Global cloud and precipitation chemistry and wet deposition: tropospheric model simulations with ECHAM5/MESSy1

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Abstract

The representation of cloud and precipitation chemistry and subsequent wet deposition of trace constituents in global atmospheric chemistry models is associated with large uncertainties. To improve the simulated trace gas distributions we apply the new submodel SCAV, which includes detailed cloud and precipitation chemistry and present results of the atmospheric chemistry general circulation model ECHAM5/MESSy1. A good agreement with observed wet deposition fluxes for species causing acid rain is obtained. The new scheme enables prognostic calculations of the pH of clouds and precipitation, and these results are also in accordance with observations. We address the influence of detailed cloud and precipitation chemistry on trace constituents based on sensitivity simulations. The results confirm previous results from regional scale and box models, and we extend the analysis to the role of aqueous phase chemistry on the global scale. Some species are directly affected through multiphase removal processes, and many also indirectly through changes in oxidant concentrations, which in turn have an impact on the species lifetime. While the overall effect on tropospheric ozone is relatively small (<10%), regional effects on O₃ can reach ~20%, and several important compounds (e.g., H₂O₂, HCHO) are substantially depleted by clouds and precipitation.

1 Introduction

Scavenging and subsequent wet deposition represent important removal processes for many trace constituents in the troposphere. They are crucial for cleansing the troposphere from aerosol particles and soluble gases and indirectly also for many less soluble species. Furthermore, multiphase chemistry can have a major impact on tropospheric composition (Ravishankara, 1997). However, the representation of wet deposition and cloud chemistry differs widely in global models, resulting in large uncertainties (Rasch et al., 2000). These uncertainties are to a large extent associated with the
coarse descriptions of cloud and precipitation processes, and also with the representa-
5 tion of scavenging and aqueous phase chemistry (Zhang et al., 2006). Even though
Levine and Schwartz (1982) and Schwartz (1986) have shown more than two decades
ago that for most gases uptake according to Henry’s equilibrium alone does not repre-
7 sent atmospheric conditions, this approximation is still commonly used in scavenging
parameterisations, in particular in global models. Alternatively, fixed scavenging co-
10 efficients are applied, implicitly assuming that the process only varies as a function of
the precipitation flux. The treatment of chemical reactions of dissolved gases in clouds
and precipitation has largely focused on sulphur species (e.g. Berglen et al., 2004).
A model inter-comparison of cloud chemical parcel models focusing mainly on the ox-
15 idation of SO\textsubscript{2} in the aqueous phase (including the basic aqueous phase transfer and
chemical reactions) has shown that large uncertainties arise from different parameteri-
20 sations (Kreidenweis et al., 2003). More detailed cloud chemistry schemes have been
applied in smaller scale models (box models, cloud parcel models, single column mod-
elts), e.g. by Chameides and Davis (1982); Jacob (1986); Lelieveld and Crutzen (1991);
Bott and Carmichael (1993); Monod and Carlier (1999); Fahey and Pandis (2001); von
Glasow et al. (2002); Ervens et al. (2003); Leriche et al. (2003); Barth et al. (2003) and
25 references therein; Kreidenweis et al. (2003) and references therein, and detailed
information about the processes involved has been gained. However, from these results
it is difficult to estimate the large-scale effects of multiphase chemistry. Lelieveld and
Crutzen (1991) have derived the global tropospheric impact by applying their chemical
box-model in combination with a global cloud data set. The application of the model of
Liang and Jacob (1997) in the regional model of Jacob et al. (1993) leads to the con-
clusion of a minor impact of multiphase chemistry on the regional scale, although this
was only valid for the simulated period and domain addressed, and the extrapolation
to the global scale is not straightforward. Fahey and Pandis (2003) have increased the
amount of detail and the model complexity, and calculated droplet size resolved cloud chemistry in a 3D regional transport
model. Yet, this model setup is quite complex and therefore it has not been applied
in an atmospheric chemistry general circulation model (AC-GCM). Global models including cloud and precipitation chemistry often lack comprehensiveness to simulate the interconnected effects in the gas and liquid phase. For example, Stier et al. (2005) calculated in-cloud SO$_2$ oxidation, but used prescribed distributions for the oxidants. Since some of these models aim at the investigation of the global sulphur cycle (e.g. Feichter et al., 1996; Rostayn and Lohmann, 2002; Dentener et al., 2006; Berglen et al., 2004), this might be appropriate, though it also introduces large uncertainties. Crutzen and Lawrence (2000) determined the effects of trace gas scavenging in a 3D global model, however, did not assess multiphase chemistry in clouds and precipitation. Global model studies explicitly considering aqueous phase chemistry may exist (e.g. Dentener, 1993), though not reported in the peer reviewed literature. Dentener and Crutzen (1993, 1994) applied several heterogeneous reactions including NH$_3$ chemistry and their influence on gas phase composition. Roelofs and Lelieveld (1995) introduced additional aqueous phase reactions including e.g., HCHO and HCOOH in their model to assess multiphase chemistry, although non-methane-hydrocarbon (NMHC) chemistry was neglected and both the gas and aqueous phase reaction mechanisms applied are not easily expandable.

In the present study the scavenging and subsequent wet deposition of trace species in the ECHAM5/MESSy1 model is presented, and the influence of more detailed cloud and precipitation chemistry on the tropospheric composition is investigated with the recent submodel SCAV (Tost et al., 2006a). Following a short model description in Sect. 2, the simulation setup is outlined in section 3. The results, presented in Sect. 4, distinguish three aspects: the evaluation of the wet deposition fluxes (4.1), the analysis of global pH values in clouds and precipitation (4.2) and the influence of liquid phase chemistry on the gas phase composition of the troposphere (4.3). The scavenging of aerosol species is not explicitly evaluated in this study. However, the relevant processes are implemented in the SCAV submodel and can be used in studies of both soluble and less soluble aerosol particles, how these are incorporated into the droplets and to what extent they take part in liquid phase chemical reactions. These latter is-
sues will be addressed in future publications. However, the contribution of scavenged particulate sulphate to the amount of dissolved sulphur(VI) is also considered in this study.

In addition to cloud chemistry effects that have been investigated in the past, the effects of the chemical processes in the liquid precipitation are directly addressed, i.e., not only the vertical downward transport within hydrometeors, but also the precipitation chemistry during the falling phase are taken into account.

2 Model description

In this study the atmospheric chemistry general circulation model ECHAM5/MESSy1 has been applied combining the 5th generation European Centre - Hamburg model (Roeckner et al., 2006; Hagemann et al., 2006; Wild and Roeckner, 2006) (version 5.3.01) and the Modular Earth Submodel System (Jöckel et al., 2005) (version 1.1). The global meteorology is calculated by ECHAM5 based on a spectral representation of the prognostic variables vorticity, divergence, temperature, and the logarithm of the surface pressure, as well as grid point representations of specific humidity, cloud water, and cloud ice. In the vertical a hybrid pressure coordinate system is applied. The processes of radiation and cloud microphysics are parameterised, as described in the ECHAM5 documentation (Roeckner et al., 2003, 2004). Advection of the prognostic tracers is calculated with the Lin and Rood (1996) flux-form semi-lagrangian transport algorithm. A first application of ECHAM5/MESSy1 as AC-GCM including a more detailed model description is presented in Jöckel et al. (2006). Since the wet deposition from the evaluation simulation from Jöckel et al. (2006) is used, the model configuration will be shortly summarised in the next section.

MESSy contains submodels for atmospheric chemistry, transport, their feedbacks on the meteorology through radiative transfer, and diagnostic tools. For the present study we applied the MESSy submodels RAD4ALL (for radiation calculations), CONVECT (convection parameterisation (Tost et al., 2006b)), CLOUD (large scale condensation),
CVTRANS (convective tracer transport (Tost, 2006)) and ONLEM, OFFLEM, TNUDGE (emissions and pseudo-emissions, described in detail by Kerkweg et al. (2006b)), LNOX (NO\textsubscript{x} emissions from lightning), TROPOP (diagnostics of tropopause and boundary layer height), PTRAC (passive tracers), MECCA (gas phase chemistry (Sander et al., 2005)), JVAL (photolysis rates for chemistry calculations), HETCHEM (reaction rates for heterogeneous chemical reactions on aerosols), M7 (aerosol microphysics (Vignati et al., 2004; Kerkweg, 2005)), DRYDEP (dry deposition of trace gases and aerosol particles (Ganzeveld et al., 1998; Kerkweg et al., 2006a)), SEDI (sedimentation of aerosol particles (Kerkweg et al., 2006a)), and SCAV (scavenging and liquid phase chemistry in clouds and precipitation (Tost et al., 2006a)). A more detailed description of the submodels can also be found on the MESSy website\footnote{http://www.messy-interface.org}. Since wet deposition and cloud and precipitation chemistry are calculated by SCAV, this submodel is the major focus of this study and its functionality is shortly described. For more details and the parameterisations used for the individual processes we refer to Tost et al. (2006a).

