



Global impacts of aviation on air quality evaluated at high resolution

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Abstract. Aviation emissions cause global changes in air quality which have been estimated to result in ~58,000 premature mortalities per year, but this number varies by an order of magnitude between studies. The causes of this uncertainty include differences in the assessment of ozone exposure impacts and in how air quality changes are simulated, and the possibility that low-resolution (~400 km) global models may overestimate impacts compared to finer-resolution (~50 km) regional models. We use the GEOS-Chem High Performance chemistry-transport model at a 50 km global resolution, an order of magnitude finer than recent assessments of the same scope, to quantify the air quality impacts of aviation with a single internally consistent, global approach. We find that aviation emissions in 2015 resulted in 21,200 premature mortalities due to particulate matter exposure and 53,100 due to ozone exposure. Compared to a prior estimate of 6,800 ozone-related premature mortalities for 2006 our estimate is increased by 5.6 times due to the use of updated epidemiological data which includes the effects of ozone exposure during winter, and by 1.3 times due to increased aviation fuel burn. The use of fine (50 km) resolution increases the estimated impacts on both ozone and particulate matter-related mortality by a further 20% compared to coarse-resolution (400 km) global simulation, but an intermediate resolution (100 km) is sufficient to capture 98% of impacts. This is in part due to the role of aviation-attributable ozone, which is long-lived enough to mix through the Northern Hemisphere and exposure to which causes 2.5 times as much health impact as aviation-attributable PM_{2.5}. This work shows that the air quality impacts of civil aviation emissions are dominated by the hemisphere-scale response of tropospheric ozone to aviation NO_x rather than local changes, and that simulations at ~100 km resolution provide similar results to those at two times finer spatial scale.

1 Introduction

Aviation is a unique source of climate and air quality impacts. For example, the contrails (ice clouds) which form in aircraft exhaust do not occur for any other major mode of transportation, but have been estimated to cause as much climate forcing as all of the carbon dioxide emitted during flight (Lee et al., 2020). Similarly, the nitrogen oxides (NO_x) and other species



emitted during flight can have long-lasting chemical consequences which result in global-scale degradation of air quality (Eastham and Barrett, 2016). It has been estimated that emissions during the cruise portion of the flight specifically contribute around 80% of the 8,000 – 58,000 premature mortalities each year attributable to aviation emissions (Barrett et al., 2010a; Eastham and Barrett, 2016; Quadros et al., 2020). When monetized, these air quality impacts are similar in magnitude to the net climate costs of aviation, including CO₂ and contrails (Grobler et al., 2019).

However, there remain several key uncertainties regarding the impacts of aviation emissions on air quality due to the practical challenges associated with simulating a global influence on a local quantity. Global models are well suited to quantify the global change in oxidative capacity due to aviation emissions, but must simulate the entire atmosphere to do so. Models used in previous studies have split the atmosphere up into grid cells which are between 2 and 5° (latitude and longitude) along each side, or roughly 200 – 500 km (Barrett et al., 2010a; Eastham and Barrett, 2016). This means that the models artificially diffuse local, airport-scale emissions over the surrounding area, potentially failing to resolve the co-location of near-airport emissions and exposed populations and underestimating the relative contribution of non-cruise emissions to air quality. A study by Pungler and West (2013) found that, due to this co-location effect alone, coarse resolution (> 250 km) global models would likely be biased low by 30-40% when estimating US population exposure to PM_{2.5}. The use of large grid cells also means that models treat areas of up to 20,000 km² over a city as being a single, well-mixed air mass, and will not be able to resolve non-linear chemical processes which could increase or decrease the air quality response to aviation emissions. If the population mean exposure to air pollution varies strongly with resolution, air quality impact estimates from coarse models could therefore provide misleading results.

Studies with nested, regional models can address this question for limited areas but incur an inconsistency at the model boundary, either in resolution or in the model being used to quantify impacts. These studies calculate global atmospheric composition changes at a relatively coarse resolution, and then use those results to provide the boundary conditions for a finer-resolution simulation. Whereas some studies using nested regional models have found air quality impacts from cruise emissions of a similar magnitude to those from global studies (Yim et al., 2015; Quadros et al., 2020), Vennam et al. (2017) found that the finer-resolution nested model produced changes in surface ozone and fine particulate matter (PM_{2.5}) which were 70 and 13 times smaller respectively than in a global model.

A related question is whether the air quality impacts of aviation are dominated by local sources (e.g. landing and takeoff (LTO) operations and flight through local airspace) or are the result of larger atmospheric changes. If the former is true, then regional regulations and emissions standards applied for only near-surface operations may be sufficient to reduce impacts. However, if impacts are dominated by large scale atmospheric responses to global aviation, then emissions standards will only be effective if applied both globally and to fuel burn beyond LTO – since LTO accounts for only 9.1% of the global total (Simone et al., 2013).

These questions urgently need to be resolved. New regulations for aviation NO_x emissions are being debated by the International Civil Aviation Organization. Since the establishment of the 1981 CAEE standard limiting NO_x emissions per unit of thrust for landing and take-off operations (LTO), subsequent CAEP regulations have continued to increase



stringency, resulting in the current CAEP/8 standard which was set in 2010 (ICAO, 2017). Since the amount of NO_x produced during take-off and during cruise are closely related for most current combustor architectures, these regulations are also the relevant limiting factor for cruise NO_x emissions. Recent studies have suggested that the climate impacts of greater NO_x emissions are sufficiently small relative to the benefits of reduced CO₂ that fuel efficiency should be prioritized over further NO_x emissions reduction (Skowron et al., 2021). However, if air quality impacts are included in this analysis then such a prioritization could cause net environmental damages rather than improvement – dependent on accurate estimation of the air quality consequences (Miller et al., 2022).

This study quantifies the global air quality response to aviation in a single, consistent modeling framework, evaluating both the role of model grid resolution and the relative contribution of local and remote aviation emissions. We use three different global resolutions to quantify how grid resolution affects the simulated outcomes, varying from 400 km to 50 km globally. To isolate the role that in-domain (“local”) and out-of-domain (“remote”) emissions have on air quality in the contiguous United States we use a perturbation approach, performing an additional simulation at each resolution in which aviation emissions over the North American domain are set to zero.

