Anthropogenic amplification of biogenic secondary organic aerosol production

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Abstract

Biogenic secondary organic aerosols (SOA) contribute to a large fraction of fine aerosols globally, impacting air quality and climate. The formation of biogenic SOA depends on not only emissions of biogenic volatile organic compounds (BVOCs) but also anthropogenic pollutants including primary organic aerosol, sulfur dioxide (SO$_2$), and nitrogen oxides (NO$_x$). However, the anthropogenic impact on biogenic SOA production (AIBS) remains unclear. Here we use the decadal trend and variability of observed OA in the southeast US, combined with a global chemistry-climate model, to better constrain AIBS. We show that the reduction in SO$_2$ emissions can only explain 40% of the decreasing decadal trend of OA in this region, constrained by the low summertime month-to-month variability of surface OA. We hypothesize that the rest of OA decreasing trend is largely due to reduction in NO$_x$ emissions. By implementing a scheme for monoterpane SOA with enhanced sensitivity to NO$_x$, our model can reproduce the decadal trend and variability of OA in this region. Extending to centennial scale, our model shows that global SOA production increases by 36% despite BVOC reductions from preindustrial period to present day, largely amplified by AIBS. Our work suggests a strong coupling between anthropogenic and biogenic emissions in biogenic SOA production that is missing from current climate models.
1. Introduction

Terrestrial vegetation emits more than 1 Pg per year of BVOCs (Guenther et al., 2012), leading to a major source of SOA in the atmosphere (Goldstein and Galbally, 2007). SOA exerts significant impacts on climate, air quality and human welfare (Shrivastava et al., 2017; Pye et al., 2021), but is not well represented in climate models. Global climate models differ largely in simulated SOA burden, variability, and radiative effects (Tsagaridis et al., 2014) due to complexity associated with emission of precursors, multiphase chemical and physical processes, aging, radiative properties, and other processes (Shrivastava et al., 2017). Many climate models simply scale SOA yield with BVOC precursors (Horowitz et al., 2020; Carslaw et al., 2013; Koch et al., 2011).

Current understanding of biogenic SOA formation has advanced far beyond this simple scaling of BVOC emissions. SOA formation from BVOC oxidation is largely dependent on its oxidants (OH/O$_3$/NO$_x$) and the yields show non-linear behavior under different NO$_x$ conditions (Ng et al., 2017; Presto et al., 2005). One advanced scheme is the Volatility Basis Set (VBS), in which intermediate semivolatile products from the oxidation of BVOCs are grouped into volatility bins and can reversibly condense onto pre-existing organic aerosols (Donahue et al., 2006; Pye et al., 2010). VBS accounts for the dependence of SOA formation on atmospheric oxidants, NO$_x$-dependent chemical regimes, POA and temperature. Some studies showed that VBS schemes underestimated observations and that photochemical aging schemes with varying complexity may improve simulation results in different regions and seasons (Zheng et al., 2015; Robinson et al., 2007; Oak et al., 2022). Another pathway is through reactive uptake of smaller molecules onto aqueous aerosols. Several isoprene oxidation products, such as epoxides (IEPOX) (Paulot et al., 2009) and glyoxal (Liggio, 2005; Li et al., 2016), though often not directly condensable due to their high equilibrium vapor pressure, can undergo aqueous phase reactions and oligomerize in the condensed phase. The detailed mechanism is complicated by aerosol acidity, composition, and coating (Shrivastava et al., 2017). These advancements highlight the role of anthropogenic emissions modulating biogenic SOA formation through nitrogen oxides (NO$_x$), SO$_2$ and primary organic aerosol (POA).

