

# How have Divergent Global Emission Trends Influenced Long-range Transported Ozone to North America?

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## Key Points

- Changes in global emission patterns have and continue to alter long-range transport and background O<sub>3</sub> air pollution levels
- Long-range transported O<sub>3</sub> to North America has a prominent contribution from open sea shipping emissions and needs improved quantification
- Stratospheric O<sub>3</sub> (ranging between 6-20ppb) constitutes 29-78% of the estimated Spring-time background O<sub>3</sub> across the continental U.S.

**Abstract:** Several locations across the United States in non-compliance with the national standard for ground-level ozone (O<sub>3</sub>) are thought to have sizeable influences from distant extra-regional emission sources or natural stratospheric O<sub>3</sub>, which complicates design of local emission control measures. To quantify the amount of long-range transported O<sub>3</sub> (LRT O<sub>3</sub>), its origin, and change over time, we conduct and analyze detailed sensitivity calculations characterizing the response of O<sub>3</sub> to emissions from different source regions across the Northern Hemisphere in conjunction with multi-decadal simulations of tropospheric O<sub>3</sub> distributions and changes. Model calculations show that the amount of O<sub>3</sub> at any location attributable to sources outside North America varies both spatially and seasonally. On a seasonal-mean basis, during 1990-2010, LRT O<sub>3</sub> attributable to international sources steadily increased by 0.06-0.2 ppb yr<sup>-1</sup> at locations across the United States and arose from superposition of unequal and contrasting trends in individual source-region contributions, which help inform attribution of the trend evident in O<sub>3</sub> measurements. Contributions of emissions from Europe steadily declined through 2010, while those from Asian emissions increased and remained dominant. Steadily rising NO<sub>x</sub> emissions from international shipping resulted in increasing contributions to LRT O<sub>3</sub>, comparable to those from Asian emissions in recent years. Central American emissions

contribute a significant fraction of LRT  $O_3$  in southwestern United States. In addition to the LRT  $O_3$  attributable to emissions outside of North America, background  $O_3$  across the continental United States is comprised of a sizeable and spatially variable fraction that is of stratospheric origin (29-78%).

**Plain Language Summary** Though implementation of control measures and technological advances have reduced ground-level ozone pollution across the United States, increasing amounts of ozone pollution inter-continently transported from other world regions with increasing emissions, combined with changing and uncertain amounts of natural ozone from variability in stratosphere-troposphere exchange processes, can confound design and implementation of local air pollution abatement strategies. “Background” or amount not produced locally, constitutes a larger fraction of the ozone pollution at a location as local control measures are implemented. Detailed model calculations are analyzed to explain the role of the dominant source regions that drive the 1990-2010 changes noted in observed ground-level ozone measurements across the U.S. In recent years, the contributions of shipping emissions to ozone imported to the U.S. troposphere are comparable to those from transport of ozone attributable to East Asian emissions and could be higher in the future if commercial shipping operations were to increase in response to anticipated growth in seaborne trade. In addition to the  $O_3$  attributable to emissions from different world regions, air-masses entering the North American domain have sizeable contributions of natural ozone of stratospheric origin, variability in which needs improved quantification to guide background ozone assessments and policy deliberations.

## 1. Introduction

Ground-level ozone ( $O_3$ ) is a key air pollutant of concern due to its harmful effects on human health, vegetation, and crop yield (Lippman, 1993; Chameides et al., 1994; Krupa et al., 1998; U.S. EPA, 2020a). Though the chemistry of  $O_3$  formation in the troposphere from its precursor oxides of nitrogen ( $NO_x$ ) and volatile organic compounds (VOCs) has long been understood and studied (Haagen-Smit and Fox, 1954), clear attribution of its levels at a location is complicated by contributions from multiple sources including domestic and international anthropogenic precursor emissions as well as natural sources including lightning, soil, wildfires, biogenic emissions, and stratosphere-troposphere exchange (STE). Substantial progress has been achieved in reducing ground-level  $O_3$  concentrations and consequent human and ecosystem exposure across the continental United States as a result of a combination of control measure implementation as well as technological advances that have significantly reduced emissions of  $NO_x$  and VOCs over the past three decades (U.S. EPA, 2020b), and is evident in numerous studies analyzing long-term trends in observed surface-level  $O_3$  mixing ratios (e.g., Cooper et al., 2012, Simon et al., 2015, Xing et al., 2015; Mathur et al., 2018; Gaudel et al., 2018). Despite this progress, attainment of the  $O_3$  National Ambient Air Quality Standard (NAAQS) continues to remain a challenge at many locations across the continental United States, with studies suggesting that a large fraction of the population is still exposed

to unhealthy levels of  $\text{O}_3$  pollution (ALA, 2020) and recommending standards lower than the current NAAQS (e.g., Cromar et al., 2019). As measures to limit domestic emissions are implemented, the contributions of natural sources and increasing  $\text{O}_3$  pollution transported from other regions become more influential in shaping local  $\text{O}_3$  levels and confound implementation of abatement measures and design of emission strategies to meet more stringent NAAQS.

Strategies for reducing pollution levels in surface air over a region are complicated not only by the interplay of local emission sources and complex physical, chemical, dynamical processes in the atmosphere, but also by the hemispheric background levels of pollutants (NRC, 2009; TFHTAP, 2010). Previous analyses of long-term observations of ground-level  $\text{O}_3$  have shown distinct regional and seasonal trends across the measured  $\text{O}_3$  concentration distributions. While ground-level  $\text{O}_3$  has generally declined at non-urban locations across the eastern United States, many locations in the western United States and at higher elevations indicate increasing trends in the lower  $\text{O}_3$  percentiles measured at least through 2010 (Lin et al., 2000; Jaffe et al., 2003; Cooper et al., 2012). This increase is suggestive of increasing  $\text{O}_3$  flowing into the western United States, which has been linked to rising amounts of long-range transported Asian pollution that reach the lower free troposphere and surface of the western United States (Jacob et al., 1999; Jaffe et al., 1999; Parish et al., 2004; Lin et al., 2017). However, the paucity of long-term records to characterize changes in  $\text{O}_3$  and precursor species aloft in the atmosphere, where transport is efficient, make it difficult to accurately attribute the noted changes in surface  $\text{O}_3$  to distant sources and the space and time variations in their contributions (cf. Mathur et al., 2018).

Large and contrasting changes in emission patterns across the globe (e.g. declining emissions in North America and Western Europe in response to implementation of control measures and increasing emissions across Asia, Central America, and Eastern Europe due to economic and population growth) have and continue to result in heterogeneous changes in the tropospheric chemical composition (e.g., Xing et al., 2015), and are likely altering long-range transport, and consequently impacting background pollution levels at receptor regions. Fiore et al. (2009) examined intercontinental source-receptor relationships for  $\text{O}_3$  pollution using simulations from multiple models that simulated impacts of a fixed emission perturbation from four continental-scale northern midlatitude source regions on surface  $\text{O}_3$  levels in the same regions. Their results illustrate that air pollution levels in any receptor region are influenced by emissions originating in multiple source regions, thereby suggesting that both the impacts of changes in emissions in the individual source regions and potential benefits of emission reductions are likely unevenly distributed across multiple receptor regions. Thus, understanding the space and time contributions of long-range transported  $\text{O}_3$  (LRT  $\text{O}_3$ ) pollution, its attribution to various source regions, and the possible changes in such attribution over time is important for guiding development of future national and local air pollution abatement strategies.

To shed light on these issues and better understand the contributions of different source regions across the Northern Hemisphere to long-range transported  $O_3$  to North America, we exercise a detailed sensitivity analysis method in the Hemispheric Community Multiscale Air Quality (HCMAQ) modeling system (Mathur et al., 2017), and analyze these in conjunction with long-term simulations for the 1990-2010 period to assess how regionally heterogeneous changes in  $NO_x$  and VOC emissions across the Northern Hemisphere influenced long-range transport of  $O_3$  and the changing contributions of the source regions in response to emission changes during this multi-decadal time period. Conducting the analyses for a historical time-period, for which model predictions of atmospheric composition have been extensively evaluated relative to available observations in previous studies, provides greater confidence in the inferred source-region contributions and changes in long-range  $O_3$  transport patterns.

## 1. Methodology

### (a) Modeling System and Simulation Details

In this analysis, we combine detailed sensitivity calculations describing the response of tropospheric  $O_3$  distributions to emissions from different regions with previously conducted multi-decadal simulations to characterize changes in tropospheric  $O_3$  across the Northern Hemisphere during the period 1990-2010 (Xing et al., 2015). These simulations utilized the Community Multiscale Air Quality (CMAQ) version 5.0 (doi:10.5281/zenodo.1079888) extended for hemispheric applications (HCMAQ) (Mathur et al., 2017). The process configuration options used in the model calculations described here follow those detailed in Xing et al. (2015) and Mathur et al. (2017). Briefly, the horizontal domain covering the Northern Hemisphere set on a polar stereographic projection (see Figure 1) is discretized with a 187x187 grid of 108 km resolution. A terrain-following vertical coordinate utilizing 44 layers of variable thickness is used to resolve the model vertical extent between the surface and 50 hPa. To minimize interpolation error and to avoid introducing mass imbalances, hemispheric simulations with CMAQ inherit the projection and grid structure from the Weather Research and Forecasting (WRF) model, which provides the driving meteorological fields. The WRF simulations, in turn, were constrained by NCEP/NCAR Reanalysis data with a  $2.5^\circ$  spatial and 6-hour temporal resolution. WRF’s simulation of hourly surface temperature, relative humidity, wind speed and direction have previously been evaluated by Xing et al. (2015) through comparison with observations and no significant bias in the meteorological fields were detected.

Year specific estimates of emissions over the Northern Hemisphere were derived from the EDGAR (Emission Database for Global Atmospheric Research, version 4.2; European Commission, 2011) database which reports emissions for 17 anthropogenic sectors and large-scale biomass burning on a  $0.1^\circ \times 0.1^\circ$  resolution grid. Biogenic VOC (based on the MEGAN model; Guenther et al., 1995) and lightning  $NO_x$  (Price et al, 1997) emissions were obtained from GEIA (Global Emission Inventory Activity; <http://www.geiacenter.org>). Xing et al. (2015) further describe the processing of global emission inventories for CMAQ, includ-

ing temporalization of the annual estimates to hourly model inputs, vertical distributions of anthropogenic and lightning emissions, and speciation of fine particulate matter ( $\text{PM}_{2.5}$ ) and non-methane volatile organic compounds (NMVOC) emissions to model primary aerosol constituents and gas-phase species. Emissions of nitric oxide (NO) from soil or sulfur dioxide ( $\text{SO}_2$ ) from volcanos were not considered in the applications presented here.

