to perfectly match the observed values.
This is clearly visible when the model results are compared to surface measurements (Fig. 10). Ethane is underestimated by the model at the surface in North America (Fraserdale, Lac la Flamme) mainly due to an underestimation of the emissions compiled in the EDGAR database (Jacob et al., 2002; Poisson et al., 2000). Furthermore, the amplitude of the seasonal cycle is not well reproduced at these locations, with problems mainly in reproducing the maximum in winter. The simulation reproduces biomass burning plumes observed in the TRACE-A campaign (African coast, Fig. 9). This campaign took place in the dry season of the southern hemisphere and some flight measurements were influenced by biomass burning. Problems occur in the upper troposphere, where the model underestimates the \( \text{C}_2\text{H}_6 \) mixing ratio by a factor of 2. Pickering et al. (1996) report that convection frequency during this campaign was unusually high, which could explain the disagreement between model and observations.

For propane, from Tables 1 and 2 we infer that the simulated vertical profiles are in good agreement with the observations (Fig. 11). This agreement is mainly achieved by the realistic representation of the emissions. As pointed out by Wang and Zeng (2004), an increase of 14±5% of the emission inventory used by Bey et al. (2001) (9.66 Tg/yr) was required to correctly match the observations, for an emission total of 11 Tg/yr. In our simulation the total emission was 11.97 Tg/yr (see Table 5). This amount is still lower than the suggested values present in literature (e.g., 13.46 Tg/yr calculated by Jacob et al., 2002).

However, even though the emissions are in the suggested range, the simulated mixing ratios are at the lower end of the measurement range (Fig. 11 and Table 1).

### 5.2 Alkenes (Ethene \( \text{C}_2\text{H}_4 \) and Propene \( \text{C}_3\text{H}_6 \))

The alkenes, in contrast to the alkanes, are generally poorly reproduced by the model. The simulated mixing ratios are usually below the measurements (Table 1). Ethene (\( \text{C}_2\text{H}_4 \)) has a poor correlation, both, with surface and aircraft measurements (Tables 2 and 4). This low correlation is due to an overestimation of the mixing ratio...
at the surface (Fig. 12), where the seasonal cycle is not reproduced with a peak in the mixing ratio during summer. The vertical profiles (Fig. 13) are mostly high biased (e.g., TOPSE-Mar, Churchill) with the largest differences between model results and observations occurring at the surface. In remote regions, where the direct influence of emissions is lower, the model is, nevertheless, at the lower end of the range of observations, with frequent underestimates (TRACE-A, Brazil Coast and South Atlantic).

The emissions from oceans appear too high (see Fig. 13, PEM-Tropics-B, Fiji) and a reduction in the model likely improves the simulation of ethene.

Propene is also not very well simulated by the model. The low correlation (Table 1) indicates a wrong representation of the vertical profiles. This poor representation cannot be reconciled with the variability of the model or measurement uncertainties, since the bias in Table 2 is much larger than one standard deviation. From Fig. 14, the erroneous description of this trace gas in the model is evident. The very high mixing ratio in the boundary layer (2–3 times the observed one) in open ocean regions (PEM-TROPICS-A, all locations) indicates a too strong emission of this tracer from the ocean. In E5/M1 the upper limit of the suggested emission from Bates et al. (1995) has been applied (1.27 Tg/yr).

This overestimation is not so evident in continental regions more strongly influenced by anthropogenic (Fig. 14, PEM-WEST-B, Japan or Fig. 15, Kosetice, Lac la Flamme) or biomass burning sources (Fig. 14, TRACE-A, West Africa Coast). Figure 14 shows that outside the planetary boundary layer (PBL), above 2–3 km, the simulated tracer is nearly depleted, in contrast to the observations.

This points to a wrong simulation of the sinks (too fast) which could explain the low values simulated outside the PBL. This tracer is not subject to either wet (due to its low solubility) nor dry deposition and is only removed by reactions with OH, NO₃, and O₃, the latter two at least two orders of magnitude slower than the first.

The rate coefficient used for the reaction C₃H₆ + OH is taken from the IUPAC recommendation of 1999 (Atkinson et al., 1999, and references therein). A new recommendation suggests a slightly different dependence of the high pressure channel on
temperature (Atkinson et al., 2005; Vakhtin et al., 2003, and references therein). As shown in Fig. 16, the new reaction coefficient is lower than the previously estimated. This might explain the strong depletion in the free troposphere of this gas in our simulation, or vice versa, our results indirectly support the revision of the reaction coefficient, as tested in Sect. 8.

6 Isoprene

The simulated isoprene emission flux, calculated for the year 1998, is 580 TgC. This is about 80 TgC larger compared to the offline calculated inventory by Guenther et al. (1995), which formed the basis for the implementation of the online calculations of biogenic NMHC emissions in E5/M1 (Ganzeveld et al., 2006; Kerkweg et al., 2006b). One reason for this significantly larger global isoprene emission flux is a large source over Amazonia, where, especially in the dry season, too high simulated surface temperatures result in a significantly larger flux from this region compared to a climatological simulation. Further analysis has revealed that simulation of the too high temperatures is mainly due a misrepresentation of the hydrological cycle of Amazonia indicated by low soil moisture and suppressed evapotranspiration. This is a feature of the nudged model simulations with ECHAM5 as noticed in a previous meteorological analysis (Bengtsson et al., 2004). Our analysis thus reveals the consequence of misrepresentation of the Amazonian dry season meteorology and hydrology for atmospheric chemistry. It has been established in various global scale atmospheric chemistry studies that using the Guenther et al. (1995) isoprene inventory of 500 TgC/yr results in a significant overestimation of the simulated tropical Planetary Boundary Layer (PBL) mixing ratios (e.g., Houweling et al., 1998). Consequently, a commonly applied approach in atmospheric chemistry studies, which do not focus on isoprene, is to use a substantially smaller global flux, ranging from 220 to 350 TgC/yr (Brasseur et al., 1998; von Kuhlmann et al., 2004). Since the main focus of the conducted simulation with E5/M1 has been the evaluation of the global ozone burden and mixing ratios, isoprene
fluxes have also in this study been scaled to achieve a reduced global annual emission flux of about 350 TgC/yr. This scaling has been based on the calculated emission burden for the year 1998. This appeared to be a particular year, with a substantially larger flux compared to the following years 1999–2004. Consequently, the actually applied isoprene emission fluxes in the presented model simulation range from 305 TgC/yr to 325 TgC/yr for the period 1999–2004.