2 Method

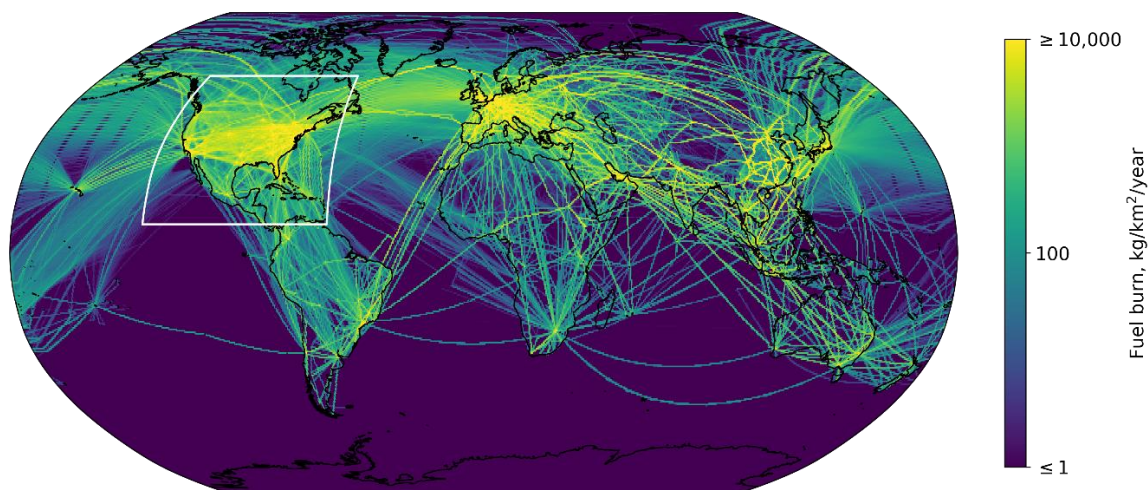
We simulate aviation’s impacts on global air quality at three resolutions, ~400 km (C24), ~100 km (C90), and ~50 km (C180), without the use of regional refinement or nesting. For each of the three resolutions, we perform three simulations. We first simulate global atmospheric composition using version 12.6.2 of the GEOS-Chem High Performance model (GCHP) (Eastham et al., 2018) with all aviation emissions enabled (AVGLOBAL) and with all aviation emissions disabled (AVOFF). The simulated differences in concentrations of ozone and particulate matter (PM_{2.5}) in the surface layer are then taken as the total air quality impacts of aviation, with health impacts calculated as discussed below.

These simulations are followed by a further simulation in which aviation emissions are included everywhere except over a “North American” domain covering 10-60°N and 60-130°W (AVNONA). This domain has previously been used in nested model simulations of regional air quality change, and can therefore be considered to be representative of models which are focused on regional change only (Kim et al., 2015; Hu et al., 2018). The difference in air quality changes between the AVGLOBAL and AVNONA is then taken as the effect of in-region emissions on regional air quality. The remaining differences in surface air quality within the region, calculated as the difference in air quality between AVNONA and AVOFF, then correspond to the effect of out-of-region emissions on surface air quality. Figure 1 shows global distribution of fuel burn and the region in which emissions are set to zero when estimating influence of “out-of-region” emissions on “in-region” air quality.

To quantify the effect of model resolution, we perform the above set of three simulations at three different global resolutions. The GCHP model uses a cubed-sphere model grid with finite volume advection (Putman and Lin, 2007). Model resolutions are therefore denoted as CN where N is the number of grid elements along each edge of the cube such that higher numbers



100 correspond to finer grid resolution. We perform our simulations at global resolutions of C180, C90, and C24, in which the average side lengths of a grid cell is 51 km, 100 km, and 380 km respectively. As such, they are also approximately equivalent to global resolutions of $0.5^\circ \times 0.625^\circ$, $1^\circ \times 1.25^\circ$, and $4^\circ \times 5^\circ$, but without distortion in grid cell size near the poles and equator. By comparing the air quality impacts calculated at each model resolution, we quantify the effect that increasing model resolution has on simulated air quality. This approach uses a single, global model to accomplish this, avoiding discrepancies due to the use of regional models or different chemical mechanisms (Yim et al., 2015; Vennam et al., 2017).



105 **Figure 1. Simulated global distribution of fuel burn in 2015. Data shown include fuel burned at all altitudes, at an approximate global resolution of $0.5^\circ \times 0.625^\circ$. The white box denotes the region in which fuel burn is set to zero when estimate the net effect of “out-of-region” aviation emissions on aviation’s air quality impacts in North America.**

2.1 Atmospheric simulation

We use the global chemistry transport model (CTM) GEOS-Chem version 12.6.2 (DOI) to simulate all scenarios, as
110 implemented in the GEOS-Chem High-Performance model (Eastham et al, 2018). GEOS-Chem includes unified tropospheric-stratospheric chemistry and has previously been used in a global, coarse-resolution configuration to estimate aviation’s impacts on surface air quality (Eastham and Barrett, 2016; Eastham et al., 2014). Multiple horizontal resolutions are used as described above, but all share a common vertical discretization using 72 non-uniform layers extending from the surface to a maximum altitude of around 80 km. Each simulation is integrated forwards in time for a total of 13 months from
115 July 1st 2014 through to July 31st 2015, of which the final 12 are used to calculate annual mean changes in surface ozone. This period is used to ensure that the northern hemisphere winter season (December through February, inclusive) is from a single, continuous period, as this is when the greatest air quality impacts from aviation are expected (Eastham and Barrett, 2016). All simulations use meteorological data from the NASA Global Modeling and Assimilation Office Modern Era Retrospective for Research and Analysis version 2 (MERRA-2) reanalysis product.



