Global Atmospheric Budget of Acetone: Air-Sea Exchange and the Contribution to Hydroxyl Radicals

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Global Atmospheric Budget of Acetone: Air-Sea Exchange and the Contribution to Hydroxyl Radicals

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Abstract Acetone is one of the most abundant oxygenated volatile organic compounds (VOCs) in the atmosphere. The oceans impose a strong control on atmospheric acetone, yet the oceanic fluxes of acetone remain poorly constrained. In this work, the global budget of acetone is evaluated using two global models: CAM-chem and GEOS-Chem. CAM-chem uses an online air-sea exchange framework to calculate the bidirectional oceanic acetone fluxes, which is coupled to a data-oriented machine-learning approach. The machine-learning algorithm is trained using a global suite of seawater acetone measurements. GEOS-Chem uses a fixed surface seawater concentration of acetone to calculate the oceanic fluxes. Both model simulations are compared to airborne observations from a recent global-scale, multiseasonal campaign, the NASA Atmospheric Tomography Mission (ATom). We find that both CAM-chem and GEOS-Chem capture the measured acetone vertical distributions in the remote atmosphere reasonably well. The combined observational and modeling analysis suggests that (i) the ocean strongly regulates the atmospheric budget of acetone. The tropical and subtropical oceans are mostly a net source of acetone, while the high-latitude oceans are a net sink. (ii) CMIP6 anthropogenic emission inventory may underestimate acetone and/or its precursors in the Northern Hemisphere. (iii) The MEGAN biogenic emissions model may overestimate acetone and/or its precursors, and/or the biogenic oxidation mechanisms may overestimate the acetone yields. (iv) The models consistently overestimate acetone in the upper troposphere-lower stratosphere over the Southern Ocean in austral winter. (v) Acetone contributes up to 30–40% of hydroxyl radical production in the tropical upper troposphere/lower stratosphere.

Plain Language Summary Acetone is widely observed in the Earth’s atmosphere, with mixing ratios ranging from parts-per-trillion levels in the stratosphere to parts-per-billion levels in polluted regions. Acetone is directly emitted from a wide variety of natural and anthropogenic sources and is also produced from the photochemical oxidation of a number of precursors. The role of the ocean is complicated;
acetone is produced in the ocean from the photolysis of colored dissolved organic materials or from biological processes but is also removed via microbial uptake. Previous studies have found that the direction and magnitude of oceanic acetone fluxes vary dramatically with seasons and locations. In this work, we use a data-oriented machine-learning approach to predict the surface seawater concentration of acetone, leveraging in situ acetone measurements in the surface seawater around the globe. This machine learning-based approach shows promising potential and can be expanded to the bottom-up oceanic emissions of other climate-relevant compounds.

1. Introduction

Oxygenated volatile organic compounds (OVOCs) influence the oxidative capacity of the atmosphere, affecting its self-cleaning capacity and hence the fate of pollutants (Monks, 2005). Acetone (CH₃COCH₃) is one of the most abundant OVOCs identified in the atmosphere, with mixing ratios ranging from ppt (parts per trillion) levels in the remote atmosphere to ppb (parts per billion) levels in polluted regions. Acetone is directly emitted into the atmosphere from a variety of natural and anthropogenic sources, including terrestrial ecosystems (Hu et al., 2013), biomass burning (Holzinger et al., 1999), vehicular exhaust (Kean et al., 2001), industrial and combustion processes (Singh et al., 1994a), and solvent use (Yuan et al., 2010). Acetone is also produced from the photochemical oxidation of a number of volatile organic compound (VOC) precursors, such as terpenes (Lee et al., 2006) and alkanes (Atkinson & Arey, 2003). Major sinks of acetone in the atmosphere include photolysis, oxidation by hydroxyl radicals (OH), and dry deposition (Fischer et al., 2012) (wet deposition contribution is likely small).

The ocean plays a key role in the global atmospheric budget of acetone. Jacob et al. (2002) showed that to reconcile existing observations, the ocean must regulate atmospheric acetone both on the global and regional scale. Jacob et al. (2002) used a global chemical transport model (GEOS-Chem) with parameterized oceanic acetone emissions. Fischer et al. (2012) included updated (slower) photolysis loss in GEOS-Chem and found that with this change to the lifetime of acetone, a large net ocean source is not needed to explain observed atmospheric mixing ratios. Fischer et al. (2012) proposed that the ocean may both emit or uptake acetone; the direction and magnitude of the oceanic acetone fluxes show substantive spatial and seasonal variations. Due to the sparse availability of acetone measurements in the surface seawater, Fischer et al. (2012) prescribed a fixed surface seawater concentration of acetone (15 nmol L⁻¹), and air-sea fluxes of acetone were calculated based on this constant value. Brewer et al. (2017) show that predictions of atmospheric acetone abundances in the tropics and over the Southern Hemisphere oceans are particularly sensitive to uncertainty associated with the concentration of acetone in surface seawater. Over the past decade, more surface seawater measurements of acetone have become available, with improved spatial and seasonal coverage (Beale et al., 2013; Dixon et al., 2013, 2014; Hudson et al., 2007; Kameyama et al., 2010; Marandino et al., 2005; Schlundt et al., 2017; Yang, Beale, et al., 2014; Yang, Blomquist, et al., 2014). Marine acetone fluxes have been estimated indirectly based on acetone measurements in both the surface seawater and the overlying atmosphere (Taddei et al., 2009; Zhou & Mopper, 1997). Direct flux measurements have also been reported using the eddy covariance method (Marandino et al., 2005; Yang, Beale, et al., 2014) and the near-surface vertical gradient method (Tanimoto et al., 2014). These observations provide much better constraints on air-sea exchange.

The photolysis of acetone leads to the production of hydrogen oxide radicals (HOₓ = OH + HO₂). Due to its relatively long global atmospheric lifetime (approximately a few weeks; Fischer et al., 2012; Khan et al., 2015), acetone may undergo long-range transport and contribute to the production of peroxyacetyl nitrate (PAN), an important reservoir of nitrogen oxides (NOₓ = NO + NO₂) (Fischer et al., 2014) and HOₓ radicals. In particular, previous studies have proposed that acetone may contribute substantially to HOₓ production in the upper troposphere and lower stratosphere, where OH production from ozone (ozone photolysis followed by the reaction between water vapor and the singlet oxygen atom, O(¹D)) is less efficient, due to lower water vapor concentrations (Arnold et al., 1986, 1997; Jaegle et al., 2001; Singh et al., 1994, 1995). The temperature and wavelength dependency of acetone photolysis quantum yield was revisited (Blitz et al., 2004) since many of these earlier studies were published, and it was found that the photolysis of acetone is ∼3 times slower than previously estimated in the upper troposphere. A recent study utilizing sensors mounted on commercial aircrafts with routes between Europe and Asia indicated that in the upper troposphere/lower
stratosphere the HO₅ production from acetone, considering the revised acetone photolysis (Blitz et al., 2004), could still be almost as important as the O₅O(1D)-H₂O mechanism (Neumaier et al., 2014). Despite the wide spatial coverage (35°–60°N and 124°W to 142°E), the commercial airlines involved in the study (Neumaier et al., 2014) sampled relatively polluted regions with high acetone abundances, as indicated by the recent satellite-borne acetone retrievals using the Infrared Atmospheric Sounding Interferometer (Franco et al., 2019). The importance of acetone to HO₅ production on the global scale, particularly in the remote atmosphere, may not be directly inferred from that study (Neumaier et al., 2014).

