



Regional and sectoral contributions of NO_x and reactive carbon emission sources to global trends in tropospheric ozone during the 2000–2018 period

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Abstract. Over the past few decades, the tropospheric ozone precursor anthropogenic emissions – nitrogen oxides (NO_x) and reactive carbon (RC) from northern mid-/high-latitude regions (e.g., North America, Europe) – have been decreasing, and those from (sub-)tropical regions (e.g., South Asia, the Middle East) have been increasing, leading to an equatorward emission redistribution. In this study, we quantify the contributions of various sources of NO_x and RC emissions to tropospheric ozone using a source attribution technique during the 2000–2018 period in a global chemistry transport model. We tag the ozone molecules with the source of their NO_x or RC precursor emission in two separate simulations: NO_x-tagged and RC-tagged. These tags include various natural (biogenic, biomass burning, lightning NO_x and RC from methane oxidation) and regional anthropogenic precursor emission sources and influx from the stratosphere. We simulate ~ 336 Tg O₃ of tropospheric ozone burden (TOB) with an increasing trend of 0.91 Tg O₃ yr⁻¹ (0.28 % yr⁻¹), largely contributed (and trend driven) by anthropogenic NO_x emissions and methane oxidation. The ozone production efficiency of regional anthropogenic NO_x emissions increases when emissions decrease (e.g., Europe, North America) and decreases when emissions increase (e.g., South Asia, Middle East, international shipping). Tropical regions, despite lower emissions, contribute more to TOB compared to emissions from higher latitudes, consistent with previous work, predominantly due to large convection (combined with intense sunlight and larger reaction rates) at the tropics, thereby lifting O₃ and its precursor molecules into the free troposphere where ozone's lifetime is longer. We simulate a smaller relative contribution from tropical regions to the global mean surface ozone compared to their contribution to the TOB. The global population-weighted mean ozone is much larger compared to global mean surface ozone, mainly due to large anthropogenic emissions from densely populated regions – East Asia, South Asia and other tropical regions – and a substantial contribution from international ship NO_x emissions. The increasing trends in anthropogenic precursor emissions from these regions are the main drivers of increasing global population-weighted mean ozone.

1 Introduction

Tropospheric ozone is a major greenhouse gas (Myhre et al., 2013), after carbon dioxide and methane. It also has an indirect influence on the lifetime of methane through its impact on the hydroxyl radical, OH (Monks et al., 2015). At the surface, ozone is a major air pollutant, causing a large number of premature deaths ($\sim 423\,100$ ozone-attributable deaths [95 % confidence interval: 223 200, 659 400] in 2019; Malashock et al., 2022) and heavy yield losses in the production of staple food crops worldwide (losses of up to 79–121 Tg annually; Avnery et al., 2011; Mills et al., 2018). Tropospheric ozone, initially thought to be mostly of stratospheric origin (Junge, 1962), is also a secondary pollutant produced from the reaction between ozone precursor gases: nitrogen oxides (NO_x: NO and NO₂) and reactive carbon (RC: CH₄, CO and volatile organic compounds (VOCs); Levy, 1972; Chameides and Walker, 1973; Crutzen, 1974). A multi-model assessment study showed that the tropospheric ozone burden has increased by ~ 44 % since the pre-industrial period, mainly driven by increasing anthropogenic emissions of ozone precursor gases (Griffiths et al., 2021). Gaudel et al. (2020) utilize the data from the In-service Aircraft for a Global Observing System (IAGOS) database and identify a remarkable increase in tropospheric ozone since 1994 over several regions in the Northern Hemisphere. Addressing the effects of tropospheric ozone both as a pollutant and a greenhouse gas requires mitigation of the anthropogenic emissions of ozone precursors.

Since the 1980s, anthropogenic emissions of tropospheric ozone precursors from North America and western Europe have decreased in response to the implementation of control measures, while emissions from Asia, Central America and eastern Europe have increased due to economic and population growth (Granier et al., 2011; Cooper et al., 2014), leading to an equatorward shift in global emission pattern. A recent study (Li et al., 2024) confirms an equatorward shift in Asian NO_x emissions between 2010 and 2017 according to bottom-up estimates. Zhang et al. (2016) found that this equatorward shift in precursor emissions is the dominant factor, compared to the change in magnitude of emissions and methane concentration, which has led to an increase in the tropospheric ozone burden between 1980 and 2010. This is mainly due to larger convection of polluted air masses from the boundary layer into the free troposphere, in addition to larger reaction rates and NO_x sensitivity over the tropical regions (Wild et al., 2001) compared to extratropical regions. Further, Wang et al. (2022) show that increasing aircraft emissions play a major role in the increasing trend in tropospheric ozone burden over the 1995–2017 period. Similarly, the increase in methane concentration during the industrial period also had a substantial impact on tropospheric ozone burden (Stevenson et al., 2020).

Source apportionment methods are used in chemical transport modeling studies to quantify the influence of specific

emission sources on ozone concentration at any given location. Typically, global modeling studies have used two different methods for source apportionment: perturbation and tagging, which will be described in more detail below.

Several previous studies have used the perturbation method of source attribution to study the impact of emission changes from a particular region/sector on tropospheric ozone (e.g. West et al., 2007; Fiore et al., 2009; Jonson et al., 2018). Zhang et al. (2021) use this method to study the influence of changing emissions from various world regions between 1980 and 2010 on tropospheric ozone burden. In their study, multiple model simulations were performed: one base simulation with all emissions at 2010 levels and other perturbed simulations with 1980 emissions from the region/sector of interest. The study found that the change in tropospheric ozone burden responded the most to perturbation in prescribed methane concentration. Of the geographical source regions, tropospheric ozone burden responded the most to perturbations in emissions of anthropogenic ozone precursor emissions at tropical regions: South Asian and Southeast Asian regions.

The second method of source apportionment is called tagging, which involves labeling the modeled ozone molecules with the name of its precursor emission sources. As opposed to the previously discussed perturbation method that delivers the sensitivity of simulated ozone to emission changes, this tagging method delivers the exact contribution to the simulated ozone concentration from each of the tagged emission sources (see Table 1 in Mertens et al., 2020, for a detailed comparison between the two methods). Several previous studies have implemented the tagging method in box (e.g., Butler et al., 2011; Coates and Butler, 2015), regional (Kwok et al., 2015; Lupascu and Butler, 2019; Zhao et al., 2022) and global models (Grewé et al., 2010, 2017; Emmons et al., 2012; Butler et al., 2018) to study the contribution of various emission sources on ozone concentrations (see Butler et al., 2018, for a review of various tagging approaches). The formation of one tropospheric ozone molecule requires two precursors (NO_x and RC); a major challenge in its emission source attribution is knowing the label of which of the precursors the ozone molecule should inherit.

Among the previous global modeling approaches that implement tagging, Grewé et al. (2010, 2017) calculate the production rate of the tagged ozone molecules, by giving equal weight to the proportion of tagged NO_x and RC molecules emitted from a given sector (“combinatorial” tagging). While this approach can deliver information on the role of each emission source in its contribution to the tropospheric ozone concentration, it does not communicate the contrasting roles of NO_x and RC from those sources. For example, Mertens et al. (2018), using the combinatorial approach, attribute ~ 8 % of tropospheric ozone burden to ozone precursor emissions from land transportation but cannot determine which emitted precursor from land transportation: NO_x or RC contributes predominantly. Butler et al. (2018) formulated the

tropospheric ozone source attribution system with tagging (TOAST) within the CAM4-chem (Community Atmosphere Model version 4 with chemistry) model that requires running two separate simulations – NO_x-tagged and RC-tagged – to separately attribute ozone to NO_x and RC sources, respectively, by labeling them with the name of their originating source region/sector. With this approach, the previously mentioned problems related to the contrasting roles of NO_x and RC are avoided as the contribution from each precursor source is separately attributed. To illustrate, O₃ attributed to biogenic emissions in the NO_x-tagged simulation would clearly mean that the contribution is from emitted biogenic NO_x only, irrespective of where the RC comes from to produce the O₃ molecule. Contribution from biogenic RC emissions can be simulated in the RC-tagged simulation. In this way, the roles of NO_x and RC emissions from a given sector are exclusively simulated in separate simulations.

Butler et al. (2020) use the TOAST tagging technique to quantify the contribution of ozone precursor emissions for the year 2010 from several regions/sectors. They demonstrated that the ozone production efficiency can directly be calculated using this method, as the ratio of tropospheric ozone attributed to a tagged emission source to the amount of precursor emission from that source. NO_x emissions from tropical regions such as South Asia, Southeast Asia and Central America were found to be the most efficient at producing tropospheric ozone compared to the emissions from other regions, consistent with the earlier work of Zhang et al. (2016). They further showed, using a methane perturbation introduced to NO_x-tagged simulation, that the contribution to tropospheric ozone burden by NO_x emitted from international shipping increases especially strongly in response to changes in methane concentration.

While Butler et al. (2020) discuss the ozone precursor contributions to tropospheric ozone burden for one year – 2010 – we use the TOAST tagging approach (further explained in Sect. 2) in this study over the 2000–2018 period to answer the following questions: (i) In Sects. 3.1 and 3.2, what is the contribution of precursor emissions from various regions/sectors to the global tropospheric O₃ burden? How do changes in these emissions during 2000–2018 affect their contribution to the trend in tropospheric ozone burden? (ii) In Sect. 3.3, how does the ozone production efficiency (OPE) of ozone precursor emissions respond to the changes in these emissions during the 2000–2018 period? And (iii) in Sect. 3.4, how do contributions of different ozone precursors to the tropospheric O₃ burden contrast with their contribution to global mean surface O₃ and population-weighted O₃? We conclude our article with a summary, limitations and future scope in Sect. 4.

2 Methods

2.1 Simulation setup

We use the ozone source attribution methods described in Butler et al. (2018) and Butler et al. (2020). We perform simulations with CAM4-chem, which is a component of the CESM (Community Earth System Model) version 1.2.2 (Lamarque et al., 2012). The model is run at a horizontal resolution of $1.9^{\circ} \times 2.5^{\circ}$, with 56 vertical levels for the 2000–2018 period with specified dynamics from MERRA2 reanalysis (Molod et al., 2015). The temperature, horizontal winds, and sensible and latent heat fluxes from MERRA2 reanalysis dataset are nudged every time step (30 min) by 10 % towards analysis fields (i.e., a 5 h Newtonian relaxation timescale for nudging).

Anthropogenic emissions of NO_x and non-methane reactive carbon (NMRC: CO and volatile organic compounds (VOCs) collectively), including land-based emissions, international shipping emissions and aircraft emissions, are taken from Hemispheric Transport of Air Pollution version 3 (HTAPv3; Crippa et al., 2023) emissions inventory. We specify aircraft emissions at various altitudes effectively representing three different flight phases (landing/take-off, ascent/descent and cruising). Biomass burning emissions are taken from GFED-v4 inventory (van der Werf et al., 2010). The biogenic NMRC emissions are taken from CAMS-GLOB-BIO-v3.0 (Sindelarova et al., 2021), and biogenic NO_x (from soil) is prescribed as in Tilmes et al. (2015). While we interpolate the emissions from HTAPv3's high-resolution dataset to our coarser model resolution, we also correct the interpolated land-based emissions over ocean grid cells by moving them to the nearest land grid cell (vice versa for ocean-based emissions) to make sure that the emissions are allocated to the correct region for the source attribution.

We impose methane concentration as a surface boundary condition. The methane concentration is taken from the 2000–2018 average mole fraction fields from the CAMS CH₄ flux inversion product v18r1 (<https://ads.atmosphere.copernicus.eu/datasets/cams-global-greenhouse-gas-inversion?tab=overview>, last access: 7 May 2025) and is specified as a zonally and monthly varying transient boundary condition. As in Butler et al. (2018) and Butler et al. (2020), we use the MOZART-4 chemical mechanism (Emmons et al., 2012) further modified to include tagged ozone tracers. This tagging system allows attribution of tropospheric ozone to reactive nitrogen (NO_x) and reactive carbon (CH₄ and NMRC) precursors in two separate simulations. Additionally, stratospheric influx and other minor production pathways of tropospheric ozone are also tagged in our model. For a complete attribution, we perform two separate simulations: (i) NO_x-tagged and (ii) RC-tagged (reactive carbon tagged) with their respective tagged emission sources.

We specify separate tag identities for biogenic, biomass burning, aircraft and surface-based anthropogenic emission sources of ozone precursors and for ozone from production in the stratosphere. We tag anthropogenic emissions according to the HTAP Tier 1 regions (Galmarini et al., 2017) from which the respective NO_x or NMRCs are emitted. We focus our study on the northern hemispheric anthropogenic emissions by individually tagging the major northern hemispheric regions (East Asia, South Asia, North America, Europe, the Middle East and the Russia–Belarus–Ukraine region) and combining the other regions under the “rest of the world” tag. Within our “rest of the world” tag, some regions (e.g. Central Asia, Southeast Asia) have been explicitly tagged in our NO_x-tagged simulation but not in our RC-tagged simulation. We also specify additional tags for emissions from international shipping and aircraft emissions (see Table 1 and Fig. S1 in the Supplement for a summary of tags used in this study).