SCAV calculates the uptake of aerosol particles and trace gases into cloud and precipitation droplets. Since the global model does not provide detailed information on droplet spectra, all calculations are based on bulk quantities, even though for the gas transfer velocities and scavenging efficiencies droplet size spectra are assumed. The aqueous phase chemistry, including the transfer of gaseous compounds, dissociation of acidic and alkaline species in the droplets and aqueous phase redox reactions are calculated by a coupled system of ordinary differential equations using the kinetic pre-processor (KPP) software (Damian et al., 2002). The chemical reaction system for both gas and aqueous phase (MECCA and SCAV) and the applied Henry’s law and accommodation coefficients are given in the supplementary material of this paper (http://www.atmos-chem-phys-discuss.net/7/785/2007/acpd-7-785-2007-supplement.pdf). The gas and the liquid phase chemical processes are fully coupled and do not require prescribed mixing ratios or pH values. For the scavenging of aerosol compounds...
we refer to the detailed submodel description by Tost et al. (2006a), noting that size dependent scavenging efficiencies for cloud and rain droplets are included.

3 Simulation setup

The results used in this study refer to the ECHAM5/MESSy1 evaluation simulation (EVAL S1), obtained with the lower-middle atmosphere model configuration described in Jöckel et al. (2006), using a spectral resolution of T42 (with a resolution of approximately 2.8° x 2.8° of the corresponding quadratic Gaussian grid) and 90 vertical levels up to 0.01 hPa (~ 80 km, as the mid of the uppermost layer) altitude, covering a simulation period from 1998 to 2005. We additionally performed a series of shorter simulations for the year 2000 with a tropospheric configuration of the model, which has a reduced vertical resolution especially in the upper troposphere and tropopause region (31 levels up to the mid of the uppermost model layer at 10 hPa). Furthermore, the sensitivity calculations include the following modifications in the setup of the SCAV submodel:

- SCM (scavenging minimum): The same small set of chemical reactions in the aqueous phase is applied as in Jöckel et al. (2006) (labelled Scm in the reaction table of the electronic supplement http://www.atmos-chem-phys-discuss.net/7/785/2007/acpd-7-785-2007-supplement.pdf).

- COM (complex aqueous phase chemistry): A relatively detailed chemical reaction set is applied in a more comprehensive simulation, including more than 150 reactions in the liquid phase (labelled Sc in the reaction table of the supplement).

- EASY: A highly simplified approach is applied, with the gas-liquid partitioning only according to absolute Henry’s law.
– EASY2: A simplified approach is applied, with the gas-liquid partitioning according to effective Henry’s law (e.g. Sander, 1999), assuming a pH value of 5 in clouds and precipitation.

– NOSCAV (no scavenging): For comparison, an additional simulation has been performed in which aqueous phase chemistry, scavenging, and wet deposition are neglected.

Furthermore, as mentioned in Jöckel et al. (2006), a correction of the precipitation liquid water content (LWC) has been applied, resulting in a more realistic scavenging representation in the four sensitivity simulations. The simulation period for the sensitivity experiments spans one year each, plus three months of spin-up, initialised with chemical data from the evaluation simulation (S1) of Jöckel et al. (2006). To approximate realistic, i.e. analysed meteorological conditions, the surface pressure is adjusted towards the surface pressure from European Centre for Medium-range Weather Forecasting (ECMWF) analyses for the year 2000 using a nudging technique (Jeuken et al., 1996; van Aalst et al., 2004). The coupling between chemistry and dynamics through radiation feedbacks has been switched off to simulate comparable meteorology in the sensitivity studies. Nevertheless, the dynamics of the various simulations are not fully identical due to interactions of water, ice and chemistry, although the meteorological patterns are very similar.

4 Results

The precipitation distribution and the hydrological cycle of the ECHAM5/MESSy1 model have been discussed in Jöckel et al. (2006). Furthermore, the influence of the convection parameterisation on the hydrological cycle is analysed in Tost et al. (2006b), especially to illustrate the uncertainties associated with convective precipitation simulations. Since the locations and rates of precipitation are essential for the wet deposition of trace species, the average rainfall is presented in the upper panel of Fig. 1.
The strongest precipitation occurs in the tropics, with the maximum over the Pacific warm pool region, the Indian Ocean, and the South Pacific Convergence Zone (SPCZ). A second maximum appears west of Central America. Both maxima mainly result from convective activity. The mid-latitude storm tracks are characterised by bands of moderate precipitation originating from both large-scale and convective cloud formation. Relative to observations the simulated precipitation is overestimated in the tropics, but compares well with the rainfall distribution in the mid-latitude storm tracks (Tost et al., 2006b).

The lower panel of Fig. 1 shows the large-scale fraction of the total precipitation. In the tropics almost all rainfall originates from convection (>80%), while this ratio decreases poleward. In the mid-latitudes the convective fraction is on average less than 30%. In the subtropical regions precipitation rates are typically low because convection is suppressed by large-scale subsidence, which is realistically simulated by the model. Overall, the precipitation distribution is reproduced quite well by the model with some limitations (Hagemann et al., 2006; Tost et al., 2006b). These limitations will have to be considered in the evaluation of wet deposition patterns by comparisons with measurement data.

4.1 Evaluation of wet deposition fluxes

4.1.1 Wet deposition fluxes in the EVAL S1 simulation

In this subsection the model calculated wet deposition fluxes from the reference simulation, as presented by Jöckel et al. (2006), for nitrate (HNO$_3$,$aq$ and NO$_3^-$), sulphate (HSO$_4^-$ and SO$_4^{2-}$) and NH$_x$ (NH$_3$,$aq$ and NH$_4^+$) are compared with measurements.
Observational data

The observational data used are from several measurement networks. The data set has been composed by Dentener et al. (2006)² (further denoted as D06). For North America observations from the North American Deposition Program (NADP)³ Hicks (2005), and for Europe from the EMEP network⁴ for the year 2000 are used. Additionally, IGAC DEBITS Africa⁵ (IDAF) measurements are used for the African continent. The data include annual averages from 1996 to 1999. Measurements for East Asia are taken from the Acid Deposition Monitoring Network in East Asia (EANET) for the year 2000 and some stations from the Integrated Monitoring Program on Acidification of Terrestrial Ecosystems (IMPACT) in China from 2001 to 2003. For India we use observations by Kulshrestha et al. (2005), whereby some stations have been selected as mentioned in D06, with data from the period 1995-2000. South American data have originally been collected by Dentener and Crutzen (1994), Filoso et al. (1999), Lara et al. (2001). Additionally, the D06 dataset includes observations described by Galloway et al. (2006)⁶ for the period 1980-1999 from a number of remote stations (several islands, Australia, South America). Some measurement stations have been excluded if a time series of one year was not available. The number of observations from the networks is listed in Table 1.

The comparison with observations is usually difficult, since the large spatial and temporal heterogeneity of precipitation and its chemical composition cannot be simulated in detail with a global model using a grid width of a few hundred kilometres. Never-

²Note, that this data is from a preliminary version, i.e. the submission state of that paper, and therefore not completely identical with respect to the stations taken into account. Therefore, small differences between the observation average values are found in Tab. 3.
³data available through the internet from: http://www.nadp.sws.uiuc.edu
⁴http://www.nilu.no/projects/ccc/emepdata.html
⁵http://www.medias.obs-mip.fr/idaf
⁶manuscript in preparation
theless, the comparison can give indications to what extent the overall patterns are captured accurately by the model.

Nitrate

The nitrate wet deposition distribution is shown in Fig. 2. The upper panel shows the average annually accumulated nitrogen content in precipitation from the EVAL S1 simulation in mg N/m². Two major wet deposition maxima are evident: in China, resulting mainly from the strong emissions of nitric oxides (NOₓ) mostly from fossil energy use, and in Central Africa, where the anthropogenic emissions are dominated by residential biofuel use and biomass burning. Furthermore, large deposition rates are also calculated for the eastern USA and western Europe, both mainly from fossil fuel related NOₓ emission sources. In Southeast Asia, Amazonia and Central Africa additionally the natural NOₓ emissions from soils and lightning significantly contribute to the nitrate content of precipitation. Over the tropical continents the large amounts of precipitation lead to an almost complete scavenging of nitric acid (HNO₃) from the gas phase, while in the Northern Hemispheric storm tracks highly efficient wet deposition of HNO₃ is most evident over the Atlantic Ocean. The latter is also a consequence of the incomplete scavenging close to the North American east coast (due to lower and less frequent precipitation compared to the tropics) and subsequent westward transport after which the atmospheric pollution is removed from the atmosphere.

