Abstract

There is growing interest in the formation of secondary organic aerosol (SOA) through condensed aqueous phase reactions. In this study, we use a global model (IMPACT) to investigate the potential formation of SOA in the aqueous phase. We compare results from several multiphase process schemes with detailed aqueous phase reactions to schemes that use a first order gas-to-particle formation rate based on uptake coefficients. The predicted net global SOA production rate in cloud water ranges from 13.1 Tg/yr to 46.8 Tg/yr while that in aerosol water ranges from -0.4 Tg/yr to 12.6 Tg/yr. The predicted global burden of SOA formed in the aqueous phase ranges from 0.09 Tg to 0.51 Tg. A sensitivity test to investigate two representations of cloud water content from two global models shows that increasing cloud water by an average factor of 2.7 can increase the net SOA production rate in cloud by a factor of 4 at low altitudes (below approximately 900 hPa). We also investigated the importance of including dissolved iron chemistry in cloud water aqueous reactions. Adding these reactions increases the formation rate of aqueous phase OH by a factor of 2.6 and decreases the amount of global aqueous SOA formed by 31%. None of the mechanisms discussed here is able to provide a best fit for all observations. Rather, the use of an uptake coefficient method for aerosol water and a multi-phase scheme for cloud water provides the best fit in the Northern Hemisphere and the use of multiphase process scheme for aerosol and cloud water provides the best fit in the tropics. The model with iron chemistry under predicts oxalate measurements in all regions. Finally, the comparison of O/C ratios estimated in the model with those estimated from measurements shows that the modeled SOA has a slightly higher O/C ratio than the observed SOA for all cases.
1. Introduction

Secondary organic aerosol (SOA) has been shown to be an important component of non-refractory submicron aerosol in the atmosphere (Zhang et al., 2007; Jimenez et al. 2009). SOA is known to form from the gas/particle partitioning of semi-volatile organic compounds produced by gas phase photochemistry (Pankow, 1994; Odum et al., 1996). However, models that only include this SOA formation mechanism typically underestimate the SOA mass as well as the oxygen-to-carbon (O/C) ratio (e.g., DeGouw et al., 2005; Heald et al., 2005; Volkamer et al., 2006; Dzepina et al., 2009). In addition, the observed O/C ratios in aged ambient organic aerosol (OA) cannot be explained using measured O/C ratios in dry smoke chamber experiments (Aiken et al. 2008; Ng et al., 2010). One method that has been used to help close the gap between measured and modeled SOA is to use a refined treatment for primary organic aerosol (POA) that allows them to evaporate and further oxidize (Robinson et al. 2007; Pye and Seinfeld, 2010; Hodzic et al., 2010; Lee-Taylor et al., 2011). However, there are large uncertainties in how to treat the evaporation rate as well as the oxidation mechanism for POA and thus the SOA yield from this source (Pye and Seinfeld, 2010; Spracklen et al. 2011).

Aqueous phase processing, as a complementary pathway to gas/particle partitioning of semi-volatile and low volatility gases, has the potential to enhance both SOA mass and the O/C ratio in atmospheric OA. Water-soluble and polar gases are taken up by the aqueous phase and can be oxidized in water thereby leading to the production of low volatility substances (e.g., organic acids and especially their corresponding salts, oligomers, and organosulfates) (Blando and Turpin, 2000; Warneck, 2003; Liggio et al., 2005; Sorooshian et al., 2007; Tan et al., 2009). These low-volatility products
predominately stay in the particle phase after water evaporation (Blando and Turpin, 2000; El Haddad et al., 2009) and tend to have higher O/C ratios than those that form in gas phase reactions (Herrmann et al., 2005; Lim et al., 2010; Ervens and Volkamer, 2010). This is because the precursors for the aqueous phase reactions tend to be small compounds with low-molecular weight (MW) that already have high O/C ratios, and the aqueous oxidation of these small compounds either tends to add O-containing functional groups to C-C bonds (thereby forming higher O/C ratio dicarboxylic acids) or to react with themselves to keep the same carbon structure and the same O/C ratio (thereby forming oligomers) (Ervens et al., 2011).

Laboratory studies have shown that aqueous phase reactions can produce SOA from C2 and C3 carbonyl compounds including glyoxal, methylglyoxal, glycolaldehyde, pyruvic acid and acetic acid (Ervens et al., 2003; Carlton et al., 2006, 2007; Altieri et al., 2008; Perri et al., 2009; Tan et al., 2009, 2010, 2012; Lim et al. 2010). The major products from the oxidation of these carbonyl compounds are carboxylic acids, of which the most abundant is oxalic acid. Oxalic acid is also observed to be part of the aerosol emitted in biomass burning (Kundu et al., 2010) and is observed to be formed from the photochemical ageing of OA (Eliason et al., 2003). In contrast to reactions in cloud water, the major products formed from the reactions of C2 and C3 carbonyl compounds in aerosol water are oligomers (Ervens and Volkamer, 2010; Lim et al., 2010).

Several models to date have been developed to estimate the amount of SOA formed in the aqueous phase (denoted aqSOA hereafter). Chen et al. (2007) used a box model to investigate the formation of SOA from aqueous phase reactions in cloud water using an explicit aqueous mechanism that included organic acid formation from glyoxal and
methylglyoxal; they found that aqueous phase processing increased the SOA concentration by 27% for a rural scenario and by 7% for an urban scenario. Carlton et al. (2008) used a simplified mechanism with a fixed yield of 4% for the conversion of glyoxal to aqSOA in a regional air quality model. The model’s ability to predict the concentration of water soluble organic carbon (WSOC) from aircraft measurements over northeastern US was improved and the bias between predicted SOA mass and observations was decreased from -64% to -15%. Myriokefalitakis et al. (2011) and Liu et al. (2012) employed a global model to study the formation of SOA in clouds and found that the global production rate could be 13-30 Tg/yr. The contribution of aerosol water to the formation of aqueous phase SOA from glyoxal and methylglyoxal has also been estimated in global models (Fu et al., 2008, 2009; Stavrakou et al., 2009; Lin et al., 2012). The amount of irreversible uptake of glyoxal and methylglyoxal into aerosol water was predicted to be much smaller than that in cloud droplets, when the same reactive uptake parameter ($\gamma = 2.9 \times 10^{-3}$) was used (Fu et al., 2008, 2009). This result was confirmed by Lin et al. (2012) using the same basic approach. Stavrakou et al. (2009), however, concluded that the source of SOA over the continents from uptake of glyoxal by aerosol particles was at least as large as that from cloud droplets or even up to 60% larger. These authors implemented an in-cloud formation parameterization similar to Ervens et al. (2008) rather than using an uptake formation scheme as in Fu et al. (2008, 2009) and Lin et al. (2012).

Uncertainties in the formation of aqSOA stem from uncertainties in how to treat its formation rate. The use of a single reactive uptake parameter $\gamma$ in aqSOA formation as in Fu et al. (2008, 2009) and Lin et al. (2012) implies a surface-limited uptake process, but
Ervens and Volkamer (2010) found no correlation between the SOA volume increase and the seed aerosol surface area for most of seed aerosols used the laboratory study by Volkamer et al. (2009). Instead, the SOA volume increase was shown to be linear to the total water mass on these seed particles, implying that SOA was formed as a result of a bulk process. On the other hand, the positive correlation between the SOA volume increase and the surface area observed for mixed sulfate/fulvic acid seed particles in the study by Ervens and Volkamer (2010) suggests that a surface process dominated the formations on these mixed particles. Waxman et al. (2013) also found that a surface limited uptake process could explain the observed gas-phase glyoxal mass in Mexico City. Finally, the use of an identical reactive uptake parameter for both cloud droplets and aerosol water, as in Fu et al. (2008) does not account for differences in the chemistry of carbonyl compounds between cloud water and aerosol water (Lim et al., 2010; Ervens and Volkamer, 2010).

Other uncertainties in treating the formation of aqSOA include the treatment of aqueous chemistry as well as the amount of cloud water represented in a model. Iron chemistry in cloud water has been shown to be a major source of aqueous OH (Ervens et al., 2003; Deguillaume et al., 2005), which is known to initiate the oxidation of glyoxal and methylglyoxal. However, no model, to our knowledge, has included the oxidation of iron in simulating the formation of aqSOA in cloud water. In addition, the amount of cloud water in a model can influence SOA production rates (Liu et al. 2012; He et al., 2013).

In this paper, we focus on the aqueous formation of SOA from C2 and C3 carbonyl compounds not only because they are highly water soluble but also because most of the existing laboratory studies use these compounds as surrogates to examine aqueous SOA
formation, so that sufficient information exists to derive their reaction mechanisms. We
use a 3-D chemical transport model to test three representations of aqueous chemistry
leading to SOA formation: two detailed mechanisms using explicit multi-phase
gas/aqueous phase chemical mechanisms and a parameterized surface-limited uptake
mechanism using a reactive uptake parameter $\gamma$ to simulate the formation of aqSOA in
both cloud water and aqueous aerosol water. We also examine the use of aqueous iron
chemistry in cloud water and the effect of changing the cloud water amount within the
model.

This paper is organized as follows. The model and chemistry are described in Sect. 2.
Global budgets and distributions of aqSOA predicted from a base mechanism are
analyzed in Sect. 3. The change in global budgets of aqSOA resulting from four other
mechanisms and one different cloud field are also analyzed in Sect. 3. We compare the
simulations with available measurements in Sec. 4. Finally Sect. 5 summarizes our
conclusions.

2. Model description

We used the IMPACT chemical transport model (Penner et al., 1998; Liu and Penner,
2002; Liu et al., 2005; Ito et al., 2007; Feng et al., 2007; Wang et al., 2009; Xu and
Penner, 2012; Lin et al., 2012) to simulate the formation of SOA. The IMPACT model
includes the microphysics of sulfate aerosol and the interactions between sulfate and non-
sulfate aerosols based on the aerosol module developed by Herzog et al. (2004) (Liu et
al., 2005). Some versions include the formation of nitrate and ammonium aerosols as
well, but here, we extended the sulfate microphysics module initially described by Liu et
al. (2005) to include the formation of SOA (Lin et al., 2012). SOA is assumed to have a log-normal size distribution with a mode radius of 0.0774 µm and a geometric standard deviation of 1.402. It becomes mixed with sulfate through the condensation of gas phase sulfuric acid, through coagulation with pure sulfate aerosols, and through aqueous formation of sulfate. We used the 1997 meteorological fields from the National Aeronautics and Space Administration (NASA) Data Assimilation Office (DAO) GEOS-STRAT model (Coy and Swinbank, 1997; Coy et al., 1997) as input to the chemical transport model. The meteorology was defined on a 4° latitude x 5° longitude horizontal grid with 46 vertical layers. Cloud water content was not available from the meteorological fields, and thus was diagnosed with a parameterization used in the NCAR CCM2 model (Hack, 1998). The large scale stratiform cloud fraction was determined based on the grid box mean relative humidity (RH) as calculated from the DAO meteorological data using the parameterization by Sundqvist et al. (1989). The convective cloud fraction was parameterized by using the convective mass flux (Xu and Krueger, 1991). The model was run for a 1-year time period with a 1-month spin up time. Global emissions of gases, aerosols and aerosol precursors and treatments of dry and wet deposition used here are the same as those used in Lin et al. (2012).