In this work, we revisit the global atmospheric budget of acetone using two global models, the Community Atmosphere Model with Chemistry (CAM-chem), and GEOS-Chem. The CAM-chem model uses a newly developed bottom-up oceanic emission inventory of acetone, with a recently developed online oceanic emission module (Wang, Hornbrook, et al., 2019) and a novel data-oriented machine-learning approach representing ocean biogeochemistry control over this process (Wang, Kinnison, et al., 2019). The GEOS-Chem model uses the configuration of Fischer et al. (2012) for oceanic acetone emissions. Modeling results are evaluated using observations obtained from a recent global-scale, multiseasonal airborne campaign, the National Aeronautics and Space Administration (NASA) Atmospheric汤ography Mission (ATom). The acetone contribution to HO₅ production is investigated on a global scale.

2. Methods
2.1. CAM-Chem Model

CAM-chem (Lamarque et al., 2012; Tilmes et al., 2015) is the atmospheric component of a widely used chemistry-climate model, the National Center for Atmospheric Research (NCAR) Community Earth System Model (CESM). In this work, CAM-chem (from CESM2.1.1) is nudged (Tilmes et al., 2015) to the NASA Modern-Era Retrospective analysis for Research and Applications, Version 2 (MERRA-2) meteorology fields with a horizontal resolution of 0.9° latitude × 1.25° longitude and 56 vertical levels (up to 3 hPa). The chemistry scheme is based on that in CESM2.1.1, which includes a detailed representation of tropospheric and stratospheric chemistry (Emmons et al., 2020) and an updated monoterpene and isoprene mechanism (Schwantes et al., 2020). This chemical mechanism includes acetone production from propane, BIGALK (including butanes and larger alkanes), BIGENE (butenes and larger alkenes), α-pinene, β-pinene, and myrcene, as well as acetone removal by OH oxidation and photolysis. We also tested acetone oxidation by chlorine radical (Burkholder et al., 2015). The temperature and pressure-dependent photolysis of acetone (Blitz et al., 2004) is implemented in CAM-chem. In this work, CAM-chem includes anthropogenic emissions from the Coupled Model Intercomparison Project Phase 6 (CMIP6) inventory (Hoesly et al., 2018), biomass burning emissions from the Fire INventory from NCAR (FINN version 1.5; Wiedinmyer et al., 2011), and biogenic VOC (BVOC) emissions calculated online using the Model of Emissions of Gases and Aerosols from Nature (MEGAN v2.1; Guenther et al., 2012). Dry deposition velocities in CAM-chem are calculated online depending on surface types (more details are given in section 6). In this work, CAM-chem is spun-up for 3 years to remove the effects of initial conditions, and we use simulations from June 2016 to May 2017 for analysis.

The oceanic fluxes of acetone are calculated using the Online Air-Sea Interface for Soluble Species (OASISS) developed for CESM (Wang, Hornbrook, et al., 2019), based on the widely used two-layer model framework (Johnson, 2010; Liss & Slater, 1974). In brief, the air-sea exchange is described by the airside and waterside transfer velocities (k_air and k_water). The k_air term is based on the National Oceanic and Atmospheric Administration (NOAA) COARE model (Jeffery et al., 2010), with the addition of the still air diffusive flux adjustment (Mackay & Yeun, 1983). The k_water term is based on Nightingale et al. (2000). This OASISS model framework calculates the bidirectional fluxes of trace gases considering the local physical state of the ocean (sea surface temperature, salinity, and waves/bubbles) and the atmosphere (temperature, pressure, and wind); the oceanic emissions are fully coupled with the atmospheric chemistry and dynamics.

The surface seawater concentrations of acetone (monthly climatology with a horizontal resolution of 0.9° latitude × 1.25° longitude; used to drive the air-sea exchange) are predicted by an observationally trained machine-learning algorithm, trained by in situ surface seawater observations of acetone around the globe. The machine-learning approach is described elsewhere (Wang, Kinnison, et al., 2019). In brief, a random
forest regression algorithm (Breiman, 2001; Pedregosa et al., 2011) is linked to the ocean model and marine ecosystem model of CESM. The random forest is trained by surface seawater observations (Table 1) and ocean physical and biogeochemical parameters (independent variables), including sea surface temperature, salinity, photosynthetically active radiation, nutrients (phosphate, nitrate, iron), oxygen, and chlorophyll contents in different phytoplankton groups. These ocean physical and biogeochemical parameters are discussed in Wang, Kinnison, et al. (2019) and references therein. The combination of these independent variables represents the ocean biogeochemical control. To reduce the chance of overtraining, a fraction of the entire data set (85%) is randomly sampled as the training data set, and the remaining (15%) is used as testing data set. Due to the sparse availability of seawater observations, the machine-learning algorithm produces a monthly climatology. We show in section 3 that this observationally trained machine-learning framework captures the seasonal variations and large-scale features revealed in the available seawater observations. The combination of the physics-based OASISS and the observationally trained machine-learning algorithm represent both the physical forcing and the biogeochemistry control on the air-sea exchange process.

2.2. GEOS-Chem Model

GEOS-Chem is a widely used global chemical transport model with a detailed representation of tropospheric chemistry (Fischer et al., 2012; Jacob et al., 2002; etc.). GEOS-Chem is driven by the NASA GEOS-5 assimilated meteorological data. The GEOS-Chem model used in this work (v11-02d) is run at a horizontal resolution of 2.0° latitude × 2.5° longitude and 72 vertical layers (up to 1 Pa). The chemical mechanism used in these GEOS-Chem simulations is largely consistent with the basic v11-02d mechanism, which uses a monoterpene oxidation scheme developed by Fisher et al. (2016) but also includes a recently updated isoprene oxidation mechanism (Bates & Jacob, 2019). The temperature and pressure-dependent photolysis of acetone (Blitz et al., 2004) is also implemented in this version of GEOS-Chem. Oceanic emissions of acetone are calculated in the same way as described in Fischer et al. (2012), based on a two-layer model framework (Johnson, 2010; Liss & Slater, 1974) but with a constant surface seawater concentration (15 nmol L$^{-1}$). The default GEOS-Chem configuration uses anthropogenic emissions from the Emissions Database for Global Atmospheric Research (EDGAR) version 4.2, with emissions in certain regions overwritten by regional emission inventories (details can be found at http://wiki.seas.harvard.edu/geos-chem/index.php/Emissions_and_Deposition_Working_Group#Recommended_Default_Emission_Inventories). In addition, GEOS-Chem uses biomass burning emissions from Global Fire Emission Database (GFED) version 4 (van der Werf et al., 2017), as well as biogenic emissions from MEGAN version 2.1 (Guenther et al., 2012). In this work, the emissions and chemistry time step was set to 20 min; the transport time step was set to 10 min. GEOS-Chem uses a fixed dry deposition velocity for acetone (0.1 cm s$^{-1}$; section 6). GEOS-Chem is spun-up for 1.5 years, and simulations from June 2016 to May 2017 are used for analysis.