We specify an additional tag for NO_x emission from lightning in our NO_x-tagged simulation and for RC from methane oxidation in our RC-tagged simulation. In both NO_x and RC tagged simulations, the sum of tagged ozone tracers is equal to the total ozone simulated by the model.

2.2 Model evaluation

CAM4-chem has been evaluated in previous studies for simulating tropospheric ozone and precursors (Lamarque et al., 2012; Tilmes et al., 2015) and was also evaluated in its ozone-tagged configuration by Butler et al. (2018) and Butler et al. (2020). Here we evaluate the configuration of CAM4-chem used in this study, which primarily differs from the earlier work through its use of the HTAPv3 global emissions. In Fig. 1 we compare our simulated surface ozone against the gridded observation dataset provided by Tropospheric Ozone Assessment Report (TOAR; Schultz et al., 2017) until year 2014. We also make the comparison using simulation output from the “CESM2-WACCM6” model, which is part of the CMIP6 ensemble (Emmons et al., 2020), using this as a standard reference model. Results are shown as monthly averages over various HTAP Tier 2 regions (Galmarini et al., 2017) at the grid cells where the TOAR data are available. We overestimate the surface ozone mixing ratio by up to ~ 4–12 ppbv over most regions where TOAR data are available, largely during the summer months. This overestimation is also simulated by the CESM2-WACCM6 model (shown in Fig. 1) and is consistent with high model bias of ~ 7 ppbv simulated by most models in various ensembles, as discussed in Young et al. (2018). We also simulate an underestimation of up to 2–6 ppbv over the northern and eastern United States during winter months, which is not simulated by CESM2-WACCM6. For the year 2010, we simulate slightly smaller surface ozone at most regions over grid cells where TOAR data are available compared to that simulated by Butler et

al. (2020), and it is closer to the observations from TOAR dataset.

We evaluate free tropospheric ozone against the ozone-sonde-based climatology compiled by Tilmes et al. (2012). Figure 2 shows the comparison of our simulated climatology (2000–2010) to the ozone sonde climatology (1995–2010) in Taylor-like diagrams (Taylor, 2001; Tilmes et al., 2012) at 900, 500 and 250 hPa pressure levels, grouped by latitude ranges. The correlation coefficient between the observed and simulated monthly regional O₃ average is usually more than 0.8, and the fractional mean difference is usually within 25 % at most regions in the troposphere. The model captures the vertical distribution of ozone derived from ozone sonde climatology very well, albeit slightly biased high particularly in the upper troposphere at most sites (Fig. S2 in the Supplement). Our free troposphere evaluation results are therefore largely consistent with previously evaluated versions of CAM4-chem (Tilmes et al., 2012; Zhang et al., 2016; Emmons et al., 2020).

2.3 Trend analysis

We express the trend for the 2000–2018 period in various quantities discussed in our study: precursor emissions, tropospheric ozone burden, global surface mean and population-weighted mean ozone concentration as a slope of the time series (in respective units yr^{−1} and in % yr^{−1} relative to the first absolute value (for year 2000)). This slope is calculated using the Theil–Sen estimator method available as a Python module (<https://docs.scipy.org/doc/scipy/reference/generated/scipy.stats.theilslopes.html>, last access: 7 May 2025), which also calculates the 95 % confidence interval of the slope. We further use the Mann–Kendall test to determine the extent to which the detected trend is monotonic (Hussain and Mahmud, 2019), expressed by *p*. In the TOAR guidelines for best statistical practices (Chang et al., 2023), it is recommended not to use a dichotomized expression such as significant/insignificant trend. Hence, we also provide *p* and the 95 % confidence interval in the Supplement for interested readers to assess the meaningfulness of the estimated trend in our study. For the sake of discussion in Sects. 3 and 4, we categorize the trends with *p* values less than 0.05 to be of high certainty, between 0.05 and 0.1 to be of medium certainty, and larger than 0.1 to be of low certainty (Table 3 in Chang et al., 2023).

3 Results

We discuss our results related to the ozone precursor emissions, contribution to tropospheric ozone burden, ozone production efficiency, and contribution to surface mean and population-weighted mean ozone from each of the tags used in our study. Each subsection is further divided into two parts to discuss the role of NO_x- and RC-tagged contributions separately. All our results are discussed as time series of annual

Table 1. List of tags used in this study. See Fig. S1 for a map of tagged regions considered in this study.

Tag name	NO _x -tagged	RC-tagged
HTAP Tier 1 regions		
International shipping	Explicit	Explicit
North America	Explicit	Explicit
Europe	Explicit	Explicit
East Asia	Explicit	Explicit
South Asia	Explicit	Explicit
Russia, Belarus, Ukraine	Explicit	Explicit
Middle East	Explicit	Explicit
Southeast Asia	Explicit ¹ → RoW	RoW
North Africa	Explicit ¹ → RoW	RoW
Mexico and Central America	Explicit ¹ → RoW	RoW
Central Asia	Explicit ¹ → RoW	RoW
Pacific, Australia and New Zealand	Southern Hemisphere regions ¹ → RoW	RoW
Southern Africa	Southern Hemisphere regions ¹ → RoW	RoW
South America	Southern Hemisphere regions ¹ → RoW	RoW
Antarctica	Southern Hemisphere regions ¹ → RoW	RoW
Arctic	Shipping	Shipping
Other tags		
Stratosphere ²	Global	Global
Aircraft	Global	Global
Biogenic	Global	Global
Biomass burning	Global	Global
NO _x from lightning	Global	N/A
RC from methane oxidation	N/A	Global
Extra production	Global	Global

¹ Although explicitly tagged in the NO_x-tagged simulation, these regions have been lumped into the “rest of the world” tag (RoW) used in this paper.

² NO_x-tagged simulation attributes ozone to influx from the stratosphere (same as RC-tagged) and also to NO produced from oxidation of N₂O.

means for the 2000–2018 period, unless otherwise specified. When referring to anthropogenic emissions and their contributions to tropospheric ozone in our discussion, we only refer to surface-based anthropogenic emissions and exclude aircraft emissions.

3.1 Ozone precursor emission trends

3.1.1 NO_x emissions

Table 2 shows the global NO_x emissions (averaged over the 2000–2018 period) from various regions and sectors and their relative contribution (in %) to the total NO_x emissions. Figure 3a and c show the time series of NO_x emissions from all sources: natural/global sources (a) and regional anthropogenic sources (c).

Over the period 2000–2018, the average annual NO_x emissions are $\sim 49 \text{ Tg N yr}^{-1}$, which includes both surface and aloft emissions. There is an increasing trend with medium certainty in the total NO_x emissions over the 2000–2018 period, with $\sim 46.6 \text{ Tg N yr}^{-1}$ emitted in 2000, peaking at $\sim$

$51.3 \text{ Tg N yr}^{-1}$ in 2011 and decreasing to $\sim 48.2 \text{ Tg N yr}^{-1}$ emitted in 2018 (Fig. 3a; Table 3).

The trend in total NO_x emissions is mainly driven by the anthropogenic NO_x emissions, which are the largest source of NO_x ($\sim 35.6 \text{ Tg N yr}^{-1}$ (72.5 %); Table 2), much larger than natural sources such as lightning, biomass burning and biogenic emissions, which contribute $\sim 7.1 \%$, 8.5% and 10% , respectively (Table 2). We also notice a slight decrease in the global anthropogenic NO_x emissions in 2008–2009, mostly related to the global financial crisis (Schneider and van der A, 2012).

Among the anthropogenic NO_x emissions, East Asian emissions are the largest ($\sim 18.5 \%$; Table 2), with an increasing trend with high certainty ($\sim 0.25 \text{ Tg N yr}^{-2}$; Table 3 and Fig. 3) over the 2000–2018 period. These emissions increase at $0.49 \text{ Tg N yr}^{-2}$, peak in 2011 and start declining after that at $-0.44 \text{ Tg N yr}^{-2}$ (Zheng et al., 2018), as shown in Fig. 3c and Table S2 in the Supplement, which largely explain the time series of anthropogenic and total NO_x emissions time series shown in Fig. 3a. The NO_x emissions increase over South Asia ($0.13 \text{ Tg N yr}^{-2}$), the

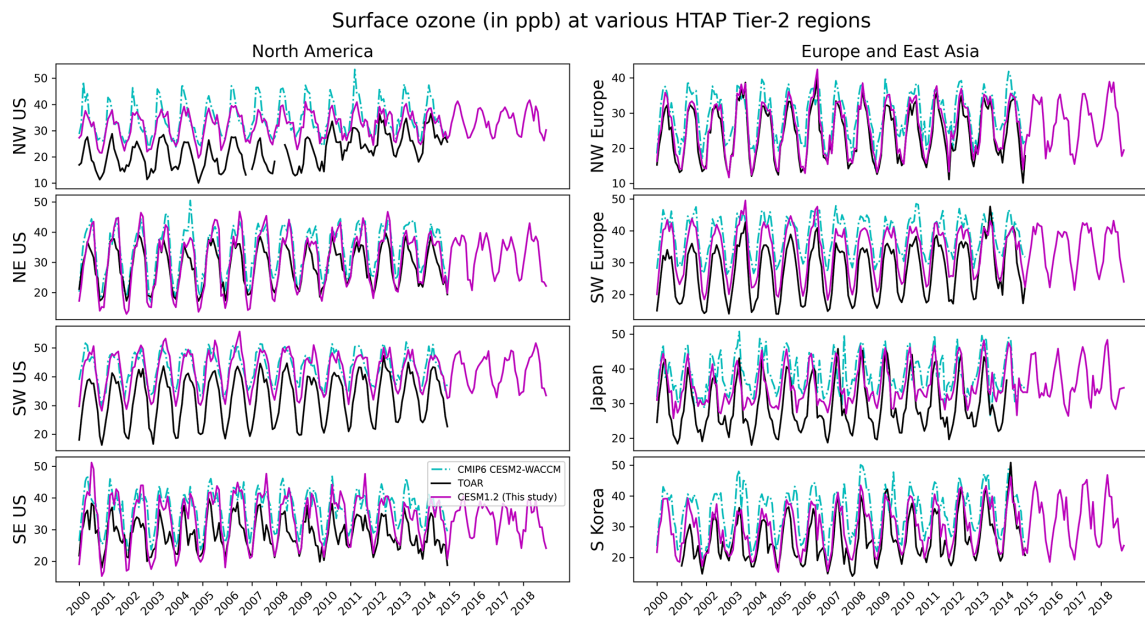


Figure 1. Comparing the time series of monthly mean surface ozone (in ppb) simulated over the 2000–2018 period in this study with the gridded observation dataset from TOAR available until 2014 (<https://toar-data.fz-juelich.de/>, last access: 7 May 2025), in various regions defined by HTAP2 (Galmarini et al., 2017). The comparison is performed for simulated surface ozone values only over those grid cells where the TOAR data are available. Also shown is the comparison with the CESM2-WACCM6 model, which is a CMIP6 ensemble member (also until 2014).

Table 2. Attribution of O₃ to NO_x-tagged precursor emissions. The 2000–2018 mean contribution from NO_x-tagged components to various metrics: precursor emissions, tropospheric ozone burden, ozone production efficiency, grid-cell-area-weighted and population-weighted global mean surface ozone. In brackets is the percentage contribution to the total, wherever applicable.