The lateral boundaries of the hemispheric domain are in the area encircling the Earth near the equator where because of the presence of the intertropical convergence zone (ITCZ), the mixing of air masses originating in the Northern and Southern Hemispheres occurs relatively slowly. The chemical lateral boundary conditions were set to represent clean tropospheric conditions and as detailed in Mathur et al. (2017) their impacts are typically confined to the lower latitudes and do not propagate into the domain. CMAQ was configured to use the Carbon Bond 2005 chemical mechanisms with updated toluene chemistry (CB05TU) gas-phase chemistry mechanism with modifications to represent multi-day chemistry associated with cycling of  $\text{NO}_x$  through reservoir organic nitrate species as described in Mathur et al. (2017). These involved updating the rate constant for the organic nitrate reaction with hydroxyl radical (OH) to that for isoprene nitrates (since on the hemispheric scale organic nitrates formed from isoprene are the largest contributor to its simulated tropospheric budget), and enhancing its physical sinks via dry deposition and wet scavenging (since the Henry’s law constants for several alkyl nitrates of biogenic emissions have been suggested to be higher than peroxyacetyl nitrate (PAN), e.g., Treves and Rudich (2003)). Additionally, to represent the impacts of STE on tropospheric  $\text{O}_3$  distribution, we specify  $\text{O}_3$  in the modelled upper troposphere/lower stratosphere (UTLS) based on estimated potential vorticity (PV).  $\text{O}_3$  at the model top is specified by multiplying the space- and time-varying model-estimated PV in the top-most model layer with a scaling factor of 20 ppb/PV-unit based on previously developed UTLS  $\text{O}_3$ -PV correlations (Mathur et al., 2008). It should be noted that while both the treatment of organic nitrates and the approach to PV scaling have been significantly enhanced in recent versions of the CMAQ modeling system, using the two specific process configuration choices described above enable us to conduct detailed sensitivity calculations (discussed in next section) with a model configuration that matches the one used to create the 1990-2010 multi-decadal tropospheric composition time-series and combine these two sets of model calculations in a complementary manner to infer changes in long-range transport of  $\text{O}_3$ .

## 1. Sensitivity Analysis Method

The decoupled direct method in three-dimensions (DDM-3D), implemented in the CMAQ modeling system (Napelenok et al., 2008), is an efficient and accurate approach for probing the sensitivity of atmospheric pollutant concentrations to changes in model inputs and parameters. DDM is a formal mathematical method that uses direct derivatives of equations governing the evolution of species concentrations in the atmosphere to calculate the local sensitivities

to a variety of model parameters and input data (Yang et al., 1997). Since the equations governing the sensitivity coefficients have a structure similar to that of the governing species conservation equation in CMAQ, the sensitivity coefficient calculations can elegantly follow, in a decoupled manner, the process parameterizations and numerical methods used to solve the governing equations for the species concentrations. Both first- and second-order sensitivity coefficients can be estimated using the current implementation of DDM-3D in CMAQ. The second order sensitivities help better capture non-linearities in the system (Hakami et al., 2003; Cohan and Napelenok, 2011), as would be useful in describing O<sub>3</sub> formation and evolution in plumes near-downwind of source regions.

The sensitivity analysis method enables one to examine the response of the modeled concentrations field to perturbations in modeled parameters. The perturbation to the parameter  $P_j$  (in the base case simulation) can be described as

$$\varepsilon_j P_j = (1 + \varepsilon_j) P_j = p_j \quad (1)$$

where,  $\varepsilon_j$  is a scaling factor and  $p_j$  is the perturbed sensitivity parameter. Following Napelenok et al. (2008), the semi-normalized first order sensitivity ( $S_j$ ) of concentration of a species in response to perturbation in  $p_j$  can then be estimated by scaling to its nominal value,  $P_j$ , as

$$S_j = P_j \frac{\partial C}{\partial p_j} = P_j \frac{\partial C}{\partial (\varepsilon_j P_j)} = \frac{\partial C}{\partial \varepsilon_j} \quad (2)$$

Note that  $S_j$  has the same units as concentrations and  $\varepsilon_j$  is unitless. Combining Equation 2 with the governing species conservation equation in the model yields the governing equation for  $S_j$  analogous to that for the species concentrations; thus, the species sensitivity calculation can be propagated through all the same processes as for the species concentration, yielding a three-dimensional species sensitivity field (Hakami et al., 2003; Napelenok et al., 2008; Cohan and Napelenok, 2011). Projecting the fractional perturbations from the base case simulations, the perturbed concentrations can be approximated using the Taylor series expansion:

$$C_{\text{perturb}} \approx C_{\text{base}} + S_j \varepsilon_j + h.o.t \quad (3)$$

where, h.o.t represents second and higher order terms. Finally, a zero-out contribution (ZOC), defined as the difference between the base-case simulation and the concentration that would occur if the sensitivity parameter did not exist (i.e.,  $p_j = 0$  or  $\varepsilon_j = 1$ ) can be approximated as (Cohan et al., 2005)

$$ZOC = C_{\text{base}} - C_{p_j=0} \approx S_j \quad (4)$$

We use the DDM-3D approach to estimate sensitivity of simulated three-dimensional tropospheric O<sub>3</sub> to emissions from different source regions across the Northern Hemisphere. We adopt the TFHTAP definitions of these source regions from Galmarini et al. (2017) but group them to larger geographic regions. Since O<sub>3</sub> production is dependent on both NO<sub>x</sub> and VOC emissions, calculating both first and second order sensitivity coefficients for O<sub>3</sub> to both

precursors and for each source region for extended simulation periods becomes computationally prohibitive. Thus, to examine the sensitivity of tropospheric  $O_3$  across the Northern Hemisphere across different seasons, we grouped the source regions and also tracked only the first order sensitivities to  $NO_x$  and VOC emissions from these regions over an entire annual cycle for the 2006 calendar year. Additionally, to more accurately project the sensitivities to other emission perturbation years while balancing the computational constraints, we also calculated second-order sensitivities only to global  $NO_x$  and VOC emissions for the 2006 calendar year. Since our primary focus is to characterize the influence of different world regions on tropospheric  $O_3$  and the long-range transported amounts to North America, we focus on sensitivity to total  $NO_x$  and VOC emissions from the regions and do not distinguish between anthropogenic and biogenic components of the contribution. It should also be noted that the sensitivity to VOC emissions from a region includes both NMVOC and carbon monoxide (CO). Though the contributions of CO to  $O_3$  formation are comparatively small, its inclusion with the VOC sensitivity fully completes all  $O_3$  precursors as they are identified in the chemical mechanism. Our previous analyses of DDM-3D calculations across the Northern Hemisphere (Itahashi et al., 2020b) show that the values of sensitivity coefficients to VOC emissions are small compared to those to  $NO_x$  emissions, and that the second-order sensitivity coefficients of  $O_3$  to VOC emissions are even smaller. This suggests that the nonlinear response of large-scale  $O_3$  distributions to VOC emissions is negligible and that  $O_3$  over the midlatitude Northern Hemisphere is in a  $NO_x$ -sensitive regime. Also, as illustrated in Section S2 of the supporting information, the estimated far field impacts of long-range pollution transport discussed in subsequent analysis is not impacted by our choice of not propagating higher-order sensitivity coefficients for each source region. We thus expect minimal impact of the simplifications in our simulation experiment design on the analysis presented here.

The entire Northern Hemisphere modeling domain was divided into seven source regions (see Figure 1) nominally representing: (1) North America (NAM), (2) Europe including all of Russia (EUR+RUS), (3) East and Southeast Asia (EAS+SEA), (4) South Asia (SAS; the Indian subcontinent), (5) Northern Africa and the Middle East (NAF+MDE), (6) Central America (CAM), and (7) Other (the rest of the geographic domain including the vast oceanic regions). In addition, the sensitivity of simulated three-dimensional  $O_3$  to the  $O_3$  specified in the top-most layer as a function of the space and time varying PV fields was also tracked. Since this PV-scaled  $O_3$  at the model top is representative of STE, this sensitivity can be interpreted to represent the modeled  $O_3$  that is of stratospheric origin. To minimize the impact of short-term meteorological variability on inferred long-range transport, in subsequent analyses we focus on seasonal-averaged values of the sensitivity coefficient which in turn were derived from the hourly model output values. Note that 2006 was chosen as the base year for the DDM-3D annual simulations based on previous applications and evaluation of the H-CMAQ modeling system for this calendar year (e.g.,

Xing et al., 2016; Mathur et al., 2017).

## 1. Results and Discussion

The CMAQ model configured with DDM-3D (Napelenok et al., 2008) as described in Section 2 was exercised over an annual cycle for the base year 2006 to estimate the sensitivity of  $O_3$  to  $NO_x$  and VOC emissions from each of the seven geographic source regions and the sensitivity of  $O_3$  to STE influences. In subsequent discussion,  $S_{NO_x,j}$  and  $S_{VOC,j}$  represent the first order sensitivity of modeled  $O_3$  to  $NO_x$  and VOC emissions from source region  $j$ , and  $S_{NO_x,G}^2$  and  $S_{VOC,G}^2$  are the second-order sensitivities of modeled  $O_3$  to global (i.e., across the entire domain)  $NO_x$  and VOC emissions, respectively, and  $S_{NO_x-VOC,G}^2$  represents the second-order cross sensitivity to global  $NO_x$  and VOC emissions (also see Itahashi et al., 2020b).