2.3. The NASA ATom Campaign

In the present work, the airborne acetone measurements from the first and second deployments of the NASA ATom field campaign (ATom-1 and ATom-2) are used for model comparison. ATom-1 and ATom-2 took place July–August 2016 and January–February 2017, respectively. During ATom, the heavily instrumented NASA DC-8 aircraft transected the lengths of the Pacific and Atlantic Oceans, regularly profiling from the marine boundary layer to the upper troposphere/lower stratosphere. During ATom-1 and ATom-2, acetone and other trace gases including hydrogen cyanide (HCN), acetonitrile (CH$_3$CN), propane (C$_3$H$_8$), i-butane (i-C$_4$H$_{10}$), and n-butane (n-C$_4$H$_{10}$) were measured with the NCAR Trace Organic Gas Analyzer (TOGA) (Apel et al., 2003, 2015), which uses fast online gas chromatography and mass spectrometry to provide mixing ratios of a large suite of VOCs. The inlet design and the analytical system of TOGA used for ATom have been described in detail elsewhere (Wang, Hornbrook, et al., 2019). In addition, O$_3$ and NO were measured using the NOAA NO$_x$O$_y$ four-channel chemiluminescence (CL) instrument (Ryerson et al., 2000); OH radical mixing ratios were measured using the Pennsylvania State University Airborne Tropospheric Hydrogen Oxides Sensor (ATHOS) (Faloona et al., 2004); photolysis frequencies for relevant compounds were calculated using CDD Actinic Flux Spectroradiometers (CAPS) measurements (Shetter & Müller, 1999). Carbon monoxide (CO) was measured using a Quantum Cascade Laser Spectrometer (QCLS; McManus et al., 2005; Santoni et al., 2014) and a Picarro analyzer (modified G2401m; Crosson, 2008; Karion et al., 2013). The CO quantity (CO.X in the merge files) uses Picarro CO data to fill calibration gaps in the QCLS CO measurements, after subtracting the low-pass filtered difference between the QCLS and the Picarro measurement. Water vapor
was measured using a Diode Laser Hygrometer (DLH) (Diskin et al., 2002). All airborne measurements are from the 2 min TOGA merge products unless otherwise noted.

Table 1

<table>
<thead>
<tr>
<th>Machine-learning training data set</th>
<th>Region</th>
<th>Month(s)</th>
</tr>
</thead>
<tbody>
<tr>
<td>Yang, Blomquist, and Nightingale (2014)</td>
<td>Atlantic (50°N to 45°S)</td>
<td>October–November</td>
</tr>
<tr>
<td>Dixon et al. (2014)</td>
<td>English Channel (50°N)</td>
<td>January–November</td>
</tr>
<tr>
<td>Beale et al. (2013)</td>
<td>Atlantic (50°N to 40°S)</td>
<td>October–November</td>
</tr>
<tr>
<td>Kameyama et al. (2010)</td>
<td>Northern Pacific (49–43°N)</td>
<td>August</td>
</tr>
<tr>
<td>Hudson et al. (2007)</td>
<td>Nordic seas (79–68°N)</td>
<td>June</td>
</tr>
<tr>
<td>Marandino et al., unpublished*</td>
<td>Pacific (2–47°S)</td>
<td>January</td>
</tr>
</tbody>
</table>

*Acetone measurements during the Knorr 06 study are not published, but the Knorr 06 study is described elsewhere (Marandino et al., 2009).

3. Seawater Concentrations and Oceanic Fluxes of Acetone

Figure 1 shows the surface water acetone concentrations (monthly mean climatology) predicted using the observationally trained machine-learning algorithm, compared to ship-borne observations. As shown in Figure 1, the machine learning captures the spatial and seasonal variations revealed from the limited seawater observational data set reasonably well. The mean absolute percentage error (\( \text{MAPE} = \frac{100}{n} \sum_{i=1}^{n} \left| \frac{OBS_i - MOD_i}{OBS_i} \right| \)) is 12.6%, and the root-mean-square error (\( \text{RMSE} = \sqrt{\frac{\sum_{i=1}^{N} (MOD_i - OBS_i)^2}{N}} \)) is 4.8 nmol L\(^{-1}\). The linear correlation between the observed and predicted surface seawater acetone is shown in supporting information Figure S1 (slope = 0.89, \( r = 0.95, n = 436 \)).

The observationally trained machine-learning algorithm reveals interesting insights. First, of all the independent variables used in this work (section 2.1), photosynthetically active radiation is found to be the most important, which is consistent with previous findings that acetone is produced from photochemical disassociation of colored dissolved organic matter (CDOM) in the surface seawater (Kieber et al., 1990; Zhou & Mopper, 1997). Second, elevated acetone is often found in oligotrophic oceans, especially the Northern Atlantic subtropical gyre, while highly productive regions (e.g., Southern Ocean) are often associated with low acetone levels. This may be due to enhanced acetone consumption in the productive regions (Dixon et al., 2013, 2014) and accumulation of acetone in the ocean gyres. As shown in Figure 1, surface seawater acetone is elevated in the tropical and subtropical Pacific Ocean all year long (>15 nmol L\(^{-1}\)). The high-latitude oceans show remarkable seasonal variations: Northern Pacific and Atlantic show low surface seawater acetone in boreal winter (<7 nmol L\(^{-1}\)), which is more than doubled in boreal summer (16–20 nmol L\(^{-1}\)). The Southern Ocean shows the same seasonal trend, with elevated surface seawater concentrations in austral summer (14–18 nmol L\(^{-1}\)) and lower concentrations in austral winter (3–6 nmol L\(^{-1}\)).

Note that in general the availability of the seawater acetone observations for the training of the machine learning is very sparse, especially in the majority of the Pacific Ocean, Indian Ocean, and the high-latitude Southern Ocean. Figure 1 implies that a machine-learning approach holds promising potential in leveraging existing in situ observations. Future ship-based studies should target these regions for a better constraint on the global distribution of the surface seawater acetone; the performance of the machine-learning approach can be further improved as more observations become available.