2.1. SOA formation through gas-particle partitioning in the gas phase

In this paper, we adopted the gas-particle partitioning mechanism for SOA formation described by Lin et al. (2012). Lin et al. (2012) use a fully explicit gas-phase photochemical mechanism to predict the formation of semi-volatile organic compounds (SV-VOCs) which then partition to an aerosol phase. These condensed SV-VOCs were assumed to further undergo aerosol phase reactions to form oligomers with an assumed
time constant (nominally 1 day). In addition, we also accounted for SOA formation due to heterogeneous reactions of epoxides on the surface of wet sulfate aerosol by assuming an uptake coefficient of 0.0029, the value measured for glyoxal by Liggio et al. (2005). Lin et al. (2012) explored the effect of OH recycling in the oxidation of isoprene on the formation of SOA using three different gas-phase chemical mechanisms. Here, we used Simulation C in Table 1 in Lin et al. (2012), since this mechanism was shown to best capture the observations of OH and the first generation products of isoprene-OH reactions. The mechanism includes the formation of epoxide proposed by Paulot et al. (2009), while HO\textsubscript{x} regeneration followed Peeters et al. (2009) but with a reduced rate for the 1,5-H and 1,6-H shifts in isoprene radicals by a factor of 10. The basic photochemistry of O\textsubscript{3}, OH, NO\textsubscript{x} and the oxidation of other VOCs utilized the chemical mechanism published by Ito et al. (2007).

2.2. SOA formation in the aqueous phase

In the presence of cloud droplets or aqueous particles, water-soluble gases (e.g., glyoxal, methylglyoxal, and glycolaldehyde) will dissolve in the aqueous phase and be further oxidized by OH and NO\textsubscript{3} radicals to form products with lower volatility (e.g., dicarboxylic acids and oligomers) (Ervens et al., 2011; Lim et al., 2010). These low volatility products are assumed to remain entirely in the particulate phase as SOA, when water is evaporated. In this paper, five aqSOA components are predicted: glyoxylic acid, pyruvic acid, oxalic acid, and two classes of oligomers formed from glyoxal and methylglyoxal.
2.2.1. Multiphase reaction scheme

The change of aqueous and gas phase species due to the photochemical reactions and the exchange between the gas and aqueous phase are expressed by the following equations,

\[
\frac{dC_g}{dt} = P_g - (L_g + k_t Q)C_g + \frac{k_t}{HRT} C_a 
\]

\[
\frac{dC_a}{dt} = P_a + k_t Q C_g - (L_a + \frac{k_t}{HRT}) C_a
\]

\(1a\)

\(1b\)

Where \(C_a\) and \(C_g\) are aqueous and gas phase concentrations (molecules cm\(^{-3}\) air), \(P_a\) and \(P_g\) are aqueous and gas phase chemical production rates (molecules cm\(^{-3}\) air s\(^{-1}\)), \(L_a\) and \(L_g\) are aqueous and gas phase pseudo-first-order chemical loss rates (s\(^{-1}\)), \(H\) is the Henry’s law coefficient (M atm\(^{-1}\)), \(R\) is the universal gas constant (L atm mol\(^{-1}\)K\(^{-1}\)), \(T\) is the temperature (K), \(Q\) is the liquid water content (cm\(^3\) H\(_2\)O cm\(^{-3}\) air), and \(k_t\) (cm\(^3\) air cm\(^{-3}\) H\(_2\)O s\(^{-1}\)) is a first-order rate constant that represents diffusion through the gas phase and across the interface of the drop (see Schwartz (1986) and Lelieveld and Crutzen (1991) for details).

We used the method described by Sillman et al. (2007) to solve Eq. (1a) and (1b). This method is based on the implicit (reverse Euler) equations but incorporates a number of nonstandard treatments as described in Sillman (1991) and Barth et al. (2003). The mass transfer rate across the gas-aqueous interface is assumed to be limited by diffusion and is determined for each gas by its molecular diffusion, mass accommodation coefficient and Henry’s law constant (Lelieveld and Crutzen, 1991). Since not all gas diffusivities are known, we used a single gas diffusivity of 0.1 cm\(^2\) s\(^{-1}\) for all gaseous species. This is unlikely to add a large uncertainty since the values of available gas diffusivities do not
differ by much from the value we used here (Schwartz, 1986; Lelieveld and Crutzen, 1991). Accommodation coefficients for each species are listed in Table S7 in the Supplement and assumed to be 0.05 for species for which no information is available. For situations in which the average concentration of an aqueous species is limited by the rate of diffusion within the aqueous phase, the average aqueous phase concentration is scaled to the surface concentration by the factor Q as shown in Eq. 2.14 in Lelieveld and Crutzen (1991). In addition, an effective cloud droplet radius of 10 µm is assumed for all clouds, while the effective radius for aqueous sulfate particles (the ratio of third to second moment of the wet sulfate aerosol size distributions) is calculated explicitly according to their relative humidity-dependent size distribution. Aerosol water was calculated based on the amount of water uptake for the modeled pure sulfate particles, which is based on the equilibrium Köhler theory (Ghan and Zaveri, 2007) using the RH and hygroscopicity of sulfate to calculate the wet volume mean radius from the dry volume mean radius of each mode. This formulation limits the aqueous SOA formation in aerosols to only include situations in which the deliquescence relative humidity (DRH) for pure sulfate aerosol (set to 0.8) has been reached. We note that the DRH would decrease in mixed sulfate and SOA particles (Brooks et al., 2002; Smith et al., 2012). Thus, the assumption that aqSOA only forms in the pure sulfate aerosols within the model, and the use of 0.80 as the DRH may underestimate the formation of aqSOA to some extent. In addition, water uptake by OA was not treated. There are two reasons for this assumption. First, the reactive uptake parameter ($\gamma=0.0029$) for glyoxal that is used for one of the mechanisms studied here was observed for wet sulfate aerosol only (Liggio et al., 2005). Second, OA is less hygroscopic than sulfate, and the uncertainty associated with neglecting the
contribution of OA water is expected to be less than uncertainties caused by the observed range of reactive uptake coefficients and the photochemical reaction rate of glyoxal. Liggio et al. (2005) reported a reactive uptake coefficient for glyoxal that varied from $8.0 \times 10^{-4}$ to $7.3 \times 10^{-3}$, and Ervens and Volkamer (2010) found that the derived first-order photochemical reaction rate for glyoxal ranges from $0.8 \text{s}^{-1}$ to $7 \text{s}^{-1}$ depending on the chemical composition of different seed aerosols. Finally, we ignored the formation of aqSOA in the water associated with sea salt aerosol. The terrestrial emissions of isoprene and aromatics are much larger than those from marine sources in the IMPACT model, so the amount of glyoxal/methylglyoxal formed from the oxidation of isoprene and aromatics over oceans is smaller than that over continents. AqSOA formed in the aerosol water associated with sea salt is expected to be small as well.

A complete list of aqueous phase reactions and their corresponding rate constants are given in Table S1 and Table S8 in the Supplement. Aqueous reactions for sulfates, nitrates, H$_2$O$_2$, O$_3$, OH and related radicals have been taken from Jacob (1986), Pandis and Seinfeld (1989), Lelieveld and Crutzen (1990) and Liu et al. (1997); Aqueous reactions for water-soluble organic compounds are based on recently published box modeling studies (Lim et al., 2005; Herrmann et al., 2005; Ervens and Volkamer, 2010; Lim et al., 2010); Aqueous reactions for iron are drawn from Deguillaume et al. (2010) and the Chemical Aqueous Phase Radical Mechanism (CAPRAM) 2.4 (Ervens et al., 2003).

While the chemistry of organic species taking place in cloud droplets is relatively well established in experiments and box modeling studies, the chemistry of organic species occurring in wet particles is only now being developed so that very few models have
been extended to include organic reactions (Ervens et al., 2011). Ervens and Volkamer
(2010) and Lim et al. (2010) proposed different schemes for SOA formation in aerosol
water. Ervens and Volkamer (2010) parameterized the SOA formation using simple first
order reaction rate constants to fit photochemical chamber experiments. Basically, gas
phase glyoxal is taken into aerosol water based on its Henry’s law coefficient and is
further hydrated to monohydrate and dihydrate glyoxal using explicit hydration
coefficients for the hydration kinetics. The dissolved glyoxal, monohydrate and dihydrate
glyoxals can react with dissolved OH radicals to form organic acids, or undergo
oligomerization using a parameterized first-order photochemical reaction rate \(k\) (unit: s\(^{-1}\)).
The reactions and kinetic coefficients used in this paper are listed in Table S4 in the
Supplement. The reactions in Table S4 extend the Ervens and Volkamer (2010) model for
glyoxal to include bulk phase reactions of methylglyoxal by adopting kinetic data from
the literature.

Lim et al. (2010) describe a second aqueous SOA formation mechanism that used a set of
detailed radical-radical reactions based on bulk aqueous phase experiments. Gas-phase
glyoxal is partitioned into aerosol water based on its effective Henry’s law constant
(implicitly accounting for its hydration) and then further reacts with dissolved OH
radicals to form radical species by H-atom abstraction, which combine with themselves
to form dimers and trimers through so called “radical-radical” reactions (Lim et al.,
2010). These “radical-radical” reactions compete with reactions of the radicals with
dissolved O\(_2\) to form organic acids. In aerosol-water relevant conditions (i.e., 1-10
mole/Liter (M) glyoxal concentrations and \(10^{-12} – 10^{-11}\) M dissolved OH radicals), over
80% of the products are oligomers. Since no kinetic data for methylglyoxal in-aerosol-
water reactions were available in Lim et al. (2010), we assumed that 80% of products
from the reaction of dissolved methylglyoxal with OH are oligomers and the rest are
oxalic acid, which is consistent with the recent work of Lim et al. (2013).

2.2.2. Surface-limited uptake process

As an alternative to the approaches described above that use a detailed gas-phase and
aqueous phase chemical mechanism coupled by gas-particle transfer, a simpler method
has also been used by Fu et al. (2008) and Lin et al. (2012) to describe the uptake of a gas
and its further reaction inside particles. In this method, the loss of gas phase glyoxal or
methylglyoxal on aqueous particles or cloud droplets is parameterized using the
following equation:

\[
\frac{dC_g}{dt} = -\frac{1}{4} \cdot \gamma \cdot A \cdot <v> \cdot C_g \quad (2)
\]

Where \( A \) is the total surface area of aqueous sulfate aerosols \([m^2/m^3]\), \( C_g \) is the
concentration of gas phase glyoxal or methylglyoxal, \( \gamma \) is the reactive uptake coefficient,
representing the probability that a molecule impacting the particle surface undergoes
reaction. The value of \( \gamma \) used in Fu et al. (2008) and Lin et al. (2012) was assumed to be
2.9 \times 10^{-3} for both glyoxal and methylglyoxal uptake on aqueous sulfate and cloud
droplets. \(<v>\) is the mean molecular speed of glyoxal or methylglyoxal in the gas phase
given by \((8RT/\pi MW)^{-1/2}\) where MW is the molecule weight of glyoxal or methylglyoxal.

2.3. Case set up

As described above, there are still large uncertainties in simulating aqSOA formation in
both cloud water and aerosol water. We thus set up six cases to study the sensitivity of
aqSOA formation to different methods representing gas-particle mass transfer and
subsequent reactions (i.e., multiphase reaction scheme vs. surface-limited uptake
method), to the different chemical schemes in cloud and aerosol water reactions, to the
cloud water content, and to the inclusion of iron chemistry in the cloud. The descriptions
of these six cases are presented here (also summarized in Table 1), and comparisons of
sensitivity test simulations with Case 1 as well as with observations are shown in Sect. 3.