We compare the simulated isoprene mixing ratios with the same observations as presented by von Kuhlmann et al. (2004), complemented with additional observations in South America, Africa, Australia and Europe.

Figure 17, which shows simulated versus observed mixing ratios, indicates that the simulated and observed isoprene mixing ratios do not correlate. However, despite the significantly reduced global annual source strength, evaluation of the range in the observed and simulated isoprene mixing ratios indicates that the model calculates maximum isoprene mixing ratios that are still significantly higher compared to the observations.

This is illustrated in Fig. 18, which shows the simulated seasonal cycle in surface layer isoprene mixing ratios over Amazonia (2° N, 60° W). Especially in the dry season (i.e., September, October) maximum simulated isoprene mixing ratios as large as 10–12 nmol/mol do not only reflect suppressed turbulent mixing conditions (early morning and late afternoon) but also the previously discussed too high surface temperatures due to underestimated evapotranspiration.

7 Oxygenated compounds

Oxygenated compounds are partly soluble and are influenced by wet deposition (Tost et al., 2006a) (e.g. CH₃OOH), as well as oxidation by OH, and some of them by photolysis (HCHO, CH₃CHO, CH₃OOH and CH₃COCH₃). Reproducing the vertical profiles of these tracers is therefore challenging, and the identification of the cause of discrepancies between model results and observations is difficult.
7.1 Hydrogen peroxide, $\text{H}_2\text{O}_2$

Hydrogen peroxide is produced by the self reaction of $\text{HO}_2$ and can photolyse to produce $\text{OH}$. Hence it is useful as indicator of $\text{HO}_x$ in the troposphere.

As shown in Fig. 19, the vertical profiles of $\text{H}_2\text{O}_2$ are reasonably well reproduced by the model at different locations and for different field campaigns. Discrepancies from the observations are difficult to define due to its high variability. In fact, from Table 2, the correlation (uncertainty weighted calculation) between model results and measurements is very high ($R^2 \approx 0.94$), mainly because the observed $\text{H}_2\text{O}_2$ shows a very high variability.

7.2 Acetaldehyde, $\text{CH}_3\text{CHO}$

This oxygenated compound is produced from the oxidation of a variety of hydrocarbons (Lewis et al., 2005) and it can produce $\text{HO}_x$ and PAN precursors.

The analysis shows that the model results agree within a factor of two (overestimation) with the observations, although the seasonal cycle is not well reproduced (Fig. 20). This is confirmed by Table 4, which shows that at the surface the amplitude of the seasonal cycle is correct (if we consider the uncertainties), but with the wrong phase ($R^2 \approx 0.298$).

Only a few aircraft measurements are available of this trace gas. Williams et al. (2001) reported a mixing ratio in Suriname in the boundary layer of about 1.7 nmol/mol and 0.8 nmol/mol in the free troposphere. The model underestimates these mixing ratios by a factor of 10. During the PEM-TROPICS-B campaign (Singh et al., 2001) over the open ocean mixing ratios between 60 and 100 pmol/mol have been measured, about 4 times higher than simulated. Singh et al. (2001) postulate some sources from the ocean (Zhou and Mopper, 1997), though more measurements are required to better constrain the abundance of this tracer.
7.3 Methanol, CH$_3$OH

Even though methanol is one of the most abundant organic trace gases in the atmosphere, its global cycle is not well understood (Heikes et al., 2002; Singh et al., 2000; Jacob et al., 2005). It influences the upper tropospheric photochemistry because of its oxidation to formaldehyde (Palmer et al., 2003; Singh et al., 2003; Jacob et al., 2005). Unfortunately, the distribution and magnitude of the sources and sinks are largely unknown (Tie et al., 2003; Singh et al., 2004; Galbally and Kirstine, 2002). Due to these high uncertainties, emissions from the ocean have not been included in our simulation, though indicated by previous studies (Jacob et al., 2005; Heikes et al., 2002).

Compared to other tracers, relatively few observations of methanol are available (PEM TROPICS-B and TRACE-P campaigns). As noted in Sect. 3, the correlation between model results and measurements is very low ($R^2=0.313$), and the vertical profiles are not reproduced by the model (Figs. 2 and 21). Wet deposition has not been taken into account for this tracer (Tost et al., 2006a), however, only small differences of $\approx 5\%$ are expected by accounting for this process (Heikes et al., 2002; Galbally and Kirstine, 2002; von Kuhlmann et al., 2003b). The simulated total dry deposition of 42.16 Tg/yr is well within the suggested values in the literature (e.g. 35–210 Tg/yr, Heikes et al., 2002, 11–43 Tg/yr, Galbally and Kirstine, 2002, 32–85 Tg/yr, Tie et al., 2003, 37 Tg/yr, von Kuhlmann et al., 2003b and 55 Tg/yr, Jacob et al., 2005). Thus, underestimated emission fluxes or incomplete description of the chemistry are more likely responsible for the wrong representation. A total emission of 77 Tg/yr for CH$_3$OH has been used, as suggested by EDGAR (with 60 Tg/yr of biogenic emissions). Note, however, that a recent study by Jacob et al. (2005) recommends a total emission of 168 Tg/yr with 151 Tg/yr of biogenic emissions.