Figure 2 shows the CAM-chem calculated oceanic acetone fluxes, using the machine learning-predicted surface seawater acetone concentrations, as well as oceanic acetone flux observations for comparison. The online air-sea exchange module is fully coupled with atmospheric dynamics and chemistry; therefore, the
calculated oceanic fluxes depend on the concentration in the surface atmosphere as well. Note that although the machine learning-predicted surface seawater acetone is a monthly climatology, the oceanic and atmospheric physical parameters that are used to determine the oceanic fluxes (such as sea surface temperature and meteorology) have interannual variations; therefore, the calculated oceanic emission fluxes can also have interannual variations (driven by the physical forcing in the ocean and the atmosphere). However, for the studied period (2013–2017), interannual variability in predicted oceanic acetone fluxes are not large on a global scale. As shown in Figure 2, the modeled and observed oceanic fluxes, although in different years, show consistent large-scale features. For instance, the Northern Pacific and Atlantic Oceans are a large net sink for acetone in boreal winter; there is a high abundance of acetone in continental outflows from the polluted regions (e.g., Asia and North America) that passes over cold water (Henry’s law constant increases with decreasing temperature). The tropical and subtropical oceans, on the other hand, are generally a net source for acetone, except for the deep tropics (10°N to 10°S) in boreal summer, where acetone from continental sources (especially from the biogenic sources in the Amazon) shows a stronger tendency to deposit into the ocean. We performed a tagged CAM-chem simulation to investigate the contribution of biogenic sources of acetone. As shown in Figure S2, biogenic acetone (including primary emission calculated by MEGAN and secondary production from terpenes) reached up to 400 ppt in the marine boundary layer over the deep tropical Pacific (10°N to 10°S) in summer and fall (June–November), nearly two thirds of the total modeled acetone. This plume of biogenic acetone largely drives the downward flux over the deep tropical Pacific (10°N to 10°S).

Spatial variations in the net fluxes in Figure 2 show similarity to those presented in Fischer et al. (2012), despite their adoption of a single surface seawater acetone concentration. Section S1 in the supporting information provides more details on the similarity to Fischer et al. (2012) and the possible reasons behind it. Note that the magnitude of the fluxes in certain regions appear to be underestimated by the CAM-chem model. For example, larger downward fluxes were reported over the Northern Pacific (Marandino et al., 2005) and Northern Atlantic (Yang, Beale, et al., 2014), while the modeled fluxes in the same seasons and regions are smaller. This may be due to interannual variability (observations and the model represent different
years), uncertainties in the air-sea exchange model (uncertainty on the order of 100%; Carpenter & Nightingale, 2015; Johnson, 2010) or an underestimation of acetone in the continental outflows from the more polluted regions (such as Asia and North America). We discuss this further in section 4 via a comparison of vertical profiles. Figure 2 also shows that there are large regions in the tropics and over the Southern Hemispheric Pacific, Indian Ocean, and high-latitude Southern Ocean where there are no comparable observations. Oceanic acetone flux measurements in these regions are highly desirable, especially flux measurements using advanced techniques such as the eddy covariance method. Overall, our approach of using an observationally trained machine-learning algorithm to represent the ocean biogeochemistry control on the air-sea exchange of trace gases is a promising alternative of the highly simplified parameterization in Earth system models.

4. Comparison to the Airborne Observations During ATom

In this section, we compare both CAM-chem (with and without oceanic acetone fluxes) and GEOS-Chem (with oceanic acetone fluxes) modeled vertical profiles of acetone in the remote troposphere in different regions over the Pacific and Atlantic to airborne observations from the ATom-1 (July–August 2016) and ATom-2 (January–February 2017) campaigns. The modeling outputs are sampled along the flight tracks unless otherwise noted. The flight tracks (over the oceans only) for ATom-1 and ATom-2 are given in Figure 3, as well as the CAM-chem and GEOS-Chem modeled monthly mean total acetone columns in August 2016 and February 2017. The large-scale features in the total acetone column are very similar between the two models. Both models predict elevated acetone column over the Amazon and Africa, due to the primary acetone emissions from the terrestrial ecosystem and the photochemical production from BVOCs. In addition, the models also predict elevated acetone columns over East Asia and North America from anthropogenic emissions of hydrocarbons (mainly alkanes). The measured and model predicted
vertical profiles are shown in Figure 4 (ATom-1) and Figure 5 (ATom-2). During ATom-1 and ATom-2, both OH radicals and the photolysis frequency of acetone are well captured by the models (Figure S3).

4.1. ATom-1: July–August 2016

During ATom-1 (July–August 2016), the acetone observations over the remote Pacific and Atlantic Oceans show remarkable spatial variation, with maximum acetone mixing ratio as high as ~2.4 ppb in the remote atmosphere. We now elaborate the observations and modeling results in different regions:

**i Over the Northern Hemispheric Pacific and Atlantic, observed acetone levels span a wide range, with peak values as high as ~2.4 ppb. This is mainly due to the elevated acetone mixing ratios in the continental outflow from Asia and North America (as indicated by CO, Figure S4). CAM-chem underestimates acetone in the Northern Hemisphere continental outflow by up to a factor of ~2 (Figures 4a and 4e). CO is reasonably well captured by CAM-chem over the Northern Hemispheric Pacific but is underestimated by up to ~1/3 over the Northern Hemispheric Atlantic (Figures S4a and S4e). With or without oceanic acetone fluxes, CAM-chem simulations do not show large discrepancies over the Northern Hemispheric Ocean during the ATom-1 period (Figure 4). GEOS-Chem acetone is in better agreement with observations in the Northern Hemisphere. CAM-chem in this work uses CMIP6 anthropogenic emissions. The CAM-chem results suggest that the CMIP6 emission inventory may underestimate the emissions of acetone and/or its precursors in the Northern Hemisphere. For instance, CAM-chem modeled propane (C₃H₈), an important precursor of acetone, is a factor of 2.6–5.0 lower than the TOGA measurements in the Northern Hemisphere. Similar model underestimation (CAM-chem) of lower tropospheric acetone was reported in the Northern Hemisphere (Emmons et al., 2020), which is...**
Figure 4. Regional vertical profiles of acetone during ATom-1. Gray points are discrete observations, and black circles and error bars show medians and quartiles of observations. Predicted median model results are shown in red (CAM-chem with no oceanic fluxes), blue (with oceanic fluxes), and green (GEOS-Chem). (a)-(h) represent different regions (Figure 3).
Figure 5. Regional vertical profiles of acetone during ATom-2. Gray points are discreet observations, and black circles and error bars show medians and quartiles of observations. Predicted median model results are shown in red (CAM-chem with no oceanic fluxes), blue (with oceanic fluxes), and green (GEOS-Chem). (a)-(h) represent different regions (Figure 3).
possibly because of underestimated or unaccounted sources of acetone precursors (such as leakage from the oil/gas industry; Jacob et al., 2002; Pétron et al., 2014). Backward trajectory analysis (https://espo.nasa.gov/atom/archive/browse/atom/DC8/BackTraj) suggests that in August 2016, the low troposphere (below 5 km) air masses probed over the Northern Hemispheric Pacific mostly originated from North America and East Asia. The anthropogenic emission inventories of acetone and its precursors show remarkable difference between GEOS-Chem and CAM-chem: GEOS-Chem anthropogenic emissions of propane (EDGAR 4.2, with emissions in certain regions overwritten by regional inventories) is nearly 2 times higher than the anthropogenic emissions of propane used in CAM-chem (CMIP6) in North America and East Asia; GEOS-Chem anthropogenic acetone emissions in North America and East Asia are approximately 11 and 4.6 times higher than CAM-chem, respectively. In light of the model underestimation in the Northern Hemisphere Pacific, a thorough evaluation of the emission inventories in the source regions (especially Asia and North America) is needed, which is beyond the scope of this work since the ATom mission is not aiming at polluted regions.