120 Emissions from civil aviation in 2015 are calculated using the Federal Aviation Administration (FAA) Aviation Environmental Design Tool (AEDT) (Wilkerson et al., 2010). This includes all emissions from civil airliners in that year during taxi, take-off, climb, cruise, descent, and landing operations. We do not account for military flights or business jets. Emissions are gridded at a resolution of $0.25^{\circ} \times 0.25^{\circ}$ globally, finer than the highest resolution model grid (C180). A fuel sulfur content of 600 ppm by mass is assumed for all scenarios, unless otherwise stated (Hileman et al., 2010). All nitrogen
 125 oxide (NO_x) and volatile organic compound (VOC) emissions are speciated as described in AEDT guidance documents (Barrett et al., 2010b). Table 1 shows the total emitted mass of each relevant compound. Black and organic carbon (BC and OC) emissions are estimated using a fixed emissions factor of 30 mgC per kg of fuel burned for each species. Emissions of a secondary organic aerosol precursor (SOAP) are calculated as 69 g of carbon per kilogram of CO emitted (Kim et al., 2015). GEOS-Chem uses a bulk aerosol parameterization and therefore does not require a size distribution or calculation of the
 130 number of particles emitted. The impacts of water vapor emissions and condensation trails on atmospheric composition are not included in this analysis.

Table 1. Total emitted mass for the fleet in 2015. The mass basis for the reported quantity is given separately for each species. All quantities are shown to two significant figures. The rightmost column shows the fraction of global emissions which occur in the simulated North American domain.

	Global	North America	% in North America
NO_x as NO, NO_2, and HONO (TgN)	1.1	0.26	24%
Carbon monoxide (Gg)	600	190	32%
Hydrocarbons (Gg CH_4 mass equivalent)	62	23	38%
Soot (black carbon, GgC)	7.2	1.8	25%
Organic carbon aerosol (GgC)	7.2	1.8	25%
SOA precursor (GgC)	41	13	32%
Sulfur dioxide (GgS)	140	36	25%
Sulfate aerosol (GgS)	2.9	0.73	25%
Total fuel burn (Tg)	240	61	25%

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Non-aviation emissions are provided by a collection of standard inventories, described in full in the Supplemental Information.

2.2 Health impact estimation

Air quality impacts are calculated based on the difference in concentrations of ozone and $\text{PM}_{2.5}$ in the lowermost simulated
 140 atmospheric layer. The lowermost layer is approximately 120 meters thick for a surface pressure of 1013.25 hPa, and the



average concentration within this surface layer is treated as the exposure-relevant value. Model predictions are not-bias corrected to observations, in part because the relatively small effect of aviation emissions on ozone and particulate matter has a different spatial pattern than background concentrations (Cameron et al., 2017). Bias correction may therefore impose a non-physical pattern on the changes in surface concentration due to aviation.

145 Impacts are calculated on the 30 arc-second ($1/120^{\text{th}}$ of a degree) global grid on which population density data is provided by the Gridded Population of the World version 4.11 (Center for International Earth Science Information Network - CIESIN - Columbia University, 2016). The concentrations of ozone and $\text{PM}_{2.5}$ are taken from whichever simulation grid cell contains that 30 arc-second cell, with no interpolation. If the population cell straddles the boundary between two simulation grid cells, the area-weighted mean concentration is used. The age distribution and baseline mortality rates within each grid cell are
150 supplied by the World Health Organization (WHO) through their 2016 Global Health Estimates (World Health Organization, 2018). For non-member countries of the WHO, we use the world region mean rate instead from the same source.

In each grid cell and for each age bracket, we calculate the relative risk of mortality due to chronic exposure to ozone and $\text{PM}_{2.5}$ with and without aviation (RR_{BASE} and RR_{NOAV} respectively). The change in the annual mortality (ΔM) due to some disease for that age bracket is then calculated for each grid cell as

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$$\Delta M = M_{\text{BASE}} \times \frac{\text{RR}_{\text{NOAV}} - \text{RR}_{\text{BASE}}}{\text{RR}_{\text{BASE}}} \quad (1)$$

where M_{BASE} is the number of mortalities due to that disease in 2016. The relative risk is calculated by comparing the simulated exposure-relevant concentration without aviation (χ_{NOAV}) to the concentration simulated when aviation is included (χ_{BASE}) using an appropriate concentration response function. In this case ΔM is expected to be negative, implying that
160 reducing aviation emissions to zero would reduce mortality rates. The mortality due to aviation reported in this paper is therefore $-1 \times \Delta M$.

For ozone, the increase in relative risk of mortality is calculated based on the association between exposure and mortality identified by Turner et al. (2015). Said study analyzed a cohort of 669,046 participants in the American Cancer Society Cancer Prevention Study II from 1982 to 2004, finding a 12% increase (95% confidence interval: 8.0-16%) in respiratory
165 mortality per 10 ppb increase in annual mean, maximum daily 8-hour average (MDA8) ozone concentration. We use this data in a log-linear concentration response function (CRF), such that the relative risk due to ozone exposure is calculated as

$$\text{RR}_{\text{NOAV}} = \exp(\beta_{\text{LL}}[\chi_{\text{NOAV}} - \chi_{\text{BASE}}]) \quad (2)$$

where the central value of β_{LL} is calculated as $\ln(1.12)/10 = 0.011 \text{ ppb}^{-1}$. Only adults over the age of 30 are included when calculating the increase in mortality, as this was the cohort in which the relationship was observed. Uncertainty in the
170 concentration response function is quantified by treating β_{LL} as a triangularly-distributed random variable. We use



$\ln(1.12)/10$ as the mode and fit a triangular distribution such that $\ln(1.06)/10$ lies at 2.5% and $\ln(1.18)/10$ as 97.5% along the cumulative distribution function.

For particulate matter, relative risks are calculated using the Global Exposure Mortality Model (GEMM) (Burnett et al., 2018). The GEMM is a set of non-linear concentration response functions which estimate the increase in relative risk of mortality based on associations with the annual-average, 24-hour mean $PM_{2.5}$ mass concentration at standard temperature and pressure. We apply the age-specific GEMM CRF for combined non-communicable disease and lower respiratory infection (NCD+LRI). The parameters for the GEMM are constructed based on a meta-analysis of 41 cohort studies worldwide examining the relationship between exposure to fine particulate matter and non-accidental mortality, and are described in detail in Burnett et al. (2018).

Impacts are calculated by performing 1,000 random draws of the CRF parameters in a paired Monte-Carlo simulation. From this, the mean and 95% confidence interval of mortality due to individual sources and mortality due to ozone and $PM_{2.5}$ combined can be estimated.