Source	Emissions (Tg N yr ^{−1})	Tropospheric O ₃ burden (Tg O ₃)	OPE (mol O ₃ (mol N) ^{−1})	Surface mean (ppbv)	Population-weighted mean (ppbv)
Total	49.12	317.93	N/A	25.39	33.01
Stratosphere	N/A	91.50 (28.78)	N/A	4.99 (19.65)	3.45 (10.44)
Aircraft	0.93 (1.90)	13.44 (4.23)	4.19	0.99 (3.91)	1.18 (3.59)
Biogenic	4.94 (10.05)	23.67 (7.44)	1.40	2.29 (9.01)	3.43 (10.40)
Biomass burning	4.18 (8.52)	12.68 (3.99)	0.88	1.32 (5.20)	1.59 (4.82)
Lightning	3.47 (7.06)	70.51 (22.18)	5.93	3.51 (13.82)	2.86 (8.68)
Extra production	N/A	0.97 (0.31)	N/A	0.08 (0.33)	0.11 (0.33)
Anthropogenic	35.60 (72.47)	116.94 (36.78)	0.96	12.20 (48.07)	20.38 (61.74)
Regional contribution to anthropogenic component					
International shipping	4.04 (8.22)	23.02 (7.24)	1.67	2.97 (11.69)	2.47 (7.47)
North America	5.51 (11.22)	9.79 (3.08)	0.54	1.43 (5.64)	1.61 (4.87)
Europe	3.54 (7.20)	4.36 (1.37)	0.36	0.84 (3.32)	1.47 (4.46)
East Asia	9.05 (18.43)	16.62 (5.23)	0.54	1.58 (6.23)	4.27 (12.92)
South Asia	2.91 (5.92)	7.64 (2.40)	0.78	0.62 (2.44)	4.60 (13.93)
Russia, Belarus, Ukraine	1.67 (3.39)	2.30 (0.72)	0.40	0.51 (2.01)	0.53 (1.62)
Middle East	1.68 (3.42)	3.64 (1.14)	0.64	0.49 (1.95)	1.15 (3.50)
Rest of the world	7.20 (14.66)	44.20 (13.90)	1.80	3.76 (14.80)	4.28 (12.97)

Middle East (0.06 Tg N yr^{−2}) and the “rest of the world” (0.15 Tg N yr^{−2}), whereas they decrease over North America (−0.33 Tg N yr^{−2}), the Russia–Belarus–Ukraine region (−0.02 Tg N yr^{−2}) and Europe (−0.11 Tg N yr^{−2}). The mag-

nitude of international ship NO_x emissions (4.04 Tg N yr^{−1}) is comparable to that of NO_x emissions from continental regions such as North America, Europe and South Asia (Table 2) and have an increasing trend (∼ 0.08 Tg N yr^{−2}; Ta-

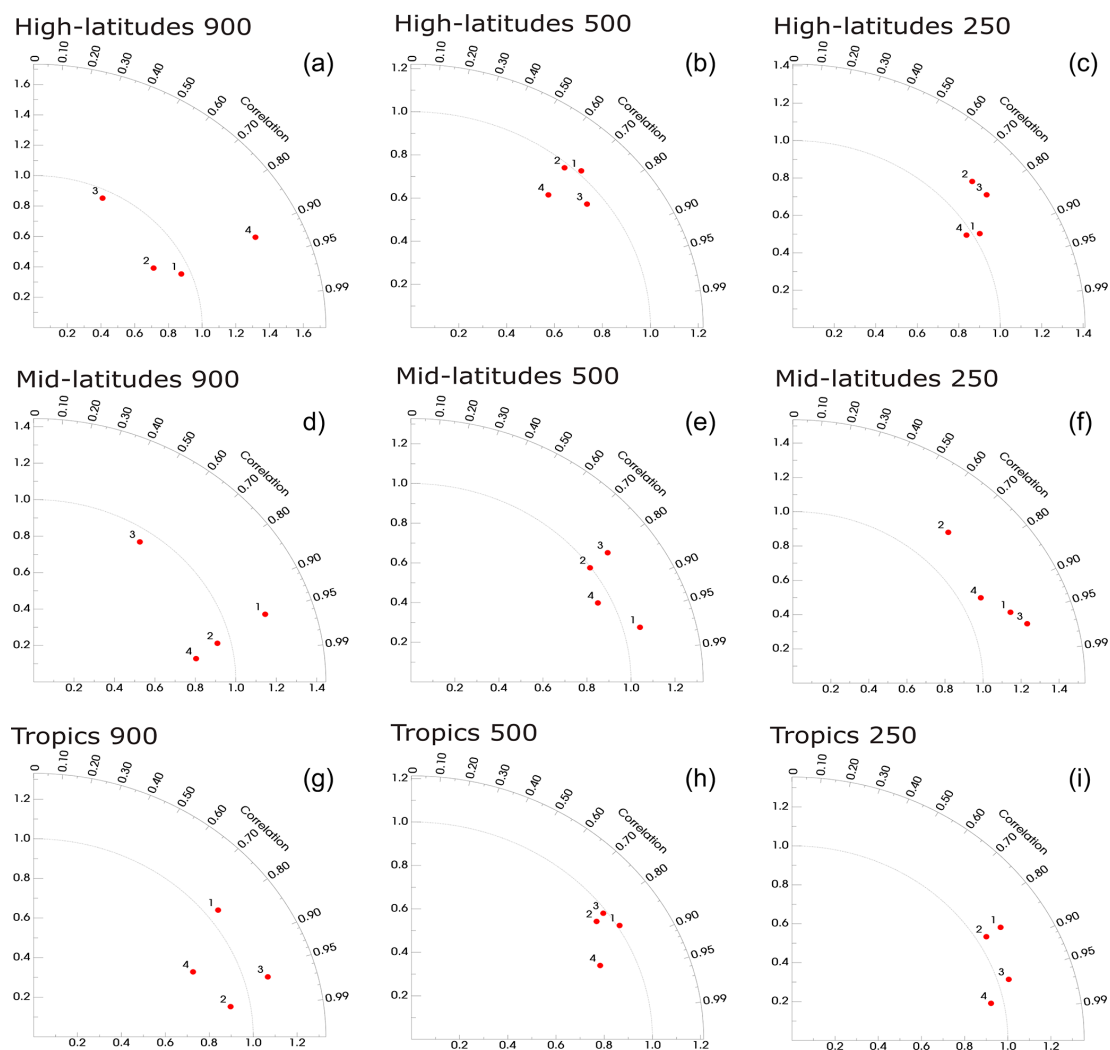


Figure 2. Taylor diagram of comparisons between modeled monthly mean ozone climatology (2000–2018) and ozone sonde climatology (1995–2010) data from Tilmes et al. (2012) in the high tropics (**g–i**), midlatitudes (**d–f**) and high latitudes (**a–c**) for three different altitude levels (900, 500 and 250 hPa) in the troposphere. The x axis shows the relative ozone-normalized bias of the simulations compared to the observations, whereas the radius in the y axis describes the correlation coefficient of seasonally averaged ozone values between simulated and observed values. Numbers indicate different regions as the difference in previous studies (Zhang et al., 2016; Emmons et al., 2020) (**g–i**) 1 – NH subtropics, 2 – western Pacific–eastern India Ocean, 3 – equatorial America, 4 – Atlantic/Africa; (**d–f**) 1 – western Europe, 2 – eastern USA, 3 – Japan, 4 – SH midlatitudes; (**a–c**) 1 – NH polar west, 2 – NH polar east, 3 – Canada, 4 – SH polar.

ble 3) with high certainty. These trends effectively also indicate an overall equatorward shift in anthropogenic emissions as discussed in several previous studies (e.g., Zhang et al., 2016, 2021; Gaudel et al., 2020). We further illustrate this equatorward shift using zonal sum profiles of deviation in anthropogenic NO_x emissions from year 2000 (Fig. S4a in the Supplement). The zonal NO_x emissions relative to year 2000 show a positive (negative) deviation south (north) of $\sim 35^\circ$ N latitude, indicating an equatorward shift in the overall global NO_x emissions.

Emissions from the aircraft sector contribute only 0.93 Tg N ($\sim 1.90\%$) to the total NO_x emissions, with an

increasing trend of $\sim 2.5\% \text{ yr}^{-1}$. This is comparable to the emissions taken from the CEDS inventory reported by Wang et al. (2022): 0.88 Tg N (contributing $\sim 3.3\%$ to total NO_x emissions) increasing at $\sim 3.46\% \text{ yr}^{-1}$ between 1995 and 2017.

3.1.2 Reactive carbon emissions

The total amount of tropospheric reactive carbon emissions averaged over the 2000–2018 period in our simulations is $\sim 1342 \text{ Tg C yr}^{-1}$ (Table 4) with a small increasing trend ($0.25\% \text{ yr}^{-2}$; Table 5) (Fig. 3b). Methane and biogenic NM-

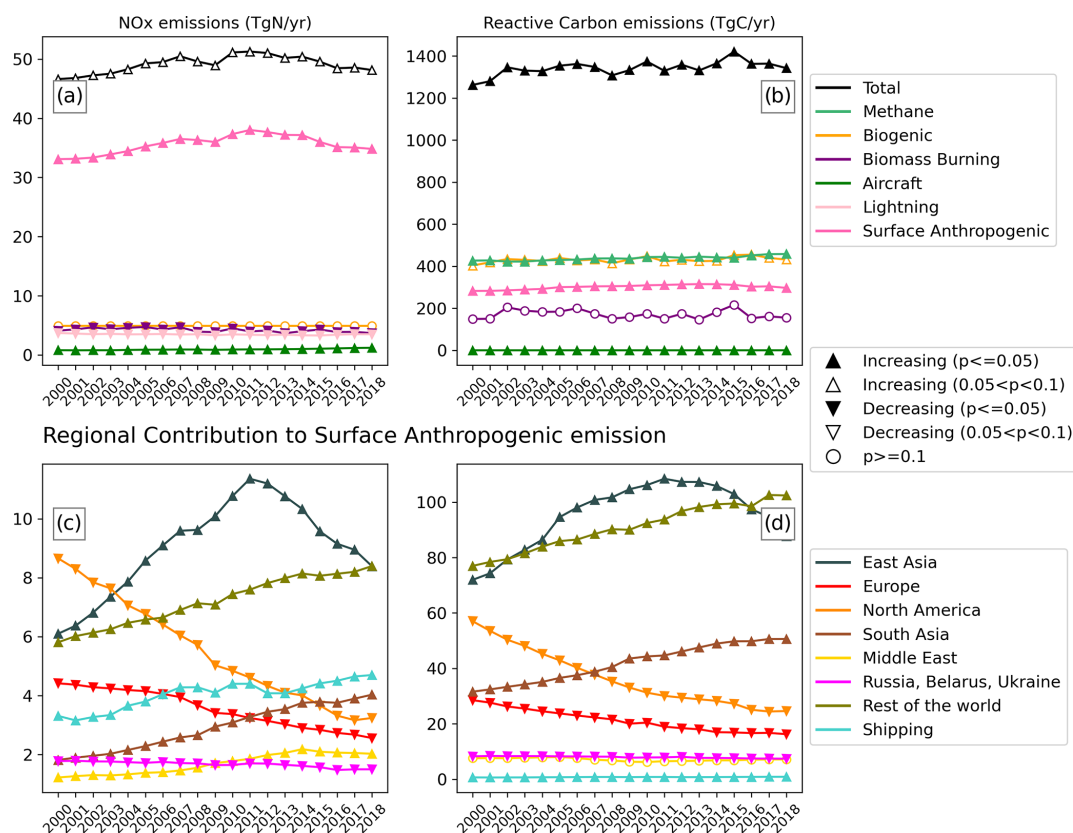


Figure 3. Global annual emissions of ozone precursors from various sources used in this study. (a, c) NO_x emissions (in TgNyr^{-1}), (b, d) reactive carbon emissions (TgCyr^{-1}). Panels (c) and (d) are the regional contributions to the total anthropogenic emission (dark-pink line) shown in (b) and (b). The symbols indicate the sign and certainty of trend in the plotted quantity. The time series of relative contributions (in %) from each of the emission sectors to the total emissions is provided in Fig. S3 in the Supplement.

RCs are the largest sources of reactive carbon (each about $\sim 430 \text{ TgCyr}^{-1}$ (32 %); Table 4; Fig. 3b; Heald and Kroll, 2020). Total anthropogenic NMRC emissions contribute $\sim 302 \text{ TgCyr}^{-1}$ (~ 22.5 %; Table 4), and biomass burning emissions contribute $\sim 171 \text{ TgCyr}^{-1}$ (~ 12.8 %). Aircraft emissions contribute very small amounts (Table 4).

Since the concentration of methane is fixed at the surface in our simulations, we consider the methane oxidation rate derived from methane's oxidation reaction with atmospheric hydroxyl radical as the effective source of RC for subsequent ozone production through rapid chemical reactions. We simulate $\sim 438 \text{ TgCyr}^{-1}$ of methane oxidation with an increasing trend of high certainty over the 2000–2018 period. The prescribed CH_4 concentrations are such that there is a plateau until the year 2006 (Lan et al., 2024) followed by a steep increase (Fig. 4a: red line). Our simulated methane oxidation rate also increases with high certainty but does not strictly follow this plateau followed by steep increase pattern and has large inter-annual variability (Fig. 4a). This could possibly be due to variability in the prescribed meteorology and OH concentration (Fig. 4b) resulting in variations in the oxidation rate of methane. We simulate slightly larger methane

oxidation for the year 2010 compared to that simulated by Butler et al. (2020), possibly due to the usage of a different anthropogenic emissions dataset, a prescribed CH_4 mixing ratio and a prescribed meteorology dataset. A detailed comparison of methane-related variables such as prescribed concentration, burden, lifetime and oxidation rate between our study and Butler et al. (2020) for the year 2010 is provided in Table S4 in the Supplement. While the absolute contributions from methane oxidation, anthropogenic and biogenic NMRC emissions to total reactive carbon emissions increase with medium–high certainty (Fig. 3b), their relative contributions show trends with low certainty (Fig. S3b).