ii Over the tropical Pacific, both CAM-chem and GEOS-chem overestimate acetone in the marine boundary layer by a factor of ~2 (Figure 4b). Both models use MEGAN for emissions from the terrestrial ecosystem. As discussed in section 3, a tagged CAM-chem simulation indicated that biogenic acetone contributes substantially in the lower troposphere over the tropical Pacific (Figure S2). CAM-chem with the oceanic acetone fluxes improves acetone predictions in the marine boundary layer (Figure 4b). Therefore, it appears that MEGAN may be overestimating the terrestrial emissions of acetone and/or other precursor BVOCs in the Amazon region.

iii Over the tropical Atlantic, a biomass burning plume from Africa is observed between 2 and 4 km, with CO up to ~250 ppb (Figure S4f) and HCN and CH$_3$CN up to ~1 ppb and 0.5 ppb, respectively (HCN and CH$_3$CN measurements can be found in Wofsy et al., 2018). Neither CAM-chem nor GEOS-Chem capture the elevated CO in the biomass burning plume, yet both models can capture the observed acetone vertical profile reasonably well (within the spreads in observations, Figure 4f), implying relatively small biomass burning contribution to the atmospheric acetone or the biomass burning emissions of acetone or of its precursors might be overestimated in the models.

iv Over the Southern Hemispheric Oceans, observed acetone is lower than that in the Northern Hemisphere. GEOS-Chem tends to overestimate acetone in the Southern Hemisphere, while CAM-chem better captures the observed acetone especially in the lower troposphere over the Southern Ocean (Figures 4c, 4d, 4g, and 4h). The overestimate of acetone in GEOS-Chem over the Southern Ocean is likely due to the higher seawater acetone in GEOS-Chem (15 nmol L$^{-1}$) than that predicted by the machine-learning algorithm (<8 nmol L$^{-1}$, Figure 1). In general, the Southern Hemisphere, including the Southern Ocean, experience less anthropogenic influences compared to the Northern Hemisphere and hence are more sensitive to natural processes, especially air-sea exchange. The better performance of CAM-chem over this region indicates that the oceanic biogeochemistry control represented by the machine-learning approach provides improved predictability for the CESM model framework. Interestingly, over the Southern Ocean, observed acetone shows clear depletion (<20 ppt) above 5–6 km, while both CAM-chem and GEOS-Chem predict >300 ppt acetone in this region (Figures 4d and 4h). This will be further explored in section 5.

4.2. ATom-2: January–February 2017

During ATom-2, the observed acetone spans a relatively narrower range (up to ~1.5 ppb) than that during ATom-1. (i) Over the Northern Hemispheric Pacific and Atlantic, observed acetone shows a clear vertical gradient in the marine boundary layer, which increases with increasing altitude. This implies the Northern Hemispheric oceans are a net sink for acetone in boreal winter, consistent with the CAM-chem predicted oceanic acetone fluxes (Figure 2). (ii) Over the tropical Pacific, the ocean is a net source for acetone (Figure 2), and GEOS-Chem and CAM-chem with oceanic acetone fluxes both well capture the acetone vertical profiles. (iii) Over the tropical Atlantic, the aircraft encountered another biomass plume between 1 and 4 km, with CO up to ~200 ppb (Figure S5f) and HCN up to ~700 ppt. Both models fail to capture the observed acetone and CO in this biomass burning plume (up to 1.5 ppb; Figures 5f and S4f). (iv) In the Southern Hemisphere, observed acetone in the marine boundary layer decreases with increasing altitude, implying that the Southern Hemispheric oceans in the austral summer are a net source of acetone, consistent with the modeled oceanic acetone fluxes (Figure 2). GEOS-Chem largely captures the observed acetone vertical
profile over the Southern Hemispheric oceans (Figures 5c and 5g). CAM-chem without oceanic acetone fluxes underestimates acetone in the lower troposphere by a factor of ~3; the online oceanic acetone fluxes improve the CAM-chem predictions in the Southern Hemisphere lower troposphere but still fall short of the measurements by a factor of 2 over the Atlantic (Figures 5c and 5g). Over the Southern Ocean, CAM-chem with oceanic acetone fluxes yields overall best agreement with observations, while GEOS-Chem tends to overestimate acetone in the marine boundary layer (Figures 5d and 5h).

To sum up, we use combined observational (TOGA measurements) and modeling (CAM-chem and GEOS-Chem) analysis to investigate the role of ocean in determining atmospheric acetone abundances over the remote oceans. As discussed above, the role of ocean in the atmospheric budget of acetone shows considerable spatial variation. Globally, the online oceanic acetone fluxes improve CAM-chem acetone predictions in the marine boundary layer (<2 km), with the mean absolute percentage error reduced from 65% to 40% during ATom-1 and 55% to 30% during ATom-2, with a mean root square error of 0.14–0.39 ppb (Table S1). GEOS-Chem, using the same oceanic acetone configuration as in Fischer et al. (2012) but with updated isoprene and monoterpane chemistry, produces a mean absolute percentage error of 75% and 41% in the marine boundary layer during ATom-1 and ATom-2, respectively, with a mean root square error of 0.15–0.43 ppb (Table S1).

5. Large Discrepancy Between the Modeled and Measured Acetone Near the Tropopause Over Southern Ocean

We now come back to the unexpected model-observation discrepancy on acetone above 5–6 km over the Southern Ocean during ATom-1 (Figures 4d and 4h). During these two flights (12 and 15 August 2016), the observed median acetone ranges from 80 to 125 ppt below 5 km, and the CAM-chem modeled acetone ranges from 80 to 185 ppt, while GEOS-Chem predicts >200 ppt throughout (Figures 4d and 4h). Above 5 km, the observed median acetone decreases strongly with increasing altitude, from 125 ppt at 5 km to <30 ppt above 8 km. Both models fail to capture the observed altitude dependency above 5 km, consistently predicting >200 ppt acetone in this region. The model-observations discrepancy is further shown in the interpolated curtains in Figure 6 (and Figure S6): CAM-chem predicts >180 ppt acetone that is widespread between 40°S and 70°S, extending from lower troposphere (2 km) all the way to the lowermost stratosphere (13 km); yet observed acetone in this region remains below 200 ppt. Such acetone overestimation is not seen in the Northern Hemisphere, possibly due to the higher acetone abundance in the Northern Hemisphere. Four reasons may be responsible for the widespread model overestimation of acetone in the Southern Hemisphere in cold seasons: (i) uncertainties in the acetone chemistry (chemical production/removal); (ii) uncertainties in large-scale transport; (iii) overestimation in the source region; and (iv) missing removal processes. We can rule out some of these reasons which are discussed as follows.