### 3.1 Trends in Surface and Aloft $O_3$

Detailed comparison of simulated surface-level predictions of  $O_3$  and related species over the 1990-2010 period from the hemispheric CMAQ model with available measurements from surface networks in the United States, Europe, and Asia have previously been reported in Xing et al. (2015) and are briefly summarized in Section S1 of the Supporting Information. Those results show that the model calculations were able to capture the significant and contrasting trends in surface-level  $O_3$  distributions across the Northern Hemisphere suggested by the observations, with declining trends across North America and western Europe resulting from control measures on combustion-related sources and increases across large parts of Asia associated with economic and population growth. Additionally, the model performance statistics were comparable with results from other studies in regions across the Northern Hemisphere. Comparisons of trends in daily maximum 8-hour average  $O_3$  (DM8O<sub>3</sub>) with corresponding trends inferred from the Clean Air Status and Trends Network (CASTNET) measurements (Puchalski, 2021) further indicated that the model captured the spatial variability in the direction of the trends across the CASTNET sites, though the hemispheric model underestimated the magnitude of the trend (Mathur et al., 2017). Further examination of the CASTNET and simulated trends in DM8O<sub>3</sub> distributions during the 1990-2010 period are shown in Figure 2, which depicts the spatial variability in trends (estimated as the slope of the linear regression) in the lower percentiles of the spring-time (March-April-May) DM8O<sub>3</sub> distributions across a portion of the model domain covering the United States at the surface and for model layers at altitudes around 1.5 km and 3.2 km above ground-level, nominally representing the upper parts of the daytime planetary boundary layer and the lower free troposphere, respectively. Increases in 5<sup>th</sup> and 25<sup>th</sup> percentile surface-level DM8O<sub>3</sub> are noticed at many sites (filled circles in Figure 2a and 2d) both in the western and eastern United States and are similar to those reported in other prior studies (e.g., Cooper et al., 2012). These increases in lower percentile DM8O<sub>3</sub> likely result from a combination of reduced  $O_3$  titration associated with declining  $NO_x$  emissions and



subsequent transport of higher  $O_3$  to rural areas as well as possible increases in baseline  $O_3$  resulting from increases in larger-scale regional and hemispheric tropospheric  $O_3$  values. Similar increases are also captured in the HCMAQ multidecadal simulations for this period, though the magnitude of the trend is underestimated. The underestimation in part results from uncertainties in global emission estimates and their changes as well as coarse model grid resolution (Mathur et al., 2017; Section S1.1 in Supporting Information). The increasing contributions of long-range transport to tropospheric  $O_3$  across the western United States is less ambiguous in the modeled DM8 $O_3$  trends at 1.5 km (Figures 2b and 2e) and 3.2 km (Figures 2c and 2f), altitudes at which trends in  $O_3$  distributions are largely driven by regional and hemispheric scale transport. Using backward sensitivity calculations with the CMAQ-Adjoint model (Hakami et al., 2007) that quantitatively link surface-level and aloft pollution, Mathur et al. (2018) show that at many locations across the U.S. more than 80% of sensitivity of ground-level  $O_3$  is controlled by  $O_3$  in layers aloft between 200-2000m. Consequently, it can be expected that downward entrainment of this increased free-tropospheric ozone contributes to the modeled (and by inference observed) increasing trends in the lower percentile DM8 $O_3$  at the surface and point to the need to better characterize the contributions of emissions from different source regions to the changing amounts of the  $O_3$  reservoir aloft.

### 3.2 Source-Region Contributions to $O_3$

To elucidate the spatially varying influence of emissions from different regions to  $O_3$  distributions across the Northern Hemisphere, Figure 3(b-h) present the estimated fractional contributions inferred from the sum of the individual sensitivities to  $NO_x$  and VOC emissions from each region following Equation 4. In constructing this figure, we first averaged computed hourly sensitivities over the entire season (March-April-May) and for all model layers between the surface and 2 km and refer to these as planetary boundary layer (PBL) and seasonal average values in subsequent discussions. The ratio of the first order sensitivity for a source region,  $i$ , (or the stratosphere) to the sum of all first-order sensitivities (or fraction  $f_i$ ) can be viewed to approximately represent the relative importance of the different emission source regions and the stratosphere in shaping the  $O_3$  distributions in the lower troposphere across the Northern Hemisphere as:

$$f_i = \frac{\overline{S}_i}{\left( \sum_{j=1}^N (\overline{S}_{NO_x,j} + \overline{S}_{VOC,j}) + \overline{S}_{STE} \right)} \quad (5)$$

where, N is the number of geographic regions and the overbars denote average values over the season and for all model layers between the surface and 2km. Figure 3a presents the simulated PBL-average  $O_3$  distribution for Spring 2006 across the Northern Hemisphere and Figures 3(b-i) illustrate the fractional contributions of the tracked sources and illustrate the relative importance of the different emission source regions and the stratosphere in shaping the  $O_3$

distributions in the lower troposphere across the Northern Hemisphere. Similar illustrations of the model estimated source signatures for the other seasons (Summer, Winter, and Fall) are presented in the supplemental information (Figure S6-S8). These figures illustrate that  $O_3$  produced from precursor emissions in a continent can undergo long range transport and impact tropospheric composition across vast remote oceans as well as far downwind continental regions. This intercontinental transport is more prominent during Spring and Summer with pronounced signatures of trans-Pacific transport of  $O_3$  from East Asia and trans-Atlantic transport of  $O_3$  from North America. During Spring and Summer,  $O_3$  attributable to emissions in East Asia undergoes long-range transport and can contribute several percent of the estimated imported  $O_3$  pollution across vast regions of continental North America. Emissions from Central America can also influence boundary-layer  $O_3$  mixing ratios over large portions of the Southwestern United States. Also evident in Figures 3 and S7 is the dominant contribution ( $>50\%$  of the total first-order  $O_3$  sensitivity) of STE processes in dictating PBL  $O_3$  abundance across the Northern Hemisphere during Spring and Winter when the height of the tropopause is lower and the stratospheric influence can penetrate far into the lower troposphere (e.g., Elbern et al., 1998; Wang et al., 2002; Langford et al., 2012; Lin et al., 2012; Xing et al., 2016; Mathur et al., 2017; Itahashi et al., 2020a). The widespread signature of the STE sensitivity depicted in these figures supports the notion that stratospheric impacts on lower tropospheric  $O_3$  are not limited only to deep intrusion events. Air masses advected into a region are also influenced by STE process upwind, resulting in the broad spatial signature of estimated STE contributions depicted in these illustrations. This highlights the need, in background  $O_3$  assessment and attribution analysis, for improved quantification of the amount of  $O_3$  in surface air that is attributable to stratospheric  $O_3$ .

### 3.3 Estimates of $O_3$ Imported to the North American Troposphere

While the terminology “background pollution” has traditionally been used to represent air pollution that is not produced locally, in context of ground-level  $O_3$  assessments, several definitions of background  $O_3$  now appear in literature and largely depend on the geographic extent of the region considered, on the distinction between controllable and non-controllable portions of the  $O_3$  imported to the region of interest, and whether  $O_3$  production from precursors emitted from natural sources (e.g., wildfires, biogenic, STE) are considered (see Jaffe et al., 2018). To avoid confusion, for the purpose of this study we use the term “imported  $O_3$ ” to represent the non-local contribution to the pollution in a region, and thus define it as the concentration attributable to sources (natural or anthropogenic) outside of the geographic region of interest. Following this definition, and the design of our sensitivity calculations described in section 2, we can estimate the imported  $O_3$  for any geographic region,  $m$ , (defined in Section 2 and depicted in Figure 1) as the sum of the first-order sensitivities of all sources other than  $m$  as follows

$$O_{3m}^{\text{imported}} = \sum_{j \neq m} (\bar{S}_{\text{NO}_x,j} + \bar{S}_{\text{VOC},j}) + \bar{S}_{\text{STE}}. \quad (6)$$

Note that this imported  $O_3$  plus the  $O_3$  attributable to natural emissions of precursors (wildfires, vegetation, soil, lightning) from within a region has been used in some studies to estimate the background  $O_3$  for a region (e.g., Emery et al., 2012; Fiore et al., 2014; Dolwick et al., 2015).

Following the definition in Equation (6), Figure 4 illustrates the estimated seasonal average  $O_3$  for Spring and Summer of 2006 over a region encompassing the United States and surrounding portions of Canada and Mexico that can be attributed to transport from the stratosphere and all other source regions across the Northern Hemisphere other than NAM. Figure 4a and 4b present the spatial variations across this region for average values between the surface and 2 km, for Spring and Summer (seasons with typically higher tropospheric  $O_3$ ), respectively, while Figures 4c and 4d show similar values for the model surface layer. Similar estimates for the cooler season (Winter) are presented in Figure S9 in the supplemental information. Significant spatial and seasonal variations are noted in the estimated imported  $O_3$  mixing ratios, with higher values in the high elevation regions of the western United States. The estimated imported  $O_3$  levels are higher during Spring than during Summer. Even though photochemical  $O_3$  production is highest in Summer leading to higher localized regional  $O_3$ , long-range transport and STE are more prominent during Spring, leading to the higher Spring imported (and background)  $O_3$  values. Additionally, the estimated imported  $O_3$  values for the model’s surface layer are lower than the estimated 0-2 km average because of  $O_3$  removal at the surface due to dry deposition and titration by surface  $NO_x$  emissions, highlighting the need for adequate vertical resolution in large scale models, not only to better represent long-range transport but also processes influencing ground-level  $O_3$ .

To further shed light on the relative contributions of the various source regions across the Northern Hemisphere on the estimated imported  $O_3$  values illustrated in Figure 4, we examined the relative contributions of the source regions to these estimated values. Figure 5 illustrates the fractional contributions of the six emission source regions and the stratosphere to the estimated average 0-2 km Spring imported  $O_3$  across the United States, while Figure 6 presents similar estimates for Summer. During Spring, cross-Pacific transport of  $O_3$  originating from Asia is estimated to contribute 10-20% of the  $O_3$  imported to the lower troposphere over North America. Transport from Central America also accounts for comparable fractions (and up to 30%) of the imported  $O_3$  in the south-central and southwestern states. Coastal regions across the western, eastern, and southern shores of the United States also show a persistent 10-20% contribution from sources across the vast oceanic regions of the Northern Hemisphere (represented by OTH or other), indicating the important role of emissions from the open ocean commercial shipping sector on tropospheric  $O_3$  distributions and transport to downwind continents. The largest contribution to imported  $O_3$  in the lower troposphere during Spring is estimated to be from  $O_3$  that is of stratospheric origin with fractional contributions ranging between 29-78% across the continental United States. High-latitude regions typically show greater influence of STE (Figure 5g) and the fractional contributions are