The lower panel of Fig. 2 depicts the comparison of nitrate wet deposition of the individual years to the observations described above, both for all stations (pentagons) and differentiated for the individual regions (other symbols) in a Taylor diagram (Taylor, 2001). The diagram relates the correlation between model results and measurements with the standard deviation of the model results, normalised to the standard deviation of the measurements. Since the observational data do not provide global coverage point-to-point comparisons between the station locations and the closest coordinates within the model grid are performed. The overall correlation for all years is \( R \approx 0.6 \) with a normalised standard deviation \( \sigma^* \sim 0.8 \) \( (\sigma^* = \sigma_{\text{mod}} / \sigma_{\text{obs}}) \), indicating an underestimation.
of the amplitude of the spatial variation. For the most comprehensive NADP data set only, the model shows a higher correspondence with a correlation of \( R > 0.8 \) for all years and a normalised \( \sigma \) between 1 and 1.2. For the data from the other networks the correlation is much lower (around 0.4 to 0.5). Furthermore, the amplitude of the variability is hardly matched accurately, \( \sigma^* \) ranges from around 2.5 for the African measurements (indicating a strong overestimation of the variation of model wet deposition in this area) to 0.3 for some data from the EMEP and Indian observations, thus indicating underestimation of the spatial variation of the wet removal of nitric acid and nitrate in these regions. This can be related to the scavenging process, but as well to an inadequate representation of other processes, mainly emissions, but also chemistry and transport. As an example, the African measurements are mostly substantially lower than the simulated values. This probably results from the biomass burning emissions and their large uncertainties, leading to too high NO\(_x\) emissions and subsequent wet deposition over the African continent in the model. For the IDAF data the interannual variability of the model is highest, whereas it is minor for the other measurement networks. We can also not exclude that some of the measurement data are not representative of the surrounding area of the size of the model grid due to the large heterogeneity of the terrain and consequently rainfall and wet deposition patterns and the sparse observatory locations. Additionally, the observations partly originate only from one specific year and are therefore not representative for all years of the simulation period. However, since it is the only available dataset for this region it can be applied, if its uncertainties are taken into account.

**Sulphate**

The wet deposition of sulphate occurs mainly in the vicinity of the emission sources of sulphur dioxide (SO\(_2\)), of which a small fraction is oxidised in the gas phase, whereas the more important oxidation pathway is in the aqueous phase (e.g. Warneck, 1999). The average annually accumulated deposition flux is shown in the upper panel of Fig. 3. The largest wet deposition fluxes are calculated for China, where the SO\(_2\) emissions
are strongest. Additional local maxima are computed for the industrialised regions in the eastern USA and Central Europe. In many of these regions energy production relies on coal types with a high sulphur content, resulting in substantial \( \text{SO}_2 \) emissions and consequent sulphate deposition.

The comparison with the observations shown in the lower panel of Fig. 3 shows a relatively high correlation between the model results and the measurements if we include all stations, \( R \approx 0.7 \), with the normalised standard deviation close to 1. As for nitrate, the NADP data is even higher correlated, and the normalised \( \sigma \) is slightly above 1. For most other networks the correlation is lower, but except for South America, always higher than 0.3. The amplitude of the variation is captured relatively well, with \( \sigma^* \) values typically between 0.6 and 1.3. For the 16 stations in South America the representation of the observed values by the model is relatively poor.

Ammonia - Ammonium

The wet deposition fluxes of \( \text{NH}_x \) compounds, shown in the upper panel of Fig. 4, are also highest in regions where the emissions are typically strongest, i.e., in China and India, and slightly lower in western Europe, eastern USA, Central Africa and the Northern part of South America. While in East Asia, North America and Europe both agricultural and industrial emissions are the main sources, in Africa and South America biofuel use and biomass burning dominate the atmospheric ammonia burden. In contrast to nitrate and sulphate higher wet deposition occurs in the tropical ITCZ, both over land and the ocean. This partly results from a neglect of the chemical loss of \( \text{NH}_3 \) in the gas phase (oxidation by OH). Furthermore, the effects of uncertainties of the simulated precipitation rates become increasingly important for moderately soluble species such as \( \text{NH}_3 \). However, since observations are very scarce over the tropical oceans, it is difficult to judge if these features are real or an artefact of the model.

The correspondence of the model results with the observations is shown in the lower panel of Fig. 4. The overall correlation by including all stations is \( R \approx 0.6 \), and the comparison indicates a slight underestimate of the spatial variation (\( \sigma^* \approx 0.6 \)). Again, for the
dense measurement network of the USA the agreement is best. While the correlation for Europe and South America are both higher than 0.7, $\sigma^*$ is either too low (for Europe) or too high (for South America). The correspondence for the African and especially the Indian stations is less good (R between 0.3 and 0.5).

It is obvious that the model representation of these three compounds of rainwater are of different quality for the different regions. For example, for South America NH$_x$ wet deposition is simulated quite accurately, but the simulated sulphate wet deposition is hardly correlated with the observations. This might be related to high sulphate burden over Chile, resulting from high values in the SO$_2$ emission dataset for this region. It must be emphasised, that not only the observations are best with respect to the operational quality and the amount of data points in time and space for NADP, but also the available emission datasets from North America and Europe are of much higher quality compared to African and South American data because of the denser observation networks. In comparison with other modelling studies the correspondence between the observations and model results is similar. In the model inter-comparison of wet deposition of nitrate, sulphate and ammonia by D06 similar correlations are achieved as shown in Table 2.

Only for the African data the representation by our model seems substantially poorer. However, these results originate from very few data points and the comparison should be weighted less as for instance with the comprehensive NADP dataset for which we obtain very good agreement. A comparison for the absolute values of the observed and simulated wet deposition fluxes is given in Table 3.

Even though the simulated precipitation is generally too strong in the tropics, this hardly affects the comparison with the observations, since the regions with the most pronounced overestimation are located over the ocean, where no wet deposition data is available. Nevertheless, the unrealistically strong rainfall in the Himalaya region (cf. Tost et al. (2006b)) leads to enhanced wet deposition in this region. Additionally, the enhanced wet deposition in Central Africa and Eastern China can also be related to
overestimations in the convective precipitation. Different model agreement for nitrate, sulphate and \( \text{NH}_x \) as well as the differences in the interannual variability indicate that this is likely not only caused by the wet deposition formulation, but also by emissions, transport and chemical processes.

4.1.2 Sensitivity on the details of the scavenging / liquid phase chemistry process description

In this section the sensitivity studies with the 31 layer tropospheric model configuration are compared for the year 2000. Since in NOSCAV wet deposition is not calculated at all this simulation is ignored in this section. Additionally, in the four simulation setups the above mentioned correction of the precipitation liquid water content has been applied. Since the chemical setup of the SCM simulation is identical to that in the EVAL S1 simulation \((\text{Jöckel et al., 2006})\) SCM differs mainly in the vertical resolution of the model, which is most significantly reduced near the tropopause, and the above mentioned correction.

The upper part of Table 4 compares the average deposition fluxes from the individual measurement networks with the four simulations. While for all three components the SCM and COM simulation yield very similar values, the EASY simulation always shows lower deposition fluxes. Replacing the absolute with the effective Henry's law coefficients (EASY2) the wet deposition of most species is larger compared to the simulations explicitly calculating aqueous phase chemistry. This is partly caused by not taking the gas phase diffusion limitation into account, and instead assuming equilibrium between gas and liquid phase.

The underestimation of the EASY setup is most obvious for \( \text{NH}_x \) for which we obtain unrealistically low wet deposition fluxes. The failure of the EASY simulation for ammonia can be explained by the highly reduced uptake of gaseous \( \text{NH}_3 \) into the cloud and precipitation droplets due to the relatively low solubility of ammonia and the neglect of the dissociation and neutralisation in water in this setup. For the more soluble compound \( \text{HNO}_3 \) the difference is less obvious. However, due to the neglect
of the dissociation the effective uptake into the droplets is still substantially reduced (EASY). In contrast, with the effective Henry’s law coefficients the uptake of the acidic species is overestimated (EASY2), since in the polluted regions the actual pH is lower than the prescribed value of five and the altered oxidation capacity of the atmosphere leads to enhanced HNO₃ production. In the EASY simulation sulphate wet deposition only results from the dissolution of sulphuric acid (H₂SO₄), since liquid phase oxidation of SO₂ does not occur. Consequently, sulphur dioxide is only oxidised to H₂SO₄ in the gas phase, which only represents a small fraction of this process, and the wet deposition of sulphate is considerably lower in EASY. This is also valid for EASY2, which therefore also underestimates the S(VI) content in the precipitation water, but removes SO₂ from the atmosphere slightly more realistically (cf. Fig. 8). For NHₓ the EASY2 simulation is able to calculate the wet removal from the atmosphere within an acceptable accuracy range compared with the observations, but similar to HNO₃ wet deposition is enhanced compared with the observations.