7.4 Formaldehyde, HCHO

Although formaldehyde is chemically formed by methanol, the low production rate from CH$_3$OH + OH $\rightarrow$ HCHO + HO$_2$ ($k\approx5\times10^{-13}$ cm$^3$ molec$^{-1}$ s$^{-1}$) indicates that this reac-
tion is not essential for a realistic simulation of formaldehyde and that the wrong representation of methanol is not strongly influencing HCHO (Stickler et al., 2006). The model, in fact, predicts the observed mixing ratios of this trace gas very well (Fig. 22 and Sect. 3).

The simulation is in good agreement with station measurements, and the amplitude of the seasonal cycle is well reproduced at the surface (Figs. 3 and 4), although with a smaller amplitude than observed.

From the comparison with the aircraft measurements (Fig. 22 and Sect. 3) we conclude that the simulation of HCHO satisfactorily reproduces the observations. Particularly good agreement is obtained for the PEM-TROPICS-B field campaign, where more than 20 measurements per location are available. A systematic underestimation of the measurements is observed at the surface for the TOPSE-A campaign, only for the location Thule (Fig. 22). This can potentially be explained by the absence of emissions of HCHO from snow (Riedel et al., 2005, and references therein). Other models (Fried et al., 2003) also simulate large disagreements compared to measurements for the TOPSE-A campaign.

7.5 Acetone, CH$_3$COCH$_3$

Acetone (CH$_3$COCH$_3$) plays a significant role in the upper tropospheric HO$_x$ budget due to its photolysis (Singh et al., 1995; McKeen et al., 1997; Müller and Brasseur, 1995; Wennberg et al., 1998; Jaeglé et al., 2001). Moreover, this tracer is essential to correctly describe the ozone enhancement in flight corridors (Brühl et al., 2000; Folkins and Chatfield, 2000).

Comparing this simulated tracer with the station observations, we conclude that at the surface this tracer is well simulated. Problems in reproducing the seasonal cycle are evident for only two locations (see Fig. 24), Zeppelin and Ispra, but Table 3 shows that acetone has one of the highest correlation coefficients and lowest biases compared to other tracers. However, the annual cycle seems to be underestimated (see Figs. 3 and 4). Folberth et al. (2006) suggested biogenic emissions of 55.93 Tg/yr,
more than the 42 Tg/yr used in this simulation.

As also seen for HCHO, acetone is not the only species with which we have problems at the Ispra location, and we cannot rule out that this site is unrepresentative for the area of one model grid-box (von Kuhlmann et al., 2003b).

Although acetone is well reproduced at the surface, large discrepancies with observations are present in the vertical profiles. As shown in Sect. 3, the correlation with the aircraft measurements is low ($R^2 = 0.385$), and the vertical profile is not reproduced (bias in Fig. 2). Large discrepancies with observations (Fig. 25) are present in the free troposphere. A comparison of simulated acetone with aircraft measurements is shown in Fig. 26, where the colour code indicates the vertical distribution of the measurements and the corresponding model results.

The differences between model results and observations appears to increase with altitude. The shape of the simulated vertical profiles suggest a potential misrepresentation of the photolysis of this tracer. Measurements made by Blitz et al. (2004) indicate that the quantum yield (and therefore the photolysis rate) of acetone is lower than previously assumed. A reduction would increase the mixing ratio of this tracer and the simulation would then agree better with the observations. Furthermore, as shown by Arnold et al. (2004), this new quantum yield may significantly change the contribution of CH$_3$COCH$_3$ to the HO$_x$ budget. Arnold et al. (2005) showed that the new quantum yield decreases the global loss of acetone by a factor of $\approx 2$ and by 80–90% in the cold upper troposphere.

The wrong vertical distribution of acetone in our model simulation can also explain the poor correlation of the simulated PAN with the aircraft measurements. As for acetone, the correlation with surface measurements is high, while the simulated vertical profiles strongly deviate from the observed profiles. However, it has to be stressed that the PAN mixing ratios are generally overestimated by the model simulation (Table 2), in contrast to acetone, for which the mixing ratio is systematically underestimated by the model.
8 Sensitivity studies

To test the hypotheses posed so far, three additional simulations have been performed for the year 2000. The first simulation (denoted as $S1a$) is based on the reference simulation $S1$, presented by Jöckel et al. (2006), with the following modifications (note that the numbers correspond to the hypotheses above):

1. Simulation of dry deposition of CO, based on few available measurements (Conrad and Seiler, 1985; Sanhueza et al., 1998). We used a constant soil deposition velocity of 0.04 cm/s limited to regions where the temperature is higher than 5 °C and the relative humidity is higher than 40% in order to exclude cold tundra and desert soils (Lawrence et al., 1999);

2. Increase of the CO fossil fuel emission over China by 45% according to Kiley et al. (2003); The new total amount of CO emission is 1130 Tg/yr, with 314 Tg/yr due to fossil fuel usage;

3. Changed reaction coefficient for the reaction $C_3H_6 + OH$ as described in Sect. 5.2 (Atkinson et al., 2005);

4. Decrease of oceanic emission of $C_2H_4$ as suggested by Plass-Dülmer et al. (1995);

5. Increase of $CH_3OH$ biogenic emissions to 151 Tg/yr (Jacob et al., 2005);

6. Increase of biogenic emissions of $CH_3COCH_3$ to 55.93 Tg/yr following Folberth et al. (2006);

7. New photolysis of $CH_3COCH_3$ as described by Blitz et al. (2004), Arnold et al. (2004) and Arnold et al. (2005).

In two more sensitivity studies, both based on $S1a$, we test the importance of dry deposition. In simulation $S1b$, dry deposition of CO (point 1 of $S1a$) is switched off.
In simulation S1c, the simulation S1a has been repeated, with a changed order of emission and deposition processes in the operator splitting approach. In this special case the dry deposition processes were calculated before the tracer emissions.