In Case 1, we used the detailed multiphase reaction scheme (Eq. 1a, b) to predict the
production of glyoxylic acid, oxalic acid and pyruvic acid in both cloud water and aerosol
water together with the kinetic uptake of glyoxal and methylglyoxal in both cloud and
aerosol water described in Sect. 2.2.1. In addition, in aerosol water we used the surface-
limited uptake method (Eq. 2) to predict the formation of oligomers from glyoxal and
methylglyoxal. The aqueous phase reactions of organic species used in this case are
shown in Table S3 in the Supplement. Reaction rate constants were adopted from Jacob
(1986), Pandis and Seinfeld (1989), Lim et al. (2005), Herrmann (2003), and Herrmann et
al. (2005). For simplicity, we did not consider the effect of ionic strength of cloud water
on the solubility of organics, though this is expected to increase aqSOA formation
(Myriokefalitakis et al., 2011). We adopted a reactive uptake coefficient $\gamma$ of $3.3 \times 10^{-3}$
from Waxman et al. (2013) to simulate the formation of oligomers due to glyoxal in the
aerosol water. The uptake coefficient of methylglyoxal was scaled to that of glyoxal by
the ratio of their effective Henry’s law constants (Table S7 in the Supplement). This
scaling is based on the reasoning that the glyoxal and methylglyoxal uptake by aerosol
water correlates with their water solubility and that glyoxal and methylglyoxal have
similar chemical reactivity and undergo similar reactions in aerosol water (Lim et al.
It might appear that the chemistry for Case 1 double counts the loss rate of glyoxal since it includes both a loss rate by reaction with OH as well as a loss rate due to a surface reaction characterized by an uptake parameter. However, the uptake parameter used in Case 1 was adopted from Waxman et al. (2013), who used a box model to test several chemical mechanisms, in order to fit the observed glyoxal concentrations in Mexico City. The best fit chemical mechanism that they found includes an uptake parameter (gamma) of $3.3 \times 10^{-3}$ as well as the reaction of OH with glyoxal in dilute water. Therefore we also included the uptake parameter together with the OH chemistry in dilute water.

In Case 2, all aqSOA formation was simulated using the multiphase reaction scheme. The chemical reactions in cloud water were the same as those used in cloud in Case 1. The parameterized reactions proposed by Ervens and Volkamer (2010) (Table S4 in the Supplement) were used for the formation of aqSOA in aerosol water replacing the reactions from Table S3 and the uptake coefficients used in Case 1.

Case 3 also used the multiphase reaction scheme, but used the bulk reactions adopted from Lim et al. (2010) to predict the formation of aqSOA in both cloud and aerosol water. The detailed bulk phase reactions of organic species used in this case are listed in Table S5 in the Supplement. In this case, we include the further oxidation of oligomers by aqueous phase OH (see the reaction R38, R42, R53, and R54 in Table S5). These further reactions were not included in the Case 1, because the uptake parameter represents the probability that a molecule impacting the aerosol surface will result in uptake and formation of a species which does not evaporate, and this already implicitly
takes into account any chemical reactions that include the loss of oligomers inside the aerosol.

For Case 4, we employed the same chemical mechanism as in Case 1, but used the cloud field output (in-cloud liquid water content and grid-box cloud fraction) from AM3, the atmospheric component of the coupled general circulation model (CM3) developed at the NOAA Geophysical Fluid Dynamics Laboratory (GFDL) (Donner et al., 2011) in place of the NCAR CCM2 parameterization. Cloud water is a prognostic variable in the GFDL AM3 while it is diagnosed in CCM2.

In Case 5, we added iron chemistry in cloud water (Table S6 in the supplementary material) to the chemistry used in Case 1. The only source of aqueous Fe in the model is the dissolution of dust aerosol particles incorporated into cloud droplets. We assumed that 3.5% of the mass of dust aerosol is composed of Fe (Taylor and McLennan, 1985), only 5% of which could be dissolved into cloud water (Ito and Xu, 2014). The initial speciation of Fe(II)/Fe(III) was set to 4 (Deguillaume et al., 2010). This scheme cannot describe the spatial and time variations in dissolved Fe that are predicted in models that include a kinetic description of Fe dissolution (Johnson and Meskhidze 2013, Ito and Feng, 2010) but is able to provide a first-order approximation of the dissolved iron content in cloud water (Table S9 in the Supplement).

For Case 6, we used the surface-limited uptake process to simulate all aqSOA formation in both cloud and aerosol water, following the method of Fu et al. (2008) and Lin et al. (2012).
3. Results & Discussion

Table 2 shows the global budget of total aqSOA and each of its components (i.e., glyoxylic acid, pyruvic acid, oxalic acid and oligomers) (if available) for these six cases. We note that the predicted oligomers in the model can consist of different numbers of monomers (e.g., dimers, trimers and tetramers), so that the total oligomers shown in different cases in Table 2 do not necessarily consist of identical species. We will focus on the detailed budget and global distributions of aqSOA for Case 1 in Sect. 3.1 and Sect. 3.2, and the difference between this case and other cases in Sect. 3.3 to Sect. 3.6.

3.1. Global budget

For Case 1, the net global aqSOA production rate totals 20.1 Tg/yr, over 95% of which is removed by wet deposition while the rest is removed by dry deposition. This rate is comparable to the SOA production rate of 28.0 Tg/yr formed from gas-particle partitioning and the rate of 26.0 Tg/yr formed from epoxide predicted in the model. The global annual mean aqSOA burden equals to 0.18 Tg, corresponding to a global mean life time of about 3.0 days due to deposition. Five aqSOA species are predicted: glyoxylic acid, pyruvic acid, oxalic acid, and two classes of oligomers formed from glyoxal and methylglyoxal. Among these five aqueous SOA components, oxalic acid accounts for about 51.7% of the total aqueous SOA source, glyoxal oligomers account for about 34.3%, glyoxylic acid for about 11.9%, methylglyoxal oligomers for 1.1% and pyruvic acid for 1.0%. While all oligomers are assumed to be formed in aerosol water, organic acids can be formed in both cloud and aerosol water. However, the contribution of aerosol water to the formation of organic acids formations is very small. The net
production rate of glyoxylic acid in the aerosol water accounts for only 0.013 Tg/yr of the
total 2.4 Tg/yr net production rate; similarly, only $2.53 \times 10^{-4}$ Tg/yr of the total 0.17 Tg/yr
pyruvic acid is formed in aerosol water; for oxalic acid, the net production rate in the
aerosol water (after consumption by reaction with OH) is -0.43 Tg/yr, compared to the
total net production rate of 10.4 Tg/yr in cloud water.

The global average reaction rates for these organic acids within cloud for case 1 are listed
in Table 3. The glyoxylic acid production rate is 15.8 Tg/yr, which is derived from the
oxidation of glyoxal, glycolaldehyde, methylglyoxal and acetic acid. Glyoxal oxidation
accounts for 77.2%, while the oxidation of glycolaldehyde, methylglyoxal and acetic acid
account for 13.9%, 0.7% and 8.2%, respectively. 85.4% of glyoxylic acid is destroyed by
reaction with OH and NO$_3$, and the rest is deposited to the surface in wet deposition. For
oxalic acid, the global production rate in cloud is equal to 15.5 Tg/yr, which is similar to
the estimate of 14.5 Tg/yr in Liu et al. (2012) but smaller than the estimate of 21.2 Tg/yr
in Simulation S1.1 of Myriokefalitakis et al. (2011), which excluded the effect of the
ionic strength of cloud water on glyoxal, glycolaldehyde and methylglyoxal, as we do
here. The reaction of glyoxylic acid with OH is a large contributor (94%) to the total
oxalic acid production while the reaction with NO$_3$ contributes the rest. Oxalic acid is
removed from the atmosphere through its reaction with OH and NO$_3$ radicals in the
aqueous phase (29.6%) and by wet and dry deposition (71.4%). The chemical destruction
rate of oxalic acid is 4.6 Tg/yr. This is smaller than the estimate in Simulation S1.1 of
Myriokefalitakis et al. (2011) but larger than that estimated by Liu et al. (2012).

Subtracting the chemical destruction rate from the production rate, the global net
production rate of oxalic acid is 10.9 Tg/yr, which is slightly smaller than the value of
13.2 Tg/yr in Simulation S1.1 of Myriokefalitakis et al. (2011) and the estimate of 13.5 Tg/yr in Liu et al. (2012). The only sources of pyruvic acid are the reactions of methylglyoxal with OH and NO₃. Over half (72.9%) of the pyruvic acid is removed by reactions with OH and NO₃ radicals. For glyoxal and methylglyoxal oligomers, no chemical destruction is included in the model, so these are only removed by wet and dry deposition.

The above analysis shows the importance of glyoxal, glycolaldehyde, methylglyoxal and acetic acid as precursors leading to aqueous SOA formation. The global budgets of these four species for Case 1 are summarized in Table 4. While all of glyoxal, glycolaldehyde and methylglyoxal are generated by the oxidation of VOCs in the gas and aqueous phase, around half of the acetic acid is directly emitted into the atmosphere through biomass burning. The global glyoxal production in the gas phase is equal to 69.6 Tg/yr, while reactions in cloud water contribute 3.4 Tg/yr from the oxidation of dissolved glycolaldehyde. About 68.5% of the total glyoxal is consumed in the gas phase, while 24.4% is oxidized in cloud and taken up by aqueous aerosol. The rest is deposited to the surface. The total source of methylglyoxal is 167.3 Tg/yr. Of this, only 0.88 Tg/yr and is absorbed and oxidized in cloud and in aqueous aerosol. Most of the methylglyoxal is destroyed in the gas phase or deposited to the surface. The net chemical production of glycolaldehyde in the gas phase is about 17.8 Tg/yr, of which 29.8% is dissolved into cloud and reacts with OH and NO₃. For acetic acid, the uptake rate by cloud is 0.82 Tg/yr, which can be compared to its total source strength of 60.9 Tg/yr. The uptake rate of acetic acid is smaller than the estimate of 6.96 Tg/yr by Liu et al. (2012). This is because of the smaller total atmospheric source strength of acetic acid (60.9 Tg/yr vs. 78
Tg/yr), the larger portion of gas-phase consumption of acetic acid (43% vs. 32%), and the smaller Henry’s law constant (3500 mol L⁻¹ atm⁻¹ vs. 8800 mol L⁻¹ atm⁻¹) in our model compared to those in Liu et al. (2012). Aqueous aerosol contributes a negligible amount to the sinks of both glycolaldehyde and acetic acid.

3.2. Global distribution and seasonal variability

Figure 1 presents the global annual mean surface mass concentrations (at approximately 970 hPa) of total aqSOA, total organic acids (i.e., glyoxylic acid, pyruvic acid and oxalic acid) which are mostly formed in cloud, and oligomers formed in aqueous aerosol. The zonal mean vertical distributions are also shown. The total aqSOA concentrations show large values over tropical Africa, the Amazon basin, Eastern Asia, Eastern United States and Europe. The SOA distributions are determined by their precursor (mainly glyoxal) distributions, oxidant (which is primarily dissolved OH radicals) distributions, and the availability of cloud water or aerosol water. The maximum SOA concentrations over tropical Africa and the Amazon basin reflect the large biogenic VOC emissions and the resulting glyoxal concentration to a great extent. The different patterns shown for organic acid concentrations (Fig. 1C) and for methylglyoxal and glyoxal derived oligomer concentrations (Fig. 1E) are due to the different patterns of cloud and aerosol water content (Fig. 2). Aerosol water content is due to sulfate, which peaks over the industrial regions in the Northern Hemisphere because the largest source of sulfate is from anthropogenic emissions. In contrast, most of the cloud water is located over the tropics and the Southern Hemisphere. This contrast is also reflected in the vertical zonal mean distributions of organic acids and oligomers (Fig. 1D and Fig. 1F). There is a hot spot in organic acids over the tropics, most of which are formed in cloud, while the peak is
located over the Northern Hemisphere for the oligomers, all of which are formed in aqueous aerosols.