Therefore, for the wet deposition calculations of the basic species, relevant e.g. for acid rain formation, the partitioning according to Henry’s law alone is not sufficient for atmospheric chemistry modelling. The attempt to partly overcome this limitation by using effective Henry’s law coefficients (e.g. Sander, 1999) that account for the dissociation of acidic and alkaline species based on a globally constant pH value leads to problems, since the H₃O⁺ concentrations in the liquid are highly variable in space and time and chemical reactions such as S(IV) oxidation cannot be such easily parameterised. Alternative effective scavenging coefficient formulations applied in other models (e.g. Yin et al., 2001; Asman, 1995), taking some basic chemical reactions into account, are often determined for a specific location and therefore also not straight-forwardly applicable on the global scale, which includes a large range of conditions from highly polluted to remote regions. For example, Mizak et al. (2005) present a different algorithm for NH₃ scavenging based on observations compared to Asman (1995). If the dependence on the pH value is applied to compute the effective Henry’s law coefficient correctly as well as the oxidant concentrations for basic.
chemistry (dependent on time and the location), these quantities must be determined appropriately by taking all relevant aqueous phase species and reactions into account, which in turn is slightly more computationally expensive and finally results in a system of coupled differential equations as it is applied in the SCM or COM simulation setup.

4.2 pH value of clouds and precipitation

As a consequence of the comprehensive treatment of dissociation reactions in the liquid phase and the coupling of all reactions that affect the solution ion balance, the cloud and precipitation pH can be calculated directly during each time step. In contrast to previous studies we do not apply assumptions about the dissociated fraction of components nor do we prescribe oxidant concentrations. The latter is associated with difficulties for oxidants that are significantly affected by aqueous phase reactions, e.g. H$_2$O$_2$. For the comparison with measurement data the aqueous phase H$_3$O$^+$ concentrations are averaged and weighted with the amount of liquid water for precipitation and clouds, and the results presented are based on our most comprehensive simulation (COM).

The left panel of Fig. 5 depicts the annual average precipitation pH at the surface, originating from both convective and large-scale rainfall, weighted with the amount of rainwater and the fraction of the precipitation type. The precipitation pH varies between values around 3.5 and 6.5, being always in the acidic regime due to the atmospheric CO$_2$. Highest pH values are found over the tropical oceans where pollution sources are small and precipitation rates are high. This is also valid for other regions where pollution is low. In some regions the neutralising effect of ammonia plays a significant role, e.g. over India, mainly neutralising the nitric acid produced from intense NO$_x$ emissions. The alkaline compounds of sea salt and mineral dust are neglected in this study, because the ionic composition of these aerosol types is not yet considered in the aerosol submodel M7.

This also explains the relatively high pH values in the southern storm tracks. Over
the tropical continents, especially Central Africa, the relatively low pH is mostly a result of the \( \text{NO}_x \) emissions by biomass burning. In several desert regions, e.g. in Africa, the pH can be rather low, mainly due to the small total amount of cloud and rain water, which concentrates the available acidity. In Europe, China and the eastern USA strong anthropogenic emissions of \( \text{SO}_2 \) and \( \text{NO}_x \) reduce the pH of clouds and precipitation, only partly compensated by the dissolution of ammonia. As mentioned earlier, the comparison with observations can only give indications to what extent the main features and average pH ranges are captured by the model. The calculated pH values are compared with observed precipitation pH for the year 2000 in the USA with data from the NADP network. Even though the main features, e.g., the east-west gradient, are captured well, the correlation is moderate \((R \approx 0.65)\) and some of the fine structures apparent from the relatively dense network are not resolved. Furthermore, the east-west gradient in the USA is stronger in the observations with values larger than 5 in the western and central USA, where the model results indicate too acidic precipitation. This may partly result from a slight underestimation of the total precipitation, especially in the Central USA, but mostly an overestimation of the \( \text{H}_3\text{O}^+ \) concentrations. Since sulphate, nitrate and ammonium deposition are calculated accurately, we suspect a contribution by mineral cations in wind blown dust and alkaline sea salt, not included in our model setup.

For Europe we use the data collected in the EMEP/CCC Report 1/2006\(^7\), and the comparison with the simulation indicates general agreement with precipitation pH distributions; however, the model is often slightly too acidic. This is partly a consequence of the underestimate of dissolved \( \text{NH}_x \) (Table 3). Even though the precipitation distribution in general is captured quite accurately by the model (Hagemann et al., 2006; Tost et al., 2006b) the local variability at specific measurement stations can show substantial differences compared to the model grid box average, which affects the rain water pH. Safai et al. (2004) use a compilation of pH samples for India with average values between 5.3 and 7.7, in relatively good agreement with the model simulated results.

\(^7\)available from: http://www.nilu.no/projects/ccc/reports.html
Additionally, data by Aikawa et al. (2001) with average precipitation pH values of 4.7 correspond to the simulation results for Japan.

The right panel of Fig. 5 depicts the annual and zonal average of cloud pH. Above 250 hPa in the tropics to 400 hPa at the poles the liquid water content of the clouds is too small to calculate cloud chemistry, and therefore pH calculations have little meaning, since the clouds consist mainly of ice. The grey shaded area near the lower boundary represents the zonal mean surface orography. The highest pH values are found in the convective regions of the tropics and in the Southern Hemisphere storm tracks. The figure shows that the deep convective towers are less acidic due to the high water content and the remoteness from acidic emissions compared to the mid-latitudes. Furthermore the composite of large-scale and convective clouds in the southern storm tracks shows a pH maximum at about 700 hPa, where the liquid water content (LWC) is typically highest, and these regions are also remote from acid precursor emissions. In the Northern Hemisphere the clouds are much more acidic from the surface up to 700 hPa due to anthropogenic emissions of SO₂ and NOₓ. The pH values slightly increase toward 600-700 hPa related to the maximum LWC. The convective activity north of 40°N coincides with slightly lower pH values at about 800 hPa compared to the subtropical regions further south where subsidence is dominant. At higher altitudes and further poleward in both hemispheres the clouds are increasingly frozen, thus reducing the cloud LWC. Therefore the relatively small amounts of HNO₃ or H₂SO₄ lead to stronger acidification under these conditions. However, it should be mentioned that the convective cloud pH is partly too low because we do not use the total convective cloud water for the scavenging calculations but rather the fraction that produces precipitation, since the currently used convection parameterisations does not explicitly provide the total convective cloud water.

Measurements of pH values in clouds result mainly from hill cap clouds and fog at elevated observatories. Marinoni et al. (2004) measure levels mainly between 4.5 and 5.5 at a remote site in France, but note that there are local NHₓ sources. However, at the same site values between 4.1 ± 0.5 and 5.7 ± 0.1 also occur (Sellegri et al., 803).
2003), indicating the high variability. Moore et al. (2004) report pH values for New York between 2.7 and 3.7, thus highly acidic. Of course, such acidic values do not appear in the zonal average, but close to the surface in polluted areas (such as the Eastern USA) a similar pH range is calculated by the model.

The pH distribution of the simplified aqueous phase mechanism (SCM) results in similar patterns. However, these are not identical due to the additional chemical processes in the liquid phase and the feedback via the gas phase (cf. Section 4.3). Since in the EASY simulation dissociation reactions in the liquid do not take place, a pH value cannot be determined, whereas in EASY2 it is prescribed to calculate the effective Henry’s law coefficients.

4.3 Influences of cloud and precipitation chemistry on gas phase constituents

To analyse the influence of comprehensive aqueous phase chemistry on gas phase constituents the sensitivity simulations described in section 3 are compared for several gas phase species, using the COM simulation as a reference and showing absolute differences to this reference for the other simulations. Furthermore, we evaluate which simulations agree best with the observational data compiled by Emmons et al. (2000, and references therein). For these comparisons only four selected locations are presented here, i.e. from the TOPSE (Atlas et al., 2003) and TRACE-P (Jacob et al., 2003) campaigns, because these measurements approximately coincide with the period for which the applied emission data are representative. We provide additional comparisons with previous campaigns in the electronic supplement (http://www.atmos-chem-phys-discuss.net/7/785/2007/acpd-7-785-2007-supplement.pdf). The general evaluation of the tropospheric and stratospheric chemistry in ECHAM5/MESSy1 has been presented by Jöckel et al. (2006) and Pozzer et al. (2006). Due to the coupling of gas phase chemistry with the cloud and precipitation chemistry we expect direct and indirect effects resulting from

8 updated dataset obtained from: http://acd.ucar.edu/~emmons/DATACOMP/camp_table.htm
a different treatment of the aqueous phase chemistry. The direct effect is caused by a difference in the uptake into the droplets which depends on the process formulation and the concentration in the liquid phase. Furthermore, the consideration of various chemical reactions in the liquid may also be conceived as a direct effect. The direct effects are especially important for the species participating in comprehensive aqueous phase chemical reactions, e.g., SO$_2$ conversion to SO$_4^{2-}$ in the droplets by H$_2$O$_2$ and O$_3$, dependent on the pH. Additionally, pH dependent acid-base equilibria are associated with changes in liquid concentrations, if more species are considered.

