

1 Quantifying the Impacts of Marine Aerosols over the Southeast 2 Atlantic Ocean using a chemical transport model: Implications 3 for aerosol-cloud interactions

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8 **Abstract.** The southeast Atlantic region, characterized by persistent stratocumulus clouds, has one of the highest
9 uncertainties in aerosol radiative forcing and significant variability across climate models. In this study, we analyze
10 the seasonally varying role of marine aerosol sources and identify key uncertainties in aerosol composition at cloud-
11 relevant altitudes over the southeast Atlantic using the GEOS-Chem chemical transport model. We evaluate simulated
12 aerosol optical depth (AOD) and speciated aerosol concentrations against those collected from ground observations
13 and aircraft campaigns such as LASIC, ORACLES, and CLARIFY, conducted during 2017. The model consistently
14 underestimates AOD relative to AERONET, particularly at remote locations like Ascension Island. However, when
15 compared with aerosol mass concentrations from aircraft campaigns during the biomass burning period, it performs
16 adequately at cloud-relevant altitudes, with a normalized mean bias (NMB) between -3.5% (CLARIFY) and -7.5%
17 (ORACLES). At these altitudes, [in the model](#) organic aerosols (63%) dominate during the biomass burning period,
18 while sulfate (41%) prevails during austral summer, when dimethylsulfide (DMS) emissions peak in the model. Our
19 findings indicate that marine sulfate can account for up to 69% of total sulfate during high DMS period. Sensitivity
20 analyses indicate that refining DMS emissions and oxidation chemistry may increase sulfate aerosol produced from
21 marine sources, highlighting [that there remains large uncertainty in the role of DMS emissions in the marine boundary](#)
22 [layer](#)~~their overall importance~~. Additionally, we find marine primary organic aerosol emissions may substantially
23 increase total organic aerosol concentrations, particularly during austral summer. ~~This study underscores the~~
24 ~~imperative need to refine marine emissions and their chemical transformations to better predict aerosol-cloud~~
25 ~~interactions and reduce uncertainties in aerosol radiative forcing over the southeast Atlantic.~~ [This study underscores](#)
26 [the imperative need to refine marine emissions and their chemical transformations, as aerosols from marine sources](#)
27 [are a major component of total aerosols at cloud-relevant altitudes and may impact uncertainties in aerosol radiative](#)
28 [forcing over the southeast Atlantic.](#)

29 1 Introduction

30 Marine aerosols are a primary contributor to natural atmospheric aerosols, and consequently influence the Earth's
31 radiative balance (Spracklen et al., 2008; Vignati et al., 2001). Aerosols in the marine boundary layer have significant
32 impact on the properties of low-altitude marine clouds, particularly their ability to reflect solar radiation and cool the
33 climate (Seinfeld and Pandis, 2016; Wood, 2012; Chen et al., 2014; Quinn et al., 2017). The southeast Atlantic (SEA)

34 is marked by a persistent deck of low-level stratocumulus (Sc) clouds. However, ~~aerosol radiative forcing in the this~~
35 region exhibits highest uncertainty ~~in aerosol radiative forcings in the AeroCom intercomparison across CMIP5~~
36 ~~general circulation models (GCMs) and chemical transport models and one of the largest intermodel spread, primarily~~
37 ~~due to the differences in modeled cloud fraction~~ (Stier et al., 2013; Zuidema et al., 2016), as well as aerosol and cloud
38 ~~properties~~ (Doherty et al., 2022). ~~Thise~~se uncertaintyies is primarily driven by challenges in accurately representing
39 ~~cloud fraction, aerosol-cloud properties, and vertical structure, both in the presence and absence of smoke~~ (Stier et al.,
40 ~~2013; Doherty et al., 2022).~~ are further compounded by poorly constrained optical properties of the absorbing biomass
41 ~~burning aerosols, the vertical distribution of aerosols relative to these clouds and the interaction of aerosols with~~
42 ~~marine boundary layer clouds~~ (Zuidema et al., 2016), and limited observations of aerosols and their precursors in
43 ~~remote marine environments~~ (Croft et al., 2021). In this study, we investigate the role of marine aerosols and sources
44 of uncertainty affecting aerosol composition ~~within the boundary layer, particularly~~ in this critical region of aerosol-
45 cloud interactions over the SEA.

46

47 The SEA region encompasses the Benguela upwelling system (BUS), renowned for its high primary production of
48 marine phytoplankton and fish populations (Shannon and Nelson, 1996; Jarre et al., 2015). This elevated
49 phytoplankton activity serves as the main natural source of the volatile organic compound dimethylsulfide (DMS),
50 thereby influencing the global tropospheric sulfur budget (Andreae, 1990; Bates et al., 1992). Once released into the
51 atmosphere through air-sea exchange, DMS undergoes complex chemical transformations. In the gas phase, it is
52 oxidized to form H₂SO₄ and methanesulfonic acid (MSA), which has implications for new particle formation (Chen
53 et al., 2015); while in the aqueous phase, it leads to the production of MSA and sulfate aerosols, impacting cloud
54 microphysical properties (Kaufman and Tanré, 1994). ~~Although DMS is a critical source of natural aerosols,~~
55 ~~contributing over 50% of natural gas-phase sulfur emissions~~ (Chin et al., 1996; Kilgour et al., 2021) ~~Despite its~~
56 ~~significance~~, the exact mechanisms of DMS oxidation and subsequent formation of sulfate and MSA aerosol remain
57 inadequately understood (Ravishankara et al., 1997; Barnes et al., 2006; Hoffmann et al., 2016), ~~leading to largest~~
58 ~~uncertainty of aerosol radiative forcing within climate models~~. ~~This gap in understanding contributes to substantial~~
59 ~~uncertainty in aerosol radiative forcing, which is highly sensitive to uncertainties in natural aerosols~~ (Carslaw et al.,
60 2013; Fung et al., 2022). Additionally, marine aerosols comprise primary aerosols such as sea spray aerosols, which
61 consist of salts, sulfate, and organic matter, released into the atmosphere primarily by the bubble-bursting process
62 (O'Dowd and De Leeuw, 2007; Russell et al., 2010; Prather et al., 2013; Brooks and Thornton, 2018; ~~Russell et al.,~~
63 ~~2023~~). Investigating the uncertainties related to marine emissions and chemistry are crucial to refine our understanding
64 of the impacts of marine aerosols on climate.

65 The SEA lies at the confluence of not only marine aerosols, but other natural and anthropogenic aerosols from local
66 and distant origin (Andreae et al., 1995; Swap et al., 1996; Formenti et al., 1999; Swap et al., 2003; Tournadre, 2014).
67 During the austral spring (August to October), seasonal fires in the neighboring southern African region contribute
68 nearly one-third of global total biomass burning emissions (van der Werf et al., 2010). This seasonal influx of biomass

69 burning aerosols aloft interacts with the underlying Sc deck, introducing considerable variability into aerosol forcing
70 assessments in the SEA region (Lindesay et al., 1996; Swap et al., 2003). To address these uncertainties, several
71 international field campaigns were conducted between 1992 and 2018 during the peak biomass burning season (Swap
72 et al., 2003; Formenti et al., 2019; Haywood et al., 2021; Redemann et al., 2021). Despite the region being a prolific
73 source of marine aerosols throughout the year, the potential impact of aerosols on regional climate dynamics through
74 interactions with the persistent low-level marine clouds outside of the biomass burning season has been largely
75 overlooked.

76 Here, we use the GEOS-Chem global chemical transport model to analyze high-resolution, seasonally varying aerosol
77 composition at the altitudes of persistent stratocumulus clouds over the SEA. We specifically focus on the role of
78 marine aerosols, analyzing their contributions to sulfate and organic aerosol concentrations. We evaluate simulated
79 aerosol optical depth (AOD) and speciated aerosol concentrations against observational data from the Aerosol Robotic
80 Network (AERONET) and the Layered Atlantic Smoke Interactions with Clouds (LASIC; Zuidema et al., 2018),
81 ObseRvations of Aerosols above CLouds and their intERactionS (ORACLES; Redemann et al., 2021), and CLOUD-
82 Aerosol-Radiation Interaction and Forcing (CLARIFY; Haywood et al., 2021) field campaigns during the year 2017.
83 We assess the sensitivity of our results to uncertainty in DMS oxidation mechanisms and emissions of DMS, SO₂, and
84 marine primary organics. Our findings aim to enhance understanding of the seasonally varying role of marine aerosols
85 in aerosol-cloud interactions in the SEA by a comprehensive evaluation of aerosol composition at cloud altitudes.

86 2 Methodology

87 2.1 Model Description

88 Here, we use the GEOS-Chem 3D atmospheric chemical transport model version 13.3.3 with detailed gas- and aerosol-
89 phase tropospheric chemistry (<https://zenodo.org/records/5748260>). The model is driven by meteorology from the
90 Modern-Era Retrospective Analysis for Research and Applications, Version-2 (MERRA2) reanalysis, from the NASA
91 Global Modeling Assimilation Office (GMAO) (Gelaro et al., 2017). We perform nested grid simulations over the
92 southwestern coast of Africa (40°W-20°E, 0-40°S) with a horizontal resolution of 0.5° by 0.625° and extending over
93 47 vertical layers from the surface to 0.01hPa. A chemical time step of 20 minutes and transport time step of 10
94 minutes is applied, as recommended by Philip et al. (2016). Prior to the target year, 2017, we conduct a 6-month spin-
95 up simulation. Boundary conditions are obtained from global simulations performed at 4° latitude × 5° longitude
96 horizontal resolution for the same year after a 6-month initialization.

97 In GEOS-Chem, carbonaceous aerosol includes organic aerosols (OA) and black carbon (BC). [BC follows Park et al. \(2003\) and Wang et al. \(2014\)](#). Organic aerosol [is simulated using](#) the “simple” scheme which treats primary organic
98 aerosol (POA) as non-volatile and includes irreversible direct yield of SOA from precursors (Pai et al., 2020). [The BC simulation follows the methodologies of Park et al. \(2003\) and Wang et al. \(2014\)](#). Sulfate (Alexander et al., 2009),
99 nitrate (Jaeglé et al., 2018), and ammonium (Fountoukis and Nenes, 2007) thermodynamic partitioning is estimated
100 using the ISORROPIA II thermodynamic model (Fountoukis and Nenes, 2007). Monthly anthropogenic emissions
101
102

103 follow the Community Emissions Data System (CEDSV2) inventory (Hoesly et al., 2018). Biomass burning emissions
104 are calculated using the Global Fire Emissions Database (GFED4.1s) at $0.25^\circ \times 0.25^\circ$ spatial resolution, with fractional
105 daily and 3-hourly scaling factors applied to the cumulative monthly data (van der Werf et al., 2017). DMS emissions
106 in the standard model use the Lana et al. (2011) climatology, which compiles DMS concentrations using data from
107 the Global Surface Seawater DMS Database (<http://saga.pmel.noaa.gov/dms/>) collected from 1972 to 2009,
108 incorporated with additional observations from the South Pacific (Lee et al., 2010). The standard DMS oxidation
109 mechanism in the model includes only three gas-phase DMS reactions, which directly yield SO_2 and MSA according
110 to the reaction mechanism outlined by Chin et al. (1996), and incorporates updated reaction rate coefficients from
111 Burkholder et al. (2015). Sea-salt aerosol (SSA) emissions from the open ocean are both windspeed (Gong et al.,
112 2003) and sea surface temperature-dependent (Jaeglé et al., 2011). Dust emissions include natural dust (Fairlie et al.,
113 2007) and anthropogenic dust from the AFCID inventory (Philip et al., 2017).

114 In this study, we carry out multiple simulations to explore the sensitivity of marine aerosols to various emission
115 sources. To quantify the impact of marine sources on sulfate aerosols within the stratocumulus cloud layer, we perform
116 a high-resolution ($0.5^\circ \times 0.625^\circ$) marine emissions only sensitivity simulation where SO_2 and SO_4 emissions from
117 anthropogenic sources, biomass burning, volcanic activity, ships and aviation were turned off. Additionally, to
118 investigate the sensitivity of DMS emission fluxes to surface ocean DMS concentrations, we perform an additional
119 simulation with DMS concentrations from Galí et al. (2018). In this dataset, DMS concentrations are estimated through
120 a remote-sensing algorithm that integrates satellite-derived estimates of chlorophyll and light penetration, along with
121 climatological mixed layer depth (Galí et al., 2018). Furthermore, we assess the impact of adding marine POA, co-
122 emitted with sea-salt aerosols (Gantt et al., 2015), on the overall organic aerosol burden, which is not included in the
123 standard model configuration. Finally, to evaluate how uncertainty in biomass burning SO_2 emissions affects the
124 relative importance of marine emissions to sulfate aerosol, we conduct two sets of sensitivity simulations using the
125 Quick Fire Emissions Dataset (QFED) (Darmenov & da Silva, 2013; Das et al., 2017), and the Global Fire
126 Assimilation System (GFAS) (Kaiser et al., 2012; Su et al., 2023). Each of these inventories differ in data sources,
127 methodology, temporal resolution and plume injection height. These sensitivity analyses were conducted for the year
128 2017, following a six-month spin-up period. Details regarding the spatial resolution used in each sensitivity analysis
129 are provided in Table A1.