One major uncertainty is to what extent anthropogenic emissions modulate biogenic SOA formation. In the southeast US (SEUS), a region largely covered by natural vegetation and also heavily populated, organic aerosol shows a decreasing trend in the recent two decades (Kim et al., 2015; Atwood et al., 2014), likely due to reductions in POA (Blanchard et al., 2016; Ridley et al., 2018), sulfate and aerosol water (Christiansen et al., 2020; Ridley et al., 2018; Marais et al., 2017; Malm et al., 2017; Blanchard et al., 2016) and NO$_x$ (Zheng et al., 2015; Xu et al., 2015; Pye et al., 2019). Several studies suggest that SO$_2$ largely modulates SOA through reactive uptake of IEPOX (Pye et al., 2013; Marais et al., 2017), but the acidity-catalyzed sulfate uptake mechanism appears to overestimate the trend of OA reduction in the SEUS (Zheng et al., 2020). The role of NO$_x$ remains unclear. While SOA yield generally decreases with NO$_x$ level due to fragmentation of large molecules (Kroll and Seinfeld, 2008), recent studies show that NO$_x$ can in fact increase SOA production (Zheng et al., 2015; Xu et al., 2015; Pye et al., 2019; Pullinen et al., 2020). The combined effect of NO$_x$, SO$_2$, and POA can be significant (Carlton et al., 2010; Hoyle et al., 2011), but remain unconstrained by ambient observations.

Here we use the decadal trend and variability of observed OA in the southeast US, combined with other observational datasets and a global chemistry-climate model (GFDL AM4.1) (Horowitz et al., 2020; Carslaw et al., 2013; Koch et al., 2011).
al., 2020), to better constrain the anthropogenic impact on biogenic SOA production (AIBS). We use three schemes (summarized in Table 1 and detailed in Methods), Simple, CMPX, and CMPX_ag, to investigate the AIBS from decadal to centennial time scales.

2. Methods

2.1 GFDL AM4.1

The Geophysical Fluid Dynamics Laboratory (GFDL)’s Atmospheric Model version 4.1 (AM4.1)(Horowitz et al., 2020) is a three-dimensional global chemistry-climate model that includes interactive simulation of stratospheric chemistry and tropospheric O\textsubscript{3}-NO\textsubscript{x}-CO-VOC and bulk aerosol chemistry, allowing explicit treatment of aerosol reactive uptake of IEPOX and glyoxal(Li et al., 2016, 2018; Mao et al., 2013). Community Emissions Data System (CEDS) are used for historical anthropogenic emissions from 1849 to 2016. Global fire emissions are based on Global Fire Emissions Database version 4 (GFED4), the Fire Modeling Intercomparison Project (FireMIP), visibility-observations and Global Charcoal Database (GCD). Biogenic isoprene and monoterpene emissions are calculated online by the Model of Emissions of Gases and Aerosols from Nature version 2.1 (MEGAN2.1), using empirical functions of plant-functional-type (PFT)-specific emission basal factors, leaf area index (LAI), temperature and light. LAI and PFTs are prescribed at the 1992 level. Dependence of soil moisture, O\textsubscript{3} and CO\textsubscript{2} are neglected due to large uncertainties. Radiative effects of SOA is calculated assuming SOA is externally mixed from other aerosols(Horowitz et al., 2020), although ISOA is formed through sulfate uptake in the chemistry module.

We perform simulations for years 1998-2016 for present day (PD) and 1870-1888 for pre-industrial period (PI). In each simulation, the first two years are discarded as spin-up. The remaining 17 years are used for analysis. The PD simulations are nudged with reanalysis winds from NCEP-DOE Reanalysis 2. The PI simulations are free running with no nudging. All simulations are driven by observed or reconstructed sea surface temperature and sea-ice(Horowitz et al., 2020). In the two PI simulations, we scale up the isoprene and monoterpene emission basal factors by 35% to account for the higher natural vegetation cover at pre-industrial period than today, equivalent to a 26% reduction of natural vegetation cover from PI to PD(Unger, 2014). We apply this single scaling factor to BVOC emission basal factors as an idealized study instead of using reconstructed land cover type and LAI to avoid uncertainties in historical vegetation reconstructions.