The column burdens of organic acids and oligomers in winter (December, January and February) and in summer (June, July, and August) are presented in Fig. 3. During the winter, the column burden of organic acids peaks over tropical land areas because of the huge biogenic emissions in these regions. During the summer, a secondary column burden maximum arises over the Northern Hemisphere land areas, which can be attributed to the enhanced photochemistry and biogenic emissions over these regions in the summer. The combination of enhanced photochemistry and larger biogenic emissions can increase the production of aqSOA precursors (e.g., glyoxal) and aqueous OH radicals, the latter due to increased H₂O₂ photolysis in cloud. For the same reason, the column burden of oligomers in the summer shows larger values that are spread over a wider area than those in the winter over the Northern Hemisphere (see the right panels in Fig. 3).

3.3. Surface-limited uptake method vs. multiphase reaction scheme

As shown in Table 2, the production rates of oligomers in Case 1 are higher than those in Case 2, which suggests that the value of the reactive uptake coefficient γ adopted from Waxman et al. (2013) is higher than that implied by the simulation with the Ervens and Volkamer (2010) aqueous phase aerosol chemical mechanism. Indeed, we can derive a global averaged γ from the Eq. (2) for the uptake of glyoxal or methylglyoxal into aqueous aerosol from Case 2. We integrated the left side and the right side of Eq. (2)
globally and annually, and thus obtained a global averaged annual mean reactive uptake probability from the following equation:

\[
\bar{y} = \frac{\left(\sum \frac{dC_g}{dt}\right)}{\left(\sum -\frac{1}{4} \cdot A \cdot < v > \cdot C_g \right)}
\]  

(3)

The derived global averaged \( \bar{y} \) for the uptake of glyoxal into the aqueous aerosol is \( 1.41 \times 10^{-5} \), while the \( \bar{y} \) for the uptake of methylglyoxal into the aqueous aerosol is \( 1.47 \times 10^{-5} \). This value for glyoxal uptake is much smaller than the value of \( 3.30 \times 10^{-3} \) suggested by Waxman et al. (2013) or the value of \( 2.90 \times 10^{-3} \) derived by Liggio et al. (2005). The value derived here for methylglyoxal uptake is comparable to the value of \( 2.92 \times 10^{-5} \) that we used for methylglyoxal uptake in Case 1. The reason for the discrepancy between measurement studies (i.e. Waxman et al. (2013) and Liggio et al. (2005)) and the values derived here might be that the actual value of the uptake parameter in remote and rural regions is smaller than that derived in urban regions (i.e. Mexico city in the case of Waxman et al. (2013)) or/and that the bulk phase reactions adopted from Ervens and Volkamer (2010) may not include the full set of reactions of glyoxal that are accounted for in the uptake parameter method. In particular, the smaller uptake parameter in rural regions than in urban regions might be due to weaker particle acidity and lower dissolved organic compound concentration in rural regions. The glyoxal uptake rate is observed to be higher in more acidic particles (Liggio et al., 2005). The more concentrated organic compounds in aerosol water might lead to a higher oligomer formation rate (Tan et al., 2009). Also, the bulk phase reactions do not include any surface reactions of glyoxal that the uptake parameter method would include. We note that the derived uptake parameter of order \( 1 \times 10^{-5} \) is a globally averaged uptake
parameter. Of course, the spatial aqSOA burden which would be predicted using this global average uptake parameter would be different than that predicted using the explicit aqueous formation mechanism. While deriving gamma in lab studies is useful, this method for determining gamma assumes that all of the gas incorporated into the aqueous phase stays in this phase. The explicit chemistry scheme is more demanding of computer time than the use of a single gamma, but it is able to capture the dynamic evolution of aqSOA formation.

3.4. The effect of cloud water content

The global near-surface distribution of grid-box averaged cloud water content for Case 4 is shown in Fig. 4(A) while Fig 4(B) shows the zonal annual mean distribution. The cloud water content in Case 1 (Fig. 2) is higher than that in Case 4 almost everywhere below about 900 hPa, leading to larger global organic acids sources and burdens (Table 2). The averaged cloud water content and the net SOA production rate in cloud below about 900 hPa in Case 1 is 2.7 times and 4 times higher than those in Case 4, respectively. Figure 4(C) shows the ratio of cloud water content in Case 1 to that in Case 4 while Figure 4(D) shows the ratio of the aqSOA production rate in cloud. Both the pattern and value of these ratios are generally similar below about 900 hPa, but above 900hPa neither the pattern nor the value is similar. He et al. (2013) studied the relationship between the aqueous SOA formation and cloud water content and proposed a parameterization in which its formation rate is linear in the cloud water content and correlates nonlinearly (concavely) with the total glyoxal and methylglyoxal precursor carbon chemical loss. At lower altitudes where the cloud water concentration in Case 1 is more than a factor of 4 larger than that in Case 4, the linear effect of the change in cloud
water content would be expected to dominate the change in the aqSOA production rate. At higher altitudes (approximately between 900 hPa and 200 hPa) where the ratio of cloud water content between in Case 1 and in Case 4 is less than 4, the effect of the change in precursor concentrations (mainly due to the change of wet deposition rates as we show below) may dominate the change in the aqSOA production rate.

The lifetime of oxalic acid with respect to deposition in Case 4 is longer than that in Case 1 (4.9 days in Case 4 vs. 2.1 days in Case 1), because in Case 4 more oxalic acid is produced at high altitudes or high latitudes and less (by a factor of over 10) is formed at low altitudes in tropical regions (Fig. 4(D)). Oxalic acid can be precipitated out more easily in the tropics than in other regions because precipitation amounts are larger, especially that due to convective precipitation. For similar reasons, we predict a longer lifetime of sulfate aerosol in Case 4 than that in Case 1. In addition, a slightly larger aqueous phase production rate of sulfate aerosol (93.3 Tg/yr) is also found in Case 4 compared to that in Case 1 (91.6 Tg/yr). This results from the more abundant cloud water in Case 4 in the Northern hemisphere, where most of sulfate is formed. The longer lifetime and larger aqueous production rate of sulfate causes a larger sulfate burden and thus a larger oligomer formation rate in sulfate aerosol water.

3.5. The effect of iron chemistry in cloud

Inclusion of iron chemistry in cloud decreases the global average net production of oxalate by 57.6%, although it increases both the chemical production and destruction of carboxylic acids in cloud (Table 2). The increase of the chemical production rate is due to the increase of the aqueous OH radical source. Figure 5 depicts the global distribution of
annual mean aqueous OH radical concentrations near 971 hPa for Case 1 and Case 5. The largest increase of aqueous OH radicals due to the inclusion of iron chemistry occurs over regions where there are abundant dust aerosols. Arakaki et al. (2013) estimated the steady-state concentrations of OH in cloud sampled over North American remote continental regions. The estimated average OH concentration is $7.2 \times 10^{15}$ mol/L. The predicted average OH concentration in this region is $9.0 \times 10^{14}$ mol/L in Case 5 and $8.8 \times 10^{14}$ mol/L in Case 1. Thus the modeled OH concentrations in cloud for both cases are higher than those estimated by Arakaki et al. (2013). Arakaki et al. (2013) also estimated that the averaged total OH production rate in cloud is $1.9 \times 10^{-9}$ mol/L/s and OH loss rate constant is $2.6 \times 10^5$ s$^{-1}$. In the model, the averaged total OH production rate in this region is $1.06 \times 10^{-8}$ mol/L/s in Case 5 and $1.03 \times 10^{-8}$ mol/L/s in Case 1. The OH loss rate is $1.18 \times 10^5$ s$^{-1}$ in Case 5 and $1.17 \times 10^5$ s$^{-1}$ in Case 1. This comparison suggests that the overestimation of OH source is the major reason for the higher OH concentration in the model. Arakaki et al. (2013) attributed the higher OH concentration in cloud water predicted in most box models to their underestimation of OH loss rates. This is not the case in our model. Our OH loss rate of about $1.2 \times 10^5$ s$^{-1}$ compares to values obtained in those box models that range from $2.0 \times 10^4$ s$^{-1}$ to $7.7 \times 10^4$ s$^{-1}$ (Table S2 in Arakaki et al. (2013)).

The global aqueous OH radical source in the troposphere (below approximately 200 hPa) in Case 5 is 2.6 times larger than that in Case 1 because of the formation of OH radical through the reaction of Fe(II) with H$_2$O$_2$ and the photolysis of Fe(III) oxalate complexes. The largest increase occurs over the Sahara desert, Northwestern China and Mongolia where there are large amounts of dust aerosol (Fig. 5). The enhancement of aqueous OH
radicals leads to an increase of 10.8 Tg/yr in the oxalate chemical production rate and partly accounts for an increase of 16.8 Tg/yr in the oxalate chemical destruction rate. Another reason for the increase in the oxalate chemical destruction rate is the fast photolysis of the iron-oxalate complex \([\text{Fe}(\text{C}_2\text{O}_4)_2]^-\), which transforms \(\text{C}_2\text{O}_4^{2-}\) to \(\text{CO}_2\). The larger increase in the chemical destruction rate compared to the chemical production rate results in a decrease in the net chemical production rate (6.0 Tg/yr) of oxalate, and thus a decrease in the burden as well. Iron-oxalate complex photolysis accounts for 21.7 Tg/yr of oxalate destruction, while the destruction rate through OH and NO\(_3\) reactions is 1.20 Tg/yr. This significant sink of oxalate by Fe-complex photolysis is consistent with the finding by Sorooshian et al. (2013) who observed that oxalate concentrations are negatively correlated with observed Fe concentrations. Sorooshian et al. (2013) used a box model to simulate oxalate formation with and without Fe chemistry. They found that the oxalate concentration would decrease when increasing the dissolved iron concentration, but gradually levels off at 15% of the concentration predicted without consideration of Fe chemistry. Our model results show that the oxalate concentration with iron chemistry is about 44% of that predicted without iron. The inclusion of iron chemistry decreases the formation rate of glyoxal oligomers from 6.9 Tg/yr in Case 1 to 6.6 Tg/yr in Case 5, which is due to the increased absorption rate of glyoxal in cloud and thus less glyoxal uptake in aqueous aerosol. The absorption rate of glyoxal in cloud is increased because the increased OH radicals in cloud cause more glyoxal to be consumed in cloud so that more gas phase glyoxal can be taken up by cloud.
3.6. SOA formation in clouds vs. SOA formation in aerosol water

On a global average basis, the fraction of SOA formed in cloud varies from case to case, ranging from 44.4% in Case 4 to 103.2% in Case 3 (Table 5). SOA formed in cloud water accounts for about 70% and 80% of total aqSOA in Case 1 and Case 6, respectively; the net SOA production rate in aerosol water is less than zero in Case 2 and Case 3, which means that aqSOA is consumed in aerosol water. This is because the oxalic acid formed in cloud subsequently dissolves into the aerosol water after cloud water evaporates and then further reacts with the high concentration of dissolved OH radicals inside the aerosol water, as shown in Section 3.1. The rate of destruction of oxalic acid by reaction with OH is larger than that of the rate of production from the reaction of glyoxylic acid with OH. Nevertheless, oligomers can still form in the aerosol water, with a production rate of $4.5 \times 10^{-2}$ Tg/yr and $3.6 \times 10^{-2}$ Tg/yr for Case 2 and Case 3, respectively. The amount of oligomers formed in the aerosol water is similar for Case 2 and Case 3 consistent with the finding by Ervens et al. (2011), who showed that their parameterized reaction system produces a similar amount of SOA as that of Lim et al. (2010) in box model simulations. The relative importance of SOA formed in cloud water decreases to 44.4% in Case 4 when using the GFDL AM3 cloud field because of the decrease in the cloud water content in the tropics together with the increased cloud water content in the Northern Hemisphere. In Case 5, which includes iron chemistry in cloud, the SOA production rate in cloud explains 52.9% of total aqSOA production rate (see the Sect. 3.5. for details).