The tropospheric methane lifetime (~ 8.54 years), calculated as the total atmospheric methane burden divided by tropospheric methane oxidation rate, decreases with high certainty (-0.01 yr^{-1} ; Table S5) in our simulations (Fig. 4a: magenta line). This result is contrary to the expectation that methane lifetime might increase due to increasing methane concentrations, leading to a smaller availability of OH radicals to oxidize methane (Prather, 1996). Oxidation of CH_4 with OH radical being the major loss pathway for CH_4 , we show the time series of OH in Fig. 4b. Here we simulate

Table 3. Theil–Sen estimator/slope of trends in contributions of NO_x-tagged metrics, wherever applicable. Trend slope and *p* estimation are summarized in Sect. 2.3. The 95 % confidence interval and *p* for each of these estimated trends are provided in Table S1. In brackets is the trend (in % yr^{−1}). Note that this trend in % yr^{−1} is the trend slope relative to the first value of the absolute contribution (for year 2000), not to be confused with the relative contribution to the total as shown in Figs. S3, S6, S7 and S8 in the Supplement.

Source	Emissions (Tg N yr ^{−2})	Tropospheric O ₃ burden (Tg O ₃ yr ^{−1})	OPE (mol O ₃ (mol N) ^{−1} yr ^{−1})	Surface mean (ppbv yr ^{−1})	Population-weighted mean (ppbv yr ^{−1})
Total	0.13 (0.29)*	0.72 (0.23)	N/A	0.08 (0.30)	0.17 (0.54)
Stratosphere	N/A	1.83E-03 (2.06E-03)**	N/A	0.02 (0.32)	*−8.59E-03 (−0.25)
Aircraft	0.02 (2.52)	0.28 (2.41)	2.74E-03**	0.02 (2.18)	0.02 (1.70)
Biogenic ¹	0 (0)**	−0.06 (−0.26)	−3.74E-03	−3.86E-03 (−0.16)	−0.02 (−0.66)
Biomass burning	−0.04 (−0.91)	−0.20 (−1.38)	−5.47E-03	−0.02 (−1.14)	−0.02 (−1.12)
Lightning	−0.01 (−0.35)	−0.43 (−0.56)	−0.01	−0.02 (−0.44)	−0.03 (−0.82)
Extra production	N/A	6.61E-03 (0.74)	N/A	3.62E-04 (0.46)	7.16E-04 (0.71)
Anthropogenic	0.19 (0.59)	1.35 (1.29)	7.35E-03	0.08 (0.69)	0.22 (1.19)
Regional contribution to anthropogenic component					
International shipping	0.08 (2.47)	0.31 (1.48)	−9.80E-03	0.05 (1.92)	0.04 (1.60)
North America	−0.33 (−3.82)	−0.34 (−2.59)	0.01	−0.04 (−2.35)	−0.05 (−2.24)
Europe	−0.11 (−2.52)	−0.11 (−2.12)	2.62E-03	−0.02 (−1.97)	−0.03 (−1.59)
East Asia	0.25 (4.04)	0.40 (3.29)	2.27E-04**	0.04 (2.87)	0.06 (1.74)
South Asia	0.13 (7.41)	0.24 (4.58)	−0.01	0.02 (3.97)	0.11 (3.09)
Russia, Belarus, Ukraine	−0.02 (−0.87)	−0.02 (−0.62)	1.34E-03	−2.64E-03 (−0.50)	−3.59E-03 (−0.63)
Middle East	0.06 (4.79)	0.07 (2.26)	−0.01	6.97E-03 (1.62)	0.01 (1.12)
Rest of the world	0.15 (2.58)	0.73 (1.87)	−8.69E-03	0.05 (1.35)	0.08 (2.09)

* 0.5 < *p* < 0.1: medium certainty.

** *p* ≥ 0.1: low certainty.

¹ Since we prescribe an annually repeating seasonal cycle of biogenic / soil NO_x emissions, the annual mean value is held constant, leading to zero slope and *p* = 1.

Table 4. Same as Table 2 but for RC-tagged components.

Source	Emissions (Tg C yr ^{−1})	Tropospheric O ₃ burden (Tg O ₃)	OPE (mol O ₃ (mol C) ^{−1})	Surface mean (ppbv)	Population-weighted mean (ppbv)
Total	1342.06	317.92	N/A	25.39	33.01
Stratosphere	N/A	76.04 (23.92)	N/A	3.99 (15.73)	3.05 (9.25)
Aircraft	0.24 (0.02)	0.03 (8.58E-03)	0.03	2.33E-03 (9.19E-03)	2.97E-03 (8.98E-03)
Biogenic	430.93 (32.11)	52.95 (16.66)	0.03	4.07 (16.04)	7.38 (22.37)
Biomass burning	171.39 (12.77)	13.85 (4.36)	0.02	1.17 (4.62)	1.34 (4.06)
Methane oxidation	437.98 (32.63)	132.45 (41.66)	0.08	11.99 (47.22)	12.36 (37.46)
Extra production	N/A	4.52 (1.42)	N/A	0.42 (1.67)	0.62 (1.88)
Anthropogenic	301.52 (22.47)	45.95 (14.45)	0.04	3.72 (14.67)	8.24 (24.96)
Regional contribution to anthropogenic component					
International shipping	0.76 (0.06)	0.17 (0.05)	0.06	0.02 (0.07)	0.02 (0.07)
North America	36.47 (2.72)	4.03 (1.27)	0.03	0.57 (2.24)	0.68 (2.07)
Europe	21.19 (1.58)	2.40 (0.75)	0.03	0.43 (1.69)	0.66 (1.99)
East Asia	95.36 (7.11)	10.87 (3.42)	0.03	1.03 (4.06)	2.52 (7.63)
South Asia	41.89 (3.12)	5.55 (1.74)	0.03	0.40 (1.59)	2.36 (7.15)
Russia, Belarus, Ukraine	7.97 (0.59)	0.92 (0.29)	0.03	0.17 (0.67)	0.21 (0.63)
Middle East	7.12 (0.53)	1.22 (0.38)	0.04	0.17 (0.68)	0.36 (1.10)
Rest of the world	90.77 (6.76)	15.65 (4.92)	0.04	0.93 (3.66)	1.43 (4.33)

an increasing trend in air-mass-weighted tropospheric OH concentration (a prominent indicator for tropospheric oxidizing capacity; e.g., Voulgarakis et al., 2013; Chua et al., 2023) but with low certainty. Nevertheless, the CH₄-reaction-weighted tropospheric OH concentration (Lawrence et al., 2001) shows an increasing trend with high certainty. This increasing trend in tropospheric OH concentration is consistent

with the trend discussed in Chua et al. (2023) and explains the decreasing lifetime of CH₄ as these quantities are inversely proportional. The increasing OH availability despite increasing CH₄ concentration could be due to increasing NO_x emissions over the simulated period that recycle HO₂ to OH (e.g. Lelieveld et al., 2008; Chua et al., 2023). The mean magnitude of our air-mass-weighted tropospheric OH

Table 5. Same as Table 3 but for RC-tagged metrics. The 95 % confidence interval and *p* for each of these estimated trends are provided in Table S3 in the Supplement.

Source	Emissions (Tg C yr ⁻²)	Tropospheric O ₃ burden (Tg O ₃ yr ⁻¹)	OPE (mol O ₃ (mol C) ⁻¹ yr ⁻¹)	Surface mean (ppbv yr ⁻¹)	Population-weighted mean (ppbv yr ⁻¹)
Total	3.18 (0.25)	0.71 (0.23)	N/A	0.08 (0.30)	0.17 (0.54)
Stratosphere	N/A	**0.03 (0.04)	N/A	0.02 (0.40)	**−4.2E−03 (−0.14)
Aircraft	3.9E−03 (1.74)	4.66E−04 (1.81)	**4.05E−05	3.60E−05 (1.62)	4.65E−05 (1.64)
Biogenic	*1.26 (0.31)	**0.06 (0.11)	**−1.38E−05	9.53E−03 (0.24)	0.03 (0.44)
Biomass burning	**−0.77 (−0.51)	**−0.08 (−0.63)	**−2.89E−05	**−1.58E−03 (−0.15)	**−1.72E−03 (−0.14)
Methane oxidation	1.81 (0.42)	0.44 (0.33)	**−7.00E−05	0.05 (0.42)	0.05 (0.38)
Extra production	N/A	0.02 (0.54)	N/A	1.92E−03 (0.48)	4.33E−03 (0.76)
Anthropogenic	1.74 (0.61)	0.45 (1.03)	2.17E−04	**4.6E−03 (0.12)	0.09 (1.23)
Regional contribution to anthropogenic component					
International shipping	0.01 (1.79)	−2.73E−03 (−1.03)	−1.38E−03	−1.91E−04 (−0.73)	**−7.06E−05 (−0.22)
North America	−1.80 (−3.16)	−0.17 (−2.83)	1.58E−04	−0.02 (−2.71)	−0.03 (−2.58)
Europe	−0.72 (−2.53)	−0.08 (−2.53)	*−4.09E−05	−0.01 (−2.46)	−0.02 (−2.26)
East Asia	1.58 (2.20)	0.23 (2.77)	2.30E−04	0.02 (2.91)	0.06 (2.85)
South Asia	1.19 (3.76)	0.15 (3.51)	**−3.11E−05	0.01 (3.54)	0.07 (3.61)
Russia, Belarus, Ukraine	−0.06 (−0.72)	−0.01 (−0.98)	−1.03E−04	−1.3E−03 (−0.71)	−1.87E−03 (−0.82)
Middle East	**−0.05 (−0.60)	−0.01 (−1.01)	−1.42E−04	*−1.3E−03 (−0.70)	*−2.62E−03 (−0.65)
Rest of the world	1.50 (1.95)	0.26 (1.87)	**2.59E−05	0.01 (1.62)	0.03 (2.31)

* 0.5 < *p* < 0.1: medium certainty.** *p* ≥ 0.1: low certainty.

concentration ($\sim 12.7 \times 10^5$ molec. cm⁻³) is slightly larger than previous studies ($\sim 10.8 \times 10^5$ molec. cm⁻³ in Chua et al., 2023, or $\sim 11.7 \times 10^5$ molec. cm⁻³ in Voulgarakis et al., 2013), which is likely why we simulate a mean CH₄ lifetime slightly smaller than the aforementioned studies (~ 9 years) but within the multi-model range (7.1–10.6 years) reported in Voulgarakis et al. (2013).

Among the anthropogenic NMRC emissions, East Asian emissions are the largest (~ 95.4 Tg C yr⁻¹ (7.11 %); Table 4), increasing at 3.54 Tg C yr⁻², peaking in 2011 and decreasing after that at −2.8 Tg C yr⁻² (Fig. 3d; Table S6 in the Supplement), a feature similar to the NO_x emissions. The second largest emitter is the “rest of the world” region (~ 90.8 Tg C yr⁻¹ (6.76 %); Table 4), with an increasing trend of high certainty (Fig. 3d). Remaining regions each contribute to less than 5 % of reactive carbon emissions. South Asian, Middle Eastern and ship NMRC emissions show an increasing trend, and North American, European and the Russia–Belarus–Ukraine region’s NMRC emissions show a decreasing trend (Fig. 3d; similar sign of trend also seen for anthropogenic NO_x emissions). We also see an equatorward shift in global anthropogenic NMRC emissions, similar to the anthropogenic NO_x emissions in our zonal sum profiles of deviation in anthropogenic NMRC from year 2000 (Fig. S4b).

3.2 Tropospheric ozone burden

The tropospheric ozone burden is calculated as the mass of ozone in the model grid cells below the ozone tropopause, defined as the highest layer in the upper troposphere with an ozone mixing ratio less than 150 ppb (e.g., Bak et al.,

2022; Liu et al., 2022). We simulate a climatological average tropospheric ozone burden of ~ 318 Tg O₃, with an increasing trend of 0.72 Tg O₃ yr⁻¹ over the 2000–2018 period. Our simulated tropospheric ozone burden is towards the lower end of values simulated by several multi-model studies for the 2005–2014 period (values from Griffiths et al., 2021): the CMIP6 ensemble at 356 ± 31 Tg O₃, ACCENT at 336 ± 27 Tg O₃, Atmospheric Chemistry and Climate Model Intercomparison Project (ACCMIP) at 337 ± 23 Tg O₃, TOAR at 340 ± 34 Tg O₃, and the Intergovernmental Panel on Climate Change (IPCC) at 347 ± 28 Tg O₃ (Szopa et al., 2021). The trend in tropospheric ozone burden is within the range of trends simulated by the CMIP6 model ensemble: 0.4 to 1.3 Tg O₃ yr⁻¹, as reported by Wang et al. (2022). When integrated from 60° S to 60° N, we simulate a climatological mean tropospheric ozone burden of ~ 287.4 Tg O₃ (lower end of the range of 287–311 Tg O₃ estimated by satellite products for 2014–2016 period; Gaudel et al., 2018), with an increasing trend of 0.68 Tg O₃ yr⁻¹ (close to the trend of 0.82 ± 0.13 Tg O₃ yr⁻¹ simulated by CMIP6 ensemble members; Griffiths et al., 2021).