2.2 Modeling of SOA formation

In GFDL AM4.1, SOA is composed of anthropogenic SOA (ASOA), isoprene SOA (ISOA) and monoterpene SOA (TSOA). ASOA is formed through the oxidation of C\textsubscript{4}H\textsubscript{10} by OH in all simulations. In the default “Simple” scheme, ISOA and TSOA are assumed to be produced with a pseudo-emission equivalent to a 10% per-carbon yield of the interactively calculated isoprene and monoterpene emissions, respectively.

In the updated “CMPX” scheme, isoprene and monoterpenes are oxidized by OH, O\textsubscript{3} and NO\textsubscript{3}. TSOA is calculated by a 4-product Volatility Basis Set (VBS) summarized in Table 1. Organic peroxy radicals (RO\textsubscript{2}) formed from OH- and O\textsubscript{3}-initiated oxidation of monoterpenes can react with NO under high-NO\textsubscript{x} conditions and with HO\textsubscript{2} under low-NO\textsubscript{x} conditions. The low-NO\textsubscript{x} pathway (RO\textsubscript{2}+HO\textsubscript{2}) has higher yields for SOA than the high-NO\textsubscript{x} pathway (RO\textsubscript{2}+NO)(Pye et
The branching ratio between the low- versus high-NO\textsubscript{x} pathways are defined as:

\[
\beta_{NO} = \frac{k_{RO2+NO} \cdot [NO]}{k_{RO2+NO} \cdot [NO] + k_{RO2+HO2} \cdot [HO2]}
\]

Where \(k_{RO2+NO}\) and \(k_{RO2+HO2}\) represent the reaction rate coefficients of RO\textsubscript{2}+NO and RO\textsubscript{2}+HO\textsubscript{2}, respectively.

At nighttime, the NO\textsubscript{3}-initiated oxidation of monoterpenes has a high yield of organic nitrates and contributes a significant amount of SOA (Ng et al., 2017).

In an additional configuration, we further implement a simplified photochemical aging parameterization to the semivolatile oxidation products of terpenes in the CMPX scheme (CMPX\textsubscript{ag}) (Zheng et al., 2015), to account for the decrease in volatility as a result of OH oxidation (Donahue et al., 2012). We apply a rate constant of \(k_{OH} = 4\times10^{-11}\) cm\textsuperscript{3} molec\textsuperscript{-1} s\textsuperscript{-1} (Robinson et al., 2007), in line with recent estimates of 2-4 \(\times10^{-11}\) cm\textsuperscript{3} molec\textsuperscript{-1} s\textsuperscript{-1} for terpene SOA (Donahue et al., 2012; Isaacman-VanWertz et al., 2018). As we show below, the aging increases the SOA burden as well as the sensitivity of SOA to NO\textsubscript{x}, improving the model underestimate of SOA by the VBS scheme.

ISOA is computed through the aqueous-phase uptake of IEPOX and glyoxal onto sulfate aerosol. The uptake coefficients for IEPOX and glyoxal are set to 0.001, different than previous studies using higher or acidity-dependent uptake coefficients (Marais et al., 2016; Lin et al., 2014a). This is supported by the OA month-to-month variability (MMV) in summer and its decadal trend over the southeast US, as a previous model with acidity-dependent uptake coefficients shows too high of MMV and too much OA in the early 2000s (Zheng et al., 2020). The uptake rate coefficients can be even lower due to the effect of aerosol-phase state (Zhang et al., 2018b). To avoid uncertainties associated with aerosol acidity, relative humidity, and coating effect, we here apply uptake coefficient of 0.001 for both IEPOX and glyoxal. This leads to good agreement between our model and observation in the SEUS on both OA magnitude and summertime MMV (Figure 1, S1, S2 and S13). The details of the Simple, CMPX, and CMPX\textsubscript{ag} schemes are summarized in Table 1.