3 Results

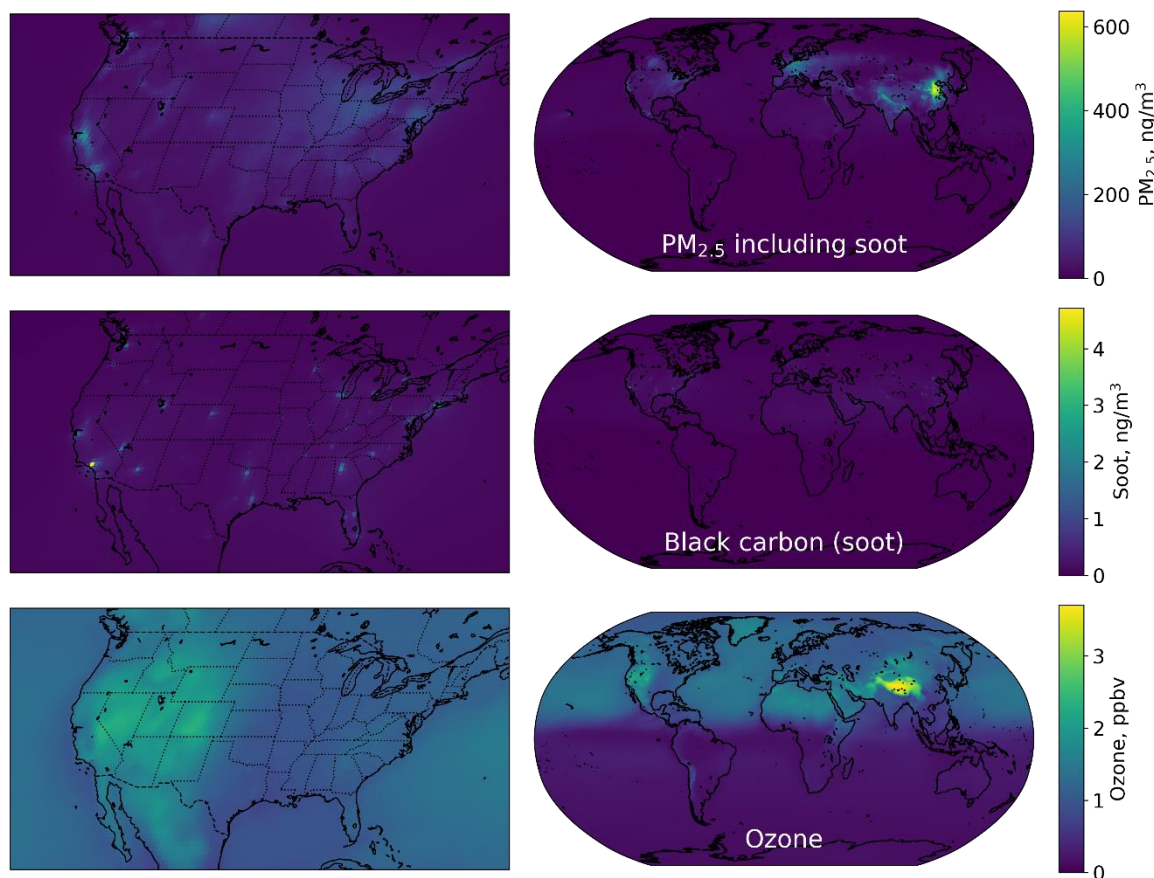
Based on results generated at ~50 km resolution globally, we find that the increase in $PM_{2.5}$ exposure results in an additional 21,200 (95% confidence interval: 19,400 to 22,900) mortalities globally each year. Aviation-attributable ozone exposure results in an additional 53,100 (36,000 to 69,900) mortalities. Combined, we estimate that aviation results in 74,300 mortalities (57,300 to 91,100), or 309 mortalities per teragram of fuel burned. These results are compared to prior studies at the end of this section.

Figure 2 shows how the changes in surface concentrations are distributed, both in the US (left) and globally (right). Changes in ozone are diffuse throughout the Northern Hemisphere, reaching a maximum over the Himalayan plateau and the Western United States. Changes in total particulate matter are more heterogeneous, with the greatest increases occurring over western China, northern India, and western Europe. This effect is likely driven by the effect of increased oxidant concentrations acting on non-aviation precursor gas emissions (Eastham and Barrett, 2016). Increases in soot (black carbon) are shown as soot is rapidly removed by precipitation, and therefore highlights near-airport regions as these are the locations most strongly affected by direct particulate matter emissions from aircraft. However, peak concentrations of aviation-attributable soot are two orders of magnitude smaller than the more diffuse increases in total $PM_{2.5}$.

Our findings suggest that direct exposure to cruise-attributable ozone is the largest air quality concern for aviation, causing 2.5 times as many premature mortalities as $PM_{2.5}$. This means that the dominant air quality impact of aviation is not localized to airports, but is also not globally or regionally homogenous. Although aviation-attributable ozone is present throughout the Northern Hemisphere with a mean mixing ratio of 1.1 ppb, elevated regions can reach annual average mixing ratios in excess of 3 ppb. In these regions air more rapidly reaches the surface from the cruise level, meaning that less of the cruise-altitude ozone is destroyed before the air reaches the population. Although the greatest increases are on the Himalayan plateau,



205 aviation-attributable ozone in the western US also exceeds 2 ppb. This is due to the descent of upper tropospheric air in the lee of mountain ranges, such that the largest impacts of aviation emissions on US air quality are just east of the Rocky Mountains.