We simulate a substantial amount of tropospheric ozone attributed to stratospheric influx in the NO_x-tagged (~ 91.5 Tg O₃ (~ 29 %); Table 2) and RC-tagged simulation (~ 76.0 Tg O₃ (~ 24 %); Table 4). The larger amount of ozone attributed to stratospheric influx in the NO_x-tagged simulation compared to the RC-tagged simulation is because of the production of tropospheric O₃ attributed to NO_x from the stratosphere formed by the oxidation of N₂O, which is then subsequently transported into the troposphere. As described in Sect. 3.1.2 of Butler et al. (2018), ozone production in the

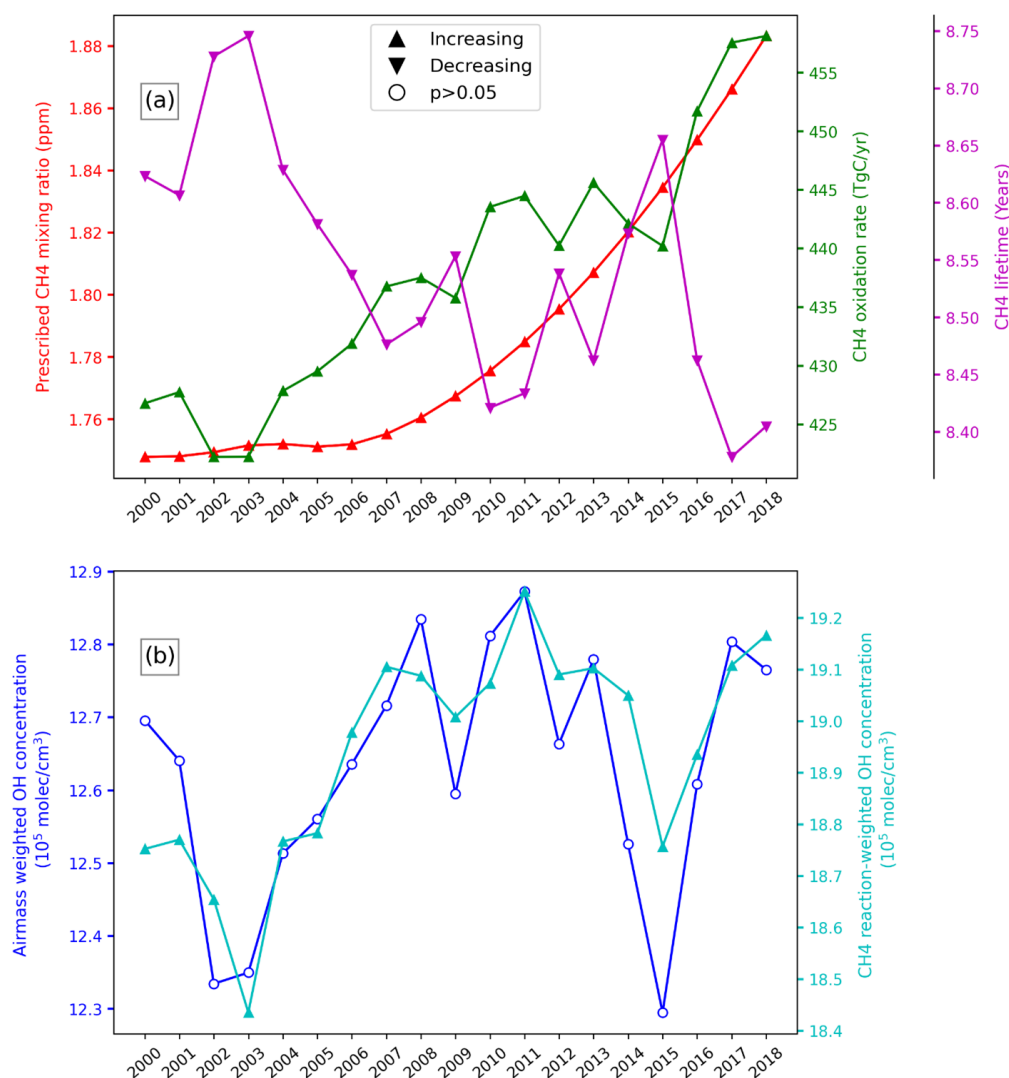


Figure 4. (a) Red line: annually varying methane mixing ratio (in ppm) prescribed in our model. Green line: methane oxidation rate (in TgCyr^{-1}) simulated by our model. Magenta line: lifetime of methane (in years). (b) Air-mass-weighted and CH_4 -reaction-weighted tropospheric OH concentration (in $10^5 \text{ molec. cm}^{-3}$). The symbols on the plotted time series indicate the sign of trend in the plotted quantity. The mean values, slope of trend, 95 % confidence interval and p for the 2000–2018 period are provided in Table S5 in the Supplement.

stratosphere is handled directly in the chemical mechanism with the addition of new reactions producing the relevant odd oxygen species. Similarly, production of NO in the stratosphere from the oxidation of N_2O by O1D is also handled directly with an additional chemical reaction.

We attribute minor ozone production pathways with the “extra production” tag. In our NO_x -tagged simulation, this category consists of O_x production from the self-reaction of OH radicals and from reactions between HO_2 and organic peroxy radicals. In our RC-tagged simulation this category consists of O_x production from the self-reaction of OH radicals and production of HO_2 from the reaction of OH with H_2O_2 . The contribution of these minor production pathways to the total tropospheric ozone burden is not very substantial (up to 2 %; Tables 2 and 4).

3.2.1 Tropospheric ozone burden attributed to NO_x emission sources

Our simulation shows that ~ 71 % of tropospheric ozone is attributed to NO_x emissions (both surface-based and aloft). The remaining ~ 29 % of tropospheric ozone burden is attributed to the stratospheric influx and a small contribution from minor production pathways, as discussed above. We note that the tropospheric ozone attributable to stratospheric influx in our NO_x -tagged simulation is larger than the stratospheric contribution in the reactive-carbon-tagged simulation (24 %, described below) due to the inclusion of tropospheric ozone production from stratospheric NO_x (as described above).

Anthropogenic NO_x emissions are the largest contributors to tropospheric ozone ($\sim 117 \text{ Tg O}_3$ ($\sim 37\%$)), followed by lightning NO_x ($\sim 71 \text{ Tg O}_3$ (22%)), biogenic NO_x emissions (24 Tg O_3 (7.44%)) and aircraft NO_x emissions (13.44 Tg O_3 ($\sim 4.23\%$); Table 2). Lightning NO_x has a large contribution despite a small amount of NO_x emissions because of large ozone production efficiency (OPE) of NO_x when emitted aloft (Table 2). Further discussion about OPE is provided in Sect. 3.3. Biomass burning NO_x emissions contribute $\sim 4\%$ to the total tropospheric ozone burden.

Among regional anthropogenic NO_x sources, the contribution from the “rest of the world” used in this study (Table 1) to the tropospheric ozone burden is largest among the anthropogenic NO_x emissions, followed by ship NO_x emissions. This “rest of the world” contribution is even larger than that of East Asian contribution, which is the region with the largest anthropogenic NO_x emission among continental regions considered in this study (Fig. 3, Table 2). This is due to more efficient production of ozone and convection into the free troposphere at the tropical regions (Zhang et al., 2016, 2021; see further discussion in Sect. 3.3.1) that are included within the “rest of the world” tag. The disaggregated contribution of explicitly tagged regions within the “rest of the world” tag in our NO_x-tagged simulation – Southeast Asia, Central Asia, North Africa and Mexico–Central America (Table 1; Fig. S1b) – are provided in the Supplement (Tables S5 and S6). We notice that the largest contributors among these regions are the NO_x emissions from the tropical regions: Mexico and Central America and Southeast Asia (Table S5). Similarly, the tropically situated South Asian contribution to the tropospheric ozone burden is larger than that of the European contribution, despite having a smaller amount of NO_x emissions (Table 2), due to the larger convection into the free troposphere as stated above.

To illustrate the predominant role of convection in transporting ozone and its precursors in the tropics into the free troposphere, we show the zonal mean vertical profiles of annual mean O₃ and NO_x (sum of NO and NO₂) mixing ratios for the 2000–2018 climatological mean attributed to NO_x emitted from a typical tropical region: Southeast Asia and a typical northern midlatitude region – Europe (Fig. 6; see Fig. S1 for the location of defined regions). The vertical gradient in the ozone attributed to European NO_x is consistent with summertime production of ozone primarily within the boundary layer with subsequent vertical transport into the free troposphere. While there is indeed some ozone attributed to Southeast Asian NO_x emissions present in the boundary layer, the mixing ratio of this attributed ozone is much higher in the free troposphere, which is consistent with ozone production aloft. This is due to emitted NO_x directly being transported aloft, eventually leading to free tropospheric ozone production. Further, we also note the increasing trend in NO_x burden attributed to anthropogenic NO_x emissions in the free troposphere (above 700 hPa; Fig. S5).

The contribution from ship NO_x is also large due being emitted in remote ocean regions where ship NO_x is the only source of NO_x emissions. This results in less competition among tagged NO_x sources in producing ozone, leading to very efficient ozone production from ship NO_x (Butler et al., 2020). Further discussion about ozone production efficiency of ship NO_x is provided in Sect. 3.3.1.

The sign of the trend in the contribution to the tropospheric ozone burden from each of the tags (Fig. 5a and c) is consistent with the sign of the NO_x emissions trend (Fig. 3a and c, Table 3). The percentage slope (ignoring the sign) of the trend in the contribution to tropospheric ozone burden by regional anthropogenic NO_x sources, however, is generally smaller than that of NO_x emission trend (Table 3). Similarly, the magnitude of the percentage slope in the trend in tropospheric ozone burden attributed to East Asian NO_x emissions for both the pre-2011 (2000–2011) and post-2011 (2011–2018) periods is smaller compared to the corresponding percentage slope in East Asian NO_x emissions trend (Table S2). The reason for this is due to changing ozone production efficiency with changing NO_x emissions, explained in detail in Sect. 3.3.1.

We simulate a trend of $\sim 0.28 \text{ Tg O}_3 \text{ yr}^{-1}$ ($2.41\% \text{ yr}^{-1}$; Table 3) in ozone burden attributed to aircraft NO_x emissions, which is comparable to $0.3 \text{ Tg O}_3 \text{ yr}^{-1}$ estimated by Wang et al. (2022) using sensitivity simulations, where the difference between a simulation with transient aircraft emissions and that with fixed aircraft emissions delivers the exclusive impact of changing aircraft emissions over the 1995–2017 period. While the trend derived from our tagged simulation only delivers the trend in contribution from aircraft NO_x emissions, a sensitivity simulation would also cover the changes in simulated ozone contributed from various sectors when their ozone production efficiency changes as a result of emissions being held constant. A combination of sensitivity simulations with tagging included would help us track these nonlinearities and compensating feedbacks that would otherwise not be seen from sensitivity-only or tagging-only simulations.

3.2.2 Tropospheric ozone burden attributed to reactive carbon emission sources

In our RC-tagged simulation $\sim 76\%$ of the total amount of tropospheric ozone burden is attributed to tropospheric reactive carbon, while the remaining $\sim 24\%$ is attributed to stratospheric influx and minor production pathways as discussed above.