### 2.3 Observational datasets

For model evaluation we use filter measurement of organic carbon from two surface aerosol measurement networks in the US: IMPROVE (the Interagency Monitoring of Protected Visual Environments) (Solomon et al., 2014) and SEARCH (the SouthEastern Aerosol Research and Characterization) (Edgerton et al., 2005). We focus on the southeast US which is both heavily vegetated and populated. We select 20 IMPROVE sites and 3 SEARCH rural sites within the southeast US region (29-37°N, 74-96°W). We apply a seasonal-dependent ratio to convert organic carbon to organic aerosol (OA) mass: 2.2 in June-July-August, 1.8 in December-January-February and 1.9 in other months (Philip et al., 2014).

We also compare modeling results to OA measurement by Aerosol Chemical Speciation Monitor (ACSM). We select 3 European sites from the ACTRIS (the Aerosol, Clouds and Trace Gases Research Infrastructure) network: Hyytiala (Finland), Puy de Dome (France) and Birkenes II (Norway); two sites from the ARM (Atmospheric Radiation Measurement) network: Southern
Great Plains (US) and Manacapuru, Amazonia (Brazil). These sites are covered by natural vegetation and have more than a year’s worth of data available.

3. Results
3.1 Decadal trend of summertime OA in SEUS and its variability
The SEUS is a region heavily influenced by both biogenic and anthropogenic emissions (Mao et al., 2018). In the last two decades, organic aerosol shows a decreasing trend, resulting from reductions in anthropogenic pollutants including SO$_2$ and NO$_x$ (Marais et al., 2017; Blanchard et al., 2016; Ridley et al., 2018). Both CMPX$_{\text{ag}}$ and CMPX schemes can reproduce the summertime surface OA concentrations measured from SEARCH and IMPROVE networks (Solomon et al., 2014; Edgerton et al., 2005), at 4-5 g/m$^3$ as shown in Figure 1, while the Simple scheme has a slight overestimate (~7 g/m$^3$).

We first examine the simulated decadal OA trend in the SEUS against filter-based measurements from IMPROVE and SEARCH networks (Methods). From 2000 to 2016, the measured summer OA declines by -0.13 g/m$^3$/year from SEARCH and by -0.09 g/m$^3$/year from IMPROVE, both at a reduction rate of -2.3%/year (Figure 1a). This decreasing trend is well reproduced by the CMPX$_{\text{ag}}$ simulation with a decrease of -0.11 g/m$^3$ (-2.0%) per year (Figure 1a), less so with the CMPX scheme (-1.4%/year). In contrast, the Simple scheme shows a slight increase (+0.7%/year) in surface OA due to lack of AIBS and little change of POA in 2000-2016 in this region (Figure 1c).

We further examine the summertime month-to-month variability of surface OA. We find that both CMPX$_{\text{ag}}$ and CMPX schemes can well reproduce the low summertime month-to-month variability of surface OA (standard deviation smaller than 2 g/m$^3$) constrained by IMPROVE and SEARCH measurements (Figure S5), using fixed uptake coefficients (γ=0.001) of IEPOX and glyoxal. This summertime month-to-month variability was found to be too high (standard deviation up to 5 g/m$^3$) in the early 2000s using an acidity-dependent IEPOX reactive uptake scheme (Marais et al., 2016, 2017), pointing to additional species besides SO$_2$ driving the decreasing OA trend.