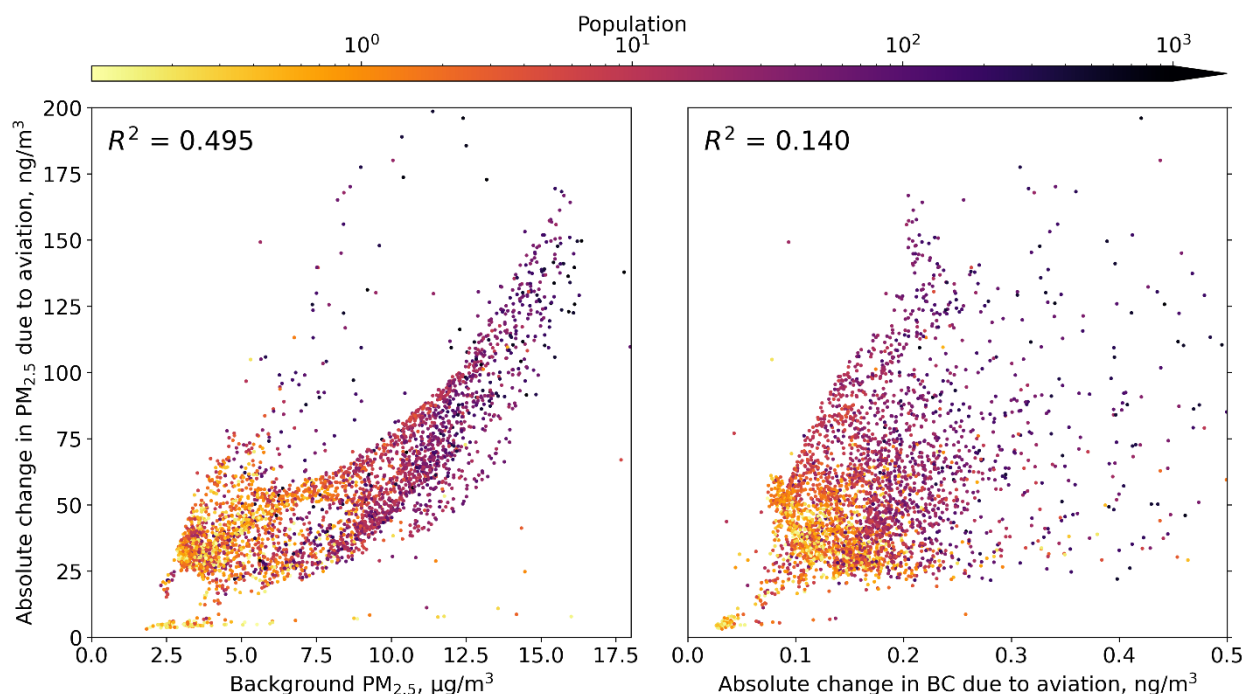
210 **Figure 2. Changes in surface air quality due to global aviation. From top to bottom: changes in PM_{2.5} (including soot), changes in soot only, and changes in ozone. The left column is a zoomed-in version of the right plot, focusing on the contiguous United States. Soot is explicitly singled out due to its potential as an indicator of the direct influence of aviation exhaust without any chemical intermediaries.**

215 The distribution of aviation-attributable PM_{2.5} follows a different pattern. Aviation-attributable soot is concentrated near airports, but this peaks at less than 5 ng/m³. The greatest changes in PM_{2.5} exceed 600 ng/m³ and are spread over larger areas in locations which already have elevated background pollution concentrations, such as western China, central Europe, and the US Northeast.

Figure 3 further explores the factors contributing to aviation-attributable increases in surface PM_{2.5}. The change in surface PM_{2.5} due to aviation is correlated with the concentration of background (non-aviation) PM_{2.5}, calculated as the concentration with all other sources included, with an R^2 of 0.495. However, aviation-attributable PM_{2.5} concentration is less strongly



220 correlated (R^2 of 0.140) with the change in aviation-attributable black carbon, an indicator of nearby aviation activity. This provides additional evidence that increases in $PM_{2.5}$ due to aviation for a location in the Northern Hemisphere occur because aviation-attributable ozone contributes to the formation of secondary particulate matter from existing precursors.



225 **Figure 3. Scatter plots of the change in $PM_{2.5}$ due to aviation as a function of other variables. Left: the absolute change in surface $PM_{2.5}$ due to aviation plotted as a function of the background (non-aviation) concentration of $PM_{2.5}$, colored by the total population in the grid cell. Right: aviation-attributable $PM_{2.5}$ plotted as a function of the absolute change in black carbon due to aviation. Each point corresponds to a single grid cell at C180 resolution. Only grid cells in the Northern Hemisphere with a population of at least 0.1 people per square kilometer are shown. R^2 values are based on a linear least-squares fit.**

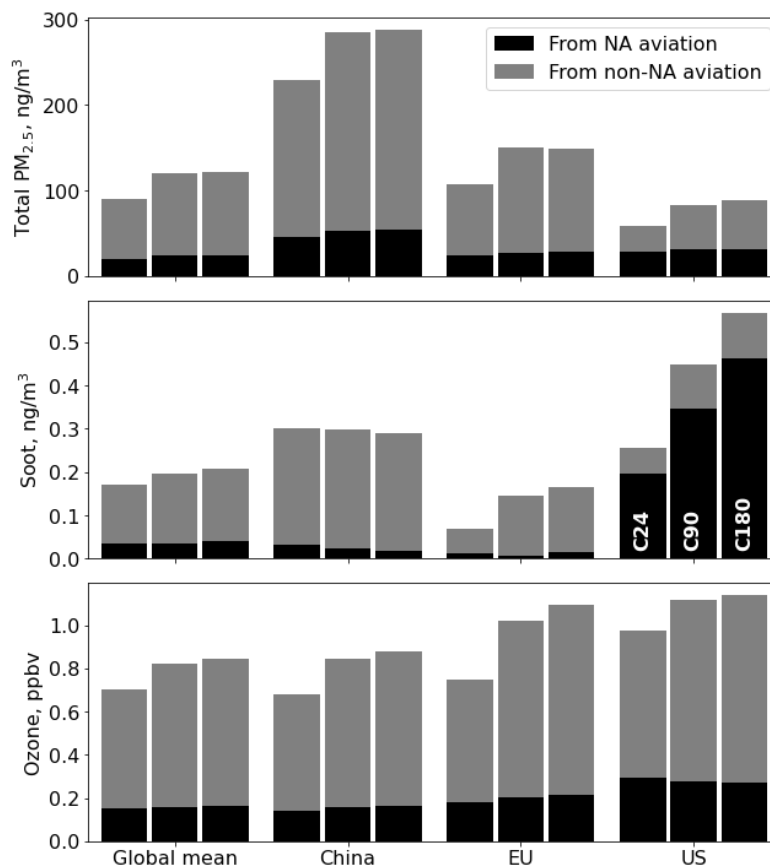
3.1 The effect of resolution and out-of-region emissions