Table 7 lists the correlation analysis between sensitivity simulation results and measurements. The same results are depicted in Fig. 27. With the exception of methanol, the model is performing better than in S1, although the improvement is, in general, not substantial.

8.1 Sensitivity studies S1a and S1b

The dry deposition of CO, as implemented in S1a, yields a total deposition of 74 Tg/yr. This is outside the range suggested in literature by Moxley and Cape (1997) and Sánchez et al. (1998) and strongly exceeds the dry deposition calculated by Horowitz et al. (2003) of 2 Tg/yr.

In Fig. 28, the relative changes between the sensitivity study S1a and the evaluation simulation (S1) are shown. For the remote regions the decrease in CO mixing ratio is very small (a few percent). No significant improvements of the simulation with respect to the measurements are obtained for remote locations and at the North America stations (see Fig. 29). From Fig. 29 we can also see that dry deposition of CO decreases the ability of the model to reproduce the CO mixing ratio at high northern latitudes (Zeppelin, (ZEP) and Alert, (ALT)).

In simulation S1a, the Chinese emissions of CO (bio-fuel and fossil fuel usage) have been additionally increased from the originally 80.81 Tg/yr to 116.9 Tg/yr. This is in agreement with the value of 116 Tg/yr calculated by Streets et al. (2006) based on the TRACE-P campaign for the year 2000. Despite this drastic increase of the CO emissions over China, the impact on the CO mixing ratios at the surface, over the ocean downwind is small (Fig. 28).

Although dry deposition is obviously higher where more CO is emitted (Fig. 30), from Table 7 we deduce that the impact of this process on CO is marginal. The simulation S1b, in fact, does not yield significantly different results compared to S1a. Moreover,
as seen in Fig. 31, $S1b$ does not show an improvement of the vertical profiles of CO compared to observations. Nevertheless, there is strong evidence that the Chinese emissions of CO are underestimated. However, the amount (and the geographic distribution) of the additional unknown sources are not clearly quantifiable (Fig. 31).

Furthermore, as shown in Table 7 and Fig. 29, the dry deposition of CO only weakly influences the results, and decreases the correlation between the model results and the measurements. We hence conclude that dry deposition of CO, as implemented for simulation $S1a$, is not suited for future studies.

The changed reaction rate for $C_3H_6$ slightly improves the correlation of this simulated trace gas (see Table 7, $R^2=0.427$) with aircraft observations. However, the model still simulates strong depletion in the free troposphere (see Fig. 32), and at relatively high temperatures (i.e. near the surface), there is virtually no difference between the simulations applying the two different reaction rates (Atkinson et al., 1999, 2005). The wrong vertical profile implies that the chemical processes related to this tracer are not completely understood; more studies on this tracer (especially on its reaction with OH) are required.

The decrease of oceanic emissions of $C_2H_4$ marginally improves the correlation at a few locations (see Fig. 33) over the ocean and in the marine boundary layer, but overall the resulting differences between $S1$ and $S1a$ are not significant ($R^2=0.413$ for $S1a$). We conclude that the major problem is the high uncertainty of the terrestrial $C_2H_4$ sources.

For methanol, Table 7 shows a decrease in the correlation between model results and observations for $S1a$ compared to $S1$. This is due to an increased scatter of the model results for $S1a$ compared to $S1$ (Fig. 34). Nevertheless, the overall bias is lower in $S1a$ (≈239 pmol/mol) than in $S1$. The dry deposition of methanol increased from 24.20 Tg/yr in the reference simulation $S1$ to 82.76 Tg/yr in the sensitivity simulation $S1a$. In comparison, Jacob et al. (2005) calculated 55 Tg/yr dry deposition of methanol and Heikes et al. (2002) and Tie et al. (2003) suggested the ranges 35–210 Tg/yr and 32–85 Tg/yr, respectively. In conclusion, we recommend to apply the increased
amount of biogenic methanol emissions (i.e., 151 Tg/yr) of the S1a simulation for future studies.

Additionally the modifications of the model setup for acetone, i.e., the altered photolysis rate calculation, between S1 and S1a improved the model results w.r.t. the observations. Fig. 35, for example, clearly shows for the TRACE-P campaign (Japan) that the results of the S1a simulation are closer to the observations than those of the S1 simulation. This is quantitatively confirmed by the correlation analysis between the different model results and the observations: for the sensitivity simulation S1a the correlation coefficient increased (from $R^2=0.385$ to $R^2=0.429$) and the overall bias decreased (from $-376.85$ to $-211.95$ pmol/mol) based on the reference simulation. However, compared with station data, the S1a simulation shows a decrease in the correlation and an increased bias (see Table 8). This implies that the value of $\approx 56$ Tg/yr suggested by Folberth et al. (2006) for the biogenic emissions is too high.