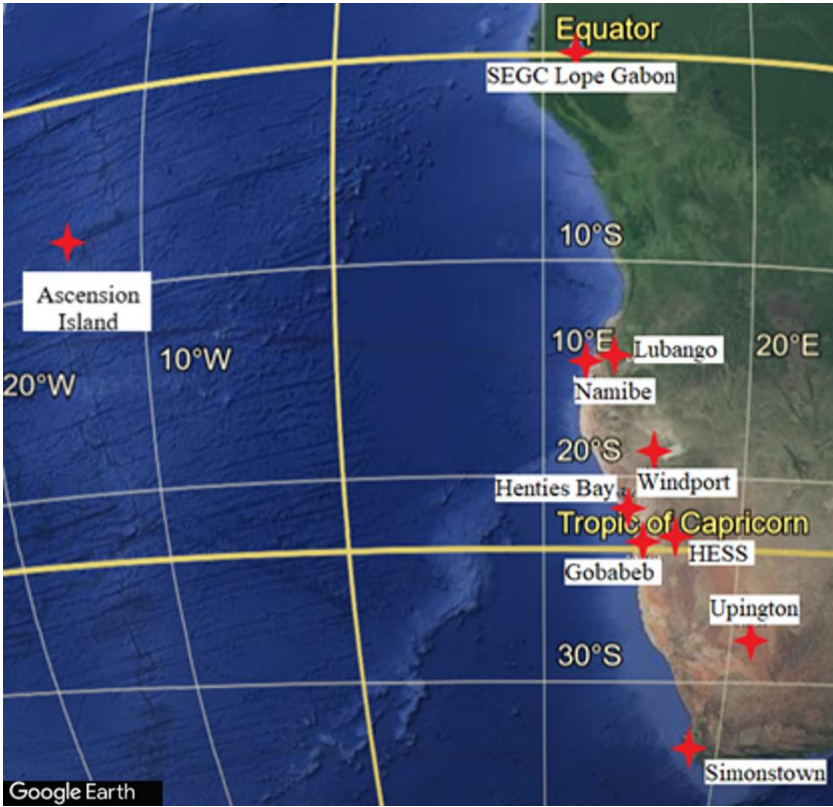
130 **2.2 Ground-based measurements**

131 We evaluate simulated aerosol optical depth (AOD) against AOD retrieved from the ground-based Aerosol Robotic
132 Network (AERONET) of sun photometers with direct sun measurements every 15 min (Holben et al., 1998). We use
133 Level 2.0 Version 3 data that have improved cloud screening algorithms (Giles et al., 2019). We strategically select
134 nine sites in the study domain along coastal and oceanic regions, as shown in Fig. 1. Site information, including the
135 coordinates, number of months with available data and the monthly average ~~daily~~ AOD for three distinct time periods,
136 is summarized in Table A2. The AERONET monthly average AOD is computed from daily averages for sites with at

137 least 3 months of observations during the model simulation period (year 2017) and months with at least 15 days of
138 measurements. These are then compared with the monthly mean AOD from the GEOS-Chem model.

139 The modeled AOD is [sampled at each AERONET site location and](#) computed at 550 nm wavelength by vertically
140 integrating scattering and absorption coefficients based on the properties of various aerosol components, such as size
141 distributions, hygroscopicity, refractive indices, and densities (Latimer and Martin, 2019). For comparison with
142 modeled monthly AOD, daily measurements at each site at 440 nm are first interpolated to the standard wavelength
143 of 550 nm using the local Ångström exponent between 440 and 870 nm channels, following the Ångström power law
144 (Eq. (1); Martínez-Lozano et al., 1998). These interpolated values are then averaged to calculate the observed mean
145 monthly AOD. The interpolation formula used is:

146
$$AOD_{(550nm)} = AOD_{(440nm)} * \left(\frac{550}{440}\right)^{-\alpha_{ext}\left(\frac{440}{870}\right)} \quad (1)$$



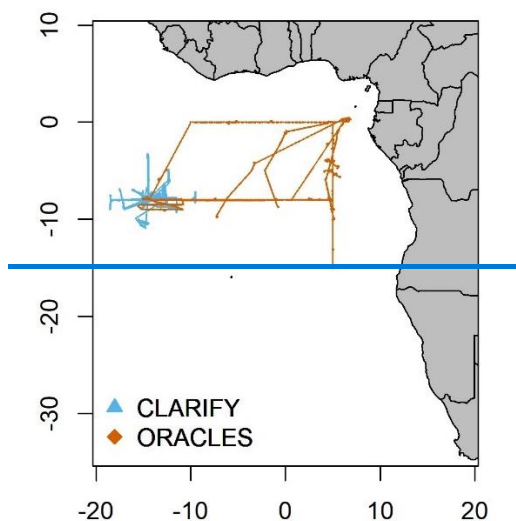
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148 **Figure 1: Map of AERONET sites used for model evaluation (© Google Earth).**

149 In addition, we evaluate the model's relative aerosol composition against measurements from the Atmospheric
150 Radiation Measurement (ARM) facility on Ascension Island during the LASIC campaign, conducted from January to
151 ~~November~~ [October](#) 2017. LASIC employed an Aerodyne aerosol chemical speciation monitor (ACSM) to [provide](#)
152 [quantitative measurement of the chemical composition of non-refractory aerosol components including quantify](#)
153 [sulfate, nitrate, ammonium, and organics aerosol mass concentrations. For comparative analysis, we use aerosol](#)

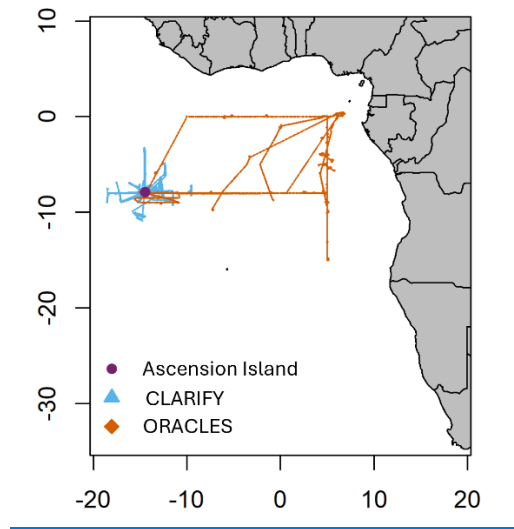
154 [concentrations corrected for composition-dependent collection efficiency \(CDCE\) obtained from the ARM Data](#)
155 [Archive](#). Barrett et al. (2022) reported that aerosol mass concentrations of individual components observed by the
156 LASIC ACSM were 2 to 4.5 times lower than those measured by the aerosol mass spectrometer (AMS) aboard the
157 CLARIFY campaign aircraft. Hence, we evaluate the relative rather than absolute aerosol speciation in GEOS-Chem
158 against the LASIC ACSM.

159 2.3 Aircraft measurements

160 We evaluate simulated aerosol composition against airborne measurements from two campaigns, NASA ORACLES
161 (Redemann et al., 2021; Ryoo et al., 2021) and UK CLARIFY (Haywood et al., 2021). The ORACLES field campaign
162 used the NASA P-3 aircraft to make measurements based out of São Tomé and Príncipe while CLARIFY used the
163 FAAM BAe-146 aircraft around Ascension Island for data collection. The ORACLES aircraft primarily conducted
164 morning sampling, between 8:00-13:00 UTC, while the CLARIFY aircraft often sampled extended hours, typically
165 from 7:00-18:00 UTC. Both campaigns occurred during the austral winter/spring (August-September), corresponding
166 with peak biomass burning events in southern Africa (Adebiyi et al., 2015). Figure 2 shows the flight tracks for these
167 campaigns. The primary instruments and references for each campaign are listed in Table 1.



168



169

170 **Figure 2: Flight tracks from the two aircraft campaigns used to evaluate the model, CLARIFY (in blue) and ORACLES**
 171 **(in orange), conducted during August-September 2017 over the southeast Atlantic region. [Ascension Island is marked by](#)**
 172 **[the purple dot.](#)**

173 To facilitate comparison between airborne measurements and the GEOS-Chem model, we sampled the model to the
 174 nearest grid box, both temporally and spatially, along the flight tracks. Observations from both campaigns are reported
 175 at 1-minute averaging intervals, while the model operates at a 10-minute temporal resolution (see Sect. 2.1). Aerosol
 176 concentrations from the campaigns are reported as mass concentrations at standard temperature and pressure (STP:
 177 273 K, 1 atm). The modeled concentrations are thus also standardized to STP conditions.

178 **Table 1: Aircraft campaigns in the southeast Atlantic used for model evaluation during the biomass burning season**

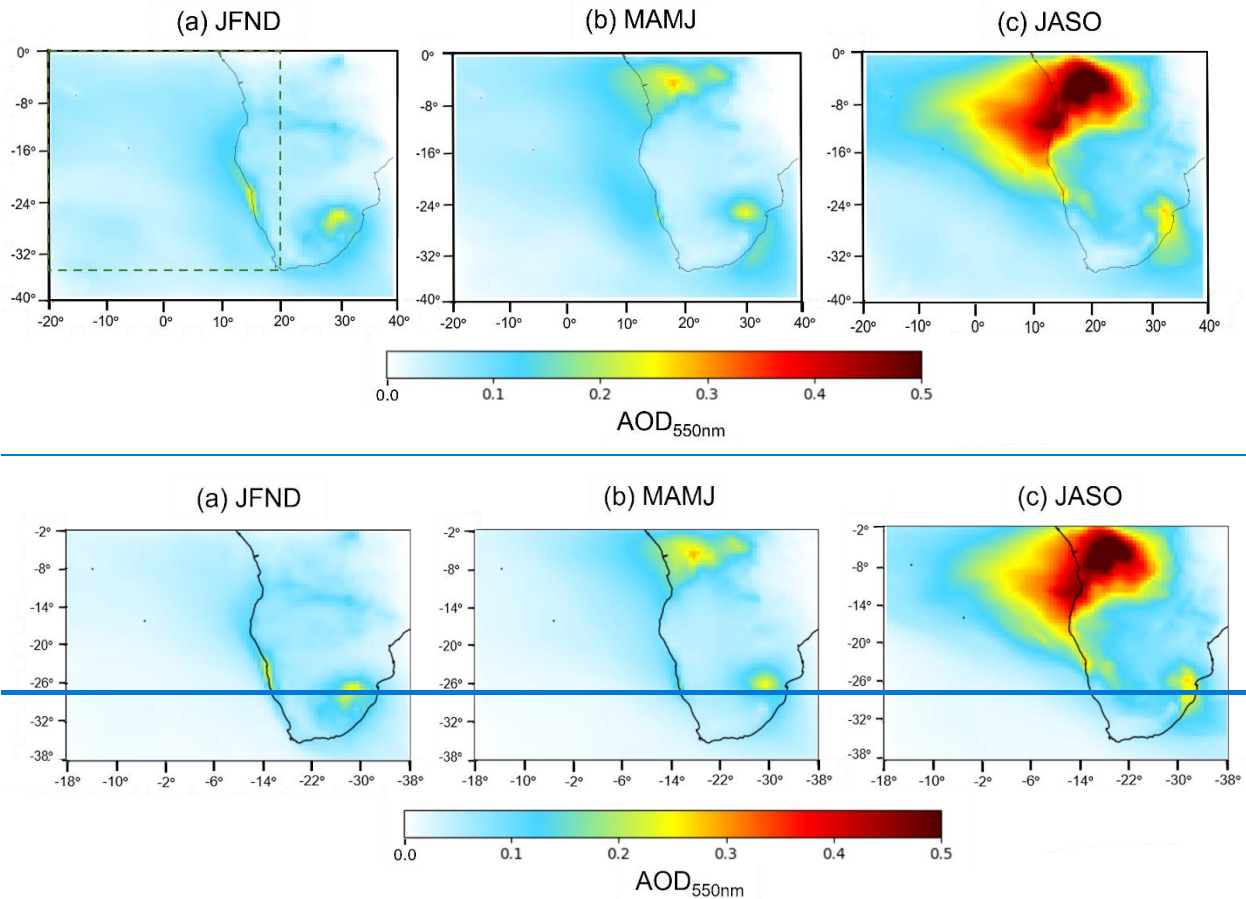
Campaign	Date range (Duration)	Instruments*	Aerodynamic Diameter (μm)	Altitude from surface (km)	Primary Reference
CLARIFY	7 th August–4 th September 2017 (99h)	C-ToF-AMS	0.05 to 0.60	0 to 8	Haywood et al., 2021
ORACLES	16 th August–6 th September 2017 (112h)	HR-ToF- AMS	0.07 and 0.70	0 to 7	Redemann et al., 2021

179 *Compact Time-of-Flight (C-ToF), High Resolution Time-of-Flight (HR-ToF), Aerosol Mass Spectrometer (AMS)

180 **3.1 Model Evaluation**

181 **3.1.1 Seasonal variation of AOD**

182 The spatial distribution of seasonal mean AOD from GEOS-Chem for the year 2017 is presented in Fig. 3. Three
 183 distinct seasonal periods reflect dominant atmospheric and oceanic processes. These include the high DMS emission
 184 period in the SEA, during the months of January, February, November, and December (JFND); the peak biomass
 185 burning season in southern Africa, spanning from July to October (JASO); and the transitional season, encompassing
 186 March, April, May, and June (MAMJ).