230 Figure 4 shows the population-mean exposure to $PM_{2.5}$ (including soot), soot, and ozone for the global mean, China, the 27 European Union member states (“EU”), and the US when simulations are carried out at three different resolutions. We find that simulations performed at the coarsest, C24 (~400 km) resolution result in a global mean ozone exposure which is 17% lower than when calculated at the C180 (~50 km) resolution, compared to 3.0% lower at C90. For $PM_{2.5}$ the exposure is 26 and 1.2% lower respectively, whereas for soot it is 18 and 6% lower. In total, the net mortalities calculated when performing simulations at a resolution of ~50 km are 2.5% higher than at ~100 km, and 25% higher than at ~400 km resolution.

235 The greater resolution has two effects: physical phenomena are more finely resolved, and changes in surface air quality which occur local to population centers can be more accurately collocated. These effects can be separated by performing exposure calculations using the same high-resolution (~50 km) output data but downgrading it to low resolution (~400 km) before calculating exposure. Doing so increases the calculated ozone exposure by 3.7% but decreases calculated $PM_{2.5}$



240 exposure by 14%. This suggests that the higher-resolution simulation of atmospheric phenomena is more important than population collocation for ozone but that the collocation effect is significant for $PM_{2.5}$, consistent with prior work focused on non-aviation exposure (Punger and West, 2013).



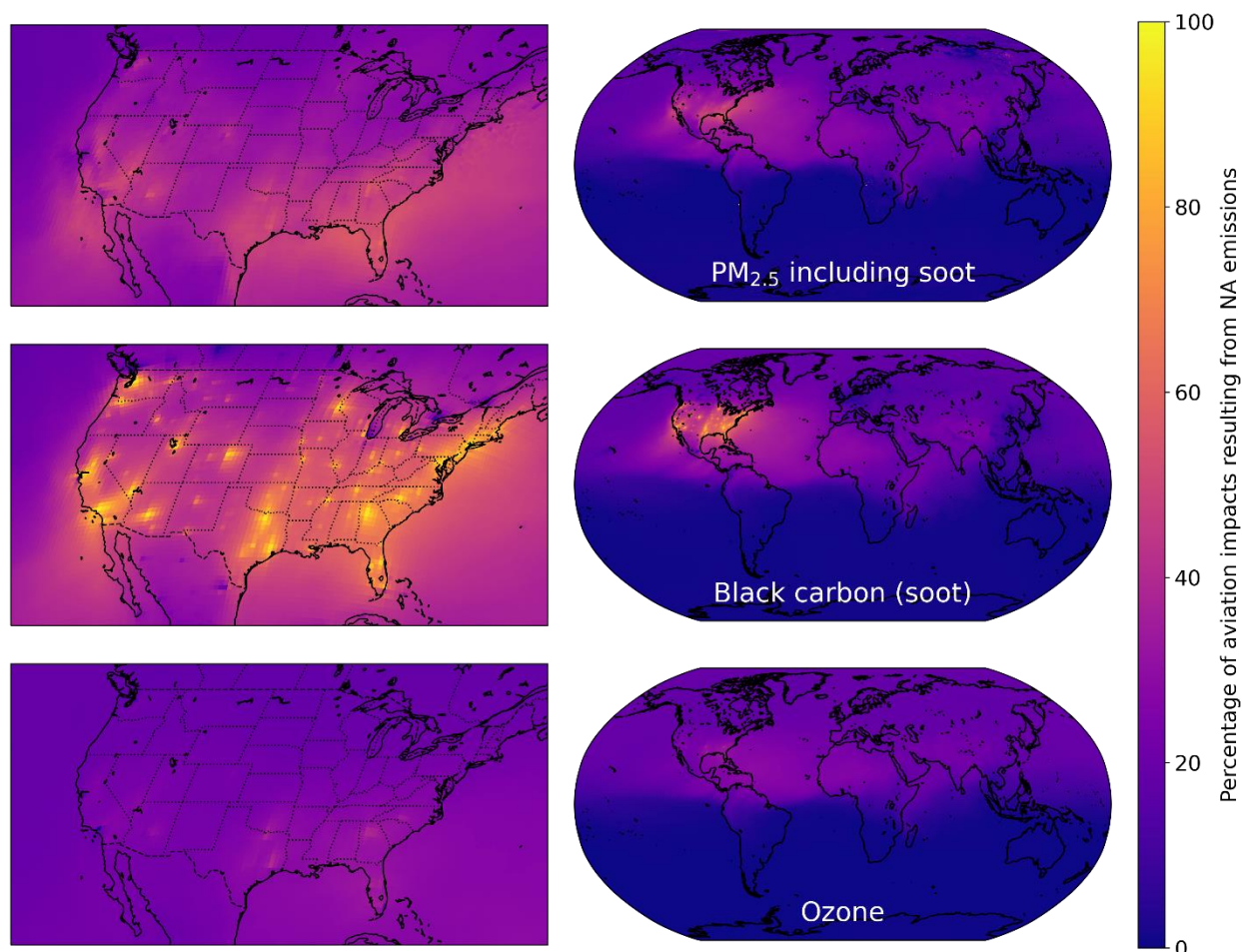
245 **Figure 4. Changes in air quality resulting from aviation emissions, calculated for three different resolutions (C24, C90, and C180 from left to right). Black bars show the contribution of emissions over North America (NA) only, while grey bars show the contribution of non-NA emissions.**

Concerning the role of long-range transport compared to in-region emissions, of the 21,200 mortalities due to aviation-attributable exposure to $PM_{2.5}$, 1,610 (1,470 to 1,740) occur in the United States, compared to 3,940 (2,670 to 5,180) of the 53,100 mortalities due to aviation-attributable ozone exposure. This yields a combined estimate of 5,550 (4,290 to 6,790) premature mortalities in the US due to global, year-2015 aviation emissions. Fuel burn in the NA domain therefore results in an additional 25 US mortalities per Tg of fuel burn, whereas fuel burn outside of the NA domain results in 22 US mortalities per Tg of fuel burn.