Of the reactive carbon sources, methane oxidation is the largest contributor to the tropospheric ozone burden ($\sim 132 \text{ Tg O}_3$ (42%); Table 4) and is consistent with previous studies (Young et al., 2013; Butler et al., 2018). We simulate an increasing trend with high certainty in the part of the tropospheric ozone burden attributed to RC from methane oxi-

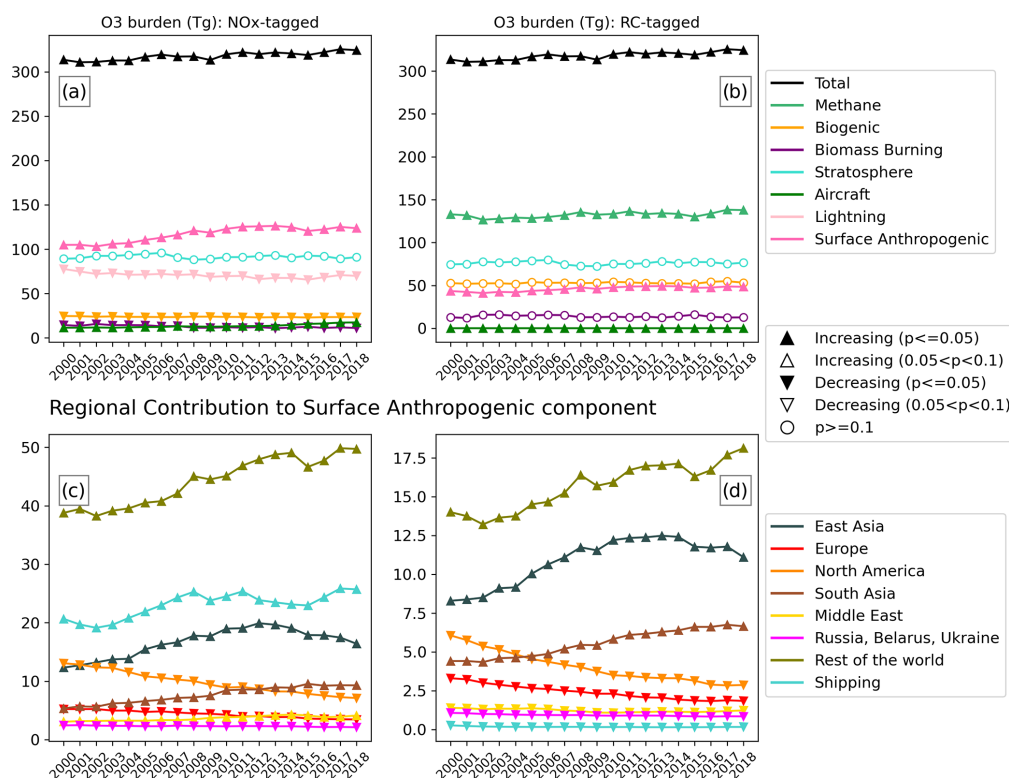


Figure 5. Simulated global annual mean tropospheric ozone burden (in Tg O₃) time series over the 2000–2018 period. Shown are the total simulated ozone burden (black line) and the contributions from the NO_x-tagged (a, c) and RC-tagged (b, d) simulations. Panels (c) and (d) are the regional contributions to the total anthropogenic component (dark-pink line) shown in (a) and (b). The symbols indicate the sign and certainty of trend in the plotted quantity. The time series of relative contributions (in %) from each of the emission sectors to the total emissions is provided in Fig. S6.

dation (Fig. 5b), consistent with the increasing methane oxidation rate during the 2000–2018 period, as shown in Fig. 4a.

The contribution to tropospheric ozone burden from biogenic NMRC emissions (53 Tg O₃ (~ 17 %); Table 4) is much smaller than that of methane, despite having a comparable mass of RC emissions from both sources. This is consistent with the smaller number of oxidizable bonds per carbon atom in biogenic emissions, mainly isoprene (2.8) compared with methane (4) and the subsequently lower number of NO to NO₂ conversions possible (per carbon atom) during isoprene oxidation compared with methane oxidation (Edwards and Evans, 2017). We simulate much smaller contributions of NMRC emissions from biomass burning and aircraft NMRCs compared to biogenic and anthropogenic sources (consistent with Butler et al., 2020; Table 4). Contribution to tropospheric ozone attributed to anthropogenic NMRC emissions is relatively low at ~ 46 Tg O₃ (14 %; Table 4).

Regional anthropogenic tags each contribute less than 5 % to the total tropospheric ozone burden. The sign of the trend in tropospheric ozone attributed to most of the tags, including regional anthropogenic NMRC tags, in our RC-tagged simulation is consistent with that of the trend in the corresponding reactive carbon emissions (Table 5). We also

note that the percentage trend in contribution to tropospheric ozone burden from anthropogenic NMRC (1.03 % yr⁻¹) is more than that of the trend in anthropogenic NMRC emissions (0.61 % yr⁻¹), whereas the trend in contribution from methane oxidation (0.33 % yr⁻¹) is less than that of trend in methane oxidation rate (0.42 % yr⁻¹). While the biogenic NMRC emissions show an increasing trend with medium certainty, the trend in its absolute contribution to tropospheric ozone burden shows a small trend with low certainty.

3.3 Ozone production efficiency (OPE)

We estimate ozone production efficiency of emissions of any given sector as the ratio of annual mean tropospheric ozone burden (in mol O₃) attributed to that sector to the amount of ozone precursors emitted in a given year (either NO_x (in mol N) or reactive carbon (in mol C)). Several previous studies report OPE as the ratio of response in tropospheric ozone burden to change in emissions introduced in the model or as a ratio of the production rate of ozone to the loss rate of NO_x (e.g. Lin et al., 1988; Kim et al., 2016; Miyazaki et al., 2021; Archibald et al., 2020). Since our model directly calculates the ozone attributed to emissions from tagged emission

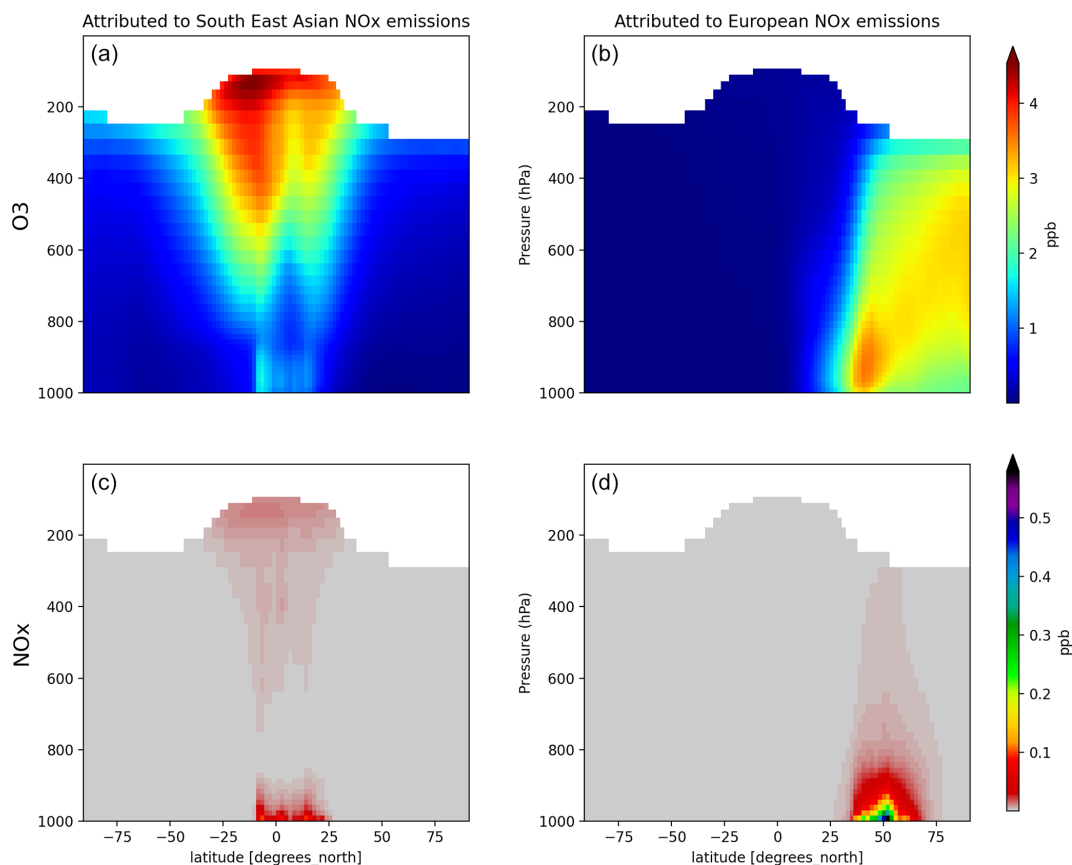


Figure 6. Vertical profiles of zonal climatological (2000–2018) mean mixing ratio within the troposphere: O₃ (a, b) and NO_x (c, d) attributed to Southeast Asian (a, c) and European (b, d) anthropogenic NO_x emissions.

sources, the ratio of the attributed tropospheric ozone burden to the emitted amount of ozone precursor is calculated as OPE of that tagged precursor source. This can only be made possible when the NO_x and reactive carbon precursors are tagged in two separate simulations, as explained in Butler et al. (2020). A similar metric – ozone burden efficiency (OBE) – has been defined by Mertens et al. (2024) as the ratio of ozone attributed of precursor emissions (both NO_x and RC) from a given source using the combinatorial tagging approach (in TgO₃) to the emitted NO_x (Tg NO) for various tagged sectors.

3.3.1 OPE of NO_x emissions

Lightning NO_x (5.93 mol O₃ (mol N)^{−1}) and aircraft NO_x (4.19 mol O₃ (mol N)^{−1}) are most efficient at producing ozone (Fig. 6a, Table 2) directly into the free troposphere, due to being emitted aloft (Hoor et al., 2009; Dahlmann et al., 2011). NO_x emissions aloft are highly efficient at producing ozone due to the relatively low quantities of NO_x at higher altitudes. Low NO_x concentrations increase the ozone production efficiency of NO_x due to a lower frequency of radical termination reactions compared with

higher NO_x concentrations so that each molecule of NO_x can produce more molecules of ozone before being lost by reaction with OH (Seinfeld and Pandis, 2016). In contrast, biogenic NO_x (1.4 mol O₃ (mol N)^{−1}), biomass burning NO_x (0.88 mol O₃ (mol N)^{−1}) and anthropogenic NO_x (0.96 mol O₃ (mol N)^{−1}) are comparatively less efficient at producing ozone as they are emitted at the surface into regions with comparatively large NO_x concentrations.

Among regional anthropogenic NO_x emissions, the “rest of the world” is the most efficient (1.8), as it mainly consists of tropical regions where the convection into the free troposphere is large. We further elaborate the explicitly tagged regions in our NO_x-tagged simulation within the “rest of the world” tag in Table S7 in the Supplement. The tropical regions are the most efficient among all the explicitly tagged regions: Mexico and Central America (1.81 mol O₃ (mol N)^{−1}) and Southeast Asia (1.98 mol O₃ (mol N)^{−1}). The larger sensitivity of tropospheric ozone to emission changes in these tropical regions has also been noted in Zhang et al. (2021). Similarly, OPE of South Asian NO_x (0.78 mol O₃ (mol N)^{−1}) and to some extent Middle Eastern NO_x (0.64 mol O₃ (mol N)^{−1}) is much larger compared to that of North American

(0.54 mol O₃ (mol N)^{−1}), European (0.36 mol O₃ (mol N)^{−1}), the Russia–Belarus–Ukraine region's (0.4 mol O₃ (mol N)^{−1}) and East Asian (0.54 mol O₃ (mol N)^{−1}) NO_x, for being more tropically situated (e.g., Wild et al., 2001; Butler et al., 2020; Zhang et al., 2021). International shipping is the second most efficient (1.67 mol O₃ (mol N)^{−1}) source of anthropogenic NO_x emissions in producing ozone, due to their presence at remote pristine regions where there is a much smaller availability of other sources of NO_x (Lawrence and Crutzen, 1999; Butler et al., 2020). We note that the chemistry occurring within the expanding plume emitted by ships is not considered in our coarse-resolution model, leading to an instantaneous dilution of emitted NO_x into the large grid cells (Vinken et al., 2011), thereby making ozone production from ship NO_x more efficient than observed (Kim et al., 2016).

The increasing trend in anthropogenic NO_x emissions from South Asia, the Middle East and international shipping (Fig. 3c and Table 3) results in less efficient production of tropospheric ozone from these sources over time (Fig. 7c and Table 3). Similarly, there is an increasing trend in OPE of anthropogenic NO_x emissions from regions with decreasing emissions (North American, European and the Russia–Belarus–Ukraine region; Fig. 7c). The decreasing (increasing) NO_x emissions from regional anthropogenic sources become more (less) efficient at producing ozone, and this leads to a dampening effect where there is a smaller-percentage slope (ignoring the sign) in tropospheric ozone burden compared to the slope in NO_x emissions (Table 3). This result is also consistent with Mertens et al. (2024), where an increase in NO_x emission from various tagged sectors is associated with a decrease in OBE in both present-day and future scenarios.

Figure 7c shows that the OPE of East Asian NO_x has a trend of low certainty over the 2000–2018 period. However, when shorter periods are considered, it decreases with high certainty during the pre-2011 period when emissions increase (2000–2011 period) and increases with high certainty for the post-2011 period when emissions decrease (2011–2018 period; Table S2).

In the case of total anthropogenic NO_x emissions, we do not simulate a decrease in their OPE despite their increasing emission trend. We rather simulate an increase in OPE of anthropogenic NO_x with high certainty. This is due to an equatorward shift in emissions over the 2000–2018 period (Fig. S4), where the emitted NO_x from the tropics and the subsequently formed ozone and NO_y (NO_x and reservoir species for NO_x (e.g., NO₃, N₂O₅) molecules are rapidly lifted into the free troposphere as discussed previously (Zhang et al., 2016). We also notice that the increase in anthropogenic emissions leads to a decreasing trend in OPE of other natural sources of NO_x such as biogenic, biomass burning and lightning sources, as these natural sources must compete with more NO_x emissions from anthropogenic sources in order to produce ozone.