5.1. Uncertainties in the Acetone Chemistry

Localized acetone production in the model is unlikely to be significant, since the atmosphere over the Southern Ocean remains mostly dark during austral winter, and major acetone precursors remain low in this region. For instance, during ATom-1 above 6 km over the Southern Ocean, the TOGA measured propane, i-butane, and n-butane are $31 \pm 12$, $2.8 \pm 1.2$, and $3.8 \pm 1.6$ ppt, respectively (limits of detection [LOD]: 20, 0.5, and 1.0 ppt for propane, i-butane, and n-butane, respectively. Data below LOD are replaced by two thirds of the LOD). α-Pinene remains below its LOD (0.1 ppt) at all times. Photochemistry in the atmosphere is a major sink of acetone, including OH oxidation, photolysis, and reaction with chlorine atoms. The OH and photolysis sinks of acetone are both well captured by the models (Figures S3, S7, and S8); therefore, OH oxidation and photolysis appear to be unlikely culprits for the missing acetone depletion in the models. Chlorine oxidation appears to be plausible, as indicated by the unique chemical signature in Figure 7 (TOGA observed i-butane/n-butane ratio vs. i-butane/propane ratio). Propane and butanes are often coemitted from anthropogenic sources (e.g., oil and gas industry, fuel evaporation, Jacob et al., 2002 and references therein). i-Butane and n-butane have similar OH reactivities (within 9% between 200 and 300 K), yet the chlorine reactivity of n-butane is 44–46% higher than that of i-butane between 200 and 300 K (Burkholder et al., 2015). On the contrary, i-butane and propane have similar chlorine atom reactivities (<3% between 200 and 300 K) but very different OH reactivities (a factor of 2–3.5 between 200 and 300 K) (Burkholder et al., 2015). Therefore, if the data points in Figure 7 are spread out horizontally on this
ratio-ratio plot, it implies OH oxidation dominates the aging process of the air mass. However, where the data points are spread vertically, deviating from the horizontal cluster implies a possible influence of chlorine chemistry. Figure 7 implies that, during these two ATom-1 flights over the Southern Ocean, the observed acetone depletion may be associated with chlorine chemistry signature, or any process that is co-located with chlorine chemistry. Both CAM-chem and GEOS-Chem include the chlorine oxidation of acetone (Burkholder et al., 2015), which cannot explain the acetone depletion observed in the lower stratosphere over the Southern Ocean during ATom-1 due to its slow rate and the low chlorine radical concentrations in the Southern Hemisphere in this season. The tropospheric and stratospheric halogen chemistry in CAM-chem has been evaluated extensively in the literature (Fernandez et al., 2014; Navarro et al., 2015, 2017; Saiz-Lopez & Fernandez, 2016; Saiz-Lopez & von Glasow, 2012; Wang, Kinnison, et al., 2019), as has the tropospheric halogen chemistry in GEOS-Chem (Parrella et al., 2012; Schmidt et al., 2016; Sherwen et al., 2016; Wang, Jacob, et al., 2019). A CAM-chem sensitivity test with 100 times faster Cl + acetone rate coefficient and another 4 year spin-up still cannot explain the observed acetone depletion. Therefore, the chlorine signature as revealed in Figure 7 implies the acetone depletion may be partially related to stratospheric processes (where chlorine chemistry is active), but is unlikely to be due to direct chlorine oxidation of acetone. To sum up, the combined observational and modeling analysis implies that the persistent model overestimation of acetone in this region is unlikely to be due to uncertainties in modeled photochemistry.
5.2. Uncertainties in the Large-Scale Transport

During austral winters, the southern high-latitude air mass in the upper troposphere is influenced by the interplay of two large-scale processes. One is the downward transport from the stratosphere driven by the season’s strong polar vortex descent, and the other is the isentropic mixing from low latitudes driven by the poleward Rossby wave breaking. To examine the model representation of transport and large-scale circulation, two specific diagnostics were performed, and the results are shown in Figure 8 (also Figures S7 and S8) and an animation in the supporting material (Movie S1).

Figure 8 shows a diagnostic of troposphere-stratosphere transition using tracer-tracer relationship. The \( \mathrm{O}_3 \)-CO and \( \mathrm{O}_3 \)-H\(_2\)O relationships are most commonly used pairs (Hoor et al., 2002; Pan et al., 2004, 2007). The stratosphere-influenced air (i.e., a mixture of tropospheric air and subsiding stratospheric air) sampled by the aircraft is characterized by elevated \( \mathrm{O}_3 \) (~100 ppb) and low water vapor (H\(_2\)O < 40 ppm). Figure 8 (Figures S7 and S8) shows that CAM-chem reproduced the observed relationships between the stratospheric tracer (\( \mathrm{O}_3 \)) and tropospheric tracers (H\(_2\)O and CO) which demonstrate the overall good model representation of the large-scale circulation. The stratosphere-influenced air (200–280 hPa) probed during these two flights contains 45–65 ppb CO, consistent with the CO measurements reported in the lowermost stratosphere but certainly toward the upper end (de Reus et al., 2003). The tropopause locations and the structures are also reasonably well captured by the CAM-chem (Figure S7 and S8). The modeled and observed acetone versus \( \mathrm{O}_3 \)/CO relationships, on the other hand, are drastically different in the tropopause region (Figure 8). CO, \( \mathrm{O}_3 \), and H\(_2\)O in this region are reasonably well captured by the models (Figure 8, Figures S4d, S4h, S7, and S8), indicating that the large-scale circulation (Brewer-Dobson circulation) are well represented by the model, and hence the model overestimate of acetone is unlikely due to uncertainties in the model transport or large-scale descent.

Additional diagnostic of large-scale transport from isentropic mixing is shown by CAM-chem models’ \( \mathrm{O}_3 \) and acetone evolution on the 310 and 320 K potential temperature surface for the observation period. As indicated in Figure 6, these isentropes represent the lower troposphere in the subtropics and the tropopause region in the high latitudes. The two animations, using \( \mathrm{O}_3 \) as a transport tracer, highlight the active isentropic transport contribution to the tropopause level acetone in the Southern high latitude from the subtropical lower troposphere. This transport behavior is consistent with the dynamical analysis of the region where active poleward Rossby wave breaking during austral winter was identified (Peters & Waugh, 2003). These wave-breaking activities are responsible for trace gas transport to the tropopause region as indicated by the low PV streamers (Movies S1 and S2).