3.7 Discussion of uncertainties

While we have discussed a number of uncertainties concerning the production of aqSOA,
there are additional uncertainties that need to be explored. Liu et al. (2012) used a global model to study the sensitivity of their predicted aqSOA in cloud to cloud lifetime, cloud droplet size and cloud fraction. They found that the predicted aqSOA was sensitive to the cloud lifetime and cloud fraction while it depended only weakly on the cloud droplet size.

Waxman et al. (2013) studied the sensitivity of aqSOA formation in aerosol to aerosol size distributions. Little size dependence was reported for the case using the multiphase reaction scheme, but the uptake method showed a strong sensitivity to size distribution because of changes in the surface to volume ratio.

Furukawa and Takahashi (2011) used X-ray absorption fine structure spectroscopy (XAFS) to show that most of the oxalate within aerosol particles is present as metal oxalate complexes, especially as Ca and Zn oxalate complexes. Xing et al. (2013) also suggest the formation of a stable Zn oxalate complex in the aerosol phase in urban aerosols over China. These complexes are weakly water soluble and very stable, which prevents oxalate from being oxidized by OH in aerosol water. We did not include these effects in our model due to the lack of detailed information, but it is clear that their inclusion could increase the amount of oxalate in aerosol water.

Many studies have suggested a 2~3 order of magnitude enhancement in the effective Henry’s law constant for glyoxal for aerosol conditions compared to cloud conditions (Kroll et al., 2005; Volkamer et al., 2009; Kampf et al., 2013). To test the potential effect of this enhancement, we added a sensitivity test based on Case 3 but increased the effective Henry’s law constants of glyoxal and methylglyoxal for aerosol water by 3 orders of magnitude. In this sensitivity case, the chemical production rate of glyoxal oligomer is $4.65 \times 10^{-1}$ Tg/yr, and the chemical destruction rate is $1.95 \times 10^{-1}$ Tg/yr,
leading to a net production rate that is roughly a factor of 10 larger than that in Case 3, \(1.95 \times 10^{-1}\) Tg/yr. The net production rate for methylglyoxal oligomer is increased by more than a factor of 10, to \(8.68 \times 10^{-2}\) Tg/yr. These increases in net production rates lead to an increase in oligomer burden by a factor of 4.4. However, these enhanced net production rates and global burdens are still smaller than those predicted in the Case 1 which adds the Waxman et al. (2013) uptake method for the production of aqSOA in aqueous aerosol.

The formation and loss of oxalate by aqueous phase OH depends strongly on the pH value (see reactions R2-6 in Table S3), because the rate constants of oxidations of glyoxylic acid and oxalic acid are smaller than those of their anions (i.e., glyoxylate and oxalate) and the relative abundance of these acids and their anions depends on pH values. We present the predicted annual mean pH in cloud near 971 hPa in Figure S1 in the Supplement and compare our predicted pH values with observations from the literature (Table S10 in the Supplement). The results agree with predictions at some locations, but show that the model is too acidic in other locations. We would expect the oxalate concentrations to vary little if pH values were increased, because both the formation rate and the destruction rate of oxalate would increase. This weak dependence of aqSOA formed in cloud has been observed in a parcel model that included a multiphase reaction mechanism similar to that which we use in this work (Ervens et al., 2008).

The comparison of Case 2 and Case 3 with Case 1 shows that the use of only the multiphase reaction scheme for aerosol water decreases the global total aqSOA burden by around 50%. In the case which increases the effective Henry’s law constant by 3 orders magnitude, the aqSOA burden remains lower than that predicted in Case 1 by 42%. The
replacement of diagnostic cloud field with GFDL AM3 cloud field causes the burden to increase by 16%. In the case including iron chemistry in cloud, the burden is predicted to be smaller than that in Case 1 by 11%. The use of the uptake parameter method for both cloud water and aerosol water in Case 6 predicts a 185% larger burden than Case 1.

4. Comparison with measurements

In this section, we compare model results to measured oxalate, AMS measurements of SOA, and SOA O/C ratios. Oxalate is a major component of aqSOA formed in cloud, and thus the comparison of oxalate with observations gives us a direct evaluation of the modeled aqSOA, while comparison of SOA measured by AMS data and O/C ratios can indirectly constrain the modeled aqSOA.

4.1. Oxalate

Figure 6 compares the modeled oxalic acid near 971 hPa with the measured oxalate compiled in Table S3 of Myriokefalitakis et al. (2011). We only show this comparison for Case 1, Case 4, and Case 5, since the oxalic acid concentrations in the other cases are similar to those in Case 1 (Table 2). Although oxalate measurements are sparse around the world, the observations listed here cover values over most continents: United States, Europe, China, Amazon basin and Africa. These measurements have time sampling durations which span from several days up to 2 years. For this comparison, the monthly simulation data were sampled for the specific month and at the specific location corresponding to the measurements. It should be noted that due to the coarse resolution used in the model (4 degrees by 5 degrees), we do not expect the model to be able to predict the high concentrations seen in urban regions and hence we leave these out. Over
rural areas, the model performance varies in different regions and different seasons. In Europe, the results for Case 1 underestimate almost all observations, although the results for summer are relatively better (green triangles in Fig. 6) than those in winter (black dots in Fig. 6). The difference in the model performance between in summer and in winter may be attributed to different oxalate sources in these two seasons. According to Legrand et al. (2007), the major oxalate sources at the CARBOSOL surface stations in winter are fast secondary production in wood burning plumes and secondary production through the rapid oxidation of toluene and ethane emitted from vehicles. In summer these anthropogenic emissions decrease and biogenic emissions (e.g., isoprene) increase and make an important contribution to the oxalate source via multiphase photochemical reactions. These four CARBOSOL sites are included in the comparison here and we suspect the other European sites that we included have similar oxalate sources to those at the CARBOSOL sites. Unlike what is expected for the measurements, the model can not represent the rapid secondary formation of oxalic acid from local sources. Other factors which might also contribute to the model under-estimation include a low cloud water content or high deposition rates in the model. When using the GFDL AM3 cloud field, the model prediction improves at some sites.

As shown by the red squares in Fig. 6, the model also significantly underestimates the measured oxalate concentrations at 2 of the 3 sites in the Amazon basin. One possible reason for this underestimation is that the model does not include the oxalate source from biomass burning either through direct emission or through secondary formation from carbohydrate species in smoke aerosols (e.g., levoglucosan) during aerosol aging (Gao et al., 2003). At these two sites, biomass burning is the major source of oxalate formation.
(Falkovich et al., 2005; Kundu et al., 2010; Granham et al., 2002). At the site where the aerosol samples were believed to be out of the influence of biomass burning (Talbot et al. 1988), the modeled oxalate concentration in Case 1 is higher than the observation, but in Case 4 (using the GFDL AM3 cloud field) the model still underestimates the observation.

There are only two sites in China and three sites in US with oxalate measurements for comparison to the model. The model underestimates the observations in China, especially at the Mangshan site, 40 km north of Beijing, polluted by air mass transported from Beijing (He and Kawamura, 2010). This site has a high oxalate concentration of 760 ng/m$^3$. The model can’t represent the local emissions and thus only captures less than 10% of the observation. Using the GFDL AM3 cloud field doesn’t improve the model performance. Overestimation of oxalate deposition in the model may also contribute to this under-prediction. The three blue circles in Fig. 6 show the comparisons for the sites in United States, and indicate that the model does reasonably well, except for the site in Sydney, Florida, at which the model still under-predicts the oxalate source or/and overestimates its sink. For the two sites in Africa (red stars in Fig. 6), the model fails to predict the observations. The modeled oxalate concentration in Case 1 is 3 times higher than the observation at the Central Africa site and is only about 1/5 of measured concentration at the South Africa site. This might suggest that the cloud water content in this simulation is inconsistent with the real cloud water content or that the model does not represent the sources and/or sinks of oxalate over this region very well. In Case 4, the model under-predicts the measurements at both of these two sites.

When including the iron chemistry in cloud in Case 5, almost all of the modeled oxalate concentrations are lower than the measurements by over a factor of 2. As noted above,
this underestimation may be caused by an overestimation of the photolytic rate of the iron-oxalate complex and/or an underestimation of oxalate production rate.

The global oxalate burden in case S1 reported by Myriokefalitakis et al. (2011) predicted is 5 times larger than the burden reported here for Case 1. This is partly because Myriokefalitakis et al. (2011) increased the solubility of glyoxal and methylglyoxal in cloud water by 2 orders of magnitude, which is at the high end of measured values, causing a larger oxalate source. Myriokefalitakis et al. (2011) also included a set of additional reactions proposed by Carlton et al. (2007) which were based on experiments that were performed under higher glyoxal concentration conditions than those present in real cloud water. These additional reactions result in more oxalate formation. Myriokefalitakis et al. (2011) lumped these additional reactions to a single reaction (see R21 in Table 1 in Myriokefalitakis et al. (2011)). We note that the reaction rate constant of $3.1 \times 10^9$ L mol$^{-1}$ s$^{-1}$ for that lumped reaction was a typo (M. Kanakidou, personal communication, 2014). The reaction rate constant should be $3.1 \times 10^{10}$ L mol$^{-1}$ s$^{-1}$. In addition, a longer oxalate lifetime with respect to deposition was predicted by these authors. The larger oxalate burden in Myriokefalitakis et al. (2011) leads to a closer agreement with measurements over rural sites. However, the simulations in Myriokefalitakis et al. (2011) did not include iron chemistry in cloud which, as we have shown, can decrease oxalate concentrations significantly (compare Case 1 and Case 5).

Figure 6 also shows a comparison of oxalate with observations at marine sites. Most of the modeled oxalate is lower than the measurements. Myriokefalitakis et al. (2011) considered an extra glyoxal source of 20 Tg/yr over the oceans to explain the gap between the glyoxal measured from satellite and that predicted in their model. The
underestimation of oxalate at marine sites in our model would also be improved by
adding an extra marine source of glyoxal, although the origin of glyoxal in the marine
boundary layer is still matter of debate (Rinaldi et al., 2011). As shown above and in
Sec. 3.4, cloud water has an important effect on oxalate formation. Therefore it is also
valuable to compare the cloud water content in the model with that in measurements.
However, there are only a few sites where measured oxalate concentrations were reported
together with cloud water content (e.g., Sorooshian et al., 2006; Wonaschuetz et al.,
2012). Thus, this comparison is unlikely to be able to discern which source might explain
the underestimate.

4.2. AMS measurements

Although there are no measurements available that can separate aqSOA from the SOA
formed in the gas phase, we can compare the modeled total SOA with the observed total
SOA. The predicted ratio of the global average aqSOA concentration to total SOA
concentration ranges from 9.5% in Case 2 to 33.5% in Case 6. Zhang et al. (2007) present
observational SOA data (measured by aerosol mass spectrometry, AMS) from a series of
surface measurements at multiple sites in the Northern Hemisphere, which were made in
different seasons and different years between 2000 and 2006 and were reported for the
average of varying durations spanning from 8 to 36 days. Here, we compare the model
data at the corresponding grid in the corresponding month. Fig. 7 shows the comparisons
of SOA between observations and predictions in Case 1, Case 2, Case 4, Case 5 and Case
6. The normalized mean bias (NMB) and correlation coefficient (R) for these
comparisons are listed in Table 6. We do not show the comparisons for Case 3 in Fig. 7
because the SOA concentrations for this case are close to those for Case 2 (Table 2).
Again, we do not expect the model with its low horizontal resolution to capture the POA emissions at urban sites, nor would we capture high local VOC and NOx emissions, which have a very complex effect on SOA formation due to non-linear chemistry (Stroud et al., 2011). The NMB in Case 1 is -32.4% for rural sites (see the Table 6). The multiphase reactions scheme used for aqSOA in aerosol water in Case 2 leads to a larger underestimation (a NMB of around -42.5%); Using the GFDL AM3 cloud fields allows the model predictions to more closely match the measurements; The model in Case 6, which adopts the reactive uptake method for both aerosols and clouds, overestimates the observations by around 20%.