much higher during Spring and Winter (refer to Figure S10 in supplemental information) when height of the tropopause is lower (e.g., Elbern et al., 1998) and the stratospheric influence can penetrate far down to the lower troposphere. Contrasting Figures 5, 6, and S10, it is also apparent that the relative contributions of the different sources to imported (and thus background)  $O_3$  vary seasonally and are dictated by prevalent large-scale flow patterns. Notably during Summer, even though the estimated imported  $O_3$  values are comparatively lower, they are influenced to a larger extent by source regions in closer proximity as indicated by the larger fractional contributions from Central America and the ocean region. Nevertheless, the role of cross tropopause transport of  $O_3$  (i.e., STE) is seen to be a significant contributor to the non-local or imported  $O_3$  budget in most seasons and is further illustrated in Figure 7 which presents spatial distributions across the contiguous U.S. and surrounding portions of the estimated contributions of stratospheric  $O_3$  to seasonal mean surface-2km average  $O_3$  mixing ratios for Spring and Summer. The magnitude and spatial variability of stratospheric  $O_3$  inferred from the DDM-3D sensitivities are comparable to prior brute-force estimates reported in Xing et al. (2016) and Mathur et al. (2017). Additionally, as reported in Itahashi et al. (2020a) the interpretation of STE events as represented by the  $O_3$ -PV correlation in the model, is not strongly influenced by the model’s horizontal grid resolution since the WRF meteorological model employs assimilation of analyzed meteorological fields in the UTLS. Consequently, the estimated PV fields in the UTLS and the representation of STE events are similar across the different resolutions. Highest amounts of stratospheric  $O_3$  are seen in the high elevation locations of the western U.S. with significantly higher values during Spring compared to summer as also suggested by the fractional contributions illustrated in Figures 5g and 6g. As suggested by our results, both the amounts and fractions vary spatially and seasonally in response to the tropopause height, and perhaps even more episodically, from deep intrusion events associated with weather patterns and frontal movement. Our estimated STE fractional contributions are comparable with estimates presented in Roelofs and Lelieveld (1997), who, using a global chemistry-transport model, estimated that stratospheric contributions to surface  $O_3$  varied between 10-60% depending on season and location. Since our estimates represent STE contributions to the imported  $O_3$  to the North American troposphere and excludes the locally produced amounts, the estimated 29-78% range is expectedly higher. Tightening  $O_3$  NAAQS and decreasing amounts of photo-chemically derived  $O_3$  due to continuously declining anthropogenic precursor emissions across large parts of North America, further emphasize the need for accurately characterizing this fraction of  $O_3$  in the troposphere, especially at the surface, that is of stratospheric origin. Finally, it should be noted that the fractional contributions of the different source regions to the estimated imported  $O_3$  for the 0-2 km altitude average are relatively similar to the fractional contributions for the surface (contrast Figures 5 and S11) even though the absolute imported  $O_3$  mixing ratios at the surface are lower (Figure 4). This is because non-local  $O_3$  at the surface largely originates from the long-range transported  $O_3$  aloft (e.g., Mathur et al., 2018).

### 3.4 Trends in Long-Range Transported O<sub>3</sub>

Observations from a variety of in-situ and space-borne platforms have suggested that the global tropospheric O<sub>3</sub> burden has increased over the past several decades (Gaudel et al., 2018). Impacts of changes in anthropogenic emissions, particularly increases in East and South Asia and decreases in North America and Europe, on regional and global tropospheric O<sub>3</sub> burden have been assessed through modeling analysis (e.g., Xing et al. 2015, Zhang et al., 2021). These spatially heterogeneous changes in O<sub>3</sub> precursor emissions are not only modulating the tropospheric O<sub>3</sub> budget, but also likely altering long-range transport patterns and contributions at distant receptor regions. To investigate the impact of these contrasting changes in NO<sub>x</sub> and VOC emissions across the Northern Hemisphere over a multi-decadal period on long-range transported O<sub>3</sub> and the changing source region contributions, we formulated a reduced-form model combining the detailed sensitivity calculations analyzed in preceding sections with previously conducted HCMAQ simulations for the 1990-2010 period. The reduced form model utilizes three key pieces of information from these detailed model calculations: (1) sensitivity coefficients for the base year 2006, (2) emission changes for each source region for each year during the 1990-2010 period relative to the base year (estimated on a regional aggregate and seasonal basis), and (3) three-dimensional O<sub>3</sub> distributions simulated by HCMAQ for the 1990-2010 period as described in Xing et al. (2015).

The long-range-transported O<sub>3</sub> to a region  $m$  from all other emission source regions for the base year (i.e., 2006) can be estimated as the sum of the zero-out contributions from all regions other than  $m$  following Equation 6 as:

$$C_{LRT, m}^{\text{base}} = \sum_{j \neq m} (\bar{S}_{\text{NO}_x, j} + \bar{S}_{\text{VOC}, j}). \quad (7)$$

Note that the definition of long-range transported O<sub>3</sub> above is equivalent to the imported O<sub>3</sub> (estimated by Equation 6) without the STE contribution. Given the perturbations in regional-total emissions for a specific year relative to the base year, we can estimate the specific perturbation year concentrations for year  $N$  ( $C_{\text{est}}^N$ ) using the first-order sensitivities in the Taylor series approximation as

$$C_{\text{est}}^N = C^{\text{base}} + \sum (\bar{S}_{\text{NO}_x, j} \varepsilon_{\text{NO}_x, j}^N + \bar{S}_{\text{VOC}, j} \varepsilon_{\text{VOC}, j}^N). \quad (8)$$

Following the above equations, the long-range transported contributions to a region  $m$  in the perturbation year can then be approximated as

$$C_{LRT, m}^N = C_{LRT, m}^{\text{base}} + \sum_{j \neq m} (\bar{S}_{\text{NO}_x, j} \varepsilon_{\text{NO}_x, j}^N + \bar{S}_{\text{VOC}, j} \varepsilon_{\text{VOC}, j}^N) \quad (9)$$

Substituting Equation (7) in Equation (9) yields,

$$\begin{aligned}
C_{LRT,m}^N &= \sum_{j \neq m} (\bar{S}_{NO_x,j} + \bar{S}_{VOC,j}) + \sum_{j \neq m} (\bar{S}_{NO_x,j} \varepsilon_{NO_x,j}^N + \bar{S}_{VOC,j} \varepsilon_{VOC,j}^N) \\
&= \sum_{j \neq m} (\bar{S}_{NO_x,j} E_{NO_x,j}^N + \bar{S}_{VOC,j} E_{VOC,j}^N)
\end{aligned} \tag{10}$$

where  $E_{NO_x,j}^N = 1 + \varepsilon_{NO_x,j}^N$  and  $E_{VOC,j}^N = 1 + \varepsilon_{VOC,j}^N$ .

Note that since we did not explicitly track the sensitivity to initial conditions, boundary conditions,  $O_3$  production from methane (which currently is a prescribed global value), higher order sensitivities for each individual source region, and since the computed single-year sensitivity coefficients cannot represent impacts of possible changes in year-to-year meteorology, the reconstructed  $O_3$  field using a Taylor series approximation will not precisely match the modeled  $O_3$  field. To normalize for this discrepancy and to incorporate impacts of year-to-year meteorological variability on the seasonal average estimates, we scale the year-specific long-range transported estimates by the ratio of the modeled  $O_3$  mixing ratio for the specific year ( $C^N$ ) to its reconstructed value ( $C_{recon}^N$ ) as detailed in Equations 11-12 below:

$$C_{LRT,m}^N = \left( \sum_{j \neq m} (\bar{S}_{NO_x,j} E_{NO_x,j}^N + \bar{S}_{VOC,j} E_{VOC,j}^N) \right) \left( \frac{C^N}{C_{recon}^N} \right) \tag{11}$$

where,

$$C_{recon}^N \approx C^{base} + \sum (\bar{S}_{NO_x,j} \varepsilon_{NO_x,j}^N + \bar{S}_{VOC,j} \varepsilon_{VOC,j}^N) + \bar{S}_{STE} + \frac{1}{2!} \bar{S}_{NO_x,G}^2 (\varepsilon_{NO_x,G}^N)^2 + \frac{1}{2!} \bar{S}_{VOC,G}^2 (\varepsilon_{VOC,G}^N)^2$$

Recall that in Equation 12,  $S_{NO_x,G}^2$ ,  $S_{VOC,G}^2$  and  $S_{NO_x-VOC,G}^2$  are the second order sensitivities to domain-wide  $NO_x$  and VOC emissions, and  $\varepsilon_{NO_x,G}^N$  and  $\varepsilon_{VOC,G}^N$  represent the change in domain wide  $NO_x$  and VOC emissions, respectively for year  $N$  relative to 2006. It may also be noted that the estimated  $C_{recon}^N$  values are typically within 6% of the  $C^N$  for majority of the grid cells across the Contiguous U.S. as further detailed in section S3 in the Supporting Information.

Based on the reduced form model encapsulated by Equations 11-12, one can then estimate the long-range transported  $O_3$  to a specific region (defined in

Figure 1) that is attributable to  $O_3$  originating from all other source regions for each year of the 21-year analysis period of 1990-2010. Figure 8 illustrates the distribution and changes in this long-range transported  $O_3$  across the United States for Spring; Figures 8a-c illustrate the estimated long-range transported amounts for the specific years of 1990, 2000 and 2010 while Figure 8d illustrates the change in this long-range transported  $O_3$  depicted as the trend expressed in ppb/yr over the 21-years spanning 1990-2010. Figure S12 presents similar estimates for changes in long-range transported  $O_3$  for Summer. All values shown in Figures 8 and S12 are computed from 0-2 km average mixing ratios. As illustrated in Figures 8 and S12, the amount of long-range transported  $O_3$  to North America has steadily increased during the 1990-2010 period both during Spring and Summer. Several ppb of  $O_3$  within the boundary layer across the continental United States can be attributed to this long-range transport during both seasons. In general, long-range transport contributions to boundary layer and surface  $O_3$  are larger in the western and south-western states, which also show greater rates of increase in such influences over this time-period. Though not directly comparable due to differences in time-periods, averaging periods and metrics examined, the estimated magnitude of the trend across portions of the Western United States are comparable to those suggested in prior measurement and modeling studies (Cooper et al., 2012; Nopmongcol et al., 2016; Gaudel et al., 2018).

To further examine the changing influences of the different source regions to long-range transported  $O_3$  to North America, Figure 9 depicts the seasonal mean contributions during Spring for the years 1990 and 2010, contrasting conditions at the start and end of our analysis period. Figure 10 further illustrates the changes in the relative contributions of the various source regions during the 1990-2010 period. In constructing this figure, we examined the contribution of each source region to the estimated long-range transported  $O_3$  at each model grid cell within the region shown in Figures 8 and S12. To illustrate the changing contributions of these source regions to the long-range transported  $O_3$  we plot the changes in the median value of these contributions across all these grid cells over the 21-year time-period in Figure 10.

Collectively, Figures 8-10 reveal several important features of long-range transported  $O_3$  to North America: (1) changes in LRT  $O_3$  during the 1990-2010 periods across the continental United States vary spatially and so do the contributions from the different source regions, (2) the relative contributions of the different source regions to LRT  $O_3$  vary seasonally and have changed (and will likely continue to change) in response to spatially heterogeneous changes in anthropogenic precursor emissions across the globe, (3) cross-Atlantic transport of Europe's contributions have steadily declined (median contributions from ~30% in 1990 to 13% in 2010) in response to reductions in  $NO_x$  and VOC emissions in that region, (4)  $O_3$  attributable to emissions in East and South-east Asia is a dominant contributor (median values from 25-38%) to LRT  $O_3$ , with increasing contributions over the 1990-2010 period associated with increasing emissions in the region, (5) emissions from Central America also contribute a sizeable

fraction of the imported  $O_3$  across portions of the south western United States (greater than 3 ppb on seasonal mean basis), (6)  $O_3$  attributable to emissions from the large oceanic regions across the Northern Hemisphere (large fraction of the OTH in Figure 1) has also been a significant contributor to LRT  $O_3$ , with a distinct increasing contribution (median contributions up to 32%) in recent years, (7) emissions from sources in closer proximity to North America (i.e., Central America and the Oceans) expectedly have a larger contribution to  $O_3$  imported to the North American troposphere during Summer and contribute to the noted higher values of LRT  $O_3$  in Summer relative to Spring (contrast Figures 8a-c and S12a-c).