The two diagnostics remove the model deficiency of representing large-scale transport as the cause for acetone high bias in the southern high-latitude UTLS (upper troposphere/lower stratosphere). The result, however, pointed out that the isentropic mixing is responsible for modeled high acetone in the tropopause region, indicating either the overrepresentation of subtropical acetone source or the missing removal processes in the cold polar tropopause region are the directions of further investigation.

Acetone measurements in the upper troposphere/lower stratosphere have been reported in the literature (Arnold et al., 1986; Singh et al., 2000); generally acetone in the stratosphere decreases strongly with increasing ozone (Scheeren et al., 2003; Singh, 2004). de Reus et al. (2003) examined the relationship between acetone and CO for a number of air mass types in a number of airborne campaigns and found systematically lower acetone/CO ratio in the lowermost stratosphere (compared to tropospheric air masses), due to “the increased photolytic loss of acetone and the decreasing OH concentration with altitude.” However, the acetone removal processes near the tropopause has not been quantitatively examined in previous studies, especially in the high-latitude regions during winter. The large discrepancy of modeled and observed acetone in the upper troposphere may imply a currently missing acetone sink in the lower stratosphere, although we
cannot rule out the possibility that acetone may be overestimated in the source regions, which will be elaborated in the following section.

5.3. Acetone Overestimation in the Source Regions

As shown in Figures 6 and S6 and Movie S1, the rapid isentropic transport connects the tropical/subtropical regions with the high-latitude regions. Therefore, we identify the tropical/subtropical (20°S to equator) midtroposphere/upper troposphere (5–12 km) as the potential source regions of the upper troposphere over the Southern Ocean, where large model-measurement disagreement is observed. Although the large-scale circulation and transport is well capture by the models (section 5.2), any overestimation of acetone in the source region may lead to an overestimation of acetone in the upper troposphere/lower stratosphere in the high-latitude regions. As shown in Figure 6, the observed acetone level in the tropical/subtropical midtroposphere/upper troposphere (150–350 ppt) over the Pacific Basin are only slightly overestimated by CAM-chem (200–450 ppt); however, Figure S6 shows that CAM-chem overestimates acetone in the subtropical midtroposphere over the Atlantic Basin (500–900 ppt) relative to observations (400–600 ppt). It appears that acetone is indeed overestimated in the source regions. The driver of the acetone overestimation in the source regions, and the degree to which such overestimation in the lower latitude regions affect the acetone transport and accumulation in the high-latitude regions, will be investigated in a separate work.

Figure 8. Tracer-tracer relationships over the Southern Ocean Pacific during 12 August 2016: (a) O₃ versus CO; (b) O₃ versus H₂O; (c) O₃ versus acetone; and (d) acetone versus CO. Airborne observations are shown as filled circles, color coded by the pressure altitude (tropopause: ~330 hPa). Flight track is given as a solid line in the embedded map in Panel (b). CAM-chem modeling results are shown as black dots. Note that CAM-chem modeling results in this figure are sampled in the dashed region in the embedded map, not along the flight track.
To sum up, the large discrepancy of modeled and observed acetone in the upper troposphere/lower stratosphere over the Southern Ocean is unlikely due to known photochemistry or large-scale transport but likely due to currently missing removal processes or acetone overestimation in the tropical/subtropical troposphere. Notably, this is not an episodic event. We have examined WACCM model (similar to CAM‐chem but extends in altitude to the lower thermosphere) outputs for previous years, and the meridional curtains always show similar acetone “hot spots” in this region during austral winter. Similar results are consistently seen in GEOS‐Chem, and the atmospheric component of the Geophysical Fluid Dynamics Laboratory coupled model (GFDL AM4.1, V. Naik and J. He, NOAA/GFDL, personal communication, December 2019), despite the very different treatment of chemistry and physics among these models. The impacts of such model high bias remain unclear and will be further investigated in a separated manuscript.

However, the high model bias is unlikely to affect the HO\textsubscript{x} budget over the Southern Ocean, since acetone contributes to HO\textsubscript{x} production mainly in the tropics and subtropics (section 7).

### 6. Global Budget of Acetone

The global atmospheric budget of acetone calculated using both CAM-chem and GEOS-Chem is shown in Figure 9. The global burden of acetone calculated by CAM-chem is 3.5 Tg, very similar to that calculated using GEOS-Chem (3.8 Tg). The global atmospheric lifetime of acetone against all sinks is estimated to be 11.0 and 12.7 days by CAM-chem and GEOS-Chem, respectively. In this work, we calculate the global atmospheric lifetime of acetone against all its chemical and physical sinks; for the ocean uptake, the gross uptake term is included, consistent with previous studies (Brewer et al., 2017; Fischer et al., 2012). Globally, both CAM-chem and GEOS-Chem predict that biogenic sources are the dominant acetone source, with primary terrestrial emission a factor of ~4–5 higher than the secondary production from other BVOCs. Anthropogenic emissions contribute 38% (CAM-chem) and 37% (GEOS-Chem) of the total acetone source globally, which is dominated by secondary production from alkanes. Biomass burning contributes a small
fraction of atmospheric acetone, yet in certain regions/seasons the biomass burning contribution can be substantial (Figure 4 and Figure 5). Photochemical loss is the major acetone sink, and both CAM-chem and GEOS-Chem predict that photolysis contributes to more acetone loss than OH oxidation. Recent GEOS-Chem studies report an opposite trend: for instance, Brewer et al. (2017) and Fischer et al. (2012) both suggest OH oxidation is the major sink of acetone. Another recent study (Khan et al., 2015) reports that OH oxidation and photolysis are equally important for acetone removal. The pressure/temperature-dependent acetone photolysis (Blitz et al., 2004) was not available in early studies (e.g., Jacob et al., 2002), leading to potential overestimate of photolysis contribution. Terrestrial uptake over land shows a factor of ~2 discrepancy between the two models. In CAM-chem, both dry deposition and wet deposition are considered for acetone, with dry deposition contributing to over 98% of the total deposition. GEOS-Chem does not have wet deposition for acetone. CAM-chem online calculates dry deposition velocity ($V_d$) of acetone for different surface types. For example, the mean $V_d$ is 0.008 cm s$^{-1}$ for bare ground/nonvegetation surface (leaf area index, LAI = 0); 0.18 cm s$^{-1}$ for LAI between 0 and 1; 0.27 cm s$^{-1}$ for LAI larger than 2. Using the surface-dependent dry deposition velocity of acetone, CAM-chem predicts that the South America (mostly Amazon region) alone accounts for 36% of the global acetone dry deposition. GEOS-Chem uses a fixed dry deposition velocity of 0.1 cm s$^{-1}$ for all surface types (Brewer et al., 2017; Fischer et al., 2012). The limited measurements of $V_d$ for acetone range from 0.02 to 0.46 cm s$^{-1}$ (Karl et al., 2004, 2005). Given the large discrepancy between CAM-chem and GEOS-Chem, and the potential importance of dry deposition especially over the forest regions, the dry deposition of acetone warrants further exploration.