In addition to the AMS measurements made in the Northern Hemisphere extra-tropics reported in Zhang et al. (2007), we also show the comparison with measurements from three different campaigns in tropical forested areas (Table 7). As SOA dominated the total submicron OA in Amazon basin (Chen et al., 2009) and in Malaysian Borneo (Robinson et al., 2011), the comparison of OA in these two sites reflects the model performance of SOA to a large extent. Comparing with the measurements at West Africa reported by Capes et al. (2009), the model predicts both a higher SOA concentration and a higher NOx concentration. The NOx concentration has been shown to have a large effect on SOA formation (e.g., Ng et al., 2007), thus, we expect that improving the NOx predictions might lead to a better SOA simulation. There are two measurements available in Amazon region. Gilardoni et al. (2011) reported average PM2.5 organic aerosol concentrations during the wet season (February-June) of 1.7 µg /m³, larger than the wet season concentrations measured at the same site by Chen et al. (2009) by about 60-80%. In comparison with the measurements reported by Gilardoni et al. (2011), our simulated
concentrations in Case 1 to Case 5 are around 60% too high. At the Malaysian Borneo site, the model simulations of SOA are higher than the observation by less than 40% for Case 1 to Case 5. The simulation in Case 6 overestimates the measurements by a factor of 3.8 at the West Africa site and by a factor of over 2 at the Amazon and the Malaysian Borneo site.

It should be noted that that good agreement between observations and model predictions does not imply a robust understanding of underlying processes. However, the relative trends in how these comparisons change between different cases (Table 8) helps one to understand the benefits of different underlying processes and to determine the best current approach. For the Northern Hemisphere, the uptake method (Case 1 and Case 6) provides an approach that agrees better with the observations while the use of the GFDL cloud field also improves the model performance. In the tropics the use of the multiphase reaction scheme helps to close the gap between the simulation and the observations, and including iron chemistry also decreases the model bias. On the other hand, including iron chemistry degrades the model’s ability to predict the oxalate concentrations. Thus there is no single approach that is able to capture all the observations well. Further mechanism development and/or aerosol transport model and general circulation model development of cloud fields are needed to improve the agreement with observations. For example, the inclusion of stable Ca and Zn oxalate complex formation would increase both oxalate and SOA concentrations and thus help to close the gap between the simulation and the observations, as we noted above in Sect. 3.7.

4.3. O/C ratios
It is useful to compare the O/C ratio of our modeled OA to observations, since the O/C ratio in aqueous formation mechanisms is expected to be larger than that in gas phase formation mechanisms. Aiken et al. (2008) derived a significant correlation between the O/C ratios of OA and their f44 signal (the ratio of m/z 44 to the total signal in the mass spectrum) and Ng et al. (2010) used this correlation to estimate the O/C ratio of oxygenated OA (OOA) obtained from a factor analysis of the Northern Hemisphere AMS dataset. Here, we compare the O/C ratios estimated in the model with those reported by Ng et al. (2010) (Fig. 8).

In the model, we have four different SOA components: SOA from the gas-particle partitioning of semi-volatile organic compounds (SVOCs), SOA from aerosol phase reactions of these condensed SVOCs, SOA from the uptake of epoxide on sulfate aerosol, and SOA formed in the aqueous phase as described above (aqSOA). We used the following methods to estimate the predicted O/C ratio of these SOA components:

1. For the SOA from the gas-particle partitioning of SVOCs, there are 26 explicit SVOCs that contribute to the SOA formed from gas-particle partitioning. We calculated the O/C ratio for each species based on their chemical formulas (See Table S1 in Lin et al. 2012).

2. For the SOA from aerosol phase reactions of condensed SVOCs, however, it is not as straightforward to calculate O/C ratio. First, the aerosol phase reactions of condensed SVOCs were simply treated in the model as first-order reactions to form oligomers with an assumed time constant (nominally 1 day), without any information on the products from these aerosol phase reactions. However, Chen et al. (2011) and Liu et al. (2012)
suggested specific aerosol phase reactions for organic hydroperoxides and organic nitrate, respectively, both of which are major components of SVOCs in the model (Lin et al., 2012). Chen et al. (2011) proposed a set of new aerosol phase reactions for organic hydroperoxides: decomposition, followed by radical-radical oligomerization to explain the smaller measured O/C ratios than those predicted in their model. This aerosol phase reaction was shown to remove 1–2 O atoms from organic hydroperoxides. Organic nitrate is thought to undergo hydrolysis in the particle phase (Liu et al., 2012), which removes two O atoms from organic nitrate. Despite the simple treatment for oligomer formation in the model, we assume that the oligomers from organic hydroperoxides and organic nitrate have 1.5 O atoms and two O atoms less than condensed organic hydroperoxides and organic nitrate have, respectively, based on Chen et al. (2011) and Liu et al. (2012). For the oligomers formed from other condensed SVOCs, we assume they have the same O/C ratio as their corresponding SVOCs. An additional complexity is that in the model the 26 SOA species formed from the oligomerization of condensed SVOCs are lumped together as one species when they are transported in the atmosphere. Therefore, we used their global-averaged SOA formation rates (listed in Table S1 in Lin et al. 2012) rather than their mass concentrations to weight their relative contributions to the O/C ratio. Based on these assumptions, we estimate the average O/C ratio for the SOA from aerosol phase reactions of condensed SVOCs to be around 0.687, with values for some of the SVOCs with smaller burdens ranging from 3.0 to 0.3. (3). For SOA from the uptake of epoxide, the O/C ratio is estimated to be 0.6 since the O/C ratio of epoxide from isoprene oxidation is 0.6 in the model, and its oligomerization would not be expected to change the O/C ratio (Surratt et al., 2010).
The O/C ratio for oxalic acid is 2, for glyoxylic acid is 1.5, and for pyruvic acid is 1.0.

The O/C ratio for oligomers from glyoxal and methylglyoxal is assumed to be 1.5, which is consistent with the measurements by Lim et al. (2010).

We then averaged the O/C ratios of these four SOA components by weighting their mass concentrations to obtain the average O/C ratio for total SOA. As shown in Fig. 8, the modeled O/C ratios are higher than those reported by Ng et al. (2010) and have a normalized mean bias of 33%. The higher O/C ratios estimated in the model might suggest an overestimation of the contribution of aqueous SOA formation or multiple generation oxidation products to SOA formation or a missing aerosol phase reaction of condensed SVOCs that would lead to products with lower O/C ratios. Alternatively, the AMS instrument might underestimate O/C ratios.

5. Conclusions

In this paper, we simulated the formation of SOA in both cloud and aerosol water using multiphase processes with different chemical reactions as well as a surface-limited uptake process. We also conducted a simulation using the GFDL AM3 cloud fields and a simulation including iron chemistry in cloud.

The annual average organic acid concentration (i.e. the sum of oxalic acid, glyoxylic acid, and pyruvic acid) peak over the tropical regions due to the large biogenic emissions and abundant cloud water there, while oligomers generally show a maxima over industrialized areas in the Northern Hemisphere due to formation within aqueous aerosols and the large sulfate aerosol concentrations located in these regions. During the summer, large organic acid concentrations are also predicted in the Northern Hemisphere resulting
from seasonally enhanced biogenic emissions and photochemistry. Similarly, the oligomer concentrations increase in the summer.

Using the surface-limited uptake process scheme with the reactive uptake parameter adopted from the laboratory studies (Case 6) leads to higher aqSOA production rates both in cloud and in aerosol water than the use of only the multiphase process scheme does (Cases 2 and 3). The use of a multiphase reaction scheme for aerosol water decreases the global total aqSOA burden by around 50% (compare Case 1 and Case 2). The use of the uptake parameter method for both cloud water and aerosol water in Case 6 predicts a 185% larger burden than Case 1. An increase of 3 orders magnitude for the effective Henry’s law constants of glyoxal and methylglyoxal in aerosol water leads to an increase of 1 order magnitude in oligomer net production rate and an increase in aqSOA burden by 8%. When we changed the diagnostic cloud field to that simulated by the GFDL AM3 model, the organic acid production rate decreased by around 60% while sulfate formation rates increased slightly, because the GFDL AM3 cloud field has a much smaller cloud water content in tropical regions but a higher cloud water content in the Northern Hemisphere. The replacement of diagnostic cloud field with GFDL AM3 cloud field causes the aqSOA burden to increase by 16%. The aqSOA formation rate in cloud water is slightly less that in aerosol water in the simulation with the GFDL AM3 cloud field, while in the other simulations with the diagnostic cloud field the aqSOA formation rate in cloud water dominates over that in aerosol water. The introduction of Fe chemistry in cloud has a large impact on the aqueous phase OH and aqSOA budget, increasing the global average tropical aqueous OH radical source by a factor of about 2.6 and decreasing the net source and burden of aqSOA by 31% and 11%, respectively.
We also compared the oxalic acid predicted from these different schemes and chemical mechanisms with measurements obtained in Europe, Amazon, Africa, China and US region. Overall, the model tends to underestimate observations, probably because it does not account for the direct emission of oxalic acid from primary sources (e.g., wood burning, meat cooking and biomass burning) or the oxalic acid formed in the aging process associated with emissions plumes. However, this underestimation might also be due to a deposition rate that is too high or to a cloud water content that is too low in the model. In fact, using the GFDL AM3 cloud field in Case 4 improved the model predictions at some sites in the Northern Hemisphere. Consistent with the findings of Myriokefalitakis et al. (2011) and Rinaldi et al. (2011), the comparison of oxalate for marine sites suggests that there may be a missing source of oxalic acid over the ocean. When including the iron chemistry in cloud, the model under-predicts all of the measured concentrations of oxalate by over a factor of 2.

Comparisons of the total SOA (aqSOA combined with the SOA formed in the gas phase) for all cases with the SOA measured by AMS in the Northern Hemisphere show reasonable agreement, although the NMB varies between +20% in Case 6 (which used a surface-based uptake coefficient method for aqSOA both in cloud and aerosol water) to around -42% in Cases 2 and 3 (which used a complete or a parameterized multiphase reaction scheme for aqSOA). While Case 6 over-estimates the observations by a factor of over 3 at a West Africa site and by over a factor of 2 at the two other tropical sites, adopting the multiphase reaction scheme for aqSOA decreases the disagreement to within a factor of 2 at all three sites. In addition to the SOA mass measured by AMS, we compared the O/C ratio of OOA estimated based on a factor analysis of AMS
measurements to the O/C ratio of modeled SOA based on some simple assumptions. The estimated O/C ratios from the model are somewhat higher than those estimated from the measurements, which might indicate that the model somewhat overestimates the contribution of aged organic species and/or aqueous SOA formation to the total SOA formation.

In our study, we were unable to find a single mechanism that is able to capture all observations well. The cases that use an uptake coefficient method for aerosol water perform better than the cases using the multiphase reaction scheme in comparison to the Northern Hemisphere AMS measurements. However, in the tropics, the use of a multiphase process scheme for both aerosol water and cloud water perform best, while the case using the uptake coefficient method in cloud water clearly overestimates the observed OA by more than a factor 3. The multiphase reaction scheme including iron chemistry under predicts the observed oxalate concentrations at all sites. Since the inclusion of the formation of stable metal-oxalate complexes in the model is expected to enhance both oxalate and aqSOA concentrations, a multiphase reaction scheme that included this complex formation might work best.