Discrepancies between the model results and the observations remain, especially over the oceans and in the upper troposphere (Fig. 35). This is probably related to an oversimplified representation of the deposition/emission of acetone from oceans, an issue under debate (Singh et al., 2001; Jacob et al., 2002; Singh et al., 2003; Marandino et al., 2005).

8.2 Sensitivity study S1c

Because we found a remarkably low sensitivity of CO mixing ratios to the emissions from China (see Sect. 8.1), we want to rule out potential systematic errors connected to the operator splitting as applied in our model. In the standard E5/M1 set up, emission tendencies are calculated before the dry deposition tendencies of trace gases. Therefore, we performed an additional sensitivity simulation (S1c) in which we changed the order of processes, such that the dry deposition tendency is calculated before emission tendencies and chemical tendencies. The model setup is the same as for S1a.

As shown in Table 9, the change in the annually averaged dry deposited tracer mass is 5% at maximum. From this we conclude that the systematic error due to the ap-
plied operator splitting is small (as for instance compared to uncertainties in emission estimates) and that the low sensitivity mentioned above is a robust result.

9 Conclusions

We presented the second part of the evaluation of the new atmospheric chemistry general circulation model ECHAM5/MESSy1 focusing on organic compounds, including CO, C$_2$H$_4$, C$_2$H$_6$, C$_3$H$_6$, C$_3$H$_8$, CH$_3$CHO, CH$_3$OH, HCHO, CH$_3$COCH$_3$, PAN and H$_2$O$_2$. We compared the model results with observational data obtained from aircraft campaigns and from various sampling stations. We applied regression analyses between the model results and the observations and summarise the results in Taylor diagrams for a quantitative statistical evaluation.

The seasonal cycle of carbon monoxide is well reproduced by the model, with a very high correlation with observations in remote regions. Our results support the recent finding (Horowitz et al., 2003) that dry deposition of CO has been overestimated in many modelling studies so far. The agreement of CO simulations with observations is best if the dry deposition of CO represents none or only a negligible contribution to the global CO budget. Although the seasonal cycle of CO is well reproduced, the model results in the northern hemisphere tend to be lower than the observations. This is presumably due to underestimated CO emissions from fossil fuel combustion in winter.

Overall, the model reproduces most of the observations of alkanes. The simulation of alkenes (C$_2$H$_4$ and C$_3$H$_6$), however, shows large discrepancies compared to observations, which could not be resolved by the conducted sensitivity studies applying reasonable changes in the model setup. The results strongly indicate that the terrestrial emissions of C$_2$H$_4$ are significant (though largely unknown) and that the photochemistry of C$_3$H$_6$ is not yet fully understood.

The quality of the model results for the oxygenated organic compounds is highly dependent on the specific species. While formaldehyde (HCHO) is very well reproduced, the correlations to observations of methanol (CH$_3$OH), acetaldehyde (CH$_3$CHO) and
acetone (CH$_3$COCH$_3$) are rather low. The few measurements available for acetaldehyde suggest that the model emissions should be increased by a factor of $\approx 5$, as noticed also by von Kuhlmann et al. (2003b). An increase of the methanol biogenic emissions to 151 Tg/yr decreases the discrepancies between model results and observations, however, the results are still not satisfactory.

An update of the photolysis rate calculation for acetone according to more recent estimates (Blitz et al., 2004) results in a higher correlation of model results with measurements, and partly solves the problem in the upper troposphere. However, the model calculated mixing ratios remain lower than the observations, especially over the Pacific Ocean. This supports the hypothesis of a missing oceanic source.

Finally, the simulation of isoprene remains challenging; model results and observations are uncorrelated. For this trace gas more work is clearly required (modelling and observations).

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**Table 1.** Summary of the correlation coefficients ($R$) and linear regression analyses of model results versus aircraft observations (model = $m \times$ measurement + $b$). Bias and $b$ are in pmol/mol; bias = model results minus observations.

<table>
<thead>
<tr>
<th>trace gas</th>
<th>num. obs.</th>
<th>bias</th>
<th>$m$</th>
<th>$b$</th>
<th>$R^2$</th>
</tr>
</thead>
<tbody>
<tr>
<td>C$_2$H$_4$</td>
<td>454</td>
<td>–23.87</td>
<td>0.26</td>
<td>9.975</td>
<td>0.409</td>
</tr>
<tr>
<td>C$_2$H$_6$</td>
<td>473</td>
<td>–174.03</td>
<td>0.69</td>
<td>78.692</td>
<td>0.799</td>
</tr>
<tr>
<td>C$_3$H$_6$</td>
<td>332</td>
<td>–11.50</td>
<td>0.14</td>
<td>0.267</td>
<td>0.410</td>
</tr>
<tr>
<td>C$_3$H$_8$</td>
<td>472</td>
<td>–18.82</td>
<td>0.92</td>
<td>–5.755</td>
<td>0.768</td>
</tr>
<tr>
<td>CH$_3$COCH$_3$</td>
<td>246</td>
<td>–376.85</td>
<td>0.42</td>
<td>–28.717</td>
<td>0.385</td>
</tr>
<tr>
<td>CH$_3$OH</td>
<td>116</td>
<td>–447.82</td>
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<td>255.18</td>
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<td>CH$_3$OOH</td>
<td>366</td>
<td>–13.19</td>
<td>0.71</td>
<td>94.598</td>
<td>0.718</td>
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<td>HCHO</td>
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<td>H$_2$O$_2$</td>
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<td>HNO$_3$</td>
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<td>63.115</td>
<td>0.337</td>
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<tr>
<td>O$_3$</td>
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<td>1835</td>
<td>1.78</td>
<td>–28464</td>
<td>0.544</td>
</tr>
<tr>
<td>PAN</td>
<td>395</td>
<td>141.97</td>
<td>0.71</td>
<td>188.99</td>
<td>0.268</td>
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<tr>
<td>CO</td>
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<td>0.51</td>
<td>36381</td>
<td>0.633</td>
</tr>
<tr>
<td>CH$_4$</td>
<td>334</td>
<td>–1103.6</td>
<td>0.66</td>
<td>588746</td>
<td>0.808</td>
</tr>
</tbody>
</table>
Table 2. Correlation coefficients and linear regression analyses between the model results and the aircraft observations, both weighted with relative uncertainties (bias = model results minus observations).