3.3.2 OPE of reactive carbon emissions

Consistent with Butler et al. (2020), we simulate methane oxidation as the most efficient (0.08 mol O₃ (mol C)^{−1}; Table 4) among the tagged reactive carbon sources in our RC-tagged simulation. Similarly, we simulate smaller OPEs for biogenic (0.03 mol O₃ (mol C)^{−1}), biomass burning (0.02 mol O₃ (mol C)^{−1}) and anthropogenic (0.04 mol O₃ (mol C)^{−1}) reactive carbon sources. As noted in Sect. 3.2.2, RC from methane oxidation is expected to be more efficient at producing ozone per unit of carbon due to the higher number of oxidizable bonds per carbon atom. Anthropogenic reactive carbon emissions might be expected to be more efficient than biogenic reactive carbon emissions at producing ozone due to their proximity to anthropogenic NO_x emissions, but we see no large differences in OPE in our study (Table 4).

We simulate a trend with low certainty in the OPE RC from natural sources such as biogenic, biomass burning and methane oxidation (Fig. 7; Table 5). We do simulate an increasing trend with high certainty in the OPE of global anthropogenic reactive carbon emissions. Among the regional anthropogenic sources, we simulate trends with low–medium certainty for European, South Asian and “rest of the world” RC emissions. We simulate an increasing trend for North American and East Asian emissions and a decreasing trend for the OPE of the remaining tagged regional anthropogenic reactive carbon emissions. OPEs of East Asian anthropogenic RC emissions have a small trend with low certainty over the 2000–2011 period and then increase with high certainty during the 2011–2018 period (Table S6).

We would expect that the ozone production by reactive carbon emissions should mainly depend on the availability of NO_x, especially in pristine environments. The trends in the OPE of regional anthropogenic NMRC emissions largely, but not always, follow the trends in the emissions of NO_x in their vicinity (Tables 3 and 5). However, the trends in the emission of the RC from a given sector itself could influence the OPE of its own emissions and the OPE of the NMRC emissions from other sources. For example, Butler et al. (2020) illustrate how an increase in prescribed methane concentration leads to smaller OPE of not just RC from methane oxidation but also of other tagged RC emission sectors. To better understand the behavior of changes in OPE of reactive carbon emissions, further studies could investigate the OPE changes from perturbation simulations, by enabling ozone source attribution with tagging in the simulations. Mertens et al. (2018) demonstrate that the combination of tagging and perturbation can be used to explain the compensating feedbacks induced by perturbation, from various tags used in the simulations.



Figure 7. Time series (2000–2018) of the ozone production efficiency (OPE; in mol O₃ (mol N)^{−1}: NO_x-tagged, **a**, **c**, mol O₃ (mol C)^{−1}: RC-tagged, **b**, **d**) of various ozone precursor emission regions/sectors tagged in this study. Panels (**c**) and (**d**) are the OPE values of regional anthropogenic emissions tagged in this study. The symbols indicate the sign and certainty of trend in the plotted quantity.

3.4 Contributions from tagged precursor emission sources to surface ozone

In this subsection, we discuss the contribution of various NO_x and reactive carbon sources to both the global surface mean and population-weighted mean ozone. We further contrast this contribution with that of their contribution to the tropospheric ozone burden (discussed in Sect. 3.2).

We select the population count for the year 2020 Gridded Population of the World (Center for International Earth Science Information Network – CIESIN – Columbia University, 2018) and scale the surface ozone according to this distribution to calculate the global population-weighted mean ozone, which is related to the ozone exposure. Choosing only 1 year for population weighting gives the exclusive effect of changing emissions and not the effect of changing population. Future studies may consider scaling according to a transient population over the analysis period rather than fixing it at 1 year (year 2020 in this study). Table S9 provides the population and their fraction of world population (in %) for the regions considered in this study.

We simulate ~ 25.4 ppbv of global mean surface ozone with an increasing trend of high certainty (0.08 ppbv yr^{−1} (0.3% yr^{−1}); Fig. 7 and Table 3). Similarly, we simulate ~ 33.0 ppbv of global population-weighted mean ozone with

an increasing trend and slightly larger slope (0.17 ppbv yr^{−1} (0.54% yr^{−1}); Fig. 8 and Table 3) compared to global surface mean ozone. We will explain the main contributors responsible for this increasing trend and the larger slope in global mean population-weighted ozone in the subsequent discussion. DeLang et al. (2021) showed that the increasing trend in global ozone exposure is mainly driven by increasing ozone exposure at highly populated regions such as Asia and Africa. In our study, we quantify the contribution of various ozone precursor emission sources to the global population-weighted surface ozone and its trend.

The share of the stratospheric contribution to the global mean surface ozone (4 – 5 ppbv ($\sim 15\%$ – 20%); Tables 2 and 4) is smaller than its contribution to tropospheric ozone burden (24% – 29% ; Tables 2 and 4). We simulate an increasing trend with high certainty ($\sim 0.3\%$ – 0.4% yr^{−1}; Tables 3 and 5) in this contribution mainly over the Southern Hemisphere (not shown). The stratospheric contribution to the population-weighted surface ozone is much smaller (~ 3 ppbv (9% – 11%); Tables 2 and 4) compared to the quantities discussed above, as it mainly contributes over remote regions such as polar and oceanic regions (not shown).

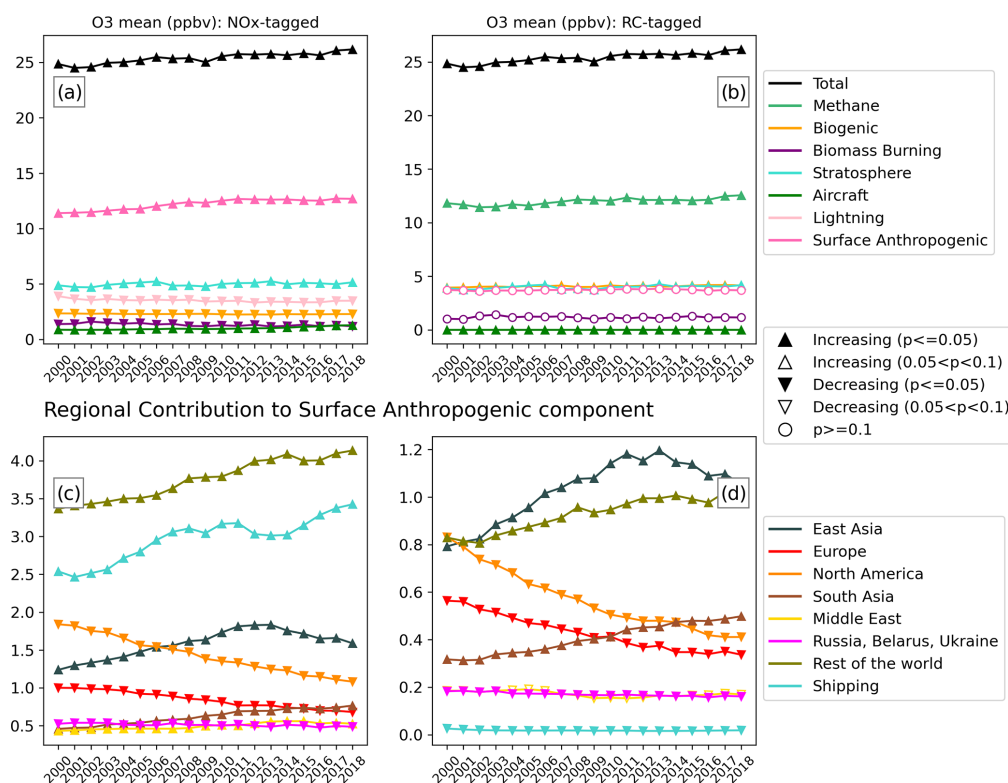


Figure 8. Same as Fig. 5 but for global annual area-weighted mean surface ozone (in ppb). The time series of relative contributions (in %) from each of the tagged components to the global annual area-weighted mean surface ozone is provided in Fig. S7.

3.4.1 Surface ozone attributed to NO_x emissions

The percentage contribution to global mean surface ozone from sources aloft (lightning, aircraft, stratosphere) is smaller compared to the contribution to tropospheric ozone burden (Table 2). Consequently, the surface-based NO_x sources (biogenic, biomass burning and anthropogenic) contribute a larger share to the global mean surface ozone compared to their share of the tropospheric ozone burden.

The share of the anthropogenic NO_x contribution has the largest difference in its contribution from $\sim 37\%$ (to tropospheric ozone burden) to $\sim 48\%$ (to global mean surface ozone; Table 2). All the tagged regional anthropogenic NO_x sources contribute a larger percentage share to global mean surface ozone than to the tropospheric ozone burden (Table 2). However, the tropically situated explicitly tagged regions within the “rest of the world” tag in our NO_x-tagged simulation – Mexico and Central America and Southeast Asia – have a smaller share in their contribution to global mean surface ozone compared to that of tropospheric ozone burden (Table S7), largely due to being situated in a tropical region with large convection to the free troposphere (Fig. 6; Zhang et al., 2016, 2021). We simulate a small increasing trend ($0.08 \text{ ppbv yr}^{-1}$ ($0.30\% \text{ yr}^{-1}$); Table 3) in global mean surface ozone, mainly driven by increasing stratospheric contribution ($0.02 \text{ ppbv yr}^{-1}$ ($\sim 0.32\% \text{ yr}^{-1}$)) and increas-

ing trend in contribution from anthropogenic NO_x sources ($0.08 \text{ ppbv yr}^{-1}$ ($0.69\% \text{ yr}^{-1}$); Table 3 and Fig. 7c).

The contribution from anthropogenic NO_x to the population-weighted surface ozone ($\sim 20 \text{ ppbv}$ (62%); Table 2) is larger than its contribution to global mean surface ozone. This larger contribution is compensated by smaller remote contributions such as the stratosphere, lightning, aircraft and biomass burning. Among the regional anthropogenic sources, the NO_x emissions from highly populated regions (see Table S9 for population of world regions) are the largest contributors to global population-weighted mean ozone: South Asia (4.6 ppbv ($\sim 14\%$), East Asia ($\sim 4.3 \text{ ppbv}$ (13%)) and the “rest of the world” ($\sim 4.3 \text{ ppbv}$ (13%)). Ship NO_x emissions also have a substantial contribution ($\sim 2.5 \text{ ppbv}$ (7.5%)) to the global population-weighted surface mean ozone, larger than contributions from anthropogenic NO_x emissions by relatively sparsely populated world regions such as North America, Europe, the Russia–Belarus–Ukraine region, or the Middle East (Table 2). The increasing trend in global population-weighted surface ozone ($0.17 \text{ ppbv yr}^{-1}$ ($0.54\% \text{ yr}^{-1}$); Table 3), which is slightly larger compared to the trend in global mean surface ozone, is mainly driven by increasing anthropogenic NO_x emissions especially from highly populated regions (South Asia, East Asia and the “rest of the world”) and from international shipping. Despite NO_x emissions from international shipping

happening only over ocean grid cells, the ozone attributed to ship NO_x spreads across land areas and contributes to the total surface ozone by $\sim 3\text{--}6$ ppb (Fig. 10a). Although there is no population over the regions where international ship NO_x is emitted, the global population-weighted mean ozone attributed to ship NO_x is comparable to that of global mean surface ozone (Table 2) as large populations at several coastal areas and densely populated inland regions in East Asia and South Asia are exposed to ozone attributed to ship NO_x emissions (Fig. 10b). We notice that while there is a steep decreasing slope after year 2013 in global mean surface ozone attributed to East Asian NO_x emissions (Fig. 8c; due to emission reduction shown in Fig. 3c), we do not find a similar feature for global population-weighted mean ozone attributed to East Asian NO_x emissions (Fig. 9c and Table S2). The reason for this is unclear at this point and could be an interesting topic for future studies.

3.4.2 Surface ozone attributed to reactive carbon emissions

RC from methane oxidation contributes ~ 12 ppbv (47 %) to global mean surface ozone; biogenic and anthropogenic NMRC emissions contribute ~ 4.1 ppbv (~ 16 %) and 3.72 ppbv (~ 14.7 %; Table 4), respectively. Regional anthropogenic tags each contribute less than 5 % to the global mean surface ozone. Since the percentage contribution from stratospheric influx to global mean surface ozone is smaller than its contribution to tropospheric ozone burden (as explained above), there is a larger contribution from methane oxidation and anthropogenic and biomass burning RC emissions. We simulate a small trend with low certainty in the contribution to global mean surface ozone from anthropogenic NMRC emissions (Table 5). The increasing trend in global mean surface ozone is mainly driven by increasing contribution from the stratosphere, RC from methane oxidation and biogenic NMRC emissions. Although there is no trend with high certainty in emissions of biogenic NMRC, its increasing contribution to global mean surface ozone (Table 5) could be explained by the increasing anthropogenic NO_x emissions (Butler et al., 2018; Lupascu et al., 2022).