The relative contribution of NA aviation emissions to aviation-attributable surface air quality degradation in each region is shown in Figure 4 as the black segment of each bar. Aviation emissions from within the NA domain contribute 37% and



255 24% of aviation-attributable exposure to $PM_{2.5}$ and ozone respectively in the US. At C24 (~400 km) resolution these contributions are higher - 49% and 30% respectively - despite total exposure being 26% and 17% lower relative to C180 (~50 km).



260 **Figure 5. Relative contribution of aviation in the North America (NA) domain to all air quality impacts of aviation. Light colors indicate that emissions over the NA domain are the dominant contributor, while dark colors indicate that non-NA emissions are dominant. Data is shown for calculations performed at a global resolution of C180 (~50 km).**

The same NA emissions contribute to a lower proportion (19%) of aviation-attributable exposure to $PM_{2.5}$, but a greater absolute value of 54 ng/m^3 in China compared to 32 ng/m^3 in the US. This is consistent with the hypothesis that aviation-attributable emissions are promoting the formation of $PM_{2.5}$ from non-aviation sources, rather than direct emissions of aviation $PM_{2.5}$ being responsible.

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The exception to this is soot. 82% of US exposure to aviation-attributable soot is the result of aviation emissions from within the NA domain, compared to 6.2% of exposure in China. However, aviation-attributable soot exposure in the US at C180 makes up 1.4% of total aviation-attributable $PM_{2.5}$ exposure due to the rapid wet scavenging of soot from the atmosphere



(Wang et al., 2014). The reason for the greater concentration and resolution-sensitivity of aviation-attributable soot in the US
270 is unclear.

This result is shown globally in Figure 5. As in prior figures, we show the annual average value of $PM_{2.5}$ and ozone. The
relative contribution of NA-domain aviation emissions to US air quality is greatest for black carbon, where it reaches nearly
100% around airports. For ozone, impacts across most of the Northern hemisphere are close to the fractional contribution of
NA airspace (as defined in Figure 1) to global fuel burn, 25%. For $PM_{2.5}$ the relative contribution is more heterogeneous than
275 for ozone but less than for black carbon. This is due to the contribution of secondary particulate matter, such as ammonium
nitrate and acidic sulfate aerosols, which form in response to aviation-attributable ozone.

3.2 Comparison to previous work

These impacts of aviation emissions on global air quality and mortality are consistent with a recent analysis which found that
year-2005 aviation emissions resulted in 58,000 mortalities globally, of which 38,000 were due to ozone exposure (Quadros
280 et al., 2020). Our 2015 emissions inventory includes 33% more fuel burn by mass than was used for their study, and our
estimate of the net mortality impact is also 33% greater. The larger contribution of ozone in this work and Quadros et al.
(2020) relative to earlier assessments is due to the use of more recent epidemiological data which includes wintertime ozone
exposure (Turner et al., 2015). This is significant because aviation-attributable ozone is maximized during winter and
minimized during summer (Eastham and Barrett, 2016). If we instead use epidemiological data from an earlier study by
285 Jerrett et al. (2009) which considers only summertime ozone, ozone-related health impacts of aviation emissions are
estimated to be 68% lower and net impacts 49% lower. A more detailed assessment of the sensitivity of our conclusions to
the health impact assessment method is provided in the Supplemental Information.

Our results constitute a significant increase in the estimate air quality impacts of aviation relative to studies such as Eastham
and Barrett (2016) and Barrett et al. (2010a). The former reported 6,800 premature mortalities per year due to aviation-
290 attributable ozone exposure and 9,200 due to aviation-attributable $PM_{2.5}$ exposure, compared to 53,100 and 21,200
respectively in this work. Assuming linear or near-linear relationships the factor of 7.8 increase in ozone-related impacts is
mostly attributable to the updated epidemiological data as described above, which results in a 3.2 times increase in ozone-
related mortality. This is accompanied by a factor 1.6 increase due to the Turner et al. data being applicable to all respiratory
diseases and not just chronic obstructive pulmonary disease and asthma, yielding a net factor of 5.0 relative to the approach
used in Eastham and Barrett (2016). An additional factor of 1.3 is due to increased fuel burn, since our work examines 2015
295 whereas these previous studies examine aviation in 2006. We also estimate that the use of finer resolution of our analysis
(~50 km) compared to the ~500 km resolution of the previous studies results in an increase in the estimated ozone-related
impacts of aviation emissions by a factor of 1.2. These three factors combined imply a 7.6 times increase in ozone-related
impacts, consistent with the factor of 7.8 which is observed.

300 For $PM_{2.5}$ we estimate 2.3 times as many mortalities as were estimated by Eastham and Barrett (2016). This smaller factor is
only partially explained by the factors described above. While the same factor of 1.3 applies for fuel burn, the concentration



response function used here results in 14% fewer mortalities being attributed to aviation emissions than if the approach used in the prior study is applied (Hoek et al., 2013). The greater resolution of our work increases the estimated PM_{2.5}-related mortality by a factor of 1.35, but combined these result in a factor of 1.5 compared to the observed factor 2.3 increase.

305 The remaining factor 1.6 difference in PM_{2.5} mortality between the two studies may be in part due to growth in non-aviation emissions between 2006 and 2015. For this work we use emissions from 2014 in the Community Emissions Data System (CEDS) (Hoesly et al., 2018) as a proxy for anthropogenic, non-aviation emissions. Figure 2 shows that the highest aviation-attributable concentrations of PM_{2.5} are found in Asia, and analysis of CEDS data suggests that emissions of NO and SO₂ from non-aviation sources in Asia (defined as a region bounded by 60 to 150°E and 10°S to 55°N) increased by 20% and

310 0.16% respectively between 2006 and 2014. In the same period, NO and SO₂ emissions fell by 5.3% and 16% respectively outside of Asia. This change was accompanied by a 9.4% increase in ammonia emissions in Asia, and a 7.3% increase outside. Such changes would increase ambient concentrations of PM_{2.5} precursors, and therefore increase the amount of PM_{2.5} formed as a result of aviation-attributable ozone descending to the surface. Other possible contributors to the increase in PM_{2.5} exposure-related mortality attributable to aviation include changes in baseline mortality rates and increases in the

315 exposed population. For example, using the same geographical boundaries as were applied to calculate changes in Asia's emissions from CEDS, the total population in Asia increased by 11% from 2005 to 2015. A simple linear combination of increases – 20% in NO, 9.4% in ammonia, and 11% in population – would imply a factor 1.45 increase in exposure. However a detailed analysis would be required to fully understand the sources of the unexplained factor 1.6 difference.