Noticeable in Figures 9 and 10 is the prominent and increasing contribution of  $O_3$  in the North American troposphere attributable to emissions from the source region marked OTH in Figure 1, which is largely comprised of the vast oceanic regions of the Northern Hemisphere.  $NO_x$  emissions from this region predominantly result from marine transportation activities. Though shipping is considered a relatively energy efficient mode for transportation of goods,  $NO_x$  emissions from shipping are relatively high, because most marine engines operate at high temperatures and pressures without effective reduction technologies. Oceangoing ships are estimated to contribute around 15% of all global anthropogenic  $NO_x$  emissions and an increase in international ship traffic has raised concerns about possible impacts of their emissions on tropospheric composition, climate effects, and health impacts (Dalsøren et al., 2009; Hoor et al., 2009; Eyring et al., 2010; Sofiev et al., 2018; Zhang et al., 2021). Analyzing results from global and regional model calculations, Jonson et al. (2020) found that the largest contribution to ozone in several regions and countries in Europe was from sea areas well outside European waters associated with open sea shipping. Using a source tagging method in a global chemical transport model, Butler et al. (2002) estimated 2-5ppb contribution of shipping  $NO_x$  emissions to annual mean surface  $O_3$  across North America with larger values during Spring in the northwestern U.S. While the magnitude of our estimated contributions from shipping emissions over North America are slightly lower than those reported in Butler et al. (2020), they do show similar widespread impacts over downwind continental regions. Figure 11 presents the changes in  $NO_x$  emissions associated with this sector during our analysis period as estimated in the EDGAR inventory used in our model calculations. Evident from Figure 11 is the steady increase in emissions (~60% during 1990-2010) from this sector associated with an increase in international transportation of goods and commercial trading, highlighting the role of increasing  $NO_x$  emissions from shipping on the tropospheric  $O_3$  burden and its increasing contributions to  $O_3$  imported to North America during the 1990-2010 period as illustrated in Figure 10. These results emphasize the need for more accurate characterization of the spatial and temporal variability in  $NO_x$  emissions from ships across the vast oceanic regions to better help guide policies seeking reductions in emissions from this sector that effectively offset the anticipated growth in seaborne trade.

## 1. Summary and Conclusions



Significant and contrasting changes in emissions of tropospheric  $O_3$  precursor species have and continue to occur across the globe in response to implementation of control measures and changes in patterns of emission producing activities. These changes are not only impacting the local  $O_3$  levels and distributions that are sensitive to these emissions, but also influencing the large-scale distribution of  $O_3$  across the global troposphere. Since the lifetime of  $O_3$  in the free troposphere, depending on altitude, can vary from several days to a month, once produced and lofted to the free troposphere, it can be efficiently transported inter-continently (e.g., NRC, 2009; TFHTAP, 2010). Increasing amounts of long-range transported  $O_3$  impact background levels in receptor regions and confound development and implementation of local abatement strategies to manage pollution and mitigate effects on human and ecological health. Analysis of long-term measurements of  $O_3$  (e.g., Chang et al., 2017; Gaudel et al., 2018) highlight these changing trends in its distributions across the globe. However, since measurements represent the cumulative impacts of multiple atmospheric dynamical-chemical-physical process on emissions from a multitude of source regions, measurements alone are unable to unambiguously attribute the causes of the noted trends to emissions from a source region or uniquely differentiate contributions from intercontinental transport relative to those from STE.

Employing multi-decadal model simulations of the changes in tropospheric  $O_3$  distributions across the Northern Hemisphere from a comprehensive atmospheric modeling system in conjunction with detailed sensitivity calculations characterizing the response of tropospheric  $O_3$  to  $NO_x$  and VOC emissions from source regions across the Northern Hemisphere, we estimate that on a seasonal-mean basis, during the 1990-2010 period, LRT  $O_3$  to regions across the Continental United States steadily increased by 0.06-0.2 ppb  $yr^{-1}$  in response to changes in precursor emissions outside North America. This net trend, in turn, arises from superposition of unequal and contrasting trends in contributions from different emission source regions across the Northern Hemisphere. Both the amount of long-range transported  $O_3$  and the magnitude of its trend vary spatially and are larger in the western United States than those in the east. Even though the inferred relative contributions of the different source regions over this multi-decadal period exhibit inter-annual variability, they provide valuable information on the seasonal variations in these contributions as well as the changing relative importance of their emissions to the long-range transported  $O_3$  to North America. Specifically, the model calculations show that declining emissions in Western Europe have substantially reduced its contributions to LRT  $O_3$  across the United States; median values of the estimated contributions declined from ~30% in the early 1990's to ~15% in 2010. In contrast, cross-Pacific transport of  $O_3$  attributable to Asian emissions contributed 20-35% of the estimated LRT  $O_3$  with an increasing trend during 1990-2010, as also suggested in several prior modeling studies analyzing impacts of perturbations in Asian emissions (e.g., Jacob et al., 1999; Fiore et al., 2009). On a seasonal basis, the contributions of  $O_3$  originating from emissions in the Indian subcontinent are comparatively small throughout the analysis period.

This is consistent with results in Butler et al. (2020) that also show that impact of emissions from South Asia is relatively localized and their transport does not influence hemispheric background to the same extent as  $O_3$  produced from other source regions in the Northern Hemisphere. Emissions from Central America are shown to modulate  $O_3$  levels across portions of southwestern United States with median contributions to LRT  $O_3$  across the United States varying between 5-15% across seasons and years. Also, since  $O_3$  production is more efficient during warmer periods, sources in closer proximity to the United States, specifically Central America and shipping emissions, exhibit higher contributions during summer compared to other seasons.

Highlighted by our analysis is the prominent and increasing contribution of emissions from shipping operations across the vast oceans. Though the impacts of shipping emissions on air quality in ports and coastal areas across the world have been frequently studied, and the role of  $SO_2$  (and consequently aerosol sulfate) and carbon dioxide ( $CO_2$ ) from the sector as modulators of global climate is well acknowledged, the prominent contribution of  $NO_x$  emissions from the sector to large-scale  $O_3$  distributions and subsequent impacts on background  $O_3$  in downwind continents is only recently being recognized (e.g., Jonson et al., 2018; Mertens et al., 2018; Butler et al., 2020; Jonson et al., 2020). It should be noted that limited field measurements of ship plume chemical composition, suggest that instantaneous dilution of ship emissions over coarse grid volumes may overestimate ozone production in vicinity of ship emissions (e.g., Chen et al., 2005) and are thus suggested to represent upper limits over the ocean basins (Butler et al., 2020). While ship plume parameterizations in large scale models improved model estimated concentrations in the vicinity of the source relative to measurements of the first few of hours of aging they exhibited minor differences in simulated ozone over downwind continental regions (e.g., Vinken et al., 2011) suggesting that the effects of the parameterizations on grid resolved chemistry and estimated far field impacts are small. This could be because across the more variable conditions of the open oceans, some fraction of the ship  $NO_x$  survives beyond the plume parameterization aging (e.g., von Glasow et al., 2003; Holmes et al., 2014) which along with  $NO_x$  sequestered in reservoir species (e.g., organic nitrates), in addition to shipping  $NO_x$  transported longer distances nocturnally, then impact grid resolved chemistry and the subsequent estimated far field impacts. For instance, in their source tagging simulations Butler et al. (2020) found that significant amount of PAN is formed downwind of regions in which anthropogenic VOCs are emitted, often via reaction with  $NO_x$  emitted from shipping. Though both the horizontal resolution and vertical layer spacing used in our model calculations is much finer than most large-scale models, as in many global models (Eyring et al., 2007; Butler et al., 2020; Jonson et al., 2020), we neglect plume dilution due to a lack of consensus and availability of robust parameterizations in common modeling systems. As noted in Butler et al. (2020), future work exploring optimal grid resolution to minimize impacts of artificial ship plume dilution would be useful. Nevertheless, our results reveal that LRT  $O_3$  attributable to emissions

from shipping has increased in response to the steadily increasing  $\text{NO}_x$  emissions from the sector as captured in the EDGARv4.2 emission inventory used in our model calculations. During the 2005-2010 period and perhaps even in the contemporary atmosphere, the contributions of shipping emissions to  $\text{O}_3$  imported to the U.S. troposphere are comparable to those from transport of  $\text{O}_3$  attributable to East Asian emissions and suggest possibly even higher contributions in the future if  $\text{NO}_x$  emissions from commercial shipping operations were to increase in response to anticipated growth in seaborne trade (Eyring et al., 2010). Future studies could utilize detailed information on ship type and position from Automatic Identification Systems (AIS) transponders to better characterize spatial and temporal variability in shipping emissions (e.g., Jalkanen et al., 2012). This combined with finer horizontal resolution simulations would enable not only improved quantification of shipping emissions but potentially also improved representation of the chemistry in ship exhaust plumes both in the vicinity of the ship tracks but also as they are transported to downwind continental regions.

In addition to the  $\text{O}_3$  attributable to emissions from the source regions across the Northern Hemisphere, model calculations also suggest that air masses entering the North American domain have sizeable contributions of  $\text{O}_3$  that is of stratospheric origin. In context of our model set up and calculations, the sum of the LRT  $\text{O}_3$  and that attributable to STE represent the amount of  $\text{O}_3$  originating from sources other than emissions within the domain and thus representative of the non-local or “imported”  $\text{O}_3$  levels. The modeled sensitivities to  $\text{O}_3$  resulting from STE processes indicate large contributions to boundary layer and ground-level  $\text{O}_3$ , the magnitude of which is consistent with prior brute force estimates (Xing et al., 2016; Mathur et al., 2018). Our estimates suggest that  $\text{O}_3$  that is of stratospheric origin could constitute 29-78% of the estimated Spring background  $\text{O}_3$  across the continental United States; high-latitude regions typically show greater influence of STE. Additionally, since comparisons of HCMAQ predictions with ozonesonde measurements indicate a systematic low bias in free-troposphere  $\text{O}_3$  mixing ratios (e.g., Hogrefe et al., 2018), the estimated STE modeled contributions to boundary-layer  $\text{O}_3$  may be considered conservative. Nevertheless, the widespread spatial signature of the sensitivity to modeled  $\text{O}_3$  resulting from STE processes noted for all seasons, and especially Spring and Winter when the tropopause height is lower, supports the notion that stratospheric impacts on lower tropospheric and ground-level  $\text{O}_3$  are not limited only to deep intrusion events. Air masses advected into a region are also influenced by STE process upwind, resulting in the noted broad spatial signature of estimated STE contributions. This highlights the need for improved quantification of the variability and amount of  $\text{O}_3$  in surface air that is attributable to the stratosphere in background  $\text{O}_3$  assessments and estimation.