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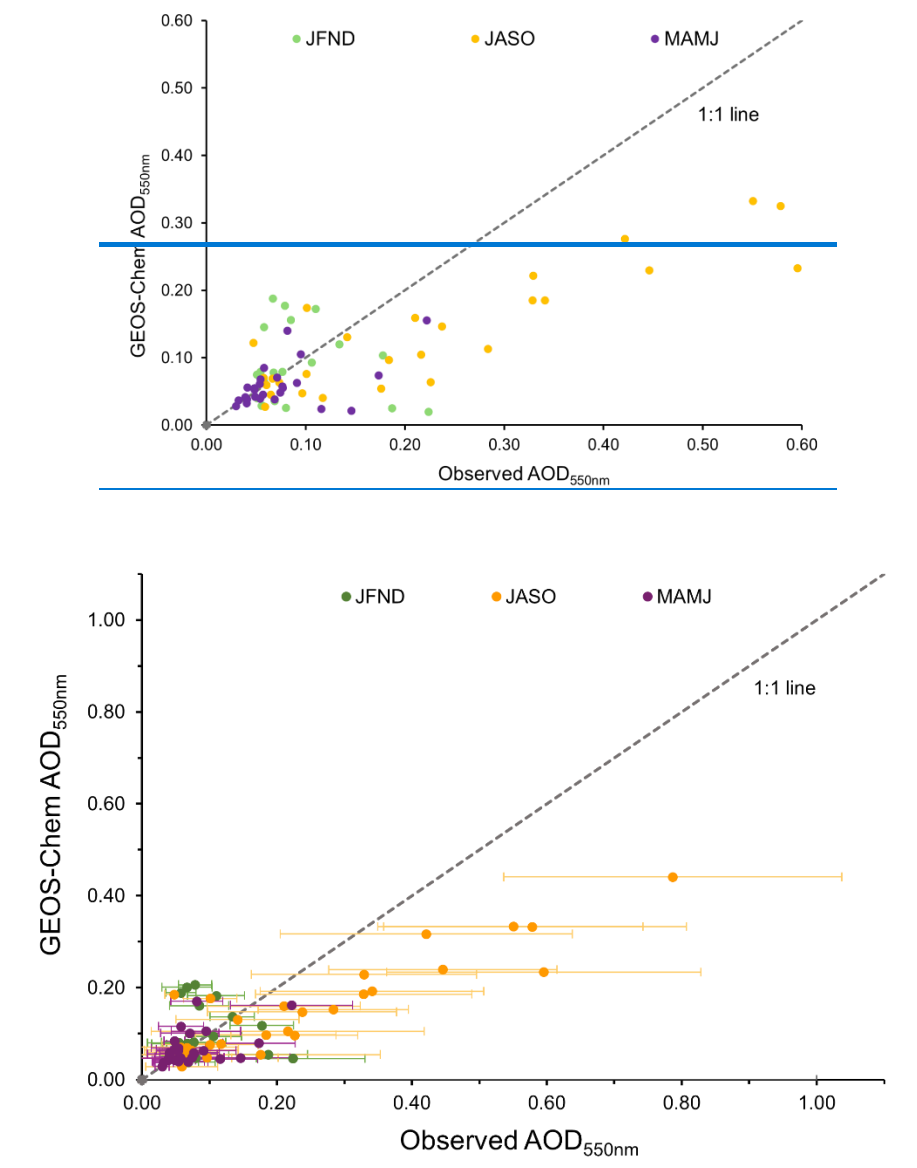
189 **Figure 3: Spatial distribution of seasonal mean modeled AOD at 550 nm for 2017. Seasons are as follows: (a) the peak DMS**
 190 **emission period (JFND), (b) the transitional period (MAMJ), and (c) the peak biomass burning period (JASO). The sub-**
 191 **domain (0–35° S, 20° E–20° W) is highlighted with a green dotted rectangle in panel (a) for reference.**

192 The simulated DMS emissions, based on Lana climatology (2011), indicates that emissions in the BUS region peak
 193 in January, leading to elevated concentrations of sulfate aerosols. This increased d sulfate (~20%), combined with dust
 194 (59%) emissions from the Namib desert, contributes to an AOD hotspot as depicted in Fig. 3a on the southwestern
 195 coast. In the JASO period (Fig. 3c), modeled AOD increases due to biomass burning aerosols, originating from
 196 savannah fires in Central and southern Africa and transported westward towards the SEA region by the southern
 197 African easterly jet (Adebiyi and Zuidema, 2016). The spatial distribution of mean transitional period AOD (Fig. 3b)
 198 features hotspots in Congo and Angola~~Namibia and southern Africa~~, which coincide with ~~dominant anthropogenic~~

199 [sources and](#) the onset of biomass burning in Central Africa. [Additionally, a year-round AOD hotspot is observed in](#)
200 [northeastern South Africa \(Gauteng province; Fig. 3\), which is associated with elevated aerosol concentrations due to](#)
201 [industrial and mining activities, as well as domestic fuel burning \(Arowosegbe et al. 2021; Zhang et al. 2021\).](#)

202

203



204

205 **Figure 4: Modeled AOD_{550nm} (Y-axis) with respect to AERONET AOD_{550nm} (X-axis). Each data point represents the**
206 **monthly mean values for each station color-coded by season (green- DMS period, yellow - biomass burning period, purple-**
207 **transitional period). [Error bars indicate the standard deviation of the AERONET AOD_{550nm} values, and \[the dotted line\]\(#\)](#)**
208 **depicts the 1:1 relationship.**

209 Figure 4 shows the correlation of monthly average AERONET and GEOS-Chem AOD across the nine selected sites
 210 (see Sect. 2.1 and Fig. 1), with the three seasonal periods ~~distinguished~~^{indicated} by color: ~~(~~green for peak DMS
 211 emission season ~~(~~JFND), yellow for biomass burning season ~~(~~JASO), and purple for the transition period ~~(~~MAMJ).
 212 ~~Each data point corresponds to the monthly mean AOD values at distinct AERONET sites. The error bars in Fig. 4~~
 213 ~~represent the ± 1 standard deviation in monthly AOD measurements at these sites, with higher deviations observed~~
 214 ~~during the biomass burning months (up to ± 0.25 at Namibe site). The comparison of monthly mean AOD across~~
 215 ~~individual sites (see Table A2 in the Appendix) shows that, with the exception of Ascension Island, Gobabeb, and~~
 216 ~~Upington, the mean AOD at the remaining sites during the biomass burning season (JASO) is atleast one standard~~
 217 ~~deviation higher than the mean AOD in other seasons (JFND & MAMJ).~~

218 Table 2 compiles the performance of monthly mean GEOS-Chem AOD with respect to AERONET AOD by season.
 219 JASO exhibits the strongest correlation ($R = 0.9012$), which is statistically significant ($p < 0.05$). The transitional
 220 period (MAMJ) shows a moderate correlation ($R = 0.4854$) with a normalized mean bias (NMB) of -94.5% . ~~A notably~~
 221 ~~low~~^{Negligible-} ~~negative~~ correlation coefficient ($R = -0.05842$) with a positive bias (29.8425%) is seen during the
 222 summer period (JFND), predominantly due to anomalies at two sites. This period witnesses a considerable
 223 underestimation of AOD at Ascension Island, alongside an overestimation of dust aerosol at Gobabeb. Excluding
 224 these two sites ~~from the analysis, improves both the model's correlation to 0.67 ($p = 0.55$) coefficient and reduces the~~
 225 ~~NMB improve to -0.61 ($p = 0.55$) and $4.7-8\%$ respectively, indicating better model performance at the remaining 7~~
 226 ~~sites. This underestimate of AOD at Ascension Island (Fig. A1 in the Appendix) during summer (JFND) suggests~~
 227 ~~potential model limitations in accurately simulating natural aerosol emissions such as sea salt and marine biogenic~~
 228 ~~emissions. Meanwhile, the AOD discrepancy at Ascension Island in the biomass burning season, may be due to the~~
 229 ~~underestimate of transatlantic transport of light-absorbing carbon aerosols (Das et al., 2017) and deviations in its~~
 230 ~~spatial distribution from typical zonal patterns over the Atlantic (Adebisi et al., 2023). Furthermore, Table 2 shows~~
 231 ~~that the model underestimates AOD during JASO by 26.5% (NMB) across the domain during. The sources of these~~
 232 ~~model biases are discussed in further detail in Section 3.1.2. This underestimate may stem from the model's bulk~~
 233 ~~aerosol scheme which inadequately captures the optical properties of aerosols and is compounded by a low relative~~
 234 ~~humidity bias (Zhai et al., 2021). The bulk scheme also assumes all aerosols are externally mixed, which contrasts~~
 235 ~~with the variable degree of particle mixing states in the atmosphere (Yu et al., 2012). Additionally, studies like Hodzie~~
 236 ~~et al. (2020) using NASA ATom aircraft data, indicate that GEOS-Chem substantially underestimates oxidation levels~~
 237 ~~of organic aerosols in remote areas, which could affect estimates of their burden and optical properties.~~

238 Table 2: Statistical parameters of monthly mean modeled AOD with respect to observed AOD at the AERONET sites
 239 by season

Time period	Number of observations	Correlation coefficient (R)	Normalized mean bias (NMB) (%)	Root-mean square error (RMSE)
JFND	20	-0.058-0.12	29.842.5	0.079

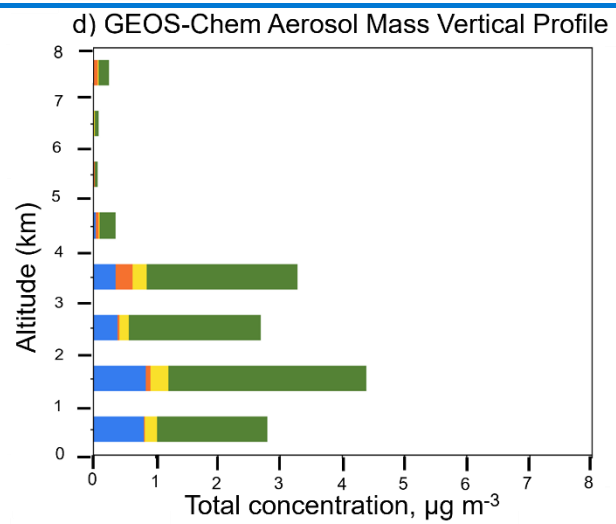
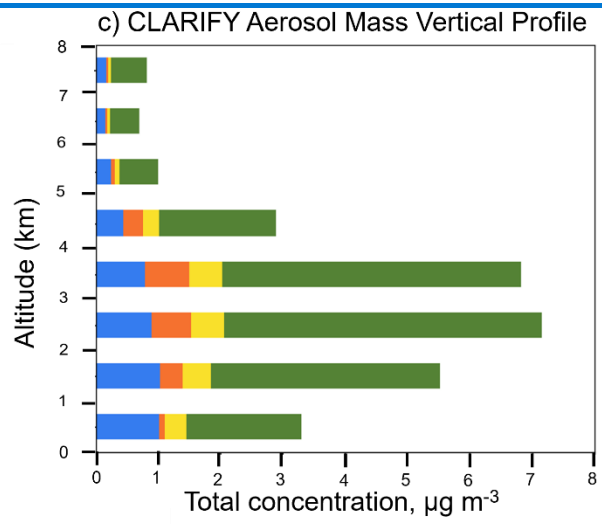
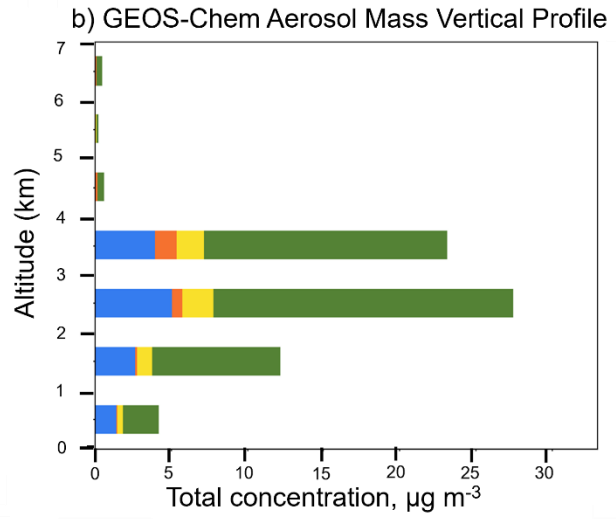
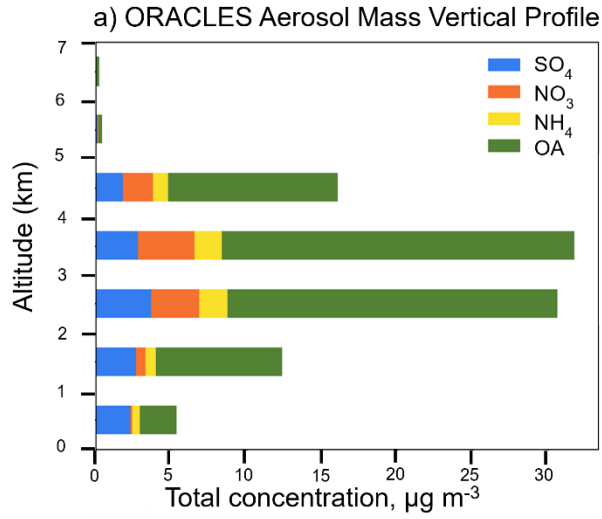
		(p = 0. 6275)		
MAMJ	26	0. 4851 (p = 0. 5915)	-94.5	0.04 34
JASO	28	0.90 12 (p = 0.0 4418)	-18.626.5	0.15