One unique feature of the CMPX$_{\text{ag}}$ simulation is the dominance of TSOA (Figure 1), mainly through enhanced sensitivity of TSOA production to NO$_x$. Such dominance of TSOA in this region is also supported by recent field observations (Xu et al., 2018; Zhang et al., 2018a). We find TSOA contributes to 60% of the surface OA trend in the CMPX$_{\text{ag}}$ scheme, mainly through NO$_x$ reduction. The NO$_x$-initiated pathway contributes to the majority of surface TSOA decrease (Figure S3), resulting from the rapid decrease of NO$_3$ (Figure 1d) (Boyd et al., 2017; Rollins et al., 2012). Compared to the CMPX scheme, the dominant contribution of TSOA is largely due to the OH aging effect, which amplifies the SOA yield from all monoterpene oxidation channels. As a result, we find that NO$_x$ reduction accounts for 60% of OA decrease in SE US. This enhanced sensitivity to NO$_x$, resonates with recent developments on monoterpene-derived highly oxygenated organic molecules or autooxidation (Pye et al., 2019), highlighting the importance of NO$_x$ in AIBS.
ISOA contributes to 40% of surface OA trend in the CMPX_ag scheme, mainly through SO2 reduction. The decrease in surface ISOA, at -0.05 g/m²/year, is associated with the strong reduction in sulfate (-7%/year). The rapidly decreasing sulfate, NOx and O3 in the model are consistent with observations over the SEUS (Figure S4) and previous studies (Zheng et al., 2020; Wells et al., 2021; Simon et al., 2015). In contrast to Marais et al. (2017), we find that this nondominant role of ISOA brings model into much better agreement with observations, especially on the annual summertime month-to-month variability of surface OA (standard deviation smaller than 2 g/m²) constrained by IMPROVE and SEARCH measurements (Figure S5) (Zheng et al., 2020).

The observed summertime month-to-month variability also implies a weaker dependence of OA to sulfate aerosols in this region than as shown in Marais et al. (2017), highlighting the importance of TSOA.

We find a similar trend of summer OA column concentration to the surface OA trend in the model. The CMPX_ag simulation suggests a decreasing trend in summer OA column concentration, driven by both TSOA (-0.13 mg/m²/year) and ISOA (-0.12 mg/m²/year) (Figure 1b). Similar to the surface, the aging effect increases the column production of TSOA in CMPX_ag and its sensitivity to changes in NOx compared with the CMPX scheme.

### 3.2 Present-day OA in vegetated regions and global budget

We further evaluated the modeled OA against measurements by Aerosol Chemical Speciation Monitor (ACSM) in other vegetated regions (Figure S6). In 3 European sites from the ACTRIS network (Methods) (Crenn et al., 2015), all model simulations underestimate measured OA. One possible reason is uncertainties associated with BVOC emissions and biogenic SOA. Jiang et al. (Jiang et al., 2019) showed that MEGAN overestimates isoprene emission but underestimates monoterpene emissions in Europe by a factor of 3. At the Amazon and US sites from the ARM network (Un et al., 2019), the CMPX_ag scheme successfully captures the measured OA magnitude and seasonal variation. The Simple scheme overestimates the surface OA in Amazon. In the SEUS compared to filter measurements, all simulations show lower OA in winter than observations (Figure S6), likely due to an underestimate of wintertime emissions of POA (Tsigaridis et al., 2014; Liu et al., 2021). In general, the updated CMPX_ag and CMPX schemes agree well with observations in the Amazon and US where biogenic emissions are high.

Globally, the SOA burden from the Simple, CMPX and CMPX_ag schemes are 0.99, 0.50 and 1.05 Tg, respectively, and their SOA production rates are 82, 40 and 69 Tg/year (Figure 2), in agreement with other global modeling studies. The AeroCom phase II model intercomparison summarizes a median SOA source of 51 Tg/year with a range between 16 to 121 Tg/year (Tsigaridis et al., 2014), although top-down methods indicate SOA source could be up to 80-380 Tg/year (Spracklen et al., 2011). Uncertainties associated with BVOC emissions contribute to the wide spread of SOA estimate by global models. In GFDL AM4.1, annual isoprene and monoterpene emissions are computed to be 505±14 and 137±5 Tg/year, respectively (Figure 2), in line with previous estimates (Guenther et al., 2012).