Furthermore, a different chemical composition of the atmosphere resulting from cloud and precipitation chemistry modifies the oxidation capacity of the atmosphere, which indirectly alters the mixing ratios of chemically active trace constituents. This is shown by the OH distributions in the supplement (http://www.atmos-chem-phys-discuss.net/7/785/2007/acpd-7-785-2007-supplement.pdf). Even though the mixing ratio differences to the reference simulation are relatively small, they indicate that the net turnover of hydroxyl radicals can increase substantially.

An additional aspect of even higher significance is the modified vertical transport behaviour in convective clouds. In case of a weaker scavenging due to less efficient uptake into the droplets, species with high mixing ratios at the surface are more effectively transported into the upper troposphere. Since the meteorology of the sensitivity runs is not completely identical, small changes in the amount of cloud and precipitation water have an impact on the trace species distribution.

For a more detailed analysis in addition to the near surface mixing ratios the annual zonal average mixing ratios can be found in the supplement, as well as figures comparing the total global tracer mass. Since the NOSCAV simulation is unrealistic, because of the neglected scavenging and wet deposition, these respective figures have been transferred to the supplement.
Hydrogen peroxide (H$_2$O$_2$) efficiently dissolves into the liquid phase, where it can oxidise sulphur components and is removed from the atmosphere by precipitation. Because of the much lower solubility of OH, NO$_3$ and other oxidants it is a main oxidant in the liquid phase. The oxidation reactions represent an additional chemical sink of this compound in clouds and precipitation. Therefore, somewhat lower values in the gas phase are calculated for the COM simulation than for SCM. EASY and EASY2, in which no aqueous phase chemical reaction are considered, show consequently much lower values than the COM simulation. Since the effective Henry’s law coefficient is almost identical to the absolute one, differences between EASY and EASY2 are mainly based on indirect effects. The NOSCAV simulation illustrates the role of clouds and precipitation in the chemical budget of hydrogen peroxide. Due to the missing wet deposition, which is only partly compensated by chemical and dry deposition sink processes, the atmospheric burden of H$_2$O$_2$ in NOSCAV is substantially higher. This is illustrated in Fig. 6, in which the annual average surface mixing ratio of H$_2$O$_2$ is displayed for the COM simulation (upper left panel). Relatively high values occur over the tropical rainforest regions, associated with abundant NMHC oxidation with high yields of the main precursor of hydrogen peroxide, i.e., hydroperoxy radicals (HO$_2$). Toward the poles a strong negative gradient is evident, related to both the lower NMHC emissions and lower photochemical activity. In the upper right panel with the reduced chemical mechanism (SCM) higher values than in COM are calculated in the tropics. Since in this setup the photolysis of H$_2$O$_2$ in the aqueous phase, representing a chemical loss process, is neglected the liquid phase concentrations are higher than in COM and the uptake into droplets is less efficient. In Eastern Europe the mixing ratio near the surface in SCM is lower than in COM, resulting mainly from the altered chemical composition (indirect effect), whereas in the polar regions with low mixing ratios the absolute differences are small. The lower left panel, presenting the EASY simulation, in which the liquid phase oxida-
tion of SO$_2$ is additionally neglected, the differences are larger. In the polluted regions of the Northern Hemisphere (eastern USA, western Europe, China) substantially higher H$_2$O$_2$ mixing ratios are calculated due to this neglect. The EASY2 simulation shows very similar average near surface mixing ratios for H$_2$O$_2$, since the processes in which hydrogen peroxide is involved are almost identical. This results in a comparable distribution as in EASY. However, especially over China in both model setups based on a Henry's law equilibrium substantially higher H$_2$O$_2$ mixing ratios are calculated due to the missing SO$_2$ oxidation.

The vertical distribution for the five simulations (Sec. 5.1 of the supplement) shows highest differences close to the surface in the tropics. In the middle troposphere these differences are smaller, and in the upper troposphere they are hardly discernable for SCM, EASY or EASY2. This is mainly based on the H$_2$O$_2$ photolysis in the gas phase which becomes more efficient with altitude. However, close to the surface the photolysis in the liquid phase has substantial effects producing the lower mixing ratios in the COM simulation compared to the sensitivity studies. Due to the vertical mixing in the troposphere, the higher mixing ratios of the sensitivity studies are transported upwards into the middle troposphere as mentioned above.

In the NOSCAV simulation (see Sec. 4 of the supplement), the missing sink of wet deposition leads to slightly higher mixing ratios compared to the reference in the polar and the marine subsidence regions, whereas in almost all other locations the values are lower (most obvious over the tropical continents, but also over parts of the oceans, e.g., the SPCZ). These regions are strongly influenced by convective activity, in which the missing wet removal results in an enhanced upward transport into the upper troposphere which is not balanced by downward transport within the aqueous phase. The vertical distribution (see supplement) indicates higher mixing ratios by more than 50% compared to the COM simulation.

Additionally, among the indirect effects it must be considered that over the tropical continents the simplified and therefore underestimated wet removal results in higher mixing ratios of organic compounds. These species can partly deplete OH and HO$_2$.
by oxidation processes (e.g., since CH$_3$OOH is not scavenged in NOSCAV the HO$_x$ mixing ratios are reduced (see reactions G4107 and G4103 in supplement). In case of a complete neglect of scavenging those effects are even stronger. Consequently, less H$_2$O$_2$ is formed from the self reaction of HO$_2$. Due to the low NO$_x$ mixing ratios in these remote regions a recycling of the oxidants is not as efficient as in the polluted regions of the Northern Hemisphere. Consequently, the total global mass of H$_2$O$_2$ is by far highest in NOSCAV throughout the middle and upper troposphere, whereas EASY, EASY2 and SCM calculate a similar atmospheric burden. Since all sensitivity simulations with simplified cloud and precipitation chemistry calculate higher mixing ratios of H$_2$O$_2$, this compound contributes to enhanced OH production through its photodissociation in the gas phase, and therefore accelerates the gas phase chemistry (contributing through indirect effects).

This is also supported by the comparison with observations in Fig. 7. The vertical profiles for COM and SCM are very similar for the TOPSE campaign. EASY2 shows slightly enhanced mixing ratios, whereas EASY and NOSCAV result in substantially higher values. However, close to the surface in NOSCAV nearly comparable low values are calculated. For the TRACE-P data in China and Hawaii, COM is slightly lower than SCM. The EASY simulation is characterised by the highest values in the lower troposphere, but NOSCAV, being similar or lower than COM close to the surface, yields significantly higher H$_2$O$_2$ values than the other simulations above 2 km altitude. As expected, the NOSCAV simulation does not reproduce the measured H$_2$O$_2$ mixing ratios, and strongly overestimates the mixing ratios at altitudes above 2 km. Similarly, in the EASY simulation the missing sinks by aqueous phase chemistry result in an overestimation relative to the observations. EASY2 calculates higher values than COM and SCM, but still lower than EASY. This is mainly restricted to the lower and middle troposphere, whereas in the upper troposphere, in which only small amounts of liquid water are available for chemical reactions in droplets, the differences between the simulations including the wet removal are smaller because of the rapid photolysitical destruction. The COM and SCM simulations both lead to too high mixing ratios
compared to the TOPSE data, although they are in reasonable agreement with the TRACE-P data, with the mean value well within the range of the measurements. We conclude that for a realistic description of lower and middle tropospheric H$_2$O$_2$ mixing ratios at least the basic cloud and precipitation chemistry of this compound must be included in the model (i.e., SCM). An enhanced scavenging coefficient, i.e. larger than expected from Henry’s law equilibrium, might be used as an approximation, but this cannot be determined easily for all conditions in the troposphere. Additionally, this approximation suppresses important chemical feedbacks.