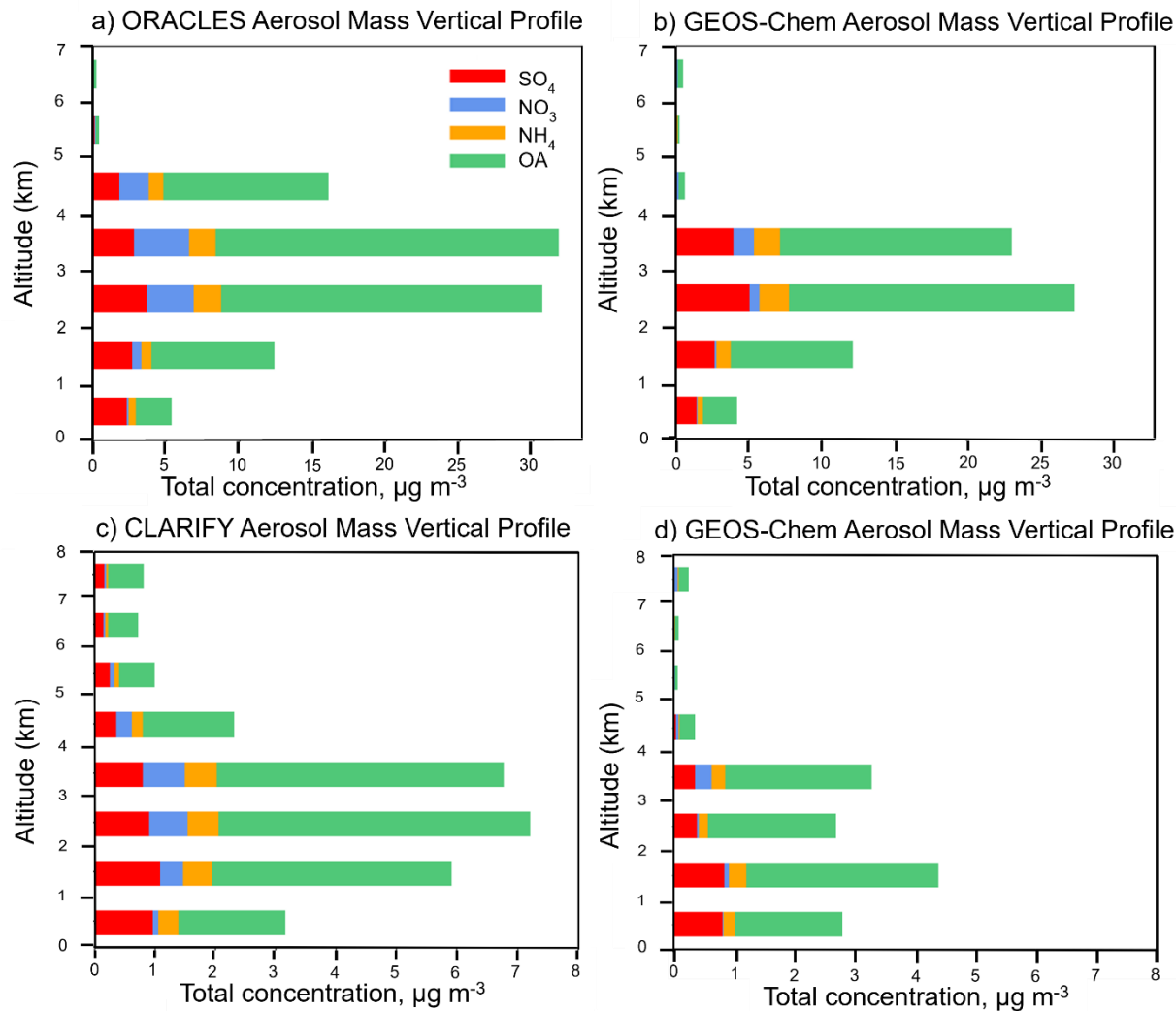
240 We evaluate the relative aerosol speciation simulated at Ascension Island against monthly mean ACSM observations
241 during the LASIC campaign (see Sect. 2.2) available for January–~~November~~~~October~~ 2017 (Fig. A2 in the Appendix).
242 The seasonality of the relative contributions of organic aerosols and sulfate are consistent between the model and
243 observations. However, the model underestimates the relative contribution of sulfate during most months, while
244 generally overestimating the proportion of organics. An increase in the transport of biomass burning organic aerosols
245 would further worsen the model underestimate of sulfate. A slight overestimate in theThe modeled relative
246 contribution of sulfate is ~~closest to that~~ observed in ~~January and~~ February, ~~and November~~, when simulated DMS
247 emissions in the region are high (Lana et al., 2011), ~~with a slight overestimate in the latter largely due to enhanced~~
248 underestimations of organics and nitrates.

249 3.1.2 Vertical profiles of aerosol composition

250 Figure 5 depicts the mean vertical profiles of speciated aerosol mass concentrations observed during ORACLES and
251 CLARIFY aircraft campaigns in August–September 2017 (the biomass burning season), compared to GEOS-Chem
252 (see Sect. 2.2 and Table 1). The cloud top height in the SEA region generally falls between 0 to 2 km (Redemann et
253 al., 2021). Findings from Diamond et al. (2018) indicate that aerosols below clouds in this lower atmospheric layer
254 can also substantially impact cloud microphysics. At these altitudes (0–2 km), GEOS-Chem performs well against
255 AMS measurements of total aerosol mass, which includes sulfate, nitrate, ammonium and organics from these
256 campaigns, with an NMB between -3.5% (CLARIFY) to -7.5% (ORACLES). At mid-altitudes (2–4 km), the model
257 is biased low, with NMB values spanning -19% (ORACLES) to -57% (CLARIFY). However, the model demonstrates
258 a pronounced bias at higher altitudes (4–7 km), where NMB values drop to -92% (ORACLES) to -93.5% (CLARIFY),
259 underscoring challenges in accurately modeling aerosol concentrations at these elevations. These significant low
260 biases in aerosol concentrations at higher altitude likely contribute to the model’s underestimation of AOD during the
261 biomass burning period (see Section 3.1.1). This underestimation may also be affected by the model’s bulk aerosol
262 scheme, which inadequately captures the optical properties of aerosols and is compounded by a low relative humidity
263 bias (Zhai et al., 2021). The bulk scheme also assumes all aerosols are externally mixed, which contrasts with the
264 variable degree of particle mixing states in the atmosphere (Yu et al., 2012; Dang et al., 2022). MoreoverAdditionally,
265 studies like Hodzic et al. (2020) –using NASA ATom aircraft data– indicate that GEOS-Chem substantially
266 underestimates oxidation levels of organic aerosols in remote areas, which could affect estimates of their burden and
267 optical properties.

268 Pai et al. (2020) [further](#) suggests that the model underestimation of organic aerosol loading at mid-tropospheric heights
269 is linked to the surface injection treatment of fire emissions in GFED4.1s. Recent studies by Wizenberg et al. (2023)
270 and Marvin et al. (2024) concur that fire injection scheme is a critical source of model uncertainty, emphasizing the
271 potential importance of accurate fire injection modeling in the free troposphere. Nonetheless, our study focuses on
272 aerosol composition within cloud-relevant altitudes to improve our understanding of aerosol-cloud interactions and
273 their climate implications. The observed vertical distribution of aerosol mass concentrations (left panels of Fig. 5),
274 indicates that 18% and 36% of the aerosol mass for the ORACLES and CLARIFY campaigns, respectively, is located
275 below 2 km, within columns extending up to flight altitudes of 7 km and 8 km. However, the model simulates elevated
276 aerosol mass at these lower altitudes, 24% and 50% of the column for ORACLES and CLARIFY, respectively.

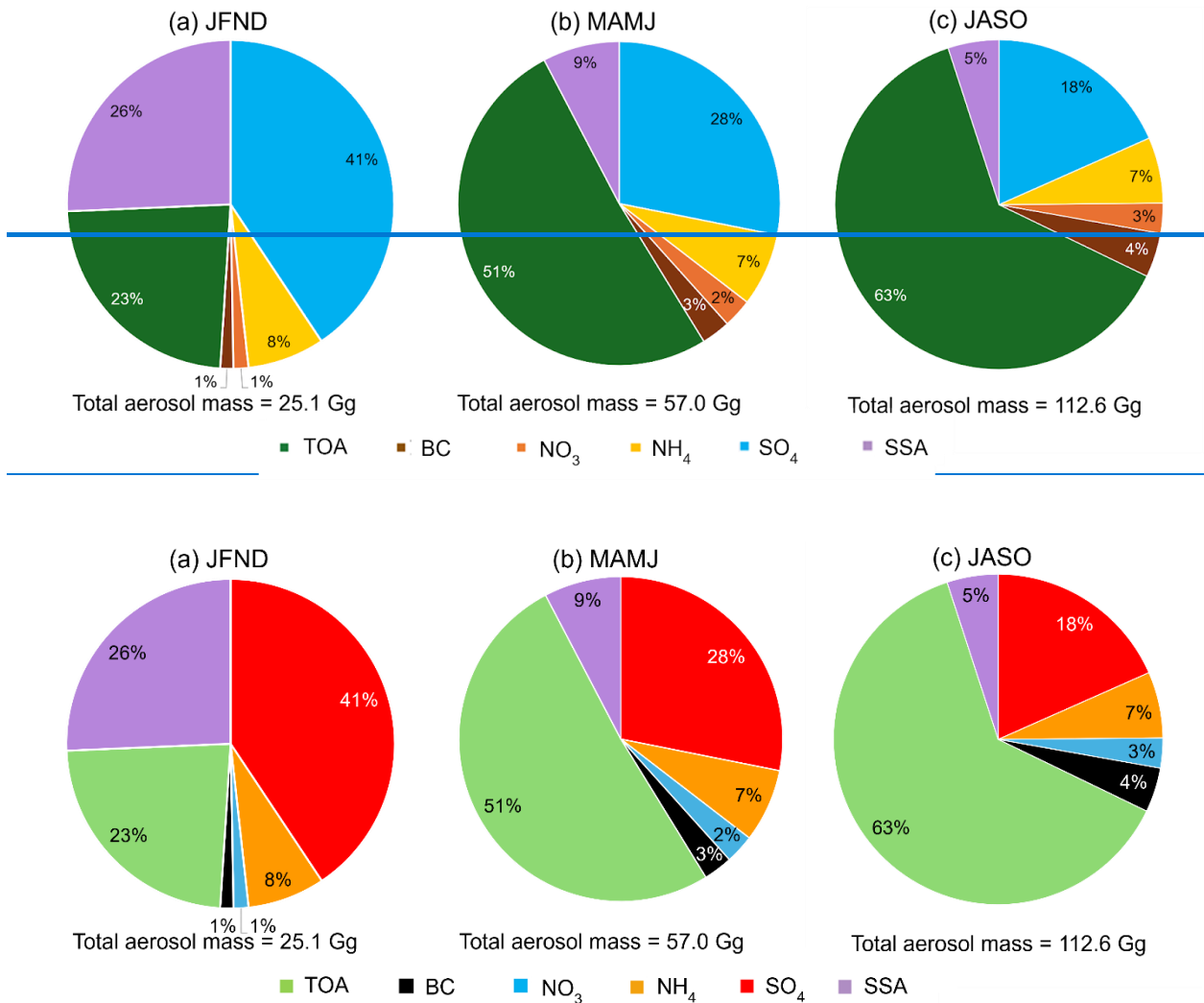




278
 279 **Figure 5: Average vertical profiles of simulated and observed aerosol mass during August–September 2017 (peak biomass**
 280 **burning season) from aircraft campaigns. The left column presents the vertical distribution of aerosols observed during the**
 281 **ORACLES flight campaign (panel a) and the CLARIFY flight campaign (panel c) at STP (see Sect. 2.3). The right column**
 282 **displays the GEOS-Chem model simulations along the respective flight tracks of each campaign (panels b and d). All data**
 283 **are averaged over 1 km vertical bins.**

284 At altitudes where clouds persist in the domain (0 to 2 km), sulfate and organic aerosols are the dominant aerosol
 285 types. Here, the model effectively captures the mass concentration of organic aerosols, with an NMB ranging from -
 286 0.40% for ORACLES to -14% for CLARIFY. However, it underestimates sulfate aerosol concentrations by 19% at
 287 cloud altitudes for both campaigns. For other aerosol types and altitudes, the model consistently underestimates
 288 concentrations, except for sulfate and ammonium aerosols between 2 to 4 km during the ORACLES campaign, which
 289 the model overestimates by 40% and 4.6%, respectively. The model captures the total aerosol mass from 0 through 7
 290 km for sulfate and ammonium aerosols during the ORACLES campaign, with only minimal underestimations of 1.5%
 291 and 0.7%, respectively. This indicates a potential discrepancy in the vertical distribution of these aerosols rather than
 292 in total mass.

293 **3.2 Seasonal variation in aerosol composition and sources at cloud altitudes**



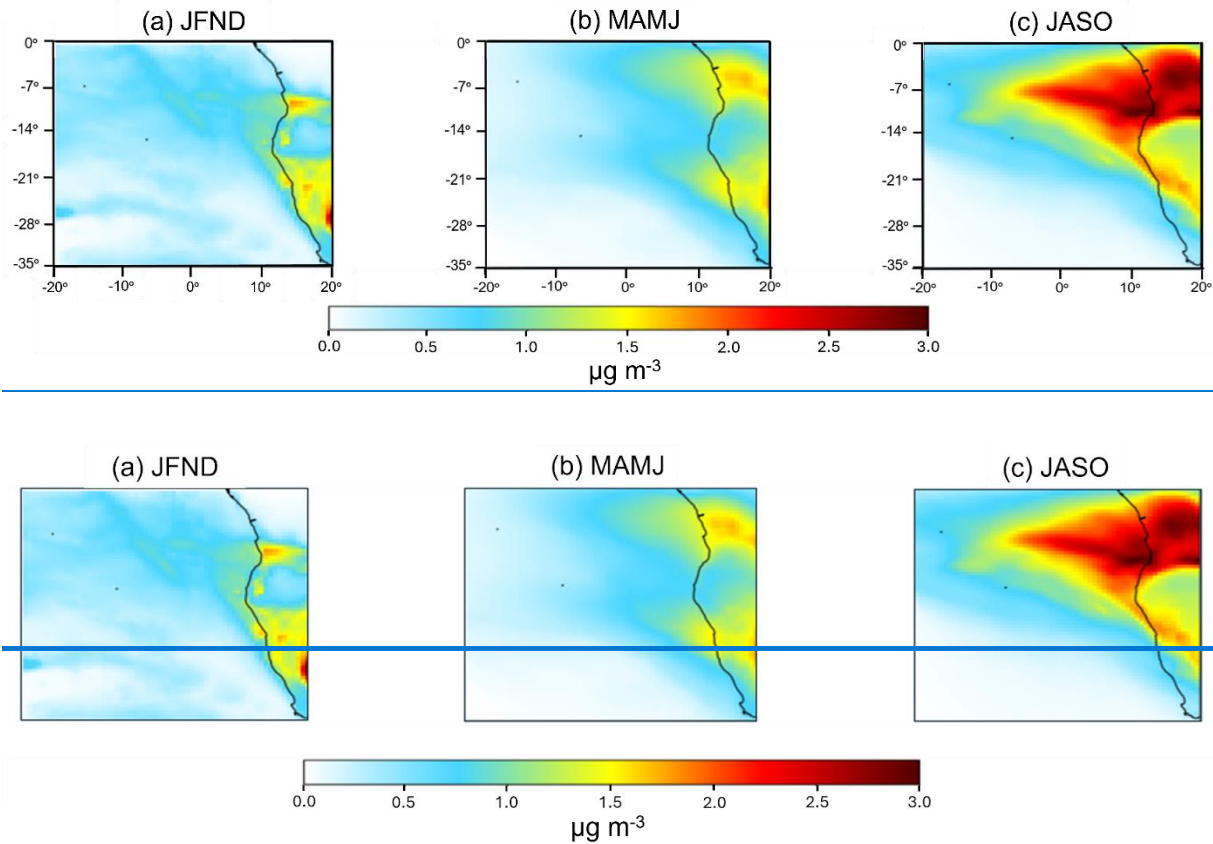
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295

296 **Figure 6: Simulated mean fractional aerosol composition at cloud heights (0–2 km) over the ocean in the stratocumulus**
 297 **sub-domain (0–35° S, 20° E–20° W) by season: (a) JFND, (b) MAMJ, and (c) JASO. Here SO₄, NH₄, NO₃, BC, TOA, SSA**
 298 **represents sulfate, ammonium, nitrate, black carbon, total organic aerosol and accumulation-mode sea salt aerosols,**
 299 **respectively.**

300 Figure 6 presents the simulated seasonal mean aerosol fractional composition within cloud-relevant altitudes (0–2 km),
 301 averaged over the ocean only across the subdomain (0–35° S, 20° E–20° W) (see the map shown in Fig. 7). This area
 302 is strategically selected to coincide with the persistent Sc cloud deck and enhance our analysis of aerosol-cloud
 303 interactions. Organic aerosols, an indicator of biomass burning, predominate during both the biomass burning (JASO)
 304 and transitional (MAMJ) periods. In contrast, sulfate aerosols dominate during austral summer, likely influenced by
 305 the high primary production from coastal upwelling that leads to DMS emissions. We investigate the model
 306 representation of sulfate and these processes further in subsequent sections. An increase in the accumulation-mode
 307 sea-salt aerosols (radius 0.01–0.5 μm) contribution (total mass of 6.7 Gg) is observed in summer (Fig. 6a) as well,
 308 compared to other seasons (5.2 Gg during MAMJ and 5.8 Gg during JASO), owing to the peak wind speeds in the

309 southern Benguela region in this season (Hutchings et al., 2009). Black carbon, ammonium, and nitrate aerosols make
310 minor contributions to simulated aerosol mass at cloud height throughout the year.