Detailed SOA budgets for the three schemes are summarized in Table 2. The CMPX and CMPX_ag schemes have much less ISOA than the Simple scheme as the latter has high pseudo emission of isoprene SOA, which is 10% in GFDL AM4.1 as compared to 3% used in other global models like GEOS-Chem (Pai et al., 2020; Henze and Seinfeld, 2006). ISOA (22.2 Tg/year) and...
TSOA (14.4 Tg/year) in the CMPX scheme are consistent with previous estimate by GEOS-Chem(Pai et al., 2020; Zheng et al., 2020). The CMPX_ag scheme has higher TSOA (44 Tg/year) than CMPX and Simple due to the aging effect of semivolatile oxidation products from terpenes (Figure 2), and is close to the high end of estimate (12.7-40 Tg/year) by AeroComII(Tsigaridis et al., 2014). ASOA is often neglected by global models despite an estimate of 13.5 Tg/year suggesting ASOA as a non-negligible source(Tsigaridis et al., 2014). In GFDL AM4.1, ASOA (3.3 Tg/year) only considers oxidation of C_8H_{16}, which does not well represent all ASOA and warrants further research.

### 3.3 Centennial change in biogenic SOA and direct radiative forcing

We now extend our analysis of AIBS from the decadal scale to the centennial scale. To represent the higher natural vegetation cover during PI, we scale up isoprene and monoterpene emission basal factor in the PI simulations by 35%, equivalent to a 26% reduction of natural vegetation cover from PI to PD(Unger, 2014). This simple scaling should be considered as an idealized study to avoid uncertainties associated with historical vegetation reconstruction and the complex role of CO₂ including both fertilization and inhibition effects. From 1870s to 2000s, the simulated isoprene emissions decrease from 632±15 to 505±14 Tg/year (-20%) and monoterpene emissions decrease from 161±5 to 137±5 Tg/year (-15%) (Figure 2), consistent with previous studies(Heald and Spracklen, 2015).

Despite the reduction in BVOC emissions from PI to PD, we show a significant increase of biogenic SOA (Figure 2), resulting from increase in anthropogenic emissions and amplified by AIBS. With an increase by 1.4, 7, and 4 for emissions of POA, SO² and NOₓ, total SOA production increases by 36% and its burden increases by 42% (in the CMPX_ag scheme). ASOA, ISOA and TSOA contribute 17%, 62%, and 21% to the changes in total SOA production, respectively. In contrast, the Simple scheme shows a decrease of SOA production following the reduction in BVOC emissions (Figure 2b). The large increase of SOA from PI to PD differs from previous estimates(Spracklen et al., 2011; Heald and Spracklen, 2015; Zhu et al., 2019; Scott et al., 2017; Lin et al., 2014b; Heald and Geddes, 2016; Hoyle et al., 2009), largely due to AIBS constrained by observations.

The total PI-to-PD SOA rise is largely dominated by ISOA (62%), resulting from the strong increase in anthropogenic SO² emissions and uptake of IEPOX and glyoxal onto sulfate aerosols. The global burden of sulfate aerosol has doubled from 0.7 Tg at PI to 1.6 Tg at PD, with large increase over the tropics, SEUS, and Eurasia (Figure S11). The increase in TSOA is due to both increased NOₓ emissions and POA emissions. In contrast to the decadal trend where βNOₓ barely changes, the PI-to-PD increase of TSOA due to the change of NOₓ is suppressed by the shift of βNO. The branching ratio βNO increases from a global average of 0.32 at PI to 0.61 at PD (Figure S12), indicating a shift from low-NOₓ pathway (higher yields) to high-NOₓ pathway (lower yields) for the OH- and O₃-initiated oxidation. These competing effects lead to a net +10% change in TSOA production and a +14% increase in burden from PI to PD. The PI-to-PD change in TSOA in the CMPX scheme is small (-7% in production and +6% in burden). Increased POA provides more organic mass for monoterpene oxidation products to condense on, especially in central Africa and central South America (Figure S11).
The large increase of biogenic SOA leads to a cooling direct radiative forcing (DRF) from PI to PD, opposed to the warming suggested by the Simple scheme. DRF is usually defined as the difference between PI and PD direct radiative fluxes at top-of-atmosphere (TOA) under all-sky conditions. We show in Figure 3 the global instantaneous DRF at top-of-atmosphere (TOA) of -26 (-44) mW/m², comparable to that of POA(-98 mW/m²). In contrast, the Simple scheme shows a warming DRF of +17 mW/m², largely due to lack of AIBS. The DRF of SOA in the updated schemes resides within reported AeroComII estimates, which ranges from -210 to -10 mW/m², with a mean value of -60 mW/m² and a median value of -20 mW/m²(Myhre et al., 2013). Due to this increase of SOA burden, our results may also imply a large indirect radiative forcing from biogenic SOA that is missing from previous work(Carslaw et al., 2013).