The largest contribution to global population-weighted mean surface ozone is from ozone attributed to RC from methane oxidation (~ 12 ppbv (37 %; Table 4). This contribution is smaller compared to its contribution to global mean surface ozone (~ 47 %), because a substantial part of this contribution to global mean surface ozone is in remote oceanic regions (not shown). Consequently, there is a larger relative contribution from biogenic (22 %) and anthropogenic (25 %) NMRC sources to global population-weighted mean ozone compared to their contribution to global mean surface ozone. Among regional anthropogenic NMRC emissions, there is a larger share of contributions from highly populated regions such as South Asian, East Asian and the “rest of the world” and a smaller share

from North American, European and the Russia–Belarus–Ukraine region’s anthropogenic NMRC emissions to global population-weighted mean ozone compared to contribution to global mean surface ozone (Table 4).

The increasing trend in population-weighted surface ozone is mainly driven by increasing trends in the contribution of RC from methane oxidation (0.05 ppbv yr^{−1} (0.38 % yr^{−1})) and anthropogenic (0.09 ppbv yr^{−1} (1.23 % yr^{−1})) and biogenic (0.03 ppbv yr^{−1} (0.44 % yr^{−1})) NMRC emissions (Table 5). McDuffie et al. (2023) show that with a ~ 100 ppb of methane pulse, the population-weighted ozone response is larger than the response in the global mean surface ozone. They further explain that this larger response was due to the larger availability of NO_x precursor emissions at populated regions, leading to larger ozone production in populated regions. In contrast, we simulate a larger trend in contribution from methane oxidation to global mean (0.42 % yr^{−1}) than to population-weighted mean surface ozone (0.38 % yr^{−1}; Table 5), despite having a comparable change in prescribed methane concentration (~ 135 ppb; Fig. 4) over the 2000–2018 period. This is due to a dominant trend in contribution from anthropogenic NMRC emissions from populated regions in our simulation, which is lacking in the methane pulse simulated by McDuffie et al. (2023). This dominant trend in anthropogenic NMRC contribution to population-weighted surface ozone can also be seen in its significantly increasing relative contribution (Fig. S8), which is not seen in other major contributors such as methane and biogenic NMRCs.

The total NO_x emissions show a positive trend leading to an increasing contribution from methane oxidation. Similarly, the increasing anthropogenic NO_x emitted at densely populated regions reacts with NMRC emitted from anthropogenic and biogenic sources (Fig. 8b; Lupascu et al., 2022). Among regional anthropogenic NMRC emissions, the contribution from highly populated regions (Table S9) – South Asia (~ 0.07 ppbv yr^{−1} (~ 0.85 % yr^{−1})) and East Asia (~ 0.06 ppbv yr^{−1} (2.85 % yr^{−1})) – Table 4) to global population-weighted ozone shows the largest increasing trend. As for the NO_x-tagged case discussed in the previous subsection, the global population-weighted ozone attributed to East Asian anthropogenic NMRC emissions shows a small trend with low certainty during the post-2011 period in response to the reduction in emissions (Table S6).

4 Summary and conclusions

In this study we quantify the contributions of ozone precursor emissions from various sources to tropospheric ozone during the 2000–2018 period. For this, we use the tagging approach using CAM4-chem introduced in Butler et al. (2018). We perform separate simulations for NO_x and RC-tagged contributors to tropospheric ozone, which allows us to quantify the absolute contribution from the tagged emission sources

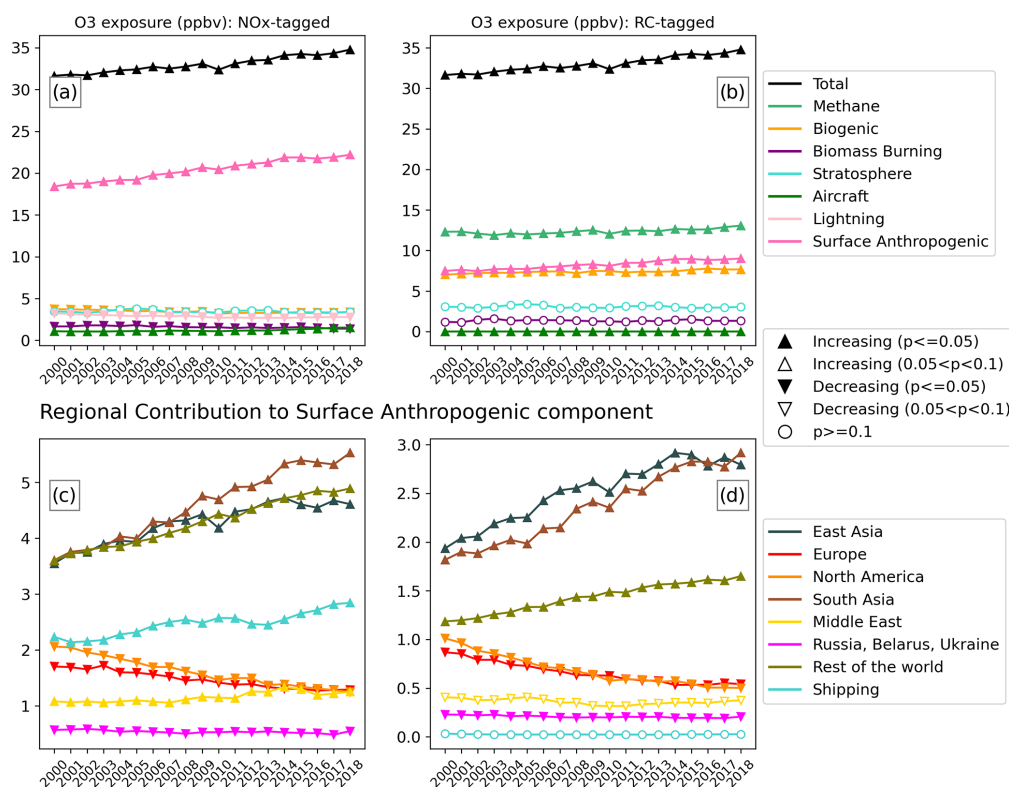


Figure 9. Same as Fig. 5 but for global annual population-weighted mean surface ozone (in ppb). The time series of relative contributions (in %) from each of the tagged components to the global annual population-weighted mean surface ozone is provided in Fig. S8.

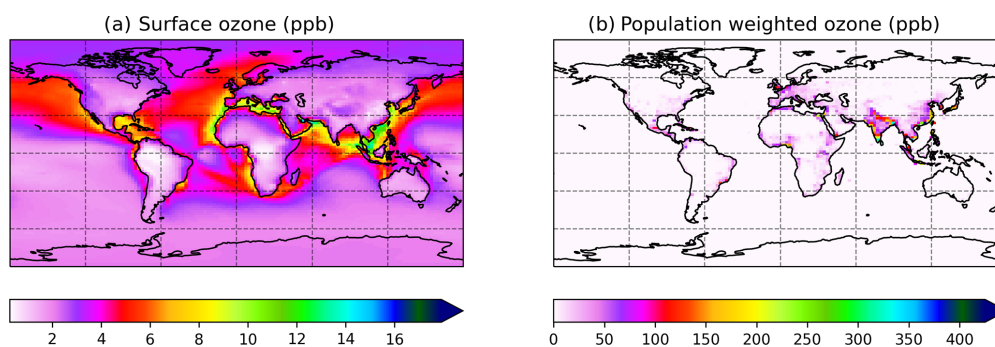


Figure 10. Spatial distribution of climatological (2000–2018) annual mean O₃ attributed to NO_x emissions from international shipping (in ppbv): (a) surface ozone and (b) population-weighted surface ozone.

to the tropospheric ozone burden and to calculate the ozone production efficiency per tagged sector/region.

We note a decreasing trend in both anthropogenic NO_x and reactive carbon ozone precursor emissions from North America, Europe and the Russia–Belarus–Ukraine region and an increasing trend from East Asia, South Asia, the Middle East, international shipping and the “rest of the world”. We simulate the largest contribution to tropospheric ozone burden from anthropogenic NO_x emissions (in our NO_x-tagged simulation) and reactive carbon from methane oxidation (in our RC-tagged simulation), both with a significantly increas-

ing trend, followed by contribution of ozone influx from the stratosphere but with no significant trend in its contribution. We simulate a relatively larger contribution to the tropospheric ozone burden from emissions at tropical regions compared to other regions, as previously discussed in Zhang et al. (2021). For example, the anthropogenic NO_x emissions from the “rest of the world” regional tag, which mainly consists of tropical and Southern Hemisphere regions, contribute more to the tropospheric ozone burden compared with East Asia despite lower NO_x emissions. We show using vertical profiles of tagged O₃ and NO_x mixing ratio fields (Fig. 6)

that it is the emitted NO_x in the tropics which predominantly gets lofted into the free troposphere and eventually forms ozone. Our tagging method only conveys information on the precursor source of the simulated ozone and associated gas molecules. It, however, does not convey information about where exactly the ozone molecule was formed: in the free troposphere or within the boundary layer. While addressing this is beyond the scope of our study, further studies could employ ozone tagging method used in Sudo and Akimoto (2007) and Derwent et al. (2015) that tag/label ozone molecules based on the location at which the ozone molecules are formed. Our study highlights the dominant effect of equatorward shifting of O₃ precursor emissions, which contributes to the tropospheric O₃ burden trend. Nevertheless, there remain other climatic factors that may play an amplifying or offsetting role to these contributions. These include, for example, changes in the general circulation of the atmosphere and monsoonal changes under a warming planet, as well as the natural variability of climate.

The trend in each tagged sector's contribution to the total tropospheric ozone burden is consistent with the sign of trend in the respective precursor emission from that tagged sector. We simulate an increasing trend in OPE of NO_x emissions from regions where the emissions are decreasing (European, North American and the Russia–Belarus–Ukraine region's emissions) and vice versa (South Asian and the Middle Eastern emissions). The anthropogenic NMRC becomes more efficient at producing ozone, following the increasing trend in anthropogenic NO_x emissions in their vicinity. We, however, do not simulate any trends with medium–high certainty in the OPE of other major reactive carbon sources such as methane oxidation and biogenic and biomass burning. We therefore recommend that further studies perform simulations combining perturbation and tagging to investigate how perturbation in the emissions of one sector induces changes in OPE of itself and other tagged sectors.

We contrast the contribution of emissions from tagged sectors to the global tropospheric ozone burden with their contribution to the global mean surface ozone and the population-weighted mean surface ozone. Except for tropical source regions such as South Asia, we simulate a larger share of contributions from regional anthropogenic NO_x emissions to the global mean surface ozone compared to the tropospheric ozone burden. We also simulate an increasing trend in contribution by RC from methane oxidation and biogenic NMRC to global mean surface ozone and population-weighted mean ozone, leading to an overall global increasing trend in both these quantities. The increasing trend in population-weighted ozone is mainly driven by the increasing trend in contribution from anthropogenic NO_x and (to a lesser extent) NMRC emissions over highly populated regions: South Asia and East Asia. We also simulate a substantial contribution of NO_x emissions from international shipping to the population-weighted mean ozone. While the focus of the present study has been the quantification of the

influences on global ozone metrics, future work using the methods employed in this study could focus on quantification of the intra- and extra-regional contributions to surface ozone in the regions studied here.

In our study, we discuss the results related to the 2000–2018 trends in global annual ozone metrics and its precursor emissions. Further studies could investigate the tagged contributions to the trends in the spatial and seasonal distribution of ozone (e.g., Wang et al., 2022; Fiore et al., 2022), which may help us enhance our understanding of various processes responsible for these trends. While previous studies used the TOAST tagging to understand the trends in surface ozone over United States (e.g., Li et al., 2023a), free tropospheric ozone over Southeast Asia (e.g., Li et al., 2023b) and over East Asia in future scenarios (e.g., Hou et al., 2023), we recommend such studies be extended to other regions of interest. Comparison of smaller time periods (2000–2011 and 2011–2018) could also be considered to understand the contrasting effect of increasing and decreasing East Asian precursor emissions, respectively. Finally, we recommend other global models implement the tropospheric ozone source attribution with NO_x and RC tagging separately, which would enable a more rigorous model intercomparison and a better understanding of the underlying processes leading to the inter-model spread in their estimation of tropospheric ozone.

Data availability. The values in the time series plotted in this study (Figs. 3, 5, 7, 8 and 9) are provided in the Supplement. Please contact tim.butler@rifs-potsdam.de for the model output of our simulations.

Supplement. Attached documents are `noxtagged_annual.csv`, `rctagged_annual.csv` and `Supplementary-figures-tables.docx`. The supplement related to this article is available online at <https://doi.org/10.5194/acp-25-5287-2025-supplement>.

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