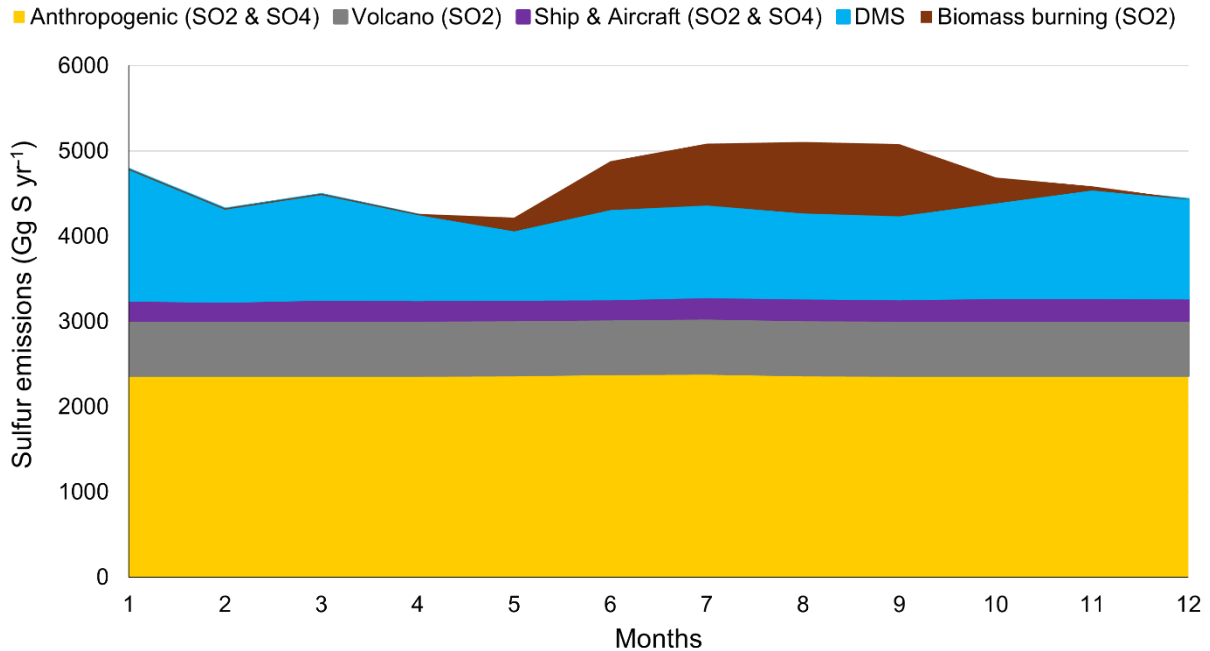


312
313 **Figure 7: Spatial distribution of simulated mean sulfate aerosol concentrations averaged over cloud altitudes (0–2 km) in**
314 **the sub-domain (0–35° S, 20° E–20° W) by season in 2017: (a) peak DMS emission season (JFND), (b) transitional phase**
315 **(MAMJ), and (c) biomass burning season (JASO).**

316 3.2.1 Drivers of sulfate aerosol and importance of marine precursor emissions

317 Sulfate aerosols are the most or 2nd most important aerosol component in cloud heights over the SEA (Fig. 6). We
318 examine the sources of sulfur emissions within the model in Figure 8A3. Within the broader domain (0–40° S, 40° E–
319 20° W), anthropogenic activities are the largest source of sulfur emissions throughout the year (Fig. 8A3). However,
320 [the model default CEDS inventory \(Hoesly et al., 2018\) fails to capture the seasonality of these emissions due to](#)
321 [absence of regional inventories and reliance on the global datasets such as the International Energy Agency \(IEA\)](#)
322 [energy statistics. The anthropogenic emissions are](#) followed by DMS emissions from the ocean, ~~DMS emissions~~
323 [which](#) become more pronounced during the austral summer, peaking in January. Additionally, biomass burning
324 contributes to SO₂ emissions seasonally, becoming the 3rd most important source of total sulfur emissions during July
325 - September (Fig. 8A3). [In contrast, sulfur contributions from volcanic, shipping, and aircraft emissions remain](#)
326 [minimal and constant year-round, reflecting assumptions of static fuel burned and emission levels across inventories.](#)

327

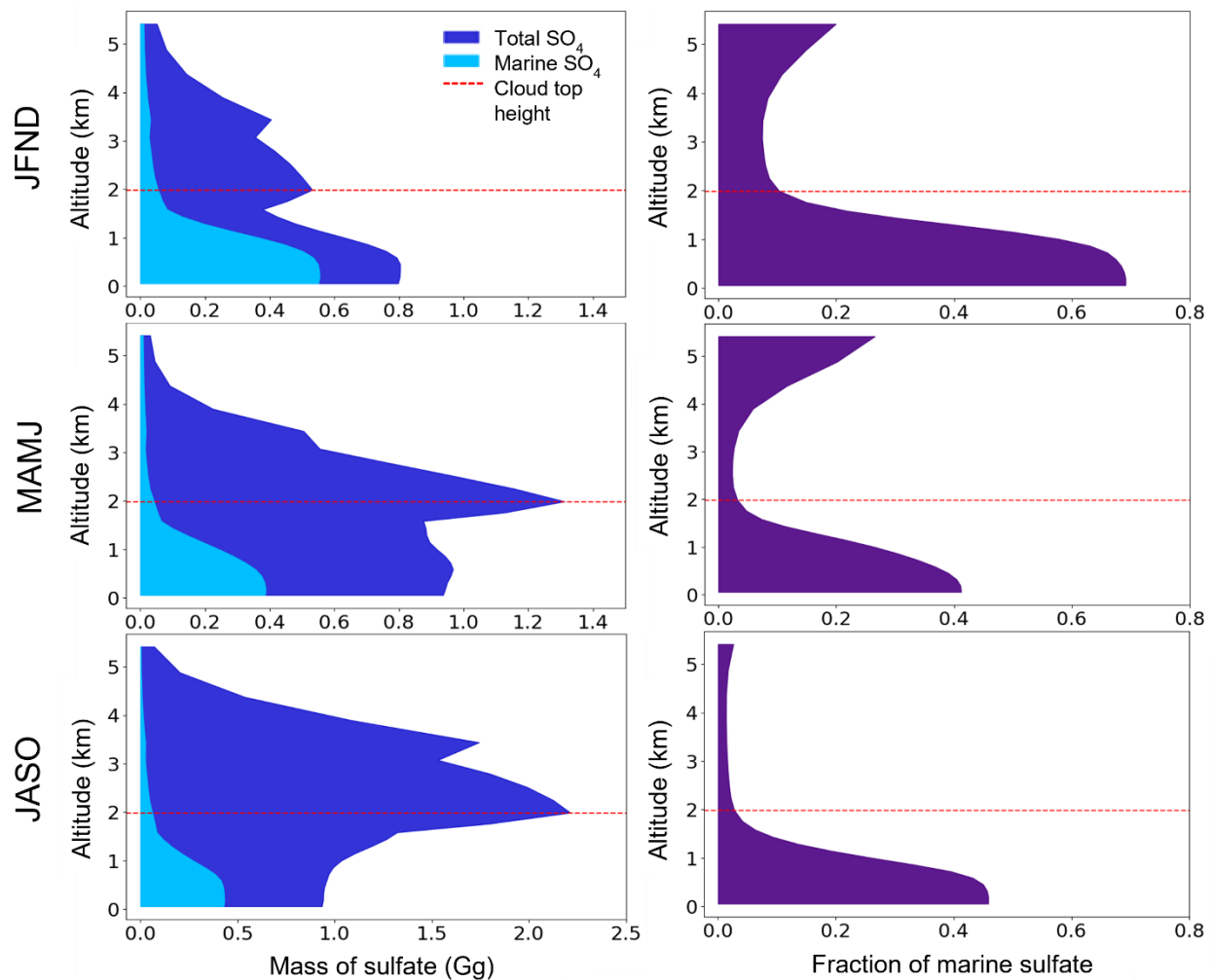


328
 329 **Figure 8A3:** Stacked area chart of monthly total sulfur emissions by source for 2017 across the study domain (0–40°
 330 S, 40° E–20° W) in gigagrams of sulfur per year (Gg S yr⁻¹). Sources are indicated by color and encompass
 331 anthropogenic activities, volcanic activity, ship and aircraft emissions, biomass burning and natural emissions of
 332 dimethyl sulfide (DMS).

333
 334 To improve understanding of the processes driving sulfate aerosol concentrations in the region, we examine its
 335 simulated spatial distribution averaged by season over the cloud height (0–2 km) in Fig. 7. Elevated concentrations of
 336 DMS, resulting from higher rates of primary production (Lana et al., 2011; Galí et al., 2018), lead to an increase in
 337 sulfate concentrations along the coastline of the Benguela region and the inner shelf of Namibia during JNFD (Fig.
 338 7a), aligning with the AOD hotspot observed in Fig. 3a. This is consistent with the simulated dominance of sulfate
 339 aerosols at cloud-relevant altitudes during JFND (Fig. 6a). During the biomass burning months (JASO), while their
 340 relative contribution decreases (Fig. 6c), sulfate aerosols display a pronounced increase in absolute concentration (Fig.
 341 7c) as a consequence of savanna fire emissions from southwestern Africa (van der Werf et al., 2010; Das et al., 2017).
 342 As outlined in the AOD evaluation (Sect. 3.1.1), the model underestimates the transport of emissions to remote sites
 343 (Fig A1), resulting in a steep gradient in sulfate concentrations from the eastern landmass towards the western open
 344 ocean.

345 To quantitatively estimate the contribution of marine precursor emissions to sulfate aerosols, we compare the sulfate
 346 mass between the standard and marine emissions only sensitivity simulations (Sect. 2.1). Figure 98 shows seasonally
 347 averaged vertical profiles over the ocean region of the Sc sub-domain (0°–35° S, 20° E–20° W). The figure presents
 348 the marine-only sulfate mass and the total sulfate mass from the standard simulation (left panels), and the ratio of
 349 marine sulfate to total sulfate (right panels). Vertical profiles were computed by summing the sulfate mass within each

350 grid box, scaled by the grid box ocean fraction, across latitude and longitude within each vertical layer of the model,
 351 and then averaged temporally across each season.



352

353 **Figure 98:** Simulated vertical profiles of sulfate aerosol mass over oceanic regions within the sub-domain (0° – 35° S, 20° E–
 354 20° W) by season. The left panel shows the mass of total and marine sulfate aerosols, and the right panel indicates the sulfate
 355 fraction from marine sources. The top row corresponds to the peak dimethyl sulfide (DMS) emission period (JFND); the
 356 middle row to the transitional period (MAMJ); and the bottom row to the peak biomass burning period (JASO) (note: the
 357 bottom left panel displays a higher x-axis scale). The upper red dashed line denotes the typical maximum cloud top height
 358 (Redemann et al., 2021).

359 Our analysis highlights the substantial influence of marine sulfur sources on sulfate during JFND, as evidenced in the
 360 top left panel of Fig. 98. During this period the proportion of marine sulfate reaches up to 69.1% within-cloud (from
 361 surface to 2 km). The contribution of marine sulfate within the cloud in the subsequent periods is reduced (ranging
 362 between 2.7–45.9%; Fig. 98). We find that marine-sourced sulfate mass remains fairly consistent throughout the year
 363 (Fig. 98, left panels), with variations in the marine sulfate fraction (Fig. 98, right panels) mainly due to changes in
 364 land-based sulfate sources. Total sulfate mass during seasons influenced by biomass burning (MAMJ and JASO)

365 peaks at 2 km, with greater mass above 2 km during peak biomass burning (JASO) in contrast to JFND where mass
366 peaks within clouds (0–2 km).

367 Table A3 summarizes the monthly mean percent contribution of marine sulfate averaged across cloud altitudes (0–2
368 km). The annual average total sulfate mass and marine sulfate mass is 16.2 Gg and 3.5 Gg, respectively. The within-
369 cloud marine sulfate contribution peaks in January (57.7%) and is smallest in September (10.3%). Thus, our analysis
370 suggests that DMS emissions influence sulfate aerosol formation and their interactions with clouds in the region
371 throughout most of the year, excepting only the peak biomass burning season. This emphasizes that constraining
372 marine sulfur sources and chemistry both in chemical transport and climate models may improve representation of
373 aerosol-climate dynamics in the SEA region. Limited available observations suggest the model is biased low in AOD
374 throughout most of the year (Sect. 3.1.1), and underestimates sulfate aerosol concentrations in August and September
375 at cloud altitudes (Sect. 3.1.2, Fig. 5). We explore related uncertainties and their implications in the following
376 sections.