4. Summary

Our work suggests a strong coupling between anthropogenic and biogenic emissions in biogenic SOA production. Constrained by observations in SEUS, we show that the summertime OA decreasing trend is likely driven by reduction in both NOx and SO2 emissions, through TSOA and ISOA. SO2 alone cannot explain this trend. Our results also point to the importance role of NOx on modulating biogenic SOA, in line with recent understanding on autooxidation(Pye et al., 2019), although further studies are warranted. The success of the updated schemes in capturing the observed OA trend and month-to-month variability provides confidence in model simulations over longer time scales. At centennial scale, ISOA dominates the total SOA change as a result of a significant rise in global sulfate aerosol from PI to PD, especially in the fast-developing regions like Africa, Middle East, India, and China. POA increases greatly in central Africa and central South America as well as India and east China, which enhances TSOA production. The significant increase in SOA due to AIBS in these regions poses new challenges to meet the World Health Organization’s recommendation on annual fine particulate matter exposure (5 μg/m²)(Pai et al., 2022). Under future scenarios with reduced emissions of SO2, NOx and POA, the AIBS may indicate larger reductions in SOA than current model predictions.

The updated SOA scheme in GFDL AM4.1 shows an advance in representing vegetation-chemistry-climate interactions than the default model which assumes fixed yields of SOA from biogenic hydrocarbons, although a variety of uncertainties still exist in the evaluation of SOA and its climate impact. First, the model likely underestimates wintertime POA in the US, total OA in Europe and anthropogenic SOA globally, and the model does not consider absorbing SOA or brown carbon which could form from biomass burning and aging(Tsigaridis and Kanakidou, 2018). The model applies the same optical parameters for all SOA as hydrophilic POA. Second, we use a simplified aging parameter to represent the enhanced SOA sensitivity of NOx. This should be improved by more explicit schemes on autooxidation (Pye et al., 2019) and organic nitrates(Takeuchi and Ng, 2019). Properties that influence the multiphase growth of SOA, including coating and viscosity, are also not implemented in our model(Shrivastava et al., 2017). The model does not consider nucleation of extremely low volatile compounds from BVOC oxidation, which may increase SOA in pristine environments in the pre-industrial period, thus reducing the PI-to-PD radiative forcing of SOA(Gordon et al., 2016; Zhu et al., 2019). These uncertainties warrant further research in studies on anthropogenic-influenced SOA in climate models.
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Methodology: YZ, LWH, RM, DJP, VN, JM
Investigation: YZ, LWH, JM
Writing—original draft: YZ, JM
Writing—review & editing: YZ, LWH, RM, DJP, VN, JL, JM

Data availability

Competing interests
The authors declare no competing interests.
References:


Table 1. Comparison of the SOA schemes used in this study. Further details and discussions are included in Methods.

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<thead>
<tr>
<th>Scheme</th>
<th>ASOA</th>
<th>ISOA</th>
<th>TSOA</th>
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<tr>
<td>Simple</td>
<td>C4H10+OH</td>
<td>10% yields from isoprene emissions</td>
<td>10% yields from monoterpane emissions</td>
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<td>CMPX</td>
<td>Same as Simple</td>
<td>Heterogeneous uptake of IEPOX ($\gamma=0.001$) and glyoxal ($\gamma=0.001$)</td>
<td>4-bin VBS $^2$</td>
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<td>$\alpha$ for C* (C* in $\mu$g/m$^3$)</td>
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<td>Yield at 10 $\mu$g/m$^3$</td>
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<td>CMPX$_{ag}$</td>
<td>Same as Simple</td>
<td>Same as CMPX</td>
<td>Same as CMPX, with aging k$_{OH}$ = $4 \times 10^{-11}$ cm$^3$ molec$^{-1}$ s$^{-1}$</td>
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$^1$ $\gamma$ represents uptake coefficients of IEPOX or glyoxal onto aqueous sulfate aerosol.