<table>
<thead>
<tr>
<th>trace gas</th>
<th>num. obs.</th>
<th>bias(^1)</th>
<th>bias(^2)</th>
<th>(R^2)</th>
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<td>C(_3)H(_6)</td>
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<td>-4818.76</td>
<td>-78982</td>
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<td>C(_3)H(_8)</td>
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<tr>
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<td>-648.6</td>
<td>0.767</td>
</tr>
<tr>
<td>CH(_3)OH</td>
<td>116</td>
<td>-2.03</td>
<td>-724.54</td>
<td>0.599</td>
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<tr>
<td>CH(_3)OOH</td>
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<td>10.72</td>
<td>0.941</td>
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<td>O(_3)</td>
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<td>0.14</td>
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<td>PAN</td>
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<td>1.17</td>
<td>205.08</td>
<td>0.999</td>
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<tr>
<td>CH(_4)</td>
<td>334</td>
<td>-0.74</td>
<td>-11117</td>
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</table>
Table 3. Summary of the correlation coefficients and linear regression analyses of model results versus station observations (model = \(m \times \text{measurement} + b\)). Bias and \(b\) are in nmol/mol; bias = model results minus observations.

<table>
<thead>
<tr>
<th>trace gas</th>
<th>num. obs.</th>
<th>bias</th>
<th>(m)</th>
<th>(b)</th>
<th>(R^2)</th>
</tr>
</thead>
<tbody>
<tr>
<td>C(_2)H(_4)</td>
<td>138</td>
<td>0.20</td>
<td>0.534</td>
<td>0.504</td>
<td>0.396</td>
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<td>C(_2)H(_6)</td>
<td>150</td>
<td>0.24</td>
<td>0.828</td>
<td>0.544</td>
<td>0.539</td>
</tr>
<tr>
<td>C(_3)H(_6)</td>
<td>137</td>
<td>0.01</td>
<td>0.658</td>
<td>0.060</td>
<td>0.502</td>
</tr>
<tr>
<td>C(_3)H(_8)</td>
<td>150</td>
<td>0.61</td>
<td>1.130</td>
<td>0.508</td>
<td>0.424</td>
</tr>
<tr>
<td>CH(_3)CHO</td>
<td>77</td>
<td>0.15</td>
<td>0.197</td>
<td>0.582</td>
<td>0.082</td>
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<tr>
<td>CH(_3)COCH(_3)</td>
<td>81</td>
<td>-0.08</td>
<td>0.528</td>
<td>0.459</td>
<td>0.508</td>
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<tr>
<td>HCHO</td>
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<td>CO(^3)</td>
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Table 4. Correlation coefficients and linear regression analyses between the model results and the station observations, both weighted with relative uncertainties (bias = model results minus observations).

<table>
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<tr>
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<th>bias (^5)</th>
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<tr>
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<tr>
<td>C(_3)H(_8)</td>
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<td>0.080</td>
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<td>CH(_3)COC(_3)</td>
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<td>−0.106</td>
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<td>−0.125</td>
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</tr>
<tr>
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<td>48</td>
<td>5.2</td>
<td>0.381</td>
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Table 5. Budget of different trace gases in the E5/M1 simulation (this work) compared to previous studies.

<table>
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<tr>
<th>tracer</th>
<th>emission (Tg/yr)</th>
<th>dry deposition (Tg/yr)</th>
<th>wet deposition (Tg/yr)</th>
<th>burden (Tg) annual average</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>this work</td>
<td>literature</td>
<td>this work</td>
<td></td>
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<tr>
<td>C₂H₄</td>
<td>26.35</td>
<td>19.55⁹–22.35¹</td>
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<td>C₂H₅</td>
<td>12.46</td>
<td>11.95⁹–12.26¹</td>
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<tr>
<td>C₃H₃</td>
<td>9.92</td>
<td>6.95⁹–9.87¹</td>
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<td>0.01</td>
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<tr>
<td>C₃H₈</td>
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<td>10.18⁴–13.46¹</td>
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<tr>
<td>CH₃COCH₃</td>
<td>47.96</td>
<td>95¹</td>
<td>24.48</td>
<td>2.23</td>
</tr>
<tr>
<td>CH₃OH</td>
<td>77.74</td>
<td>312¹</td>
<td>42.16</td>
<td>2.46</td>
</tr>
<tr>
<td>HCHO</td>
<td>7.78</td>
<td>6.97¹</td>
<td>43.35</td>
<td>1.05</td>
</tr>
<tr>
<td>H₂O₂</td>
<td>–</td>
<td>–</td>
<td>142.12</td>
<td>4.07</td>
</tr>
<tr>
<td>CO</td>
<td>–</td>
<td>–</td>
<td>375.93</td>
<td></td>
</tr>
</tbody>
</table>

[^a]: von Kuhlmann et al. (2003b)
[^b]: Jacob et al. (2002)
[^c]: Jacob et al. (2005)
[^d]: Heikes et al. (2002)
[^e]: Arnold et al. (2005)
[^f]: Horowitz et al. (2003)
[^g]: Park et al. (2004a)
[^h]: Folberth et al. (2006)
[^i]: Jacob et al. (2002), only terrestrial
Table 6. Different estimates of CO emissions in Tg/yr. The total includes also other sources (e.g. biogenic).