377 Table A3: Seasonal variation of percentage of monthly mean percent contribution of marine sulfate within cloud
378 height

<u>Month</u>	<u>Percentage of marine sulfate</u>
<u>January</u>	<u>57.7</u>
<u>February</u>	<u>54.8</u>
<u>March</u>	<u>25.3</u>
<u>April</u>	<u>26.6</u>
<u>May</u>	<u>15.3</u>
<u>June</u>	<u>15.0</u>
<u>July</u>	<u>14.8</u>
<u>August</u>	<u>14.7</u>
<u>September</u>	<u>10.3</u>
<u>October</u>	<u>22.4</u>
<u>November</u>	<u>39.1</u>
<u>December</u>	<u>44.3</u>

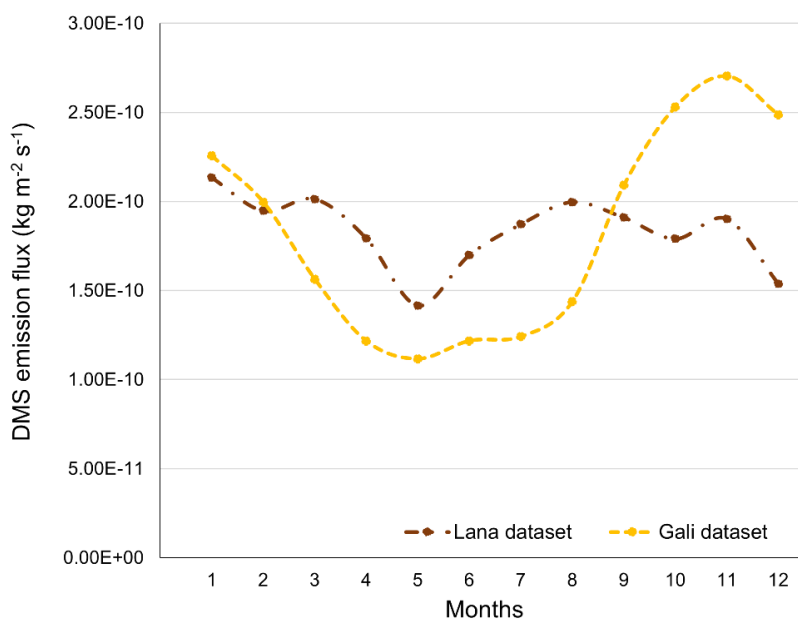
379

380 3.3 Uncertainties

381 3.3.1 Assessing variations in DMS emission rates and oxidation mechanism on sulfate aerosol formation

382 The Benguela region has substantial uncertainties in DMS concentrations in surface seawater (Asher et al., 2011;
383 Tortell et al., 2011) and the corresponding emission fluxes owing to the limited availability of biogenic sulfur
384 measurements. To investigate the sensitivity of DMS emission fluxes to changes in surface seawater DMS
385 concentrations, we conducted two simulations with DMS concentrations from Lana et al. (2011) and Galí et al. (2018)
386 (see Sect. 2.1). The standard results presented thus far were conducted using the Lana dataset.

387 In the southern Benguela, south of approximately 27° S, marked upwelling during the austral summer (Shannon and
388 Nelson, 1996; Hutchings et al., 2009) promotes phytoplankton growth and elevates DMS emissions. Figure 10
389 indicates that Although the Lana dataset aligns with this phenomenon, displaying peak DMS emission
390 fluxes over the Sc sub-domain peak in January (Fig. A4 of the Appendix), coinciding with this phenomenon However,
391 it lacks clear seasonality for the remaining months. In contrast, satellite-based DMS estimates from Galí show
392 pronounced emissions throughout the austral summer (JFND), as shown in Fig. 10A4. Both datasets concur in
393 magnitude for January and February, a period with better data coverage in the Lana et al. (2011) climatological data
394 set over the domain. However, the Lana dataset DMS emissions are up to 38% less during December, while 51%
395 higher in July relative to the Galí dataset (Ghahreman et al., 2019). This suggests the marine contribution to sulfate in
396 our standard simulation using the Lana dataset may be underestimated from October through December
397 (encompassing two months of the peak DMS season) and overestimated from March through August (Fig. 10A4).



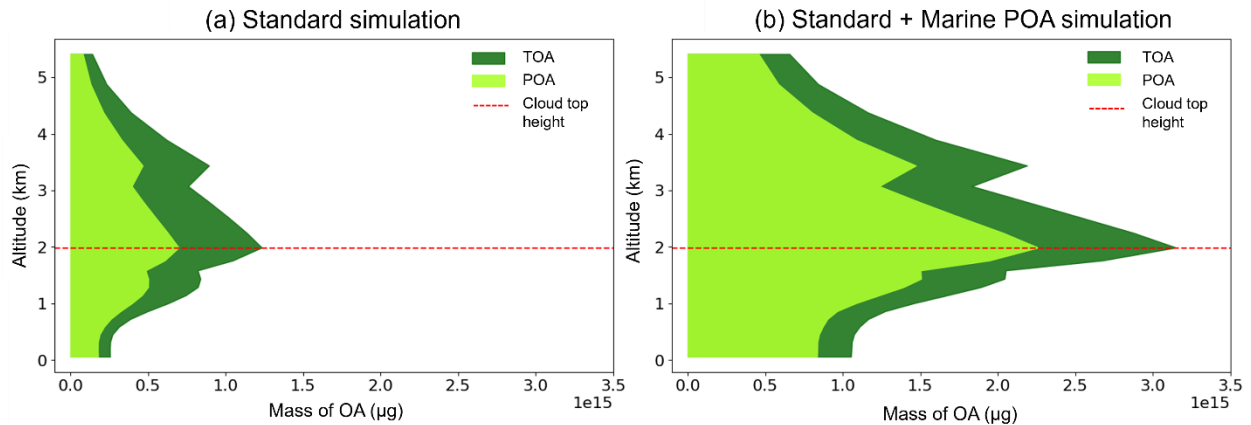
398

399 [Figure 10: Monthly DMS emissions over the stratocumulus sub-domain \(0–35° S, 20° E–20° W\) using two distinct](#)
400 [datasets for surface seawater DMS concentrations. The brown dashed line presents emissions calculated using Lana](#)
401 [et al. \(2011\) climatology, which compiles data across 1972-2009 from multiple sources. In contrast, the yellow dashed](#)
402 [line depicts emissions based on satellite-derived estimates of surface seawater DMS concentrations \(Galí et al., 2018\).](#)

403 The ongoing discovery of complexities within DMS oxidation mechanisms, along with the incomplete incorporation
404 of these mechanisms into atmospheric chemistry models, further contributes to uncertainties in predicting the impact
405 of DMS emissions on aerosols and climate (Faloona, 2009; Quinn and Bates, 2011; Carslaw et al., 2013). Chen et al.
406 (2018) highlighted the impacts of changes to DMS chemistry in the GEOS-Chem model, integrating a series of
407 multiphase sulfur oxidation mechanisms and two DMS intermediates, which led to a decrease in the global DMS
408 burden, thereby decreasing SO₂ and sulfate levels. On the other hand, Novak et al. (2021) found that the cloud uptake
409 of hydroperoxymethyl thioformate (HPMTF), a newly identified oxidation product of DMS (Wu et al., 2015; Veres
410 et al., 2020), lowers near-surface SO₂ concentration while elevating sulfate concentration in the model. Most recently,
411 Tashmim et al. (2024) implemented an advanced DMS oxidation mechanism in GEOS-Chem that incorporates the
412 latest developments in DMS chemistry, including those previously mentioned, which led to a lower SO₂ mixing ratio
413 (~70%) and a higher SO₄ mixing ratio (~35%) over the SEA during austral summer. Thus, an improved representation
414 of DMS emissions and oxidation chemistry in the model could enhance the sulfate aerosol estimations during the peak
415 DMS season. This refinement may address model underestimates of aerosol concentrations during this period (Sect.
416 3.1.1).

417 **3.3.2 Exploring the impact of marine organic aerosol emissions on organic aerosol concentrations**

418 Beyond marine sulfate and sea-salt aerosols, organic matter also makes a significant contribution to marine aerosol
419 mass (Middlebrook et al., 1998; Oppo et al., 1999; Russell et al., 2010). Notably, substantial concentrations of organic
420 carbon aerosols have been observed in marine regions, particularly during periods of intense biological activity
421 (O'Dowd et al., 2004). These aerosols can also increase CCN, affecting cloud properties and radiative balance (Arnold
422 et al., 2009; Gantt and Meskhidze, 2013). However, the standard GEOS-Chem model does not account for these
423 [marine](#) organic aerosol emissions. [Here, ~~W~~we](#) analyzed the impact of marine POA on cloud-altitude aerosols over the
424 SEA by incorporating [marine](#) POA emissions based on satellite-derived chlorophyll-a concentrations (Gantt et al.,
425 2015; See Sect. 2.1) in the model.



426

427 **Figure 119:** Vertical distribution of organic aerosol mass during November 2017, the month of maximum discrepancy
 428 between the standard and MPOA simulations, over the Sc sub-domain (0–35° S, 20° E–20° W). Left: mass profile for total
 429 organic aerosols (TOA) and primary organic aerosols (POA) under standard simulation conditions (Std); right: when
 430 marine primary organic aerosol (MPOA) emissions are included (Std + MPOA). The red dashed line indicates the typical
 431 maximum cloud top height.

432 We find that the inclusion of MPOA emissions consistently resulted in higher organic aerosol mass, with the greatest
 433 increase in November. Figure 119 shows the vertical distribution of total organic aerosols (TOA) mass and POA mass
 434 (including MPOA and other POA sources) with and without MPOA emissions during this month. Similar to our earlier
 435 vertical profile analysis (refer to Sect. 3.2.2), we find that the maximum organic aerosol mass occurred at the highest
 436 cloud top height (2 km). The Standard + MPOA simulated peak total organic aerosol mass was approximately three
 437 times higher than that in the Standard simulation, highlighting the potential contribution of marine sources to total
 438 organic aerosol mass concentrations. However, during the biomass burning season, the sensitivity simulation showed
 439 only a minimal increase, indicating that it does not adequately address the model’s underestimation (refer to Fig. 5).
 440 Gantt et al. (2015) demonstrated that including MPOA emissions in GEOS-Chem reduced the normalized mean bias
 441 (NMB) of surface organic aerosol concentrations at coastal sites by 67%. Additionally, Pai et al. (2020) noted that
 442 without a marine POA source, the model fails to accurately reproduce lower-tropospheric concentrations over oceans,
 443 although the marine POA scheme might be biased high. Despite the limitations of a chlorophyll-based
 444 parameterization like the one used here in providing mechanistic understanding of the seasonal and geographical
 445 variability of organic matter emissions from sea spray (Burrows et al., 2022), our findings suggest that MPOA may
 446 play a role in aerosol-cloud interactions outside of the biomass burning season, in addition to marine-derived sulfate
 447 from DMS (Sect. 3.2).

448 3.3.3 Impacts of uncertainties in biomass burning emissions of SO₂

449 To assess the impact of uncertainty in biomass burning emissions of SO₂ on the relative contribution of marine vs.
 450 land sources to aerosol, we performed a sensitivity analysis using two alternative inventories, QFED and GFAS (see
 451 Sect. 2.1 and Fig. Table_A32). The standard simulations, as detailed in Sect. 2.3, use the default biomass burning
 452 inventory in GEOS-Chem, GFED. The GFAS inventory SO₂ and CO emissions over the domain are constant in time,

453 aligning with QFED during the non-biomass burning months (Fig. A35). We find that CO emissions from GFED and
454 QFED align closely; however, there is a notable difference in SO₂ emissions between the two inventories (Fig. A35).
455 These discrepancies likely originate from variations in SO₂ emission factors employed by each inventory. In July,
456 which exhibits the largest difference between the two inventories, peak SO₂ emissions in QFED are almost five-fold
457 higher than those in GFED. This discrepancy leads to a 25% increase in sulfate aerosol concentrations at cloud
458 altitudes relative to the standard results using GFED (not shown). Consequently, the contribution of marine sulfate to
459 total sulfate (see Sect. 3.2.1) may further decrease during the peak biomass burning season if QFED is used,
460 highlighting the sensitivity of aerosol source attributions to the selected biomass burning inventory.

461 **4 Implications**

462 In this study, monthly marine sulfate constitutes between 10.3% and 57.7% of total sulfate within the cloud height,
463 peaking during the high DMS emission period. However, the default Lana et al. (2011) climatology largely
464 underestimates DMS emissions during the austral summer (November and December) by up to 38%, compared to the
465 satellite-derived estimates from Galí et al. (2018). Moreover, improvement of DMS chemistry in the model by
466 incorporating new oxidation mechanisms and intermediate products could shift the balance towards increased sulfate
467 aerosol production (with Tashmim et al., 2024 suggesting an increase of up to 35% over the SEA). Marine primary
468 organic aerosol emissions may also contribute substantially to the organic aerosol mass during the peak primary
469 production period (JNFD), highlighting the importance of marine contributions to overall aerosol concentrations.
470 Meanwhile, discrepancies in SO₂ emissions from biomass burning can increase sulfate aerosol from biomass burning
471 by up to 25%. These changes would improve the model underestimate of AOD relative to AERONET observations;
472 however, observations of aerosol composition outside of August-September are very limited and this is a large gap.
473 Our results suggest marine-sourced sulfate and organics significantly influence aerosol loading and composition in
474 the SEA, particularly during the non-biomass burning period. Accurately characterizing the seasonal dynamics of
475 aerosols within cloud heights is imperative for quantifying aerosol-cloud interactions and understanding the dynamics
476 of marine aerosols in the SEA region, where uncertainties in aerosol radiative forcing are most pronounced. This
477 understanding is essential for improving the reliability of climate models in areas critical to both regional and global
478 climate dynamics.