$^2$ In the 4-bin VBS, monoterpane (MTP) is oxidized by OH, O$_3$ or NO$_3$ to generate 4 semivolatile surrogate products, which can reversibly partition into pre-existing organic aerosol. C* represents saturation concentration of each semivolatile product and determines the partitioning of these products between gas and aerosol phase. The mass-based stoichiometric yield coefficients, $\alpha$, for each parent hydrocarbon/oxidant system are fit with a VBS using C* of 0.1, 1, 10, and 100 $\mu$g/m$^3$ (Pye et al. 2010).

$^3$ In the aging scheme, at every time step, each semivolatile product except for the lowest volatility bin (C*=0.1 $\mu$g m$^{-3}$) is assumed to be further oxidized by OH with a rate constant of k$_{OH}$ = $4 \times 10^{-11}$ cm$^3$ molec$^{-1}$ s$^{-1}$, which reduces its volatility by an order of magnitude.
Table 2. Annual mean budget of POA and SOA in all simulations. Results are averaged over 1872-1888 for pre-industrial and 2000-2016 for present-day simulations. SOA includes ASOA (anthropogenic SOA), ISOA (isoprene-SOA), and TSOA (monoterpene-SOA).

<table>
<thead>
<tr>
<th>Simulation</th>
<th>Variable</th>
<th>PI</th>
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<th></th>
<th></th>
<th>PD</th>
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<tr>
<td></td>
<td></td>
<td>Burden (Tg)</td>
<td>Production (Tg/yr)</td>
<td>Wet Deposition (Tg/yr)</td>
<td>Dry Deposition (Tg/yr)</td>
<td>Lifetime (day)</td>
<td>Burden (Tg)</td>
<td>Production (Tg/yr)</td>
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<td>-</td>
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<td>3.3</td>
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<td>0.78</td>
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<tr>
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1For POA budget, the differences between different schemes are negligible.
Figure 1. Summertime (June-July-August) values averaged in the southeast US in 2000-2016. (a) Surface concentrations of OA from the two measurement networks, IMPROVE and SEARCH, and the Simple, CMPX and CMPX_{ag} simulations. (b) Column concentrations of OA. In (a) and (b), color shades represent OA components from the CMPX_{ag} scheme. (c) Standardized emissions of isoprene, monoterpenes, POA, NO_{x} and SO_{2}. (d) Standardized surface concentrations of gases O_{3}, OH and NO_{3}, sulfate aerosol, and branching ratio. In (c) and (d), each variable has been divided by its 17-year average for standardization. In attached text, “m” represents 2000-2016 relative trend with units of %/year; numbers in parenthesis in (a) and (b) represent trends with units of μg/m³/year or mg/m²/year; “avg” represents the 17-year average with units of μg/m³ in (a), mg/m² in (b), mg/m²/hour in (c) and different units shown in (d).
Figure 2. (a) Emissions (Tg/year) of POA, SO₂, NOₓ, isoprene (ISOP) and monoterpenes (MTP). ISOP and MTP emissions have been divided by 10. (b) Simulated SOA global production (Tg/year) and burden (Tg). (c) Simulated SOA column concentration (mg/m²) at PI and PD and their difference in the CMPX_ag scheme.
Figure 3. Direct radiative forcing (RF, mW/m$^2$) of POA and SOA at top-of-atmosphere (TOA) all-sky, TOA clear-sky (TOA_cs), surface all-sky and surface clear-sky (Surface_cs) conditions. Negative RFs represent cooling effects.