4.3.2 Sulphur dioxide (SO$_2$)

Since SO$_2$ is relatively short-lived, the distribution near the surface, presented in Fig. 8, shows strong gradients with the highest values close to the emission centres, i.e. Europe, eastern China, and additionally the eastern USA, India, Saudi-Arabia and Chile. Over the ocean the DMS (dimethylsulfide) oxidation and to a lesser extent ship emissions result in average mixing ratios of less than 0.5 nmol/mol. In case of a reduced aqueous phase chemistry as in the SCM simulation, in which the main oxidation of S(IV) to S(VI) by H$_2$O$_2$ and O$_3$ is nevertheless considered, the differences to the reference are relatively small except in the heavily polluted regions near the strongest SO$_2$ sources. This results mainly from the altered H$_2$O$_2$ mixing ratios, and the consequent changes in oxidant mixing ratios (indirect effect), but also from the more detailed sulphur chemistry in the aqueous phase, e.g., interactions between oxidised sulphur and nitrogen compounds (direct effect). Overall, in SCM this leads to slightly enhanced SO$_2$ mixing ratios compared to COM, both close to the surface and at higher altitudes. In the EASY simulation the aqueous phase oxidation of sulphur dioxide is neglected, i.e., a main chemical sink is missing, and the scavenging is less effective as in SCM or COM. This leads to globally higher mixing ratios, most strongly in the regions near the emission sources. In EASY2 the dissociation of SO$_2$ in the liquid is taken into account, but the liquid phase oxidation is still not considered. This cannot be easily parame-
terised with the help of a coefficient since it depends not only on the pH, but also on the dissolved oxidant concentrations, which show large variability in time and space (e.g., H$_2$O$_2$, cf. Fig. 6). Consequently, the atmospheric content of sulphur dioxide is still overestimated and the sulphate wet deposition is underestimated (cf. Tab. 4).

In the NOSCAV simulation higher mixing ratios at the surface are calculated, too, but in this case they are not as high as in EASY, mainly related to the more effective vertical transport. The overall global mass of SO$_2$ is not highest in this simulation, since the increased oxidising capacity leads to an amplified H$_2$SO$_4$ production in the gas phase, resulting in higher S(VI) and lower S(IV) content of the atmosphere. In the EASY, EASY2 and NOSCAV simulations the absolute amounts of SO$_2$ are by a factor of up to 3 larger than in the COM and SCM simulations.

The comparison with the Emmons et al. (2000) data in Fig. 9 for the TOPSE campaign shows a strong overestimation for all simulations close to the surface over Boulder, whereas for Churchill, the COM and SCM simulations largely capture the observed SO$_2$. This is also valid for the TRACE-P data over China and Hawaii. Only in the lower two kilometres the mixing ratios at some specific locations are overestimated. However, since this occurs in all sensitivity studies it might indicate that emissions are overestimated and/or that mixing between the boundary layer and free troposphere is underestimated. Nevertheless, at typical cloud base levels between one and two kilometres altitude, both the COM and SCM simulation capture the efficient uptake into clouds and precipitation and subsequent wet deposition of sulphur compounds, indicating that scavenging processes are well represented. The weaknesses of EASY, EASY2 and NOSCAV are obvious, because the simulated gas phase mixing ratios are substantially overestimated compared to the observed profiles. The dissociation of SO$_2$ in the droplets taken into account in EASY2 does not contribute efficiently to a better representation of the S(VI) uptake and wet removal.

In summary, the scavenging and wet deposition description by the effective or absolute Henry’s law coefficients is not sufficient to simulate mixing ratios as low as observed, whereas a basic aqueous phase chemistry (such as SCM) oxidising dissolved SO$_2$
yields more realistic values.

4.3.3 Nitric Acid (HNO₃)

The surface mixing ratio of HNO₃ in the COM simulation, depicted in the upper left panel of Fig 10, shows highest values in regions with strong NOₓ emissions through the rapid chemical transformation of NO₂ by OH during daylight and heterogeneous removal of N₂O₅ during night time. The NOₓ emissions by industry and traffic (Europe, USA, China, partly India), soil emissions (tropical rain forests), biomass burning (e.g. Central Africa), all significantly contribute to the surface mixing ratios of HNO₃. Lightning as a NOₓ source is more important in the upper troposphere. Scavenging and subsequent wet deposition is the main sink of HNO₃, being much more efficient than dry deposition and photodissociation. In the SCM simulation (upper right panel of Fig 10) similar nitric acid mixing ratios near the surface are computed. Only in the eastern USA and Europe slightly higher values are calculated, whereas almost everywhere else the differences to COM are rather small. The differences mainly result from the indirect effects and to a lesser extent from interactions with sulphur compounds in the aqueous phase.

The reduced scavenging in EASY, neglecting the dissociation of HNO₃ in the liquid, leads to higher near-surface mixing ratios worldwide. However, due to the high solubility an important fraction of nitric acid is nevertheless scavenged and removed from the atmosphere by wet deposition. The largest differences occur in the regions with highest overall mixing ratios, emphasising the underestimation of precipitation loss by simplified scavenging, whereas in regions remote from the sources the differences are mostly small. Nevertheless, in this simulation the nitric acid is on average substantially higher than in the other setups. In contrast to SO₂, the near surface mixing ratios in the EASY2 simulation are very close to those of COM (lower right panel of Fig. 10). This emphasises that the concept of effective Henry's law coefficients is in general applicable for species that mainly dissociate and are only of minor importance for aqueous phase chemistry. The assumed pH of 5 is within the range of that expected (varying
from 3.5 to 6, cf. Fig. 5, that the scavenging of nitric acid, which almost completely dissociates into nitrate above a pH value of 1, is approximated sufficiently well. On the other hand it should be noted, that, even though the near surface mixing ratio and also the zonal average distribution of gaseous nitric acid in EASY2 are very close to the COM setup, the nitrate wet deposition in EASY2 is substantially higher than in COM and the observations (cf. Table 4). Additionally, the dry deposition of HNO$_3$ is larger in the EASY2 setup, both indicating that due to the altered oxidation capacity in EASY2 more HNO$_3$ is produced and subsequently removed by dry and wet deposition.

In the NOSCAV simulation almost everywhere unrealistically enhanced HNO$_3$ mixing ratios are calculated near the surface (see supplement Fig. 4c). In regions with strong sources the differences are largest although they propagate over large distances due to the enhanced lifetime. Compared to EASY, the absolute values of the differences to COM are slightly smaller resulting from the more efficient vertical transport due to the neglect of depletion by clouds and precipitation. The global mass of nitric acid is very similar for COM, SCM and EASY2, showing that the sink process of wet removal is simulated acceptably with the simplified chemistry, whereas EASY and - much worse - NOSCAV substantially overestimate the HNO$_3$ content of the atmosphere.

The calculated HNO$_3$ mixing ratios for TOPSE and TRACE-P for both the COM and SCM simulations are well within the range of the observations (Emmons et al., 2000), whereas EASY and much worse NOSCAV strongly overestimate the HNO$_3$ mixing ratios. In EASY2 the HNO$_3$ values are only slightly higher than in COM and SCM. The computed vertical distributions typically indicate a C-shape profile resulting from the NO$_x$ emissions at the surface and downward transport of HNO$_3$ from the stratosphere. Since the vertical resolution in the tropopause region in the L31 model configuration is relatively coarse, cross-tropopause transport is not represented as well as in the L90 setup of Jöckel et al. (2006), leading to an overestimation of the influx into the troposphere from above. For the TRACE-P measurements, in regions where the tropopause is located at higher altitudes and the stratospheric influence at 12 km altitude is less important, this C-shape is less evident.
For nitric acid, being hardly chemically active in the liquid phase, except for its almost complete dissociation, the atmospheric content and the wet deposition can be simulated with simplified chemistry as well as with an effective Henry’s law equilibrium.

4.3.4 Formaldehyde (HCHO)

Formaldehyde (HCHO) is moderately soluble and only marginally reactive in the liquid phase. Therefore, wet deposition is one of the sinks to consider in global budgets, even though gas-phase chemical processes are globally dominant and dry deposition is also more important. The near-surface mixing ratios are presented for the COM simulation in the upper left panel of Fig. 12. Highest mixing ratios typically occur over the tropical rain forests as a result of biogenic NMHC emissions and subsequent oxidation reactions, whereas in some regions anthropogenic NMHC are also significant. The differences of SCM to the COM simulation mainly result from the indirect effect (since only few additional reactions in the aqueous phase are considered in COM). They are most pronounced in regions where mixing ratios are highest. For instance, significant differences occur in polluted industrial regions, in which the oxidation capacity of the troposphere is modified in this simulation setup. This also applies to the rainforest regions, where hydrocarbon compounds compete for the available oxidants. The differences between EASY and COM are similar, again mainly due to the indirect effects. This is evident since there are no aqueous phase chemical reactions including HCHO in SCM and EASY. Therefore, only the altered oxidant distributions can contribute to the patterns of relative differences between simulations, as well as the neglect of the gas phase diffusion limitation. This is also valid for EASY2 (lower right panel of Fig. 12), in which HCHO is treated as in EASY. Consequently, the small differences in the average near surface distributions between EASY and EASY2 are related to the indirect effects of the modified oxidation capacity. By neglecting scavenging (NOSCAV) the formaldehyde mixing ratios near the surface over the continents are generally lower than in the COM simulation. Similar as for H₂O₂, the vertical redistribution of HCHO by convection, without the balancing
downward transport in precipitation and the convective transport of oxidants, are the main causes for these patterns. For the same reason in the middle and upper troposphere the overall mixing ratios are globally higher in NOSCAV than in COM. The representation of the mixing ratio profiles as measured during the TOPSE and TRACE-P campaigns is realistic for all simulations, with COM and SCM usually closest to the observations, and small differences between all simulations. The vertical shape is also captured relatively well by all sensitivity studies. Since wet deposition is only of minor importance for the formaldehyde budget, the differences in the vertical HCHO distributions are largely determined by differences in the gas phase chemistry (indirect effects).