<table>
<thead>
<tr>
<th>Source</th>
<th>Fossil fuel</th>
<th>Biomass burning</th>
<th>Total</th>
</tr>
</thead>
<tbody>
<tr>
<td>Brasseur et al. (1998)</td>
<td>281</td>
<td>661</td>
<td>1218</td>
</tr>
<tr>
<td>Bey et al. (2001)</td>
<td>388</td>
<td>522</td>
<td>1043</td>
</tr>
<tr>
<td>von Kuhlmann et al. (2003a)</td>
<td>400</td>
<td>748</td>
<td>1261</td>
</tr>
<tr>
<td>Park et al. (2004a)</td>
<td>384</td>
<td>746</td>
<td>1131</td>
</tr>
<tr>
<td>this work</td>
<td>281</td>
<td>702</td>
<td>1096</td>
</tr>
<tr>
<td>range</td>
<td>300–600</td>
<td>300–900</td>
<td>656–1730</td>
</tr>
</tbody>
</table>
Table 7. Correlation coefficients and linear regression of the comparison of the sensitivity simulations with aircraft observations (model = \( m \times \text{meas} + b \)). Bias and b coefficient in pmol/mol. Bias = model-observations.

<table>
<thead>
<tr>
<th>tracer</th>
<th>num. obs.</th>
<th>bias</th>
<th>m</th>
<th>b</th>
<th>( R^2 )</th>
</tr>
</thead>
<tbody>
<tr>
<td>( \text{C}_2\text{H}_4(S1a) )</td>
<td>454</td>
<td>-24.782</td>
<td>0.26</td>
<td>9.31</td>
<td>0.413</td>
</tr>
<tr>
<td>( \text{C}_3\text{H}_6(S1a) )</td>
<td>332</td>
<td>-11.562</td>
<td>0.14</td>
<td>0.18</td>
<td>0.427</td>
</tr>
<tr>
<td>( \text{CH}_3\text{COCH}_3(S1a) )</td>
<td>246</td>
<td>-211.95</td>
<td>0.57</td>
<td>52.00</td>
<td>0.429</td>
</tr>
<tr>
<td>( \text{CH}_3\text{OH}(S1a) )</td>
<td>116</td>
<td>-239.05</td>
<td>0.28</td>
<td>380.84</td>
<td>0.273</td>
</tr>
<tr>
<td>( \text{CO}(S1a) )</td>
<td>456</td>
<td>-6526.8</td>
<td>0.53</td>
<td>36585</td>
<td>0.660</td>
</tr>
<tr>
<td>( \text{CO}(S1b) )</td>
<td>456</td>
<td>-4003.1</td>
<td>0.54</td>
<td>37864</td>
<td>0.651</td>
</tr>
<tr>
<td>( \text{CO}(S1c) )</td>
<td>456</td>
<td>-6470.1</td>
<td>0.53</td>
<td>36966</td>
<td>0.656</td>
</tr>
</tbody>
</table>
Table 8. Summary of the correlation coefficients and linear regression analyses of model results versus station observations (model = \( m \times \text{measurement} + b \)). Bias and \( b \) are in nmol/mol; bias = model results minus observations.

<table>
<thead>
<tr>
<th>CH$_3$COCH$_3$</th>
<th>bias</th>
<th>( m )</th>
<th>( b )</th>
<th>( R^2 )</th>
</tr>
</thead>
<tbody>
<tr>
<td>( S1 )</td>
<td>-0.08</td>
<td>0.528</td>
<td>0.459</td>
<td>0.508</td>
</tr>
<tr>
<td>( S1a )</td>
<td>-0.15</td>
<td>0.67</td>
<td>0.526</td>
<td>0.435</td>
</tr>
</tbody>
</table>
### Table 9.
Annual average dry deposition of different tracers as calculated by the different sensitivity simulations.