479 **5 Conclusion**

480 Aerosols over the southeast Atlantic strongly influence global climate dynamics due to the presence of persistent
481 stratocumulus clouds and large uncertainties in aerosol-cloud interactions. However, precisely representing these
482 interactions in global climate models remains challenging, in part due to sparse available observations, especially
483 outside of the biomass burning season. In this study, we employed the GEOS-Chem chemical transport model to
484 assess the aerosol composition at cloud-relevant altitudes (0–2 km) and identify the sensitivities to marine emissions
485 and chemistry in the southeast Atlantic. This analysis aims to enhance our understanding of the role of marine aerosols
486 and the associated uncertainties affecting aerosol-cloud interactions within this climate-sensitive region.

487 We performed nested grid simulations with a $0.5^\circ \times 0.625^\circ$ horizontal resolution and evaluated the model against
488 ground-based and aircraft campaign observations throughout 2017. We analyzed results for three seasonal periods
489 with distinct dominant processes including (a) the high DMS emission season (JFND), (b) the peak biomass burning
490 season (JASO), and (c) the transitional season (MAMJ). Our analysis showed that simulated monthly average aerosol
491 optical depth (AOD) exhibits the strongest correlation ($R = 0.9012$) with the AERONET AOD observations during
492 the JASO season. However, the model generally underestimates AOD throughout the year, except in the JFND period,
493 ~~where an overestimate at Gobabeb site offset underestimations at other sites. These underestimations are primarily~~
494 ~~due to limitations in representing natural aerosol emissions, transatlantic aerosol transport, particle mixing states, and~~
495 ~~the oxidation levels of organic aerosols.~~ Moreover, a comparison of aerosol speciation measured at Ascension Island
496 during the LASIC campaign indicates that the model consistently underestimates sulfate aerosols. We further
497 evaluated the simulated vertical profile of aerosol mass concentrations and composition against measurements from
498 the ORACLES and CLARIFY campaigns. These comparisons showed that sulfate aerosols were underestimated by
499 19% at cloud-relevant altitudes of 0–2 km by both campaigns. ~~The underestimate of sulfate aerosols at lower altitudes~~
500 ~~(0-2 km), coupled with an underestimate of other aerosols at higher altitudes (4-8 km), likely contributes to the overall~~
501 ~~low bias in modeled AOD. The misrepresentation of natural aerosol emissions and transatlantic aerosol transport may~~
502 ~~be responsible for these underestimates. Nevertheless~~ However, discrepancies increase with altitude, reflecting
503 challenges in accurately modeling high-altitude aerosol concentrations.

504 Analysis of seasonal mean aerosol composition at cloud height showed that organic aerosols predominate during
505 JASO (63%) and MAMJ (51%), while sulfate aerosols are most prevalent (41%) during the austral summer (JFND).
506 Given the prominence of sulfate as a marine sourced aerosol in remote oceanic environments, we investigated the
507 processes influencing the sulfate aerosol concentrations in our domain. Throughout the year, anthropogenic sources
508 and oceanic DMS emissions are the primary atmospheric sulfur contributors. Spatial mapping across the sub-domain
509 ($0\text{--}35^\circ \text{ S}$, $20^\circ \text{ E}\text{--}20^\circ \text{ W}$) showed high sulfate concentrations (up to $3\mu\text{g m}^{-3}$) at cloud height during the peak biomass
510 burning season (JASO), primarily from savannah fires in southern Africa. Despite this, sulfate aerosols only account
511 for 18% of the total aerosol mass in JASO.

512 Sulfate, primarily from marine sources, is the dominant aerosol at cloud-relevant altitudes during JFND in the model
513 (up to 69% marine contribution); however, significant uncertainties regarding the treatment of DMS persist that may
514 affect this finding. To assess the impact of these uncertainties on sulfate aerosols, we compared DMS emission fluxes
515 from Lana et al. (2011) climatological data, ~~which has limited spatial and temporal coverage in our domain, with those~~
516 ~~from and~~ Galí et al. (2018), ~~which are based on~~ satellite-based estimates of surface seawater DMS concentrations. ~~We~~
517 ~~find that, within our domain, the~~ Lana dataset emissions estimates are 51% ~~higher~~ in July and a-38% ~~lower~~ in
518 December ~~compared relative to the~~ Galí ~~dataset~~. Moreover, improvement of DMS chemistry in the model by
519 incorporating new oxidation mechanisms and intermediate products could shift the balance towards increased sulfate
520 aerosol production (with Tashmim et al., 2024 suggesting an increase of up to 35% over the SEA). Additionally,
521 emissions of marine primary organic aerosols during the peak primary production period (JNFD) may substantially
522 contribute to the mass of organic aerosols which can also act as CCN. This emphasizes the critical role of marine

523 sources in influencing aerosol concentrations, even in oceanic regions impacted by large seasonal biomass burning.
 524 Variations in SO₂ emissions from biomass burning could potentially increase sulfate aerosol concentrations at cloud
 525 altitudes by up to 25%. Addressing these discrepancies is essential for improving the model's underestimation of AOD
 526 and aerosol concentrations compared to observations.

527 This study highlights the importance of constraining marine emissions and their chemical transformations by
 528 incorporating satellite-retrieved datasets and extending field campaign efforts during non-biomass burning periods.
 529 Such initiatives are essential to accurately characterize seasonal aerosol dynamics at cloud heights and to improve our
 530 understanding of aerosol-cloud interactions in regions with persistent low-altitude clouds. These advancements could
 531 substantially minimize uncertainties in model estimates of radiative forcing and enhance the reliability of climate
 532 model projections in the southeast Atlantic region.

533

534 **Appendix A**

535 **Table A1:** Configuration of sensitivity analysis simulations

Simulations	Resolution	<u>Emission inventories (References)</u>
Marine sulfur emissions only	0.5° x 0.625°	<u>Standard model inventories (see Section 2.1 in the main text)</u>
DMS emissions	4° x 5°	<u>Climatological product for seawater DMS (Lana et al., 2011); Satellite-derived DMS estimates (Galí et al., 2018)</u>
Biomass burning inventories	4° x 5°	<u>Global Fire Emissions Database (van der Werf et al., 2017); Quick Fire Emissions Dataset (Darmenov & da Silva, 2013); Global Fire Assimilation System (Kaiser et al., 2012).</u>
Marine primary organics	0.5° x 0.625°	<u>Marine primary organic aerosol emission estimates from satellite-derived chlorophyll-a concentrations (Gantt et al., 2015)</u>

536 **Table A2:** AERONET site information and the monthly average AOD₅₅₀ value (± 1 standard deviation) ~~for AOD₅₅₀~~
 537 for three distinct time periods per site are shown. An asterisk (first column) indicates that the mean AOD during the
 538 biomass burning season (JASO) at these sites is more than 1 standard deviation higher than the mean AOD in other
 539 season(s) (JFND and MAMJ).

Site	Latitude (°)	Longitude (°)	Months of data availability for 2017	<u>Monthly Average</u> <u>AOD_{550nm} ± 1 SD</u> <u>(JFND)</u>	<u>Monthly Average</u> <u>AOD_{550nm} ± 1 SD</u> <u>(MAMJ)</u>	<u>Monthly Average</u> <u>AOD_{550nm} ± 1 SD</u> <u>(JASO)</u>
Ascension Island	-7.976	-14.415	7	<u>0.320 ± 0.061</u>	<u>0.256 ± 0.046</u>	<u>0.396 ± 0.078</u>
Gobabeb	-23.562	15.041	12	<u>0.268 ± 0.033</u>	<u>0.254 ± 0.032</u>	<u>0.399 ± 0.115</u>
HESS*	-23.273	16.503	10	<u>0.162 ± 0.024</u>	<u>0.211 ± 0.033</u>	<u>0.368 ± 0.108</u>
Henties_Bay*	-22.095	14.26	3	<u>0.166 ± 0.021</u>	<u>0.138 ± 0.021</u>	<u>0.380 ± 0.114</u>
Lubango*	-14.958	13.445	9	<u>0.120 ± 0.011</u>	<u>0.251 ± 0.011</u>	<u>0.534 ± 0.139</u>
Namibe*	-15.159	12.178	8	<u>0.279 ± 0.028</u>	<u>0.314 ± 0.029</u>	<u>0.689 ± 0.192</u>
Simonstown_IMT*	-34.193	18.446	7	<u>0.246 ± 0.049</u>	<u>0.179 ± 0.018</u>	<u>N/A</u>
Upington	-28.379	21.156	8	<u>0.131 ± 0.019</u>	<u>0.193 ± 0.035</u>	<u>0.321 ± 0.115</u>
Windport*	-19.366	15.483	10	<u>0.249 ± 0.041</u>	<u>0.242 ± 0.037</u>	<u>0.514 ± 0.159</u>

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542 ~~**Table A3:** Seasonal variation of percentage of monthly mean percent contribution of marine sulfate within cloud~~
 543 ~~height~~

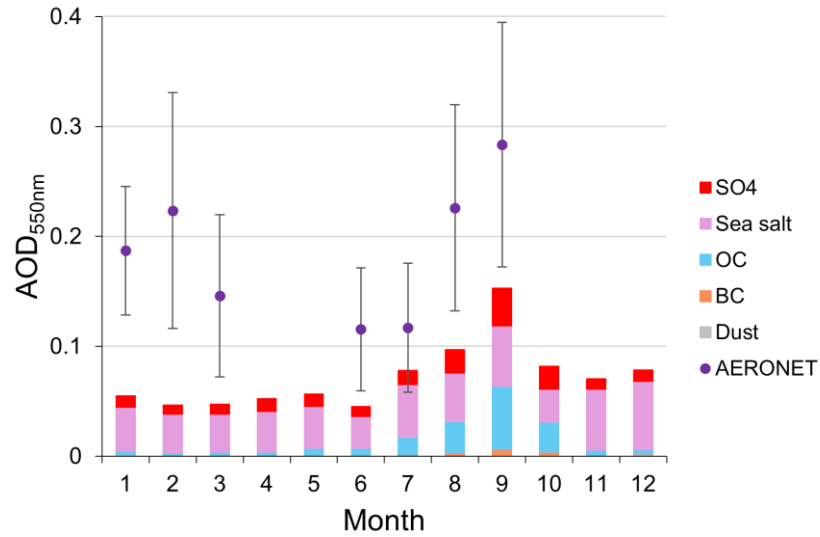
<u>Month</u>	<u>Percentage of marine sulfate</u>
<u>January</u>	<u>57.7</u>
<u>February</u>	<u>54.8</u>
<u>March</u>	<u>25.3</u>
<u>April</u>	<u>26.6</u>

May	15.3
June	15.0
July	14.8
August	14.7
September	10.3
October	22.4
November	39.1
December	44.3

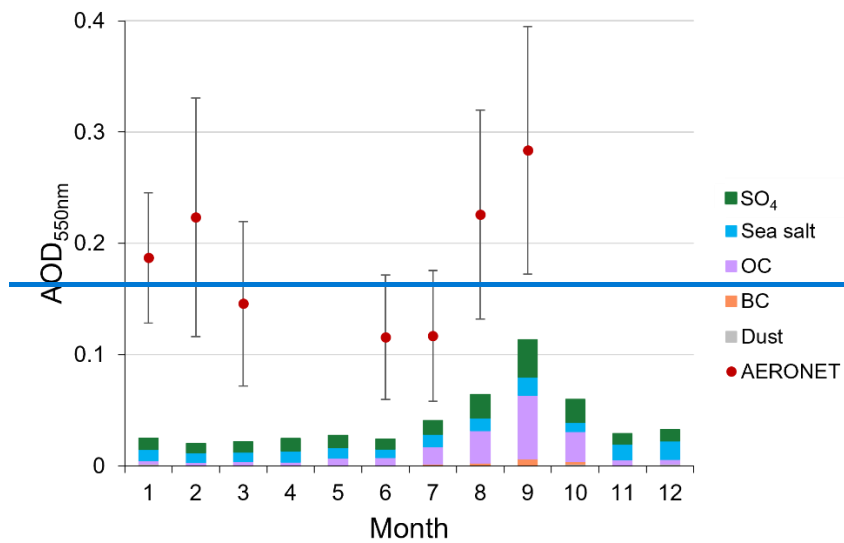
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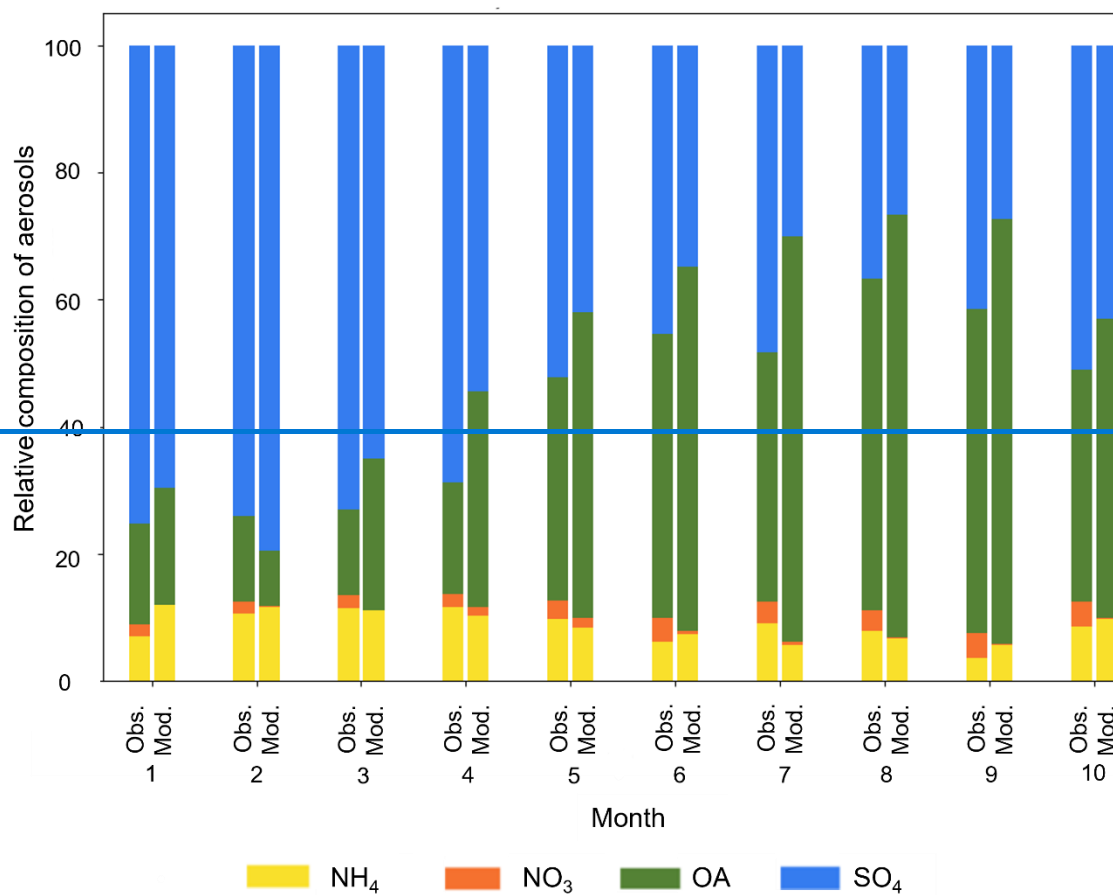