4.3.5 Ozone (O$_3$)

Figure 14 presents the annual average surface mixing ratio of ozone of the COM simulation in the upper left panel. The results from the COM simulation near the surface are very close to those presented in the O$_3$ evaluation by Jöckel et al. (2006), showing relatively high mixing ratios in California (e.g. Jacobson, 2005), the Mediterranean region (e.g. Lelieveld et al., 2002), around the Persian Gulf (resulting from high propane emissions), northern India and the Himalayas (the latter associated with the strong surface elevation). The absolute differences between the SCM and COM simulations, depicted in the upper right panel of Fig. 14, mostly result from indirect effects, in particular via the removal of nitrogen oxides. Nevertheless, direct chemical effects through reduced O$_3$ formation and enhanced O$_3$ destruction in clouds play a secondary role over parts of the eastern USA, western Europe, South America and near Japan. On the other hand, in eastern Europe and near Indonesia and the Indian Ocean cloud chemical processes enhance ozone mixing ratios near the surface, mostly because chemical destruction pathways of O$_3$ are reduced. The ozone mixing ratios computed in the EASY simulation are predominantly higher than in COM, largely resulting from the less effective removal of O$_3$ precursor gases. This results
in relative increases of up to about 20% in some locations, as shown in the lower
left panel of Fig. 14, whereas in Europe and near Indonesia the reduction of \( \text{O}_3 \) loss
is more important. With the effective Henry's law equilibrium in EASY2 (lower right
panel of Fig. 14) the differences to COM are much smaller than in EASY, but overall
slightly more \( \text{O}_3 \) is simulated in the surface layer, again resulting mainly from indirect
effects. The simulation without cloud processes and scavenging (NOSCAV) generally
results in much higher and unrealistic \( \text{O}_3 \) mixing ratios throughout the troposphere,
resulting from the drastically altered concentrations and distributions of precursor
gases. Unquestionably, the cloud and precipitation related chemistry and removal
processes yield much reduced oxidant levels.

Figure 15 presents a comparison of model results with aircraft observations. From
the TOPSE measurements an increase of \( \text{O}_3 \) mixing ratios with altitude near the
tropopause illustrates the influence of stratospheric \( \text{O}_3 \). For the lower and middle
troposphere the COM, SCM, EASY2 and EASY simulations produce rather similar
results, being close to the measurements, while the EASY results for \( \text{O}_3 \) are sys-
tematically higher. The NOSCAV simulation strongly overestimates \( \text{O}_3 \) mixing ratios
throughout the troposphere, and is unable to reproduce the observations. In general,
the observations are reproduced quite accurately by the COM and SCM simulations,
although for Hawaii all simulations overestimate \( \text{O}_3 \), with COM and SCM being slightly
lower. In the upper troposphere in the model the influence of the stratosphere appears
to be too strong, resulting from an overestimate of transport into the troposphere
due to the coarse vertical resolution. Jöckel et al. (2006) show that this process
is simulated quite realistically using the model configuration with enhanced vertical
resolution near the tropopause (90 vertical levels).

4.3.6 Summary and discussion

A statistical analysis of the sensitivity simulations for several species, comparing our
model results to observations of the updated Emmons et al. (2000) dataset, is shown
by the Taylor diagram in Fig. 16. In addition, Table 5 lists the mean biases in units of standard deviations.

With respect to ozone it appears that the overall performance of the COM and SCM simulations, with a correlation of $R \approx 0.85$ and a normalised $\sigma^* > 1.5$, is poorer than in Jöckel et al. (2006), and we emphasise that these results are strongly influenced by $O_3$ in the upper troposphere through transport from the stratosphere. Not considering the results above 9 km altitude, we achieve a better correlation ($R \approx 0.9$) and an almost ideal $\sigma^*$ close to one. The quality of the EASY and EASY2 simulations with respect to spatial variation and correlation are comparably good, whereas the NOSCAV simulation is substantially worse. If we additionally take into account the bias for $O_3$ (Table 5), the COM simulation performs best, although SCM is not much worse. The EASY setup appears less suited to adequately simulate ozone, whereas with the help of the EASY2 setup comparable agreement with the simulations explicitely calculating cloud and precipitation chemistry is achieved.

Also for $H_2O_2$ the COM simulation achieves best agreement with observations, with a very high correlation, $R > 0.98$, and a $\sigma^*$ close to one. SCM is not very different with respect to $R$; however, the temporal and spatial variation is less well captured. Again EASY seems to perform reasonably well with respect to correlation and $\sigma^*$, although the COM results are clearly superior. A bias in COM is nearly absent, small in SCM, enhanced in EASY2, and much larger in the EASY setup.

The performance of the COM model setup for $SO_2$ is similarly good, again slightly better than SCM. In contrast, in the EASY and EASY2 setup the $SO_2$ observations are poorly reproduced (out of scale in Fig. 16, $\sigma^* \approx 6$), caused by the neglect of in-cloud oxidation of $SO_2$ in these setups. On the other hand, a significant bias is found for both COM and SCM, both underestimating $SO_2$ mixing ratios, whereas the bias in EASY is similar but of opposite sign. The underestimation in COM and SCM is partly a consequence of neglecting volcanic $SO_2$ emissions globally $\approx 8 – 10$ Tg S/yr according to Graf et al. (1997), and to a lesser extent due to the uncertainty in the parameterised marine sulphur (DMS) emissions. In some polluted regions the model
SO\textsubscript{2} emissions might be overestimated (e.g., too high surface mixing ratios in China in the third panel of Fig. 9), although the wet deposition of sulphate seems to agree well with observations. Nitric acid is also represented quite well by the model, with SCM being slightly better with respect to the variability compared to COM, but with a similar correlation. Also EASY is similarly good with respect to these two parameters, because HNO\textsubscript{3} scavenging is largely controlled through its high solubility. However, from the bias it appears that both COM and SCM underestimate the gas phase HNO\textsubscript{3} mixing ratios, but less than for SO\textsubscript{2}, whereas in EASY HNO\textsubscript{3} is overestimated more strongly. EASY2 seems to agree very well with the observations, however as mentioned above, the deposition sinks are much stronger than in the observations and other simulations. The model appears to be very well suited for simulating formaldehyde mixing ratios; comparably good values with a high correlation $R>0.95$ for COM and SCM and $\sigma^*$ varying between $\approx 0.9$ and 1 are achieved. In all simulations HCHO seems slightly overestimated, though least in COM.

Formic acid (HCOOH), not discussed in detail above, is not captured very well. Even though the correlation exceeds 0.8, the variability is not captured accurately, being best in NOSCAV and EASY, indicating that the source description of HCOOH is inadequate. The biases are all negative, pointing to an underestimation of gas phase mixing ratios. Again, the bias is smallest for COM, although improvements of the processes that control HCOOH in the model are required.

In summary, the COM simulation results agree best with the observations, show the lowest biases, the highest correlations, and normalised $\sigma^*$ values closest to one. This implies that our detailed treatment of cloud and precipitation chemistry improves the simulated trace gas mixing ratios and distributions, and most realistically represents the chemical composition of the troposphere. The effective Henry’s law coefficients (as in EASY2) usually applied for wet deposition calculations in global chemistry models can achieve realistic mixing ratios for some species (e.g., HNO\textsubscript{3} or O\textsubscript{3}), whereas for
other compounds this approach is insufficient (e.g., SO₂).

Due to the multitude of direct and indirect effects of multiphase chemical processes, the overall influence of clouds and precipitation cannot easily be quantified. However, we reiterate that in contrast to most previous studies in our simulations additionally to cloud chemical effects the chemical processes in falling precipitation are taken into account.