<table>
<thead>
<tr>
<th>Tracer</th>
<th>S1a (Tg/yr)</th>
<th>S1c (Tg/yr)</th>
<th>Difference</th>
</tr>
</thead>
<tbody>
<tr>
<td>CO</td>
<td>74.0</td>
<td>73.6</td>
<td>-0.5</td>
</tr>
<tr>
<td>CH₃OH</td>
<td>82.76</td>
<td>78.14</td>
<td>-5.5</td>
</tr>
<tr>
<td>CH₃COCH₃</td>
<td>31.25</td>
<td>29.82</td>
<td>-4.57</td>
</tr>
</tbody>
</table>
Fig. 1. Taylor diagram of the comparison between aircraft measurements and simulation S1. The biases are presented in parentheses (in pmol/mol). The empty box represents the measurements.
Fig. 2. Taylor diagram of the comparison between aircraft measurements and model results, both weighted with relative uncertainties (Jöckel et al., 2006). The biases are listed in parentheses (units of standard deviation). The empty box represents the measurements.
Fig. 3. Taylor diagram of the comparison between station observations and model results. The biases are presented in parentheses (in pmol/mol). The empty box represents the measurements.
Fig. 4. Taylor diagram of the comparison between station measurements and model results, both weighted with relative uncertainties. The biases are listed in parentheses (units of standard deviation). The empty box represents the measurements.
Fig. 5. Comparison of simulated (black) and observed (red, Novelli et al., 1998) CO mixing ratios in nmol/mol (ordered by latitude). The green lines show the model results sampled from the corresponding grid-boxes upwind (see text).
Fig. 6. Taylor diagram comparing 7 year (1998–2004) monthly averages of CO from the model simulations with the surface observations from the NOAA/CMDL network (Novelli et al., 1998). The colour code denotes the geographic latitude.
Fig. 7. Relative differences of 7 year averaged CO between model and measurements ((mod-obs)/mod).
Fig. 8. Vertical profiles of CO (nmol/mol) for some selected campaigns from Emmons et al. (2000). Asterisks and boxes represent the average and the standard deviation (w.r.t. space and time) of the measurements in the region, respectively. The simulated average is indicated by the red line and the corresponding simulated standard deviation w.r.t. time and space by the dashed lines. On the right side the number of measurements is listed.
Fig. 9. Vertical profiles of C$_2$H$_6$ (in pmol/mol) for some selected campaigns. Symbols and colours as in Fig. 8.
Fig. 10. Seasonal cycle (monthly averages) of $\text{C}_2\text{H}_6$ (in nmol/mol) for some selected locations at the surface (Solberg et al., 1996). Model: red solid line; model standard deviation: red, dashed line; measurements: circles.
Fig. 11. Vertical profiles of C$_3$H$_8$ (in pmol/mol) for some selected campaigns. Symbols and colours as in Fig. 8.
Fig. 12. As Fig. 10, for C$_2$H$_4$.
Fig. 13. Vertical profiles of $C_2H_4$ (in pmol/mol) for some selected campaigns. Symbols and colours as in Fig. 8.
**Fig. 14.** Vertical profiles of $C_3H_8$ (in pmol/mol) for some selected campaigns. Symbols and colours as in Fig. 8.
Fig. 15. As Fig. 10, for C$_3$H$_6$. 
Fig. 16. Temperature dependence of the reaction coefficient (cm$^3$ s$^{-1}$) for C$_3$H$_6$ + OH according to the IUPAC recommendations (black: Atkinson et al., 2005; red: Atkinson et al., 1999).
Fig. 17. Comparison of simulated and observed surface layer and PBL isoprene mixing ratios (in nmol/mol) for a large selection of sites covering the globe.
Fig. 18. Simulated surface layer isoprene mixing ratios (in nmol/mol) in Amazonia, 60°W, 2° N, for the year 2004 based on a 5-h output frequency (black) with monthly averages overimposed (red).
Fig. 19. Vertical profiles of $\text{H}_2\text{O}_2$ (in pmol/mol) for some selected campaigns. Symbols and colours as in Fig. 8.
Fig. 20. As Fig. 10, for CH$_3$CHO.
Fig. 21. Vertical profiles of CH$_3$OH (in pmol/mol) for some selected campaigns. Symbols and colours as in Fig. 8.
Fig. 22. Vertical profiles of HCHO for some selected campaigns (unit in the labels). Symbols and colours as in Fig. 8.
**Fig. 23.** As Fig. 10, for HCHO.
Fig. 24. As Fig. 10, for CH$_3$COCH$_3$. 
**Fig. 25.** Vertical profiles of CH$_3$COCH$_3$ (in pmol/mol) for some selected campaigns. Symbols and colours as in Fig. 8.
Fig. 26. Scatter plot of simulated versus observed CH$_3$COCH$_3$. The regression line is shown in red, the one-by-one equivalence is drawn in black.
**Fig. 27.** Taylor diagram of the comparison between aircraft measurements and model results. The color code denote the tracers and the symbol the different sensitivity simulations, respectively. The empty box represents the measurements. The biases are listed in parentheses (in pmol/mol).
Fig. 28. Relative difference of the annual average CO mixing ratio at the surface (in %) between sensitivity simulation $S1a$ (including dry deposition of CO) and reference simulation $S1$ (without dry deposition of CO).
Fig. 29. Comparison of CO mixing ratios (in nmol/mol): observations are in red, model results are in black (reference simulation S1) and blue (sensitivity simulation S1a). For the locations Florida (KEY) and Palmer station, Antarctica (PSA), the model results are sampled one grid-box upwind (see Sect. 4).
Fig. 30. Annually integrated dry deposition of CO (simulation S1a) in g/m².
Fig. 31. Vertical profiles of CO (in nmol/mol) for some selected campaigns from Emmons et al. (2000). Asterisks and boxes represent the average and the standard deviation (w.r.t. space and time) of the measurements in the region respectively. Red is the reference simulation $S1$, light blue the simulation $S1a$, and dark blue the simulation $S1b$. The corresponding standard deviations w.r.t. time and space are indicted by the dashed lines.
Fig. 32. Vertical profiles of $\text{C}_3\text{H}_6$ (in pmol/mol) for some selected campaigns and for the sensitivity simulation $S1a$. Symbols and colours as in Fig. 8.
**Fig. 33.** Vertical profiles of $\text{C}_2\text{H}_4$ (in pmol/mol) for some selected campaigns. The red line shows the results of the reference simulation ($S1$), the blue line the results of the sensitivity simulation $S1a$. Symbols as in Fig. 8.
Fig. 34. Scatter plot of model results versus observations of CH$_3$OH (both in nmol/mol). The red circles represent the results of the reference simulation (S1), the black triangles those of the sensitivity simulation (S1a).
Fig. 35. Vertical profiles of CH$_3$COCH$_3$ (in pmol/mol) for some selected campaigns. The red line shows the results of the reference simulation S1, the blue line the results of the sensitivity simulation S1a. Symbols as in Fig. 8