Our results show that the intercontinental transport of pollutants has and continues to change in response to emission changes across the globe. Accurately quantifying the impacts of changes in emissions in distant continents on amounts of pollution imported is becoming as important as domestic control measures at

receptor continents. Analyzing the impact of well documented and contrasting multi-decadal (1990-2010) changes in emissions across the Northern Hemisphere helped better explain the dominant source regions that drive the noted changes in observed  $O_3$  mixing ratios across the United States and highlight the need for greater emphasis on reducing uncertainties in emissions associated with commercial shipping and in characterization of space and time variability in lower tropospheric  $O_3$  that is of stratospheric origin. While our analyses have focused on better understanding factors influencing long-range transport of  $O_3$  to North America and portions of the United States, the existing calculations and the reduced form model could easily be extended to any of the other continental source regions defined in Figure 1. Though our analyses focused on  $O_3$ , many of the issues discussed here are also applicable to other criteria pollutants (e.g.,  $PM_{2.5}$ ), and any potential actions taken to understand or reduce background  $O_3$  may also reduce background concentrations of other pollutants.

### 1. Caveats and Future Directions

Changes in  $O_3$  precursor emissions across the globe have occurred during the decade post our 1990-2010 analysis period and have likely further altered tropospheric  $O_3$  distributions and amounts transported intercontinentally. Based on review of the current state of knowledge, Jaffe et al. (2018) summarize estimates of seasonal U.S. background  $O_3$  (USB  $O_3$ ) to be 20-40 ppb depending on location with an uncertainty of  $\pm 10$  ppb. While our estimates are within the reported range, our methodology could be refined further for current time periods by leveraging recent enhancements to the HCMAQ modeling system and its evolving applications. In particular, enhancements to the representation of seasonal variations in the impacts of (1) halogen chemistry in marine environments (Sarwar et al., 2019), (2) stratosphere-troposphere exchange (Xing et al., 2016; Mathur et al., 2017), and (3) emissions from natural sources such as lightning  $NO_x$  (Kang et al., 2020) will enable more accurate representation of seasonal changes in attribution of the source regions in terms of absolute  $O_3$  mixing ratios. These model process enhancements combined with updates in global emission estimates and trends for the contemporary atmosphere are currently being used to simulate changes in tropospheric composition across the Northern Hemisphere during 2002-2017 as part of EPA’s Air Quality Time Series (EQUATES) project (Foley et al., 2020). A refined DDM-3D configuration that explicitly tracks  $O_3$  sensitivities to natural sources plus anthropogenic sources in countries outside the United States could then be deployed to estimate attributions commensurate with the U.S. background (USB)  $O_3$  definition (i.e., influence of all sources other than U.S. anthropogenic emissions (Dolwick et al., 2015)) used in policy deliberations. These simulations could also help quantify the impact of recent emission reductions in East Asia (e.g., Zheng et al., 2018) on cross-Pacific transport of  $O_3$ . In addition, expanding the set of sensitivity coefficients to also include sensitivity to initial and boundary conditions, and second-order sensitivity to  $NO_x$  emissions from individual countries would enable more accurate attribution of absolute  $O_3$  mixing ratio contributions for the 2002-2017 period, extending the reduced form model framework

developed in this study. As computational resources increase and model efficiency improves, multi-year DDM-3D calculations could also be conducted to more robustly assess possible impacts of meteorological variability on inferred long-range transport impacts of the different source regions. Multi-year average sensitivity coefficients may also yield a more robust reduced form model to project the impacts of future emission changes. The current analyses focused on regionally aggregate changes in emissions, and thus represents response to average emission changes over large geographic source regions. New features in CMAQ that allow disaggregation of input emissions (Murphy et al., 2021), in conjunction with explicit tracking of sensitivity to the sector emissions can facilitate developing inferences on impacts of changes in emissions from specific sectors if desired.

**Code and Data Availability:** CMAQ source code is freely available via <https://github.com/usepa/cmaq.git>. Archived CMAQ versions are available from the same repository. Data used to generate figures shown in this article are available from Mathur (2021). The raw model outputs are available upon request from the corresponding author. Observational data from the CASTNET network (Puchalski, 2021) are available at: <https://www.epa.gov/castnet>.

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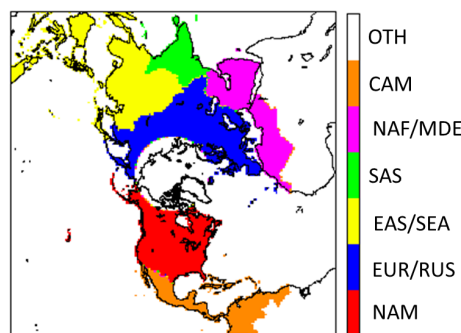
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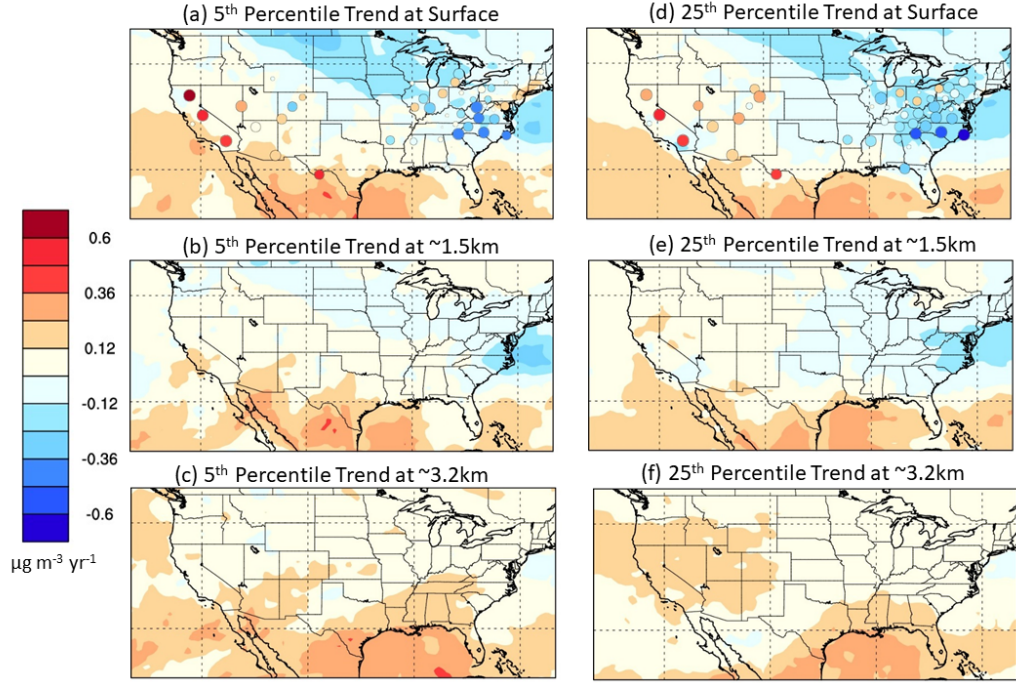
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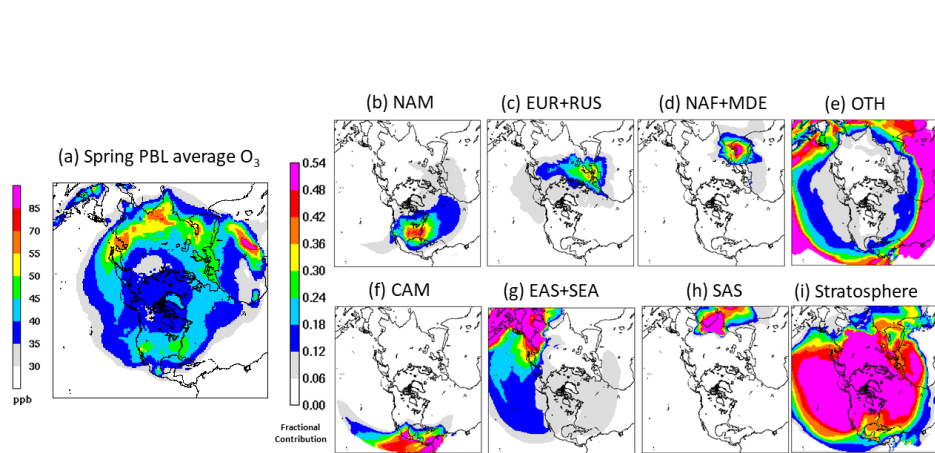


**Figure 1:** The Northern Hemisphere modeling domain and source region def-

initiation: (1) North America (NAM), (2) Europe and Russia (EUR+RUS), (3) East and Southeast Asia (EAS+SEA), (4) South Asia (SAS; the Indian sub-continent), (5) Northern Africa and the Middle East (NAF+MDE), (6) Central America (CAM), and (7) The rest of the geographic domain (OTH).