For example, Liang and Jacob (1997) rate the influence of multiphase chemistry on ozone as minor, whereas Lelieveld and Crutzen (1991) rate it to be important. It is critical to carefully distinguish which processes have actually been addressed in these studies. If cloud and precipitation processes are not considered (as in NOSCAV), the effects on many trace gases (even the poorly soluble O₃) are very large, as indicated by Lelieveld and Crutzen (1991), who compared the overall effects between simulations with and without clouds. Even by considering highly simplified aqueous phase sink processes (in EASY) the effects of clouds on ozone are significant. Liang and Jacob (1997), on the other hand, consider a different scenario on a regional scale, more comparable to the differences between the SCM and COM simulations (SCM including a realistic description of scavenging and deposition, and COM additionally considering comprehensive liquid phase chemistry). As analysed with the help of Figs. 14 and 15 (also confirmed by additional figures in the supplement) this influence on global tropospheric O₃ is small, especially also because localised positive and negative effects tend to cancel through transport processes (Barth et al., 2002). Due to the low solubility of O₃, the direct effect of relatively detailed cloud chemistry is small. For compounds with a larger solubility this direct effect gains importance as well as the indirect effects. Furthermore, transition metal chemistry in clouds, neglected in the present work, may additionally lead to ozone loss, as suggested by Matthijsen et al. (1997), although in their regional model the direct effect on O₃ is small. Zhang et al. (2006) derive a strong dependence of aqueous phase chemistry on ambient NH₃, H₂O₂ and SO₂ mixing ratios, studied on the cloud scale. This is confirmed in
our study for the global scale. Overall, most studies consistently find that cloud and precipitation processes have a significant influence on tropospheric chemistry on all scales.

5 Conclusions

Our model results, considering aqueous phase chemistry in clouds and precipitation, show that the representation of scavenging according to absolute Henry's law coefficient alone is not sufficiently accurate for tropospheric chemistry simulations. Furthermore, the applicability of effective Henry's law coefficients is limited because equilibrium constants are highly dependent on the chemical composition of the aqueous phase, which can vary strongly. By considering a basic (minimum) chemical reaction mechanism for cloud and precipitation droplets the accuracy of the simulation results improve substantially. Nevertheless, the use of the ECHAM5/MESSy1 model as in the COM setup, including comprehensive multiphase chemistry, yields superior agreement with observed wet deposition fluxes, especially in view of the uncertainties associated with the heterogeneity of precipitation and wet deposition patterns.

For the first time the prognostic treatment of ion-concentrations in the droplets (including H$_3$O$^+$) with the newly developed SCAV submodel (Tost et al., 2006a) enables the explicit computation of pH values of clouds and precipitation on a global scale. The simulated global pH distributions agree well with observations. Since observations are scarce and mostly limited to elevated measurement stations (thus sampling orographic clouds) the results indicate that some detailed regional features are not resolved in sufficient detail. Improved model simulations may be expected from a higher model resolution and improvements of the aqueous phase chemistry and scavenging model formulations. For instance, the interaction of trace species with ice clouds and falling frozen precipitation will be taken into account in detail as a next step.

The simulations of detailed cloud and precipitation chemistry points to direct and in-
direct influences on the global distributions of many trace gases. The overall effects range from relatively small, e.g., for global O$_3$ (<10%), to very large for HNO$_3$, SO$_2$, H$_2$O$_2$ and intermediate for HCHO. The detailed treatment of aqueous phase chemistry generally improves the accuracy of model simulated tropospheric trace gas concentrations, including O$_3$. Yet, a simplified and computationally much cheaper mechanism (SCM) performs sufficiently well, and may be used in long-term simulations.

Acknowledgements. We acknowledge the work of G.-J. Roelofs, who introduced the first scavenging mechanism in ECHAM4, as well as L. Ganzeveld for the implementation into ECHAM5, providing a starting point for the present study. Additionally, we thank A. Bott for the co-supervision of the PhD thesis of H. Tost on which this paper is based. We thank F. Dentener for providing the collection of deposition data and acknowledge the work of all the measurement initiatives for the observational data. Furthermore, we wish to acknowledge the use of the Ferret program for analysis and graphics in this paper. Ferret is a product of NOAA’s Pacific Marine Environmental Laboratory. (Information is available at http://ferret.pmel.noaa.gov/Ferret/). Additionally we thank all MESSy developers and users for their support, hints, proposals and discussions. This study is part of the ENIGMA project; the authors thank the Max Planck Society for support and funding.

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http://www.atmos-chem-phys.net/5/433/2005/.
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789, 791, 792, 793, 799, 804, 812, 814, 815, 816, 833


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<table>
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<th>NH$_x$</th>
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Table 1. Number of wet deposition observations from the individual networks.
Table 2. Correlation for nitrate, sulphate and NH$_x$ from this study and D06.

<table>
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<tr>
<th></th>
<th>nitrate this study</th>
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<th>NH$_x$ this study</th>
<th>D06</th>
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Table 3. Average annual deposition values for nitrate (in mg N/m²), sulphate (in mg S/m²) and NH₅ (in mg N/m²) from this study and D06 and the corresponding measured values as applied in the respective studies.

<table>
<thead>
<tr>
<th></th>
<th>nitrate (mg N / m²)</th>
<th>sulphate (mg S / m²)</th>
<th>NH₅ (mg N / m²)</th>
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</table>

This table shows the average annual deposition values for nitrate, sulphate, and ammonia from this study and D06, along with the corresponding measured values as applied in the respective studies.
Table 4. Average annual deposition values for nitrate (in mg N/m²), sulphate (in mg S/m²) and NH₅ (in mg N/m²) for the observed values and for the four sensitivity studies (upper table) and correlation of the wet deposition fluxes with the observations for the four sensitivity studies (lower table).
Table 5. Bias in units of standard deviation (geometric average of model and observations $\sqrt{\sigma^2_{\text{sim}} + \sigma^2_{\text{obs}}}$) for the individual species in the five simulations.

<table>
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<tr>
<th>Species</th>
<th>COM</th>
<th>SCM</th>
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<th>EASY2</th>
<th>NOSCAV</th>
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<td>0.25</td>
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<td>0.03</td>
<td>0.16</td>
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<td>−0.22</td>
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<tr>
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<td>−1.01</td>
<td>1.81</td>
<td>1.69</td>
<td>1.22</td>
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</tbody>
</table>
Fig. 1. Annual average daily precipitation in mm/day of the EVAL S1 (Jöckel et al., 2006) simulation (upper panel) and fraction of the large-scale to the total precipitation (lower panel).
Fig. 2. Average annually accumulated wet deposition of nitrate in mg N / m² (upper panel) and comparison with observations in a Taylor diagram (lower panel) showing the correlation and the standard deviation of the model normalised to the standard deviation of the observations. The individual years of the simulation (colour coded) are each compared with the measurement data set. The symbols denote the different measurement networks and the composite of all observations (pentagons).
Fig. 3. Average annually accumulated wet deposition of sulphate in mg S / m² (upper panel) and comparison with observations in a Taylor diagram (lower panel). For the notation see Fig. 2.
Fig. 4. Average annually accumulated wet deposition of ammonia and ammonium in mg N / m² (upper panel) and comparison with observations in a Taylor diagram (lower panel). For the notation see Fig. 2.
Fig. 5. Annual average precipitation pH at the surface from large-scale and convective precipitation (left) and annual zonal average of the cloud pH (right).
Fig. 6. Comparison of the annual average H$_2$O$_2$ surface layer mixing ratio for the four sensitivity simulations: shown is the absolute value for the COM simulation (upper left panel) and - as indicated - the absolute differences of the other simulations to the COM simulation.
Fig. 7. Comparison of the H$_2$O$_2$ vertical profiles compiled by Emmons et al. (2000) with the results of the five sensitivity simulations. The observations are depicted by the black points, their variability (standard deviation) by the black boxes and the number of points taken into account for this campaign (in space and time for this region) are listed at the right axis. The model results are represented by the solid and the respective standard variation by the dashed lines. The colours denote the simulations as indicated in the legend, i.e., red = COM, green = SCM, blue = EASY, turquoise = EASY2 and magenta = NOSCAV.
Fig. 8. Comparison of the annual average SO$_2$ surface layer mixing ratio for the four sensitivity simulations, displayed similarly as in Fig. 6.
Fig. 9. As Fig. 7, but for SO₂.
Fig. 10. Comparison of the annual average HNO₃ surface layer mixing ratio for the four sensitivity simulations, displayed similarly as in Fig. 6.
Fig. 11. As Fig. 7, but for HNO$_3$. 
Fig. 12. Comparison of the annual average HCHO surface layer mixing ratio for the four sensitivity simulations, displayed similarly as in Fig. 6.
Fig. 13. As Fig. 7, but for HCHO.
Fig. 14. Comparison of the annual average O$_3$ surface layer mixing ratio for the four sensitivity simulations, displayed similarly as in Fig. 6.
Fig. 15. As Fig. 7, but for O$_3$. 
Fig. 16. Taylor diagram depicting the correlation and the normalised standard deviation $\sigma^*$ of the simulations with the Emmons et al. (2000) database of observations. The different simulations are represented with different colours, the selected species with different symbols.