Future work is needed to close the gap between simulations and observations. In particular, more lab and model studies are needed to improve the representation of chemical reactions within the aqueous phase and at the gas/aerosol interface, since there is still a large inconsistency between the existing uptake parameter method and the multiphase reaction schemes derived from lab studies and field measurements. Also, more work is needed to improve our understanding of oxalate sources and sinks, especially the formation of stable metal-oxalate complexes, because the model tended to
underestimate observed oxalate concentrations for all cases studied. In addition, the high
sensitivity of aqSOA to cloud water content shows the importance of improving the
representation of cloud water content in general circulation models in order to improve
aqSOA formation.

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References

S., Ulbrich, I. M., Mohr, C., Kimmel, J. R., Sueper, D., Sun, Y., Zhang, Q., Trimborn,
A., Northway, M., Ziemann, P. J., Canagaratna, M. R., Onasch, T. B., Alfarr, M. R.,
Prevot, A. S. H., Dommen, J., Duplissy, J., Metzger, A., Baltensperger, U., and
Jimenez, J. L.: O/C and OM/OC Ratios of Primary, Secondary, and Ambient Organic

Altieri, K. E., Seitzinger, S. P., Carlton, A. G., Turpin, B. J., Klein, G. C., and Marshall,
A. G.: Oligomers formed through in-cloud methylglyoxal reactions: Chemical
composition, properties, and mechanisms investigated by ultra-high resolution FT-ICR

Altieri, K. E., Turpin, B. J. and Seitzinger, S. P.: Composition of Dissolved Organic
Nitrogen in Continental Precipitation Investigated by Ultra-High Resolution FT-ICR


Chen, Q., Liu, Y., Donahue, N. M., Shilling, J. E. and Martin, S. T.: Particle-Phase


Liu, X., Mauersberger, G. and Möller, D.: The effects of cloud processes on the
tropospheric photochemistry: An improvement of the eurad model with a coupled
gaseous and aqueous chemical mechanism, Atmos. Environ., 31(19), 3119–3135,

Myriokefalitakis, S., Tsigidaris, K., Mihalopoulos, N., Sciare, J., Nenes, A., Kawamura,
troposphere: a 3-D modeling study, Atmos. Chem. Phys., 11, doi:10.5194/acp-11-

Ng, N. L., Chhabra, P. S., Chan, A. W. H., Surratt, J. D., Kroll, J. H., Kwan, A. J.,
McCabe, D. C., Wennberg, P. O., Sorooshian, A., Murphy, S. M., Daleska, N. F.,
Flagan, R. C., and Seinfeld, J. H.: Effect of NOx level on secondary organic aerosol
(SOA) formation from the photooxidation of terpenes, Atmos. Chem. Phys., 7, 5159–

Ng, N. L., Canagaratna, M. R., Zhang, Q., Jimenez, J. L., Tian, J., Ulbrich, I. M., Kroll, J.
aerosol components observed in Northern Hemispheric datasets from Aerosol Mass
Spectrometry, Atmos. Chem. Phys., 10(10), 4625–4641, doi:10.5194/acp-10-4625-

Odum, J. R., Hoffmann, T., Bowman, F., Collins, D., Flagan, R. C. and Seinfeld, J. H.:
30(8), 2580–2585, 1996.

Pandis, S. N. and Seinfeld, J. H.: Sensitivity analysis of a chemical mechanism for
aqueous-phase atmospheric chemistry, J. Geophys. Res., 94, 1105-1126,

Pankow, J. F.: An absorption model of gas/particle partitioning of organic compounds in
the atmosphere, Atmos. Environ., 28(2), 185–188, doi:10.1016/1352-2310(94)90093-
0, 1994.

Paulot, F., Crounse, J. D., Kjaergaard, H. G., Kurten, A., St Clair, J. M., Seinfeld, J. H.
and Wennberg, P. O.: Unexpected Epoxide Formation in the Gas-Phase
Photooxidation of Isoprene, Science, 325(5941), 730–733,

Peeters, J., Nguyen, T. L. and Vereecken, L.: HOx radical regeneration in the oxidation
of isoprene, Physical Chemistry Chemical Physics, 11(28), 5935–5939,

Penner, J. E., Chuang, C. C. and Grant, K.: Climate forcing by carbonaceous and sulfate

Perri, M. J., Seitzinger, S., and Turpin, B. J.: Secondary organic aerosol production from
aqueous photooxidation of glycolaldehyde: Laboratory experiments, Atmos. Environ.,


<table>
<thead>
<tr>
<th>Case name</th>
<th>SOA formation in cloud water</th>
<th>SOA formation in aerosol water</th>
<th>Cloud field</th>
<th>Iron chemistry</th>
</tr>
</thead>
<tbody>
<tr>
<td>Case 1</td>
<td>Multiphase reaction scheme is used to predict carboxylic acids and a kinetic approach is used for all gas/particle transfer. Aqueous phase reactions of organics are adopted from Jacob (1986), Pandis and Seinfeld (1989), Lim et al. (2005), Hermann (2003), and Herrmann et al. (2005) (Table S3 in the Supplement)</td>
<td>Multiphase reaction scheme is used to predict carboxylic acids with the same aqueous phase reactions and gas/particle transfer as in cloud water. Surface-limited uptake process is used to predict oligomers with a reactive uptake parameter of 3.3E-3 for glyoxal and 2.9E-5 for methylglyoxal.</td>
<td>The diagnostic cloud field&lt;sup&gt;a&lt;/sup&gt;</td>
<td>N/A</td>
</tr>
<tr>
<td>Case 2</td>
<td>The same as in Case 1</td>
<td>Multiphase reaction scheme is used with the aqueous phase reactions proposed by Ervens and Volkamer (2010) (Table S4 in the Supplement)</td>
<td>The diagnostic cloud field&lt;sup&gt;a&lt;/sup&gt;</td>
<td>N/A</td>
</tr>
<tr>
<td>Case 3</td>
<td>Multiphase reaction scheme is used with the detailed chemistry of Lim et al. (2010) (Table S5 in the Supplement)</td>
<td>Multiphase reaction scheme is used with the detailed chemistry of Lim et al. (2010) (Table S5 in the Supplement)</td>
<td>The diagnostic cloud field&lt;sup&gt;a&lt;/sup&gt;</td>
<td>N/A</td>
</tr>
<tr>
<td>Case 4</td>
<td>The same as in Case 1</td>
<td>The same as in Case 1</td>
<td>GFDL AM3 cloud field</td>
<td>N/A</td>
</tr>
<tr>
<td>Case 5</td>
<td>The same as in Case 1, but including aqueous iron chemistry in cloud (Table S6 in the Supplement)</td>
<td>The same as in Case 1</td>
<td>The diagnostic cloud field&lt;sup&gt;a&lt;/sup&gt;</td>
<td>Includes aqueous iron chemistry</td>
</tr>
<tr>
<td>Case 6</td>
<td>Surface-limited uptake process is used with a reactive uptake parameter of 2.9E-3 for both glyoxal and methylglyoxal</td>
<td>Surface-limited uptake process is used with a reactive uptake parameter of 2.9E-3 for both glyoxal and methylglyoxal</td>
<td>The diagnostic cloud field&lt;sup&gt;a&lt;/sup&gt;</td>
<td>N/A</td>
</tr>
</tbody>
</table>

<sup>a</sup>The diagnostic cloud field used the parameterization for cloud water published by Hack (1998)
<table>
<thead>
<tr>
<th>Case name</th>
<th>Species name</th>
<th>Chemical production (Tg/yr)</th>
<th>Chemical destruction (Tg/yr)</th>
<th>Net production (Tg/yr)</th>
<th>Dry deposition (Tg/yr)</th>
<th>Wet deposition Burden (Tg/yr)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Case 1</td>
<td>Glyoxylic acid</td>
<td>15.9</td>
<td>13.5</td>
<td>2.4</td>
<td>0.1</td>
<td>2.3</td>
</tr>
<tr>
<td></td>
<td>Methylglyoxal oligomer</td>
<td>6.9</td>
<td>--</td>
<td>6.9</td>
<td>0.4</td>
<td>6.5</td>
</tr>
<tr>
<td></td>
<td>Glyoxal oligomer</td>
<td>2.3×10^-1</td>
<td>--</td>
<td>2.3×10^-1</td>
<td>2.4×10^-2</td>
<td>2.0×10^-1</td>
</tr>
<tr>
<td>Case 2</td>
<td>Glyoxylic acid</td>
<td>16.5</td>
<td>13.9</td>
<td>2.6</td>
<td>0.1</td>
<td>2.5</td>
</tr>
<tr>
<td></td>
<td>Methylglyoxal oligomer</td>
<td>1.7×10^-2</td>
<td>--</td>
<td>1.7×10^-2</td>
<td>4.4×10^-3</td>
<td>1.3×10^-2</td>
</tr>
<tr>
<td>Case 3</td>
<td>Glyoxylic acid</td>
<td>14.8</td>
<td>12.3</td>
<td>2.5</td>
<td>0.1</td>
<td>2.4</td>
</tr>
<tr>
<td></td>
<td>Oxalic acid</td>
<td>6.0×10^-1</td>
<td>4.4×10^-1</td>
<td>1.6×10^-1</td>
<td>7.8×10^-3</td>
<td>1.5×10^-1</td>
</tr>
<tr>
<td></td>
<td>Glyoxal oligomer</td>
<td>1.1×10^-1</td>
<td>7.5×10^-2</td>
<td>3.4×10^-2</td>
<td>2.6×10^-3</td>
<td>3.1×10^-2</td>
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<tr>
<td></td>
<td>Methylglyoxal oligomer</td>
<td>1.5×10^-3</td>
<td>--</td>
<td>1.5×10^-3</td>
<td>8.2×10^-5</td>
<td>1.4×10^-3</td>
</tr>
<tr>
<td>Case 4</td>
<td>Glyoxylic acid</td>
<td>8.4</td>
<td>7.4</td>
<td>1.0</td>
<td>0.1</td>
<td>0.9</td>
</tr>
<tr>
<td></td>
<td>Oxalic acid</td>
<td>4.5×10^-1</td>
<td>3.6×10^-1</td>
<td>9.0×10^-2</td>
<td>6.0×10^-3</td>
<td>8.0×10^-2</td>
</tr>
<tr>
<td></td>
<td>Glyoxal oligomer</td>
<td>9.0</td>
<td>4.9</td>
<td>4.1</td>
<td>0.3</td>
<td>3.8</td>
</tr>
<tr>
<td></td>
<td>Methylglyoxal oligomer</td>
<td>4.0×10^-1</td>
<td>--</td>
<td>4.0×10^-1</td>
<td>5.7×10^-2</td>
<td>3.4×10^-1</td>
</tr>
<tr>
<td>Case 5</td>
<td>Glyoxylic acid</td>
<td>23.0</td>
<td>20.6</td>
<td>2.4</td>
<td>0.1</td>
<td>2.3</td>
</tr>
<tr>
<td></td>
<td>Oxalic acid</td>
<td>1.8</td>
<td>1.6</td>
<td>0.2</td>
<td>1.0×10^-2</td>
<td>1.9×10^-1</td>
</tr>
<tr>
<td></td>
<td>Glyoxal oligomer</td>
<td>27.3</td>
<td>22.9</td>
<td>4.4</td>
<td>0.2</td>
<td>4.4</td>
</tr>
<tr>
<td></td>
<td>Methylglyoxal oligomer</td>
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<td>--</td>
<td>6.6</td>
<td>0.4</td>
<td>6.2</td>
</tr>
<tr>
<td>Case 6</td>
<td>Glyoxal oligomer</td>
<td>2.3×10^-1</td>
<td>--</td>
<td>2.3×10^-1</td>
<td>2.5×10^-2</td>
<td>2.0×10^-1</td>
</tr>
<tr>
<td></td>
<td>Methylglyoxal oligomer</td>
<td>36.9</td>
<td>--</td>
<td>36.9</td>
<td>1.6</td>
<td>35.3</td>
</tr>
</tbody>
</table>
Table 3. Predicted global average production rates in cloud water by the individual reactions for Case 1a