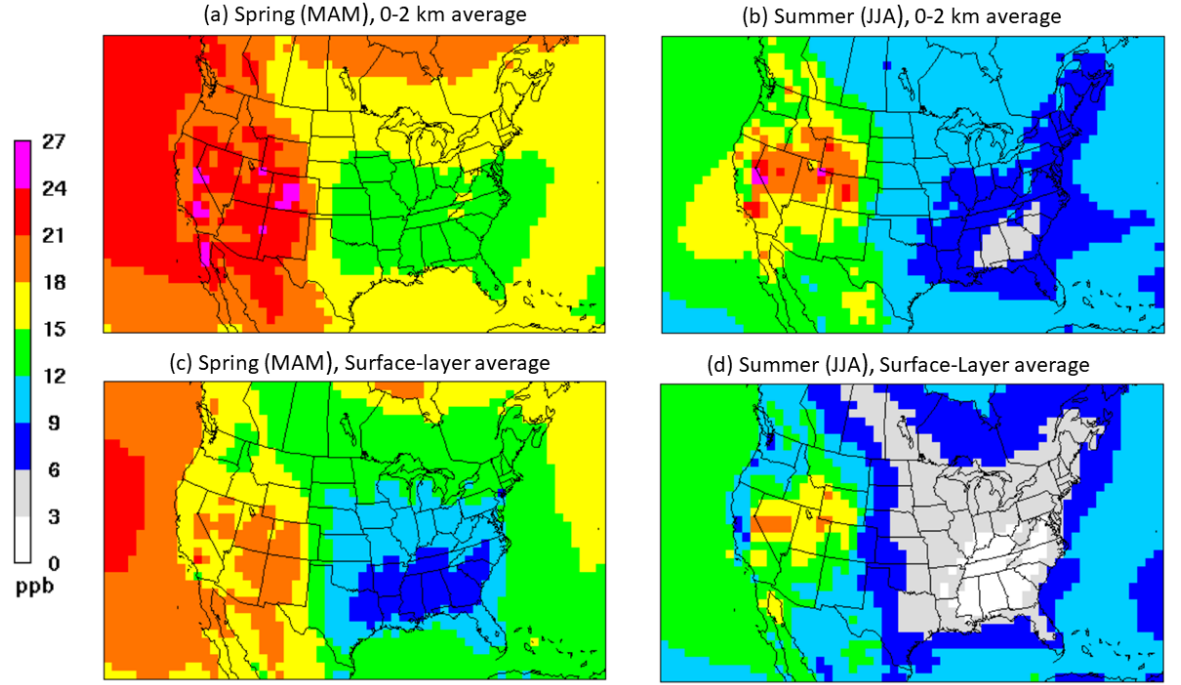


**Figure 2:** 1990-2010 trends in Spring (March-April-May) seasonal-average daily-maximum 8-hour-average  $O_3$  (DM8O<sub>3</sub>) distributions across a portion of the model domain covering the United States at the surface and at model layers at altitudes of 1.5 km and 3.2 km above ground-level. (a)-(c): trends in 5<sup>th</sup> percentile DM8O<sub>3</sub> at different altitudes. (d)-(f): trends in 25<sup>th</sup> percentile DM8O<sub>3</sub> at different altitudes. Filled circles in (a) and (d) represent corresponding observed trends at the CASTNET monitoring locations.



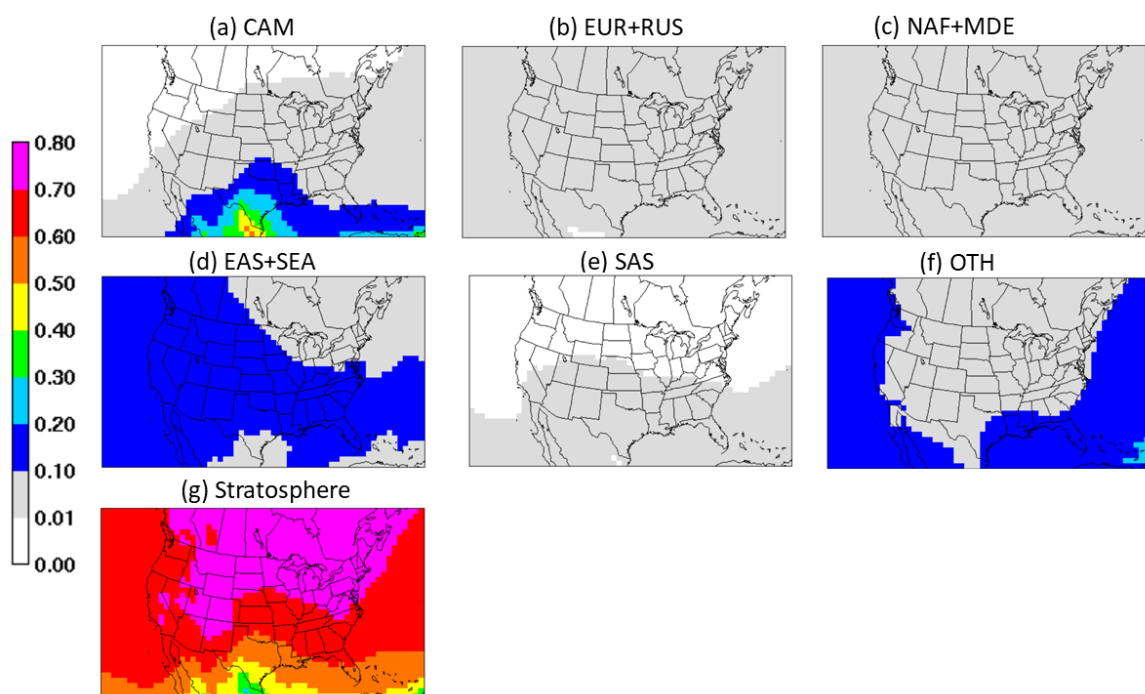
Estimated contributions of different source regions to  $O_3$  distributions across the Northern Hemisphere. (a) PBL average  $O_3$  during spring 2006. Fractional contributions of individual source regions: (b) NAM, (c) EUR+RUS, (d) NAF+MDE, (e) Other, (f) CAM, (g) EAS+SEA, (h) SAS and (i) stratosphere

**Figure 3:** Estimated contributions of different source regions to  $O_3$  distributions across the Northern Hemisphere for Spring (March-April-May) 2006. (a) PBL average  $O_3$  during spring 2006. Fractional contributions of individual source regions estimated from the first-order sensitivities: (b) NAM, (c) EUR+RUS, (d) NAF+MDE, (e) Other, (f) CAM, (g) EAS+SEA, (h) SAS and (i) stratosphere.

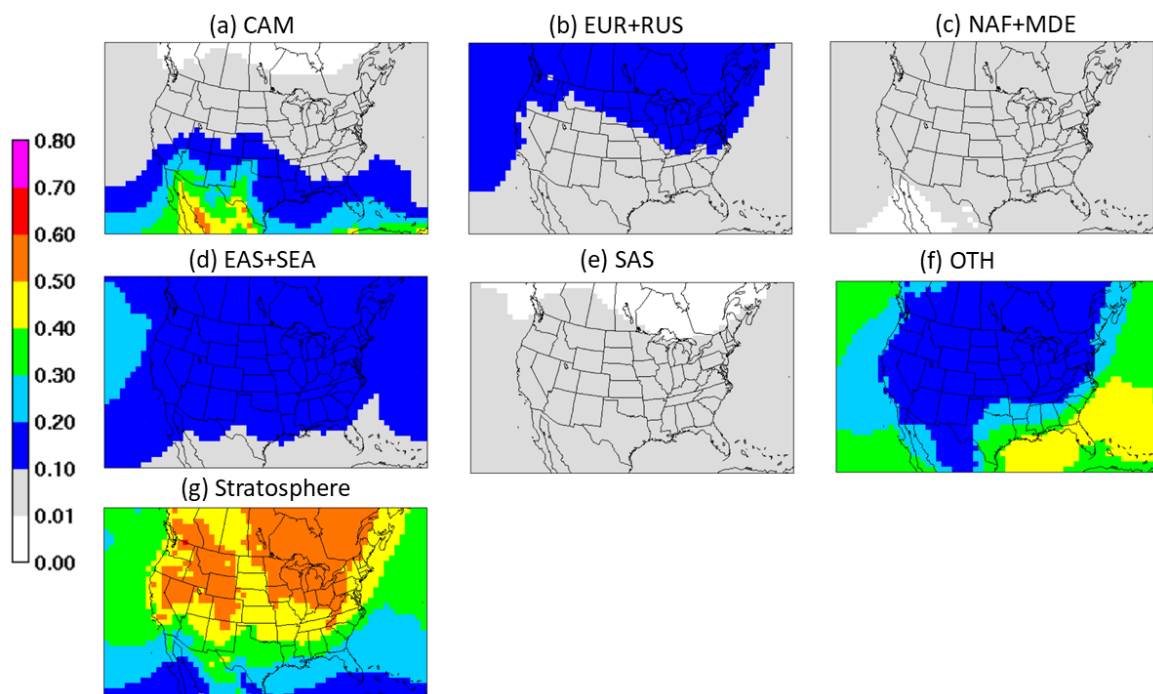


**Figure 4:** Estimated imported  $O_3$  (defined as  $O_3$  attributed to transport from the stratosphere and all other source regions across the Northern Hemisphere other than NAM) distributions across the United States and surrounding regions for the surface to 2km average for (a) Spring 2006 and (b) Summer 2006. (c) and (d) illustrate similar distributions but for the model's surface-layer.

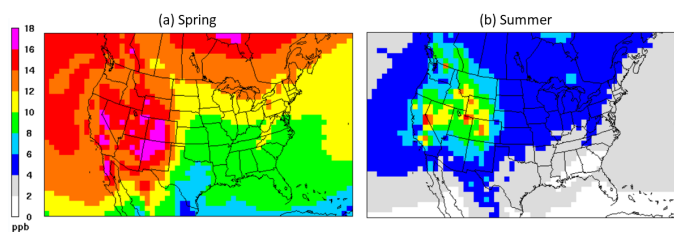




**Figure 5:** Fractional contributions of the different source regions to the 0-2km average Spring 2006 estimated imported  $O_3$  mixing ratios in Figure 4a.

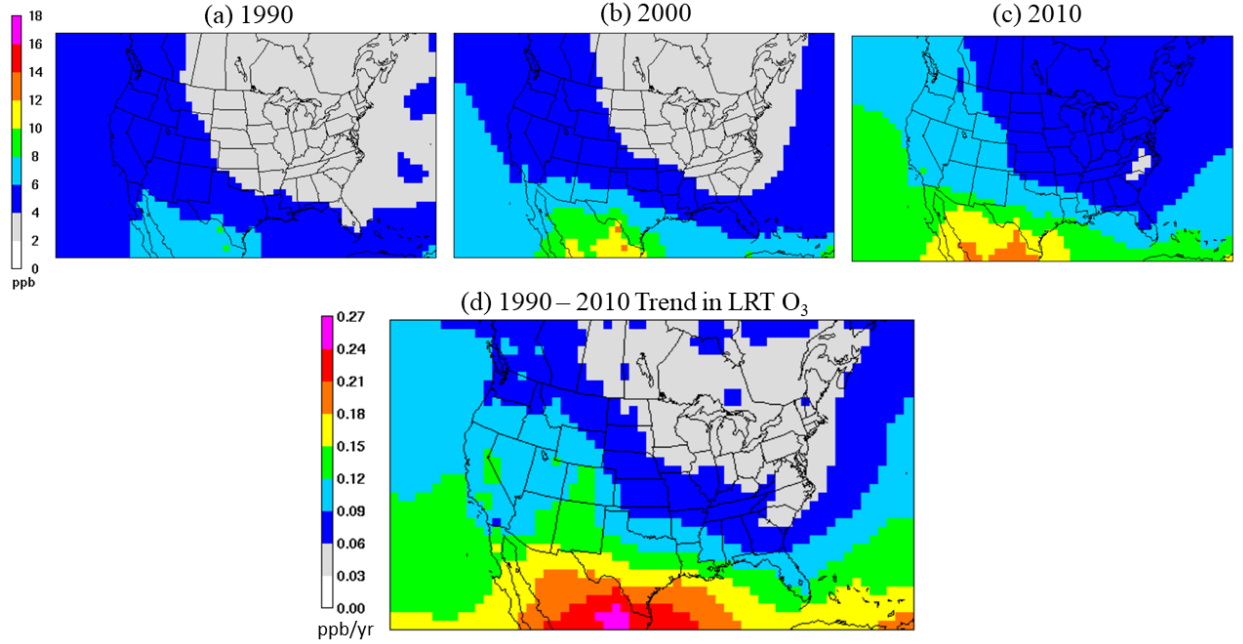


**Figure 6:** Fractional contributions of the different sources to the 0-2km average Summer 2006 estimated imported  $O_3$  mixing ratios in Figure 4b.