On the global scale, the large gross upward (emitting) oceanic fluxes of acetone are more or less balanced by the gross downward (uptaking) oceanic fluxes, implying that the atmosphere–ocean exchange of acetone is close to equilibrium. Using the machine learning-predicted surface seawater acetone concentrations and the updated BVOC chemistry, CAM-chem predicts that the global ocean is a small net sink of acetone (~8.1 Tg a$^{-1}$), which is the difference between the two large gross terms (upward and downward: 33.4 and 41.5 Tg a$^{-1}$).
−1, respectively, Figure 9). GEOS‐Chem, with a fixed surface seawater concentration of acetone (15 nM), predicted global gross upward and downward oceanic fluxes are on the same order of magnitude (45.7 and 44.4 Tg a⁻¹, Figure 9), with a very small net emitting tendency from the ocean (+1.3 Tg a⁻¹). We consider the CAM‐chem prediction of the ocean net exchange (−8.1 Tg a⁻¹) to be more robust, since the machine‐learning approach linked to CAM‐chem takes advantages of more recent seawater acetone measurements, leading to an improved agreement with observations in the global marine boundary layer (Table S1). To summarize, both models support the view that the ocean strongly regulates atmospheric acetone abundances, consistent with previous studies (Brewer et al., 2017; Fischer et al., 2012; Jacob et al., 2002).

7. HOₓ Production From Acetone

In this work, acetone contribution to total HOₓ production is calculated by counting the difference between simulated HOₓ in a base case scenario (as shown in Figures 3−5) and the simulated HOₓ when acetone is allowed to undergo chemical removal without HOₓ radical production. This approach accounts for the total HOₓ production from acetone oxidation products (mainly CH₃O₂ and CH₃CO₃), which depends on NOₓ, several photolysis channels and cycling pathways. As shown in Figure 10, the acetone contribution to total HOₓ production peaks in the tropical and subtropical upper troposphere and the lowermost stratosphere, contributing up to 24% in CAM‐chem and 27% in GEOS‐Chem (both meridional means). The acetone contribution to HOₓ production peaks in the tropical and subtropical upper troposphere and lowermost stratosphere, where the Oₓ−O(³D)−H₂O mechanism is less efficient due to the dehydration associated with updrafts. The acetone contribution to HOₓ production decreases substantially in the middle and high‐latitude regions. GEOS‐Chem predicts higher acetone‐induced HOₓ production in the Northern Hemisphere in summer (Figure 10), due to the higher acetone abundance predicted by GEOS‐Chem over
this region (Figure 3). The sharp latitude gradient is also visible in Figure 11, which shows the seasonal means of the acetone-induced HO$_x$ production at the tropopause. Figure 10 shows that near the tropopause over the Western tropical Pacific, acetone contributes to ~35–40% HO$_x$ production year long. Between September and February, the contribution of acetone to HO$_x$ production near the tropopause reaches ~40% over Africa. Between March and August however, the contribution of acetone to HO$_x$ production reaches ~36% over South Asia/Indian Ocean, Northern Africa and Western Asia. Note that the acetone contribution to the total HO$_x$ production estimated in this work may be potentially biased, since the possible contributions of higher aldehydes (Müller & Brasseur, 1999; Wennberg et al., 1998) may not be captured by the models in the remote atmosphere. Recently, Neumaier et al. (2014) estimated the acetone contribution to HO$_x$ production in the upper troposphere and lower stratosphere using airborne observations, and reported that acetone-induced HO$_x$ production can be 61–95% of the HO$_x$ production from ozone (note that other HO$_x$ sources, such as formaldehyde, are not considered in Neumaier et al.). Assuming 4% HO$_x$ production from formaldehyde (Nicely et al., 2016), the results in Neumaier et al. (2014) imply an acetone-induced HO$_x$ production of up to 36–47%, generally consistent with the estimate in this work (Figures 10 and 11).

8. Conclusions

We report the global atmospheric budget of acetone calculated using two global models, CAM-chem and GEOS-Chem, and the models are compared to observations from a recent global-scale, multiseasonal airborne campaign, NASA ATom. Both CAM-chem and GEOS-Chem include updated BVOC chemical mechanisms. CAM-chem is equipped with an online air-sea exchange model framework to calculate the bidirectional oceanic fluxes of acetone. The surface seawater concentrations of acetone are predicted using a data-oriented machine-learning algorithm. We show that the ocean strongly regulates the acetone budget over the remote oceans. The machine-learning algorithm captures the large-scale features revealed in the limited surface seawater observations. Generally, there is very sparse availability of surface seawater concentration and flux measurements for acetone, especially in the Pacific Ocean, Indian Ocean, and Southern Ocean, and future ship-based studies should target these regions. We show that together the online air-sea exchange framework and the observationally trained machine-learning approach show promising potential. The performance can be further improved when more measurements become available. This model framework can be expanded to the marine emissions for other climate-relevant gases, such as dimethyl sulfide (DMS).

We show that the newly developed marine acetone inventory, with the ocean biogeochemistry control represented by a machine-learning approach, results in improved agreement with observations in the remote marine boundary layer, such as in the Southern Hemisphere and the Southern Ocean. Both CAM-chem and GEOS-Chem capture measured acetone vertical distributions in the remote troposphere reasonably well, although the drivers behind some of the observed spatial and seasonal variability remain unclear. Despite different model configurations, the global budget, total atmospheric burden, and the atmospheric lifetime of acetone derived from the two models are quite similar. The combined observational and modeling analysis suggest that (i) the CMIP6 anthropogenic emission inventory may underestimate acetone and/or its precursors in the North Hemisphere. (ii) MEGAN may overestimate acetone and/or its precursor emissions from the terrestrial ecosystem, and/or the biogenic oxidation mechanisms may overestimate secondary acetone yields. (iii) The model overestimation of acetone in the upper troposphere-lower stratosphere in the Southern Hemisphere in winter is unlikely due to photochemistry or transport; the widespread model-measurement discrepancy may imply an unknown acetone removal mechanism or the acetone overestimation in the subtropical midtroposphere.

Lastly, we show that acetone may contribute up to 30–40% of HO$_x$ production near the tropopause, especially in the tropical regions, where troposphere-stratosphere exchange is active (e.g., strong convective transport or Asian Monsoon). The acetone contribution to HO$_x$ production derived in this work is comparable to previous work (Neumaier et al., 2014). The tropical oceans are also a net source of acetone, leading to stronger ocean biogeochemistry control on the oxidative capacity in the tropical upper troposphere and the lowermost stratosphere.
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