<table>
<thead>
<tr>
<th>Source</th>
<th>Reaction rate (Tg/yr)</th>
<th>Sink</th>
<th>Reaction rate (Tg/yr)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Glyoxyl</td>
<td>Glyoxal + OH(NO₃) 12.2</td>
<td>Reaction with OH</td>
<td>12.7</td>
</tr>
<tr>
<td></td>
<td>Methylglyoxal +OH(NO₃) 6.5x10^-2</td>
<td>Reaction with NO₃</td>
<td>7.2x10^-3</td>
</tr>
<tr>
<td></td>
<td>Acetic acid + OH 1.3</td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>Glycolaldehyde +OH 2.2</td>
<td></td>
<td></td>
</tr>
<tr>
<td>Oxalic</td>
<td>Glyoxylic acid +OH 14.6</td>
<td>Reaction with OH</td>
<td>3.5</td>
</tr>
<tr>
<td></td>
<td>Glyoxylic acid + NO₃ 8.7x10^-1</td>
<td>Reaction with NO₃</td>
<td>1.1</td>
</tr>
<tr>
<td>Pyruvic</td>
<td>Methylglyoxal + OH 7.0x10^-1</td>
<td>Reaction with OH</td>
<td>5.2x10^-1</td>
</tr>
<tr>
<td></td>
<td>Methylglyoxal + NO₃ 1.5x10^-2</td>
<td>Reaction with NO₃</td>
<td>1.4x10^-3</td>
</tr>
</tbody>
</table>

a organic acids formed in cloud water account for around 67% of total aqSOA formation rate, while the rest is attributed to the oligomer formation in aerosol water.

Table 4. Global budgets of aqSOA precursors (Tg/yr) for Case 1

<table>
<thead>
<tr>
<th></th>
<th>Glyoxal</th>
<th>Methylglyoxal</th>
<th>Glycolaldehyde</th>
<th>Acetic Acid</th>
</tr>
</thead>
<tbody>
<tr>
<td>Emission</td>
<td>0</td>
<td>0</td>
<td>0</td>
<td>31.4</td>
</tr>
<tr>
<td>Gas phase production</td>
<td>69.6</td>
<td>167.3</td>
<td>81.5</td>
<td>29.5</td>
</tr>
<tr>
<td>Aqueous phase production</td>
<td>3.4</td>
<td>0</td>
<td>0</td>
<td>3.8x10^-1</td>
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<tr>
<td>Gas phase consumption</td>
<td>47.7</td>
<td>158.0</td>
<td>63.7</td>
<td>25.9</td>
</tr>
<tr>
<td>Aqueous phase consumption</td>
<td>17.0</td>
<td>0.88</td>
<td>5.3</td>
<td>1.2</td>
</tr>
<tr>
<td>Deposition</td>
<td>8.3</td>
<td>9.3</td>
<td>12.5</td>
<td>34.2</td>
</tr>
</tbody>
</table>
Table 5. aqSOA formation in cloud vs. aqSOA formation in aerosol water

<table>
<thead>
<tr>
<th>Case name</th>
<th>SOA production rate in cloud (Tg/yr)</th>
<th>SOA production rate in aerosol water (Tg/yr)</th>
<th>Total aqSOA production rate (Tg/yr)</th>
<th>Fraction of SOA production in cloud</th>
</tr>
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<tbody>
<tr>
<td>Case 1</td>
<td>13.4</td>
<td>6.7</td>
<td>20.1</td>
<td>66.7%</td>
</tr>
<tr>
<td>Case 2</td>
<td>13.8</td>
<td>-0.4</td>
<td>13.4</td>
<td>103.0%</td>
</tr>
<tr>
<td>Case 3</td>
<td>13.1</td>
<td>-0.4</td>
<td>12.7</td>
<td>103.2%</td>
</tr>
<tr>
<td>Case 4</td>
<td>6.8</td>
<td>8.5</td>
<td>15.3</td>
<td>44.4%</td>
</tr>
<tr>
<td>Case 5</td>
<td>7.3</td>
<td>6.5</td>
<td>13.8</td>
<td>52.9%</td>
</tr>
<tr>
<td>Case 6</td>
<td>46.8</td>
<td>12.6</td>
<td>59.4</td>
<td>78.8%</td>
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</tbody>
</table>

Table 6. Normalized mean bias (NMB) and correlation coefficient (R) between the predicted SOA for the simulations and observations reported by Zhang et al. (2007). The number of sites in the comparison is in parentheses.

<table>
<thead>
<tr>
<th>Case name</th>
<th>Urban sites (N=14)</th>
<th>Urban downwind sites (N=6)</th>
<th>Rural sites (N=17)</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>NMB</td>
<td>R</td>
<td>NMB</td>
</tr>
<tr>
<td>Case 1</td>
<td>-40.3%</td>
<td>0.69</td>
<td>-37.6%</td>
</tr>
<tr>
<td>Case 2</td>
<td>-49.2%</td>
<td>0.71</td>
<td>-48.0%</td>
</tr>
<tr>
<td>Case 3</td>
<td>-49.5%</td>
<td>0.72</td>
<td>-48.3%</td>
</tr>
<tr>
<td>Case 4</td>
<td>-28.1%</td>
<td>0.63</td>
<td>-25.9%</td>
</tr>
<tr>
<td>Case 5</td>
<td>-41.2%</td>
<td>0.69</td>
<td>-38.4%</td>
</tr>
<tr>
<td>Case 6</td>
<td>-12.5%</td>
<td>0.72</td>
<td>+8.9%</td>
</tr>
</tbody>
</table>
Table 7. Comparison of simulated OA and NO₅ with observations in tropical forested regions.

<table>
<thead>
<tr>
<th></th>
<th>West Africa (Below 2 km)</th>
<th>Amazon Basin (surface)</th>
<th>Malysian Borneo (surface)</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>NO₅ (ppt)</td>
<td>SOA (µg/m³)</td>
<td>Total OA (µg/m³)</td>
</tr>
<tr>
<td>Observations</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td></td>
<td>210 (Capes et al., 2009)</td>
<td>1.18 (Capes et al., 2009)</td>
<td>0.7 (Chen et al., 2009)</td>
</tr>
<tr>
<td>Simulations</td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Case 1</td>
<td>354</td>
<td>2.54</td>
<td>2.54</td>
</tr>
<tr>
<td>Case 2</td>
<td>354</td>
<td>2.36</td>
<td>2.19</td>
</tr>
<tr>
<td>Case 3</td>
<td>352</td>
<td>2.20</td>
<td>2.02</td>
</tr>
<tr>
<td>Case 4</td>
<td>367</td>
<td>2.69</td>
<td>2.86</td>
</tr>
<tr>
<td>Case 5</td>
<td>347</td>
<td>1.90</td>
<td>1.64</td>
</tr>
<tr>
<td>Case 6</td>
<td>352</td>
<td>4.47</td>
<td>4.45</td>
</tr>
</tbody>
</table>
Table 8. Comparison of the normalized mean bias (NMB) between observations and the model results for different cases.

<table>
<thead>
<tr>
<th>Case</th>
<th>AMS measurements at rural sites in Northern Hemisphere(^1)</th>
<th>AMS measurements in tropical regions(^2)</th>
<th>Oxalate measurements(^3)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Case 1</td>
<td>-32.4%</td>
<td>70.7%</td>
<td>-63.1%</td>
</tr>
<tr>
<td>Case 2</td>
<td>-42.5%</td>
<td>49.7%</td>
<td>-61.2%</td>
</tr>
<tr>
<td>Case 3</td>
<td>-42.9%</td>
<td>38.7%</td>
<td>-62.2%</td>
</tr>
<tr>
<td>Case 4</td>
<td>-13.8%</td>
<td>82.9%</td>
<td>-76.2%</td>
</tr>
<tr>
<td>Case 5</td>
<td>-33.2%</td>
<td>19.3%</td>
<td>-88.2%</td>
</tr>
<tr>
<td>Case 6</td>
<td>+20.0%</td>
<td>189.7%</td>
<td>N/A</td>
</tr>
</tbody>
</table>

\(^1\) reported by Zhang et al. (2007)

\(^2\) reported by Capes et al. (2009), Gilardoni et al. (2011), and Robinson et al. (2011)

\(^3\) compiled by Myriokefalitakis et al. (2011)
Figure 1. Annual mean simulated concentrations of total aqSOA (panel A and B), organic acids (the sum of glyoxylic acid, pyruvic acid and oxalic acid) (panels C and D), and oligomers from glyoxal and methylglyoxal (panels E and F). The left column shows the global distributions at the level of 971hPa in the model; the right column depicts the zonal mean distributions. All plots are for Case 1. Units: μg/m³.
Figure 2. Annual mean grid-box averaged cloud liquid water content (LWC) in mg/m$^3$ at approximately 971hPa (A) and zonal mean content (B), and annual mean aerosol LWC in µg/m$^3$ at approximately 971 hPa (C) and zonal mean content (D). Plots are for Case 1, but are the same for all cases except Case 4 (compare Figure 4).
Figure 3. Seasonally averaged column concentrations (mg/m$^2$) of organic acids (left column) and oligomers (right column) in December, January and February (DJF) (top row) and in June, July and August (JJA) (bottom row). All plots are for Case 1.
Figure 4. Annual mean grid-box averaged cloud liquid water content (LWC) in mg/m$^3$ from the GFDL AM3 cloud field at approximately 971 hPa (A) and zonal mean content (B) for Case 4. Panel C shows the zonal mean distributions for the ratio of grid-box averaged cloud LWC in Case 1 to that in Case 4; panel D shows the zonal mean distributions for the ratio of the source of aqSOA formed in cloud water in Case 1 to that in Case 4.
Figure 5. The predicted aqueous OH concentration (units in mol/L) in cloud water near 971 hPa in Case 1 (without Fe chemistry) and Case 5 (with Fe chemistry). The ratio of annual mean cloud water OH radical production rates in Case 5 to that in Case 1 is also shown (2nd row). The value shown is the logarithm of the ratio.
Figure 6. Comparison of the oxalate mass concentrations observed at rural and marine sites adopted from Table S3 of Myriokefalitakis et al. (2011) with the oxalic acid concentrations simulated in Case 1, Case 4 and Case 5. Solid lines show the 1:1 ratio, and dashed lines show the 1:2 and 2:1 ratios. The measurements at the various sites were made in different seasons and different years between 1980 and 2007 and most of them were reported with several days of sampling duration. The model results are the average values over the same months as the observations.
Figure 7. Comparison of SOA mass concentrations observed at the urban, urban-downwind and rural sites reported in Zhang et al. (2007) with those simulated in Case 1, Case 2, Case 4, Case 5 and Case 6. Solid lines show the 1:1 ratio, and dashed lines show the 1:2 and 2:1 ratios. The measurements at the various sites were made in different seasons and different years between 2000 and 2006 and were reported for the average of different durations spanning from 8 to 36 days. The model results are the average values over the same months as the observations.
Figure 8. Comparison of O/C ratio estimated by Ng et al. (2010) for urban downwind, rural and remote sites with those simulated in Case 1. The values for other cases are similar to those in Case 1. Solid lines show the 1:1 ratio, and dashed lines show the 10:7 and 7:10 ratios, which roughly corresponds to the uncertainty of the AMS measurements, stated as 30% for the O/C ratio (Aiken et al. 2008). The measurements at the various sites were made in different seasons and different years between 2000 and 2009 and were reported for the average of different durations spanning from 8 to 36 days. The model results are the average values over the same months as the observations.