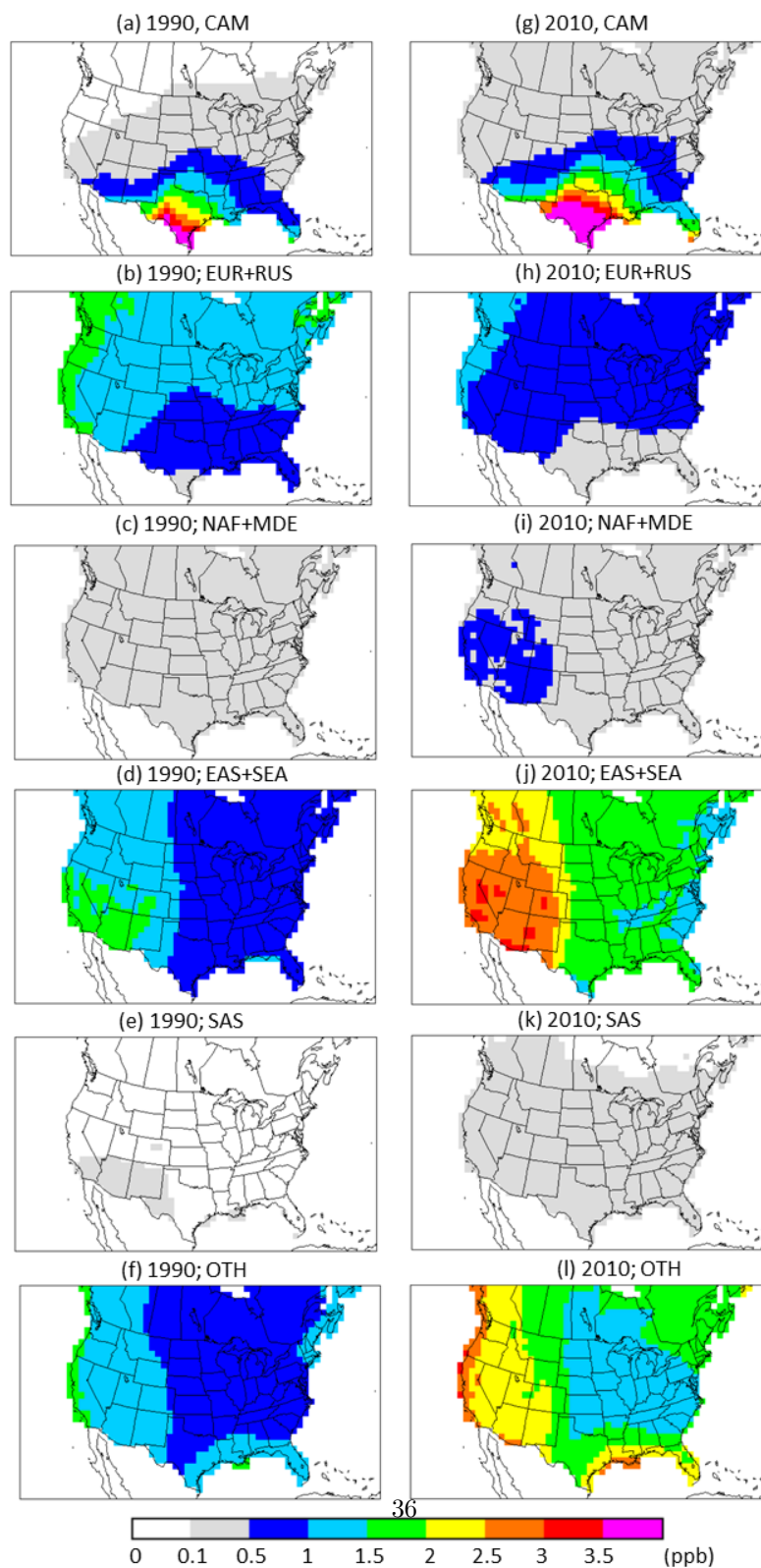


Estimated contributions of stratospheric  $O_3$  to seasonal-mean 0-2km average  $O_3$  mixing ratios during (a) Spring 2006, and (b) Summer 2006 (add after fig6?)

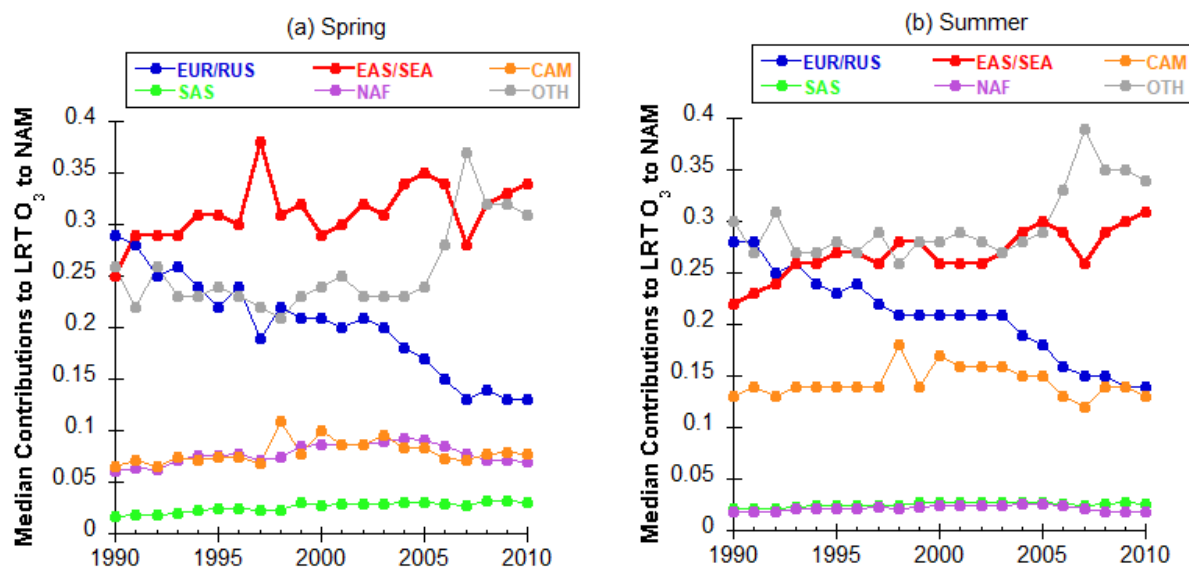
**Figure 7:** Estimated contributions of stratospheric  $O_3$  to seasonal-mean 0-2km average  $O_3$  mixing ratios for (a) Spring 2006 and (b) Summer 2006.



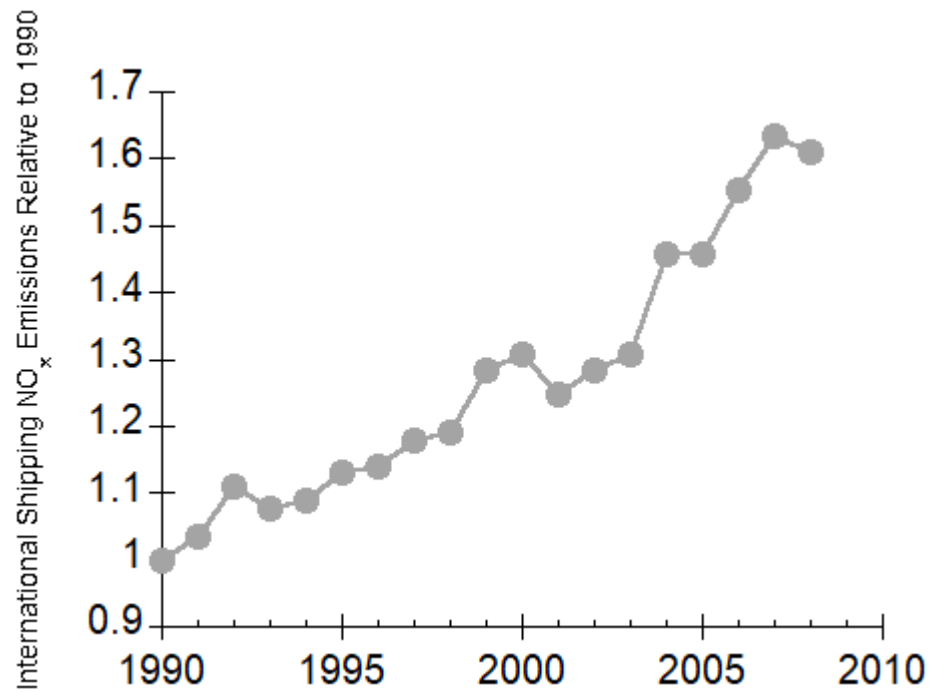
**Figure 8:** 1990-2010 changes in long-range transported  $O_3$  to North America during Spring. Estimated long-range transported  $O_3$  to North America from the different source regions for (a) 1990, (b) 2000, and (c) 2010. The estimated trend in long-range transported  $O_3$  to North America for the 1990-2010 period is illustrated in (d). The trend is estimated as the slope of the linear regression of the estimated seasonal-mean long-range transported  $O_3$  mixing ratios for each of the individual years in the 1990-2010 period. All estimates are for the 0-2km average  $O_3$  mixing ratios.



**Figure 9:** Estimated contributions of different source regions to long-range transported  $O_3$  to North America during Spring for 1990 (a-f) and 2010 (g-l). All estimates are for the 0-2km average  $O_3$  mixing ratios. Values are only shown for grid cells in the NAM region in Figure 1.



**Figure 10:** Changes in median fractional contributions to long-range transported  $O_3$  to North America from different source regions during 1990-2010.



**Figure 11:** Changes in NO<sub>x</sub> emissions from international shipping operations across the Northern Hemisphere during 1990-2008; Individual year emissions are normalized to the 1990 value.