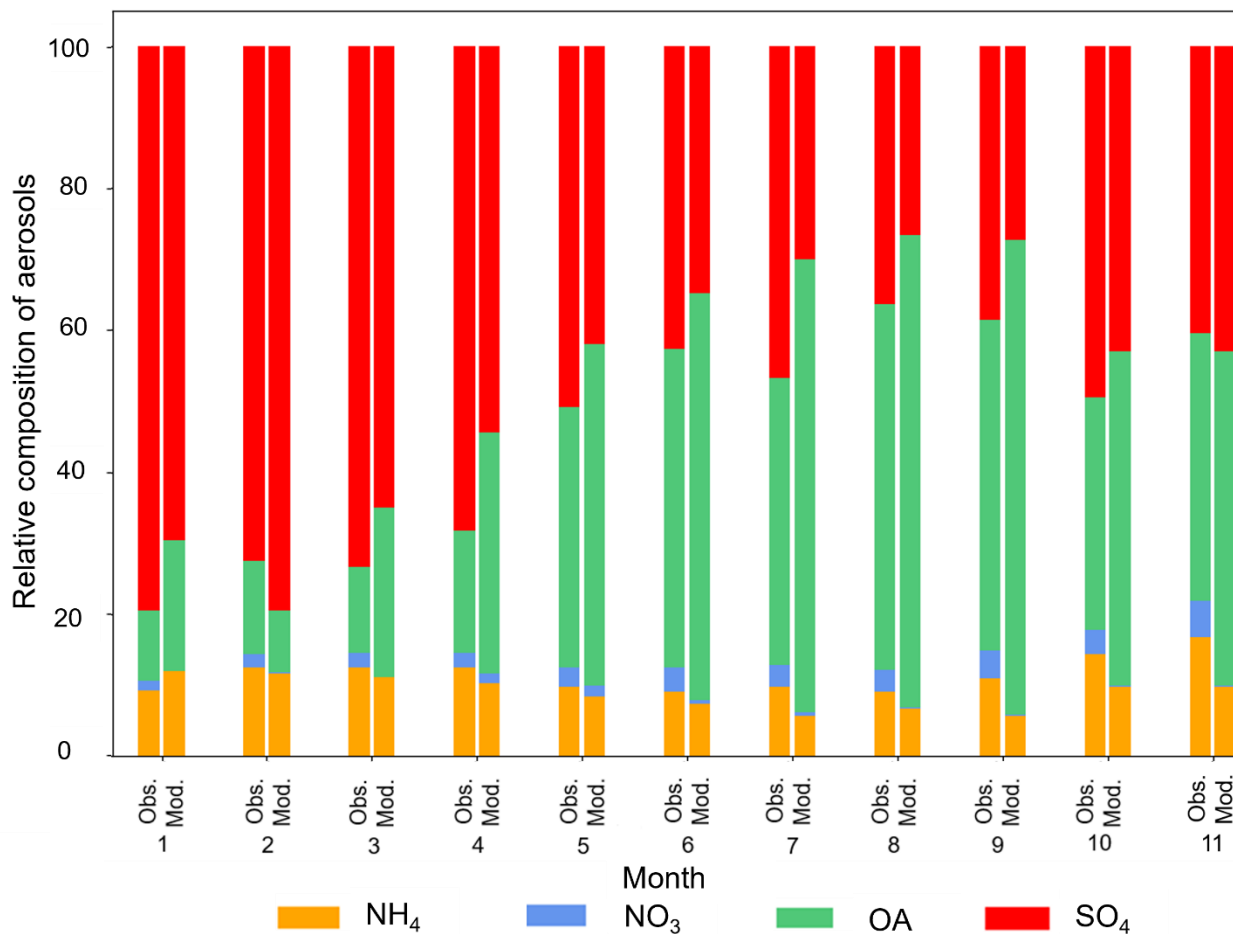
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549 **Figure A1:** Comparative analysis of aerosol optical depth at 550 nm (AOD_{550nm}) for Ascension Island in 2017. The
 550 ~~purple~~ dots present the measured mean monthly AOD values, with vertical error bars illustrating the range of
 551 AOD_{550nm} measurements captured by the AERONET ground station. The stacked bars represent the GEOS-Chem
 552 model's simulated AOD values, with each layer corresponding to the major aerosol components, such as sulfate (SO₄),
 553 sea salt, organic carbon (OC), black carbon (BC), and dust, providing insight into the model's aerosol composition
 554 representation.





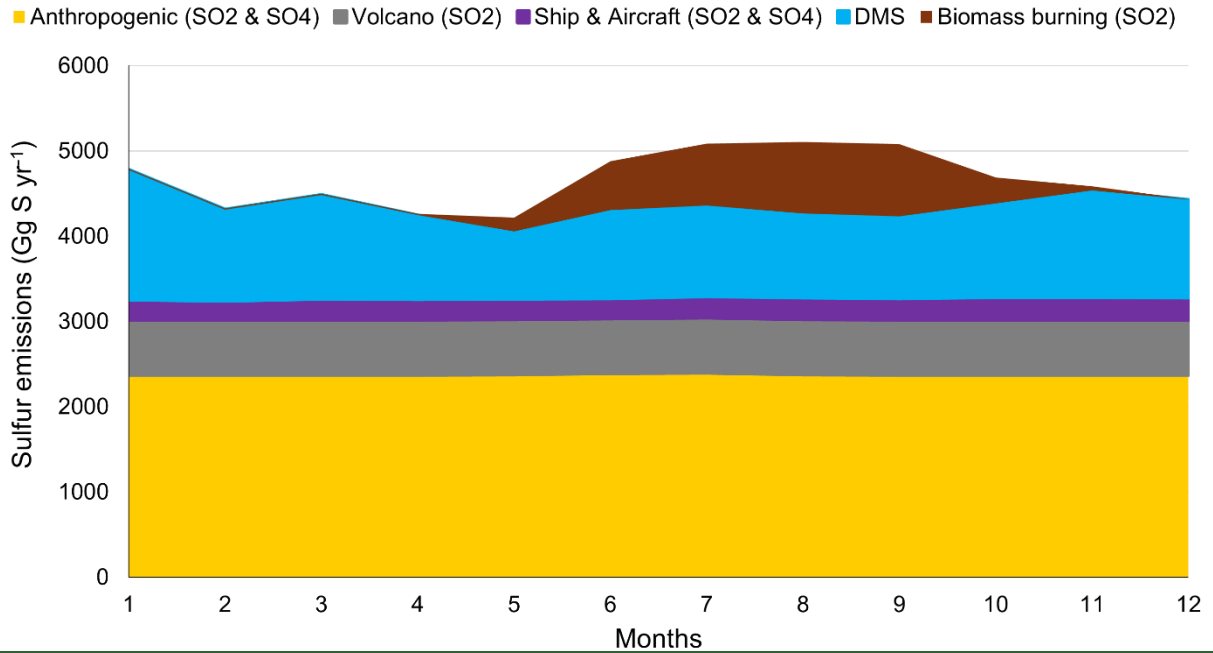
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557 **Figure A2:** Comparative analysis of the relative aerosol composition at Ascension Island in 2017. The stacked bars
 558 on the left depict observations of chemical composition taken during the LASIC campaign at the ARM facility on
 559 Ascension Island, utilizing an aerosol chemical speciation monitor (ACSM) at 341 meters. The bars on the right
 560 illustrate the GEOS-Chem simulated aerosol composition at the same [coordinates/altitude](#). Each segment of the stack
 561 represents different aerosol components: ammonium (NH₄), nitrates (NO₃), organic aerosols (OA), and sulfate
 562 (SO₄).

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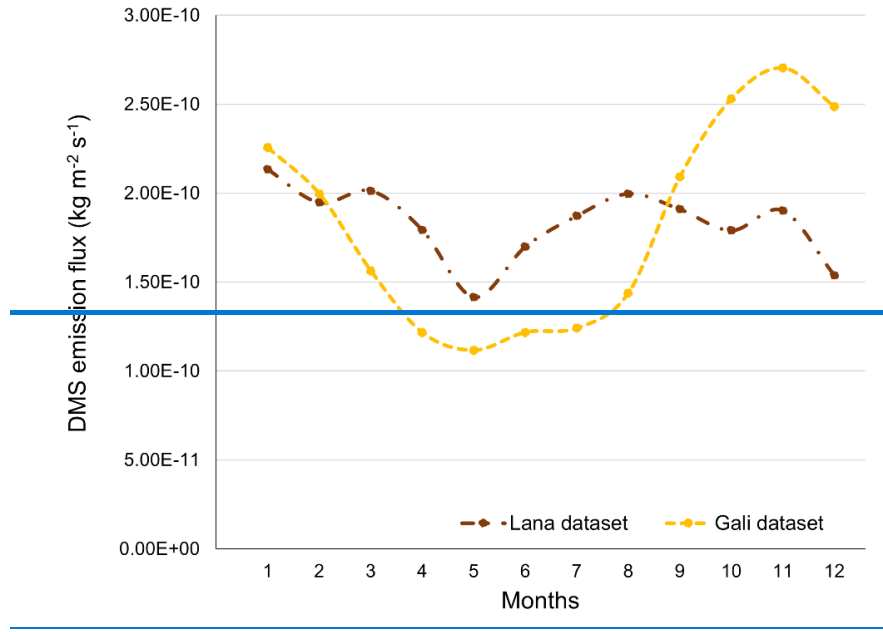
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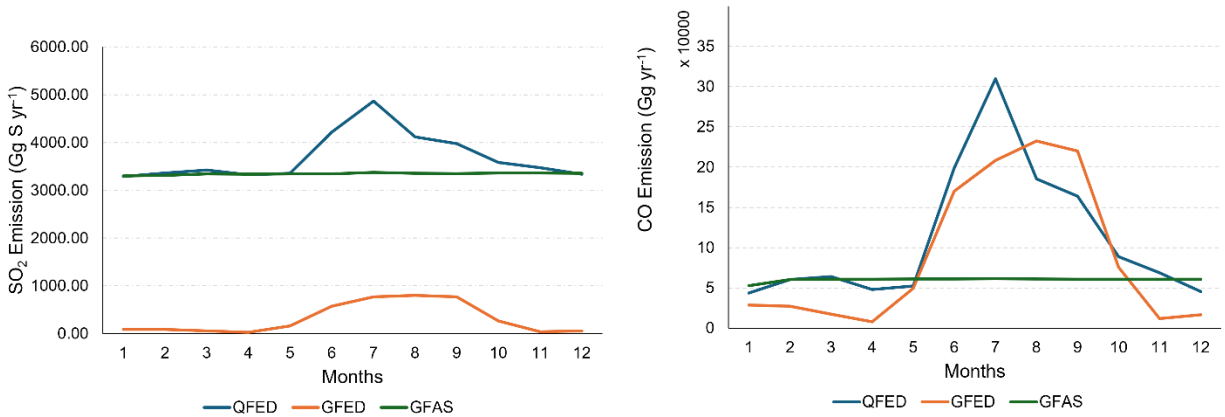
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 567 ~~Figure A3: Stacked area chart of monthly total sulfur emissions by source for 2017 across the study domain (0–40°~~
 568 ~~S, 40° E–20° W) in gigagrams of sulfur per year (Gg S yr⁻¹). Sources are indicated by color and encompass~~
 569 ~~anthropogenic activities, volcanic activity, ship and aircraft emissions, biomass burning and natural emissions of~~
 570 ~~dimethyl sulfide (DMS).~~

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Figure A4: Monthly DMS emissions over the stratocumulus sub-domain (0–35° S, 20° E–20° W) using two distinct datasets for surface seawater DMS concentrations. The brown dashed line presents emissions calculated using Lana et al. (2011) climatology, which compiles data across 1972–2009 from multiple sources. In contrast, the yellow dashed line depicts emissions based on satellite-derived estimates of surface seawater DMS concentrations (Gali et al., 2018).



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Figure A35: Comparison of biomass burning emissions across various inventories, namely GFED, QFED, and GFAS across the domain (0–40° S, 40° E–20° W). The panels depict the interannual variability of biomass burning emissions, with the left panel illustrating sulfur dioxide (SO₂) emissions, and the right panel displaying carbon monoxide (CO) emissions. Both GFED and QFED indicate similar emission trends; however, GFED exhibits lower SO₂ emission magnitudes compared to QFED. GFAS presents emission magnitudes similar to QFED during non-biomass burning period.

Author contributions

MH, HMM, and RMG designed the research. MH conducted the model simulations, analysis, and visualization, with expert advice from HMM. MH drafted the manuscript, which was then revised by all co-authors.

Competing interests

At least one of the (co-)authors is a member of the editorial board of Atmospheric Chemistry and Physics.

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