Discussion

320 This work finds that aviation's air quality impacts are greater than has been previously estimated. Since prior assessments have shown aviation's impacts on air quality to be similar in magnitude to its impacts on the climate (Grobler et al., 2019), this work suggests that impact mitigation options which do not address NO_x emissions-attributable air quality impacts will therefore not address one of the largest environmental impacts of aviation. This may change the balance of cost-effectiveness when considering trade-offs between CO₂ emissions and NO_x emissions.

325 These results also imply that coarse-resolution global simulations may underestimate the impacts of aviation on surface air quality. However, moderate-resolution studies at C90 (around 100 km or 1° resolution) resolve 98% of the impacts calculated by a C180 simulation at one quarter of the computational cost (or one eighth if a smaller time step is required). This includes 99% of PM_{2.5} impacts and 97% of ozone impacts.

This work explains differences in the literature regarding aviation's air quality impacts. Yim et al. (2015) found that the use of finer resolution nested modeling increased estimated ozone exposure by 12% but decreased exposure to PM_{2.5} by 29%,

330 whereas Vennam et al. (2017) found reductions of more than 90% in both. We hypothesize that the relatively smaller differences with the former study are due to their use of two inconsistent models to represent the global and nested regions. The latter study used boundary conditions from a single global simulation which included aviation emissions for both their



aviation and non-aviation regional simulations of North America, inherently removing the effect of aviation NO_x on
335 hemispheric-scale tropospheric ozone and therefore neglecting the influence of a larger atmospheric response. Our work
shows that this global response drives the majority of the change in surface air quality, as evidenced by the finding that each
kilogram of fuel burned in US air space causes only 14% more health impact for the US population than each kilogram
burned outside US air space.

A remaining important uncertainty regards the response of human health to ozone exposure. Although several recent
340 epidemiological analyses have findings consistent with Jerrett et al. (2009) and Turner et al. (2015) that ozone exposure
increases morbidity and mortality (Zhao et al., 2021; Lim et al., 2019; Rhee et al., 2019), including at low concentrations
(Yazdi et al., 2021), other studies have suggested that ozone exposure may not be significantly associated with mortality
(Atkinson et al., 2016; Brunekreef et al., 2021; Huangfu and Atkinson, 2020). This would not affect the finding that over
20,000 mortalities per year due to $\text{PM}_{2.5}$ exposure are attributable to aviation.

345 There has also been research suggesting that exposure to certain specific components of $\text{PM}_{2.5}$ may be more harmful than
other constituents of $\text{PM}_{2.5}$, which may indicate greater impacts related to soot, ultrafine non-volatiles, or organic carbon
(Verma et al., 2015). If so, our finding of a greater relative sensitivity of soot exposure to model resolution suggests that the
more localized impacts of aviation soot emissions will require higher-resolution simulations or localized modeling
approaches such as those in Yim et al. (2015) to quantify.

350 **Conclusions**

Our findings show that 74,300 premature mortalities each year are attributable to aviation emissions, based on 2016 data.
This is 4.6 times greater than a previous assessment for 2005 finding 16,000 premature mortalities each year. A factor of
1.29 is due to our use of a global model with greater spatial resolution; a factor 1.84 is due to the availability of an ozone
concentration response function which includes wintertime ozone; a factor 1.24 is due to the inclusion of a broader set of
355 diseases in those impacted by ozone exposure; and a factor 1.28 is due to increases in aviation fuel burn. Accounting for the
different effects each of these factors has on ozone and $\text{PM}_{2.5}$ -attributable mortality, these alone explain a factor 4.1 increase
in the estimate, 87% of the total. We hypothesize that the remaining 13% discrepancy may be the result of regional increases
in non-aviation emissions.

We find no evidence of a decrease in air quality impact with increasing model resolution. Instead we find that the simulated
360 impacts of aviation emissions on surface air quality increase by 24% when using a 50 km model compared to a 400 km
resolution model. Impacts simulated at a resolution of 100 km globally are within 2% of those at 50 km, suggesting that
moderate-resolution simulations are capable of accurately simulating aviation's air quality impacts.

Finally, we show that the impacts of aviation on air quality are global in nature. Emissions from aircraft flying over and near
North America cause 25 mortalities per Tg of fuel burn in the US, compared to 23 mortalities in the US per Tg of fuel
365 burned outside this region. We also find that $\text{PM}_{2.5}$ concentrations in China are more strongly affected in absolute terms by



North American aviation emissions than concentrations of $PM_{2.5}$ in the US, and that concentrations of aviation-attributable $PM_{2.5}$ are more strongly correlated with the concentration of non-aviation $PM_{2.5}$ than with a marker of local aviation emissions.

370 Previous work has shown that the monetized air quality impacts of a unit of aviation fuel burn are similar in magnitude to aviation's monetized climate impacts, including effects of contrails. The further increase in estimated air quality impacts we find suggest that aviation-attributable air quality degradation is a significant contributor to aviation's environmental impacts, and that mitigation of full-flight NO_x emissions should be considered alongside ongoing efforts to reduce the effects of aviation on the climate.

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Code availability statement

This work was completed using version 12.6.2 of the GEOS-Chem High Performance (GCHP) code. The specific codebase used can be found at https://github.com/geoschem/gchp_legacy/releases/tag/12.6.2.

385 Author contributions

SDE, RLS, and SB conceived and designed the study. GC generated the aircraft emissions data. SDE performed all model simulations and analysis. DJJ provided computational resources for high-fidelity simulation. All authors contributed to the manuscript writing and editing.

Competing interests

390 The authors declare no competing interests.



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