Investigating the response of China's surface ozone concentration to the future changes of multiple factors

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Abstract

Climate change and associated human response are supposed to greatly alter surface ozone (O_3), an air pollutant generated through photochemical reactions involving both anthropogenic and biogenic precursors. However, a comprehensive evaluation of China's O_3 response to these multiple changes has been lacking. We present a modelling framework under Shared Socioeconomic Pathways (SSP2-45), incorporating future changes in local and foreign anthropogenic emissions, meteorological conditions, and BVOCs emissions. From the 2020s to 2060s, daily maximum 8-hour average (MDA8) O_3 concentration is simulated to decline by 7.7 ppb in the warm season (April-September) and 1.1 ppb in non-warm season (October-March) over the country, with a substantial reduction in exceedances of national O_3 standards. Notably, O_3 decreases are more pronounced in developed regions such as BTH, YRD, and PRD during warm season, with reductions of 9.7, 14.8, and 12.5 ppb, respectively. Conversely, in non-warm season, the MDA8 O_3 in BTH and YRD will increase by 5.~~4~~5 and 3.~~4~~3 ppb, partly attributed to reduced NO_x emissions and thereby weakened titration effect. O_3 pollution will thus expand into the non-warm season in the future. Sensitivity analyses reveal that local emission change will predominantly influence future O_3 distribution and magnitude, with contributions from other factors within ±25 %. Furthermore, the joint impact of multiple factors on O_3 reduction will be larger than the sum of individual factors, due to changes in the O_3 formation regime. This study highlights the necessity of region-specific emission control strategies to mitigate potential O_3 increases during non-warm season and under climate penalty.

1 Introduction

Surface ozone (O_3) is a secondary air pollutant generated by photochemical reactions in the presence of two main kinds of precursors NO_x (NO_x=NO+NO_2) and volatile organic compounds (VOCs). It has been reported to be a non-negligible threat to both human health and crop yield, and also a short-lived climate forcer with warming effect (Finlaysonpitts and Pitts, 1997; Jerrett et al., 2009; Avnery et al., 2011; Lelieveld et al., 2015; von Schneidemesser et al., 2015; Tai and Val Martin, 2017; Yin et al., 2020; Feng et al., 2022; Niu et al., 2022). Given the abundant emissions of anthropogenic NO_x and VOCs, China has suffered from extremely high and continuously increasing O_3 pollution from 2013 to 2019 with the peak season daily maximum 8-hour average (MDA8) O_3 concentration over 95 μg m^{-3}. The rising trend has been reversed since 2020, along with the national annual NO_x and non-methane VOCs (NMVOCs) emissions reduced by 28.3 % and 3.8 %, respectively during 2013–2020 (Zheng et al., 2018; Xiao et al., 2022; Liu et al., 2023a; Wang et al., 2023). However, current O_3 concentration over China is still much higher than the global air quality guidelines (60 μg m^{-3} for the averaged peak season MDA8 O_3, WHO, 2021). This presents a great challenge for the country to meet the criteria for public heath welfare in the future (Feng et al., 2023; Jiang et al., 2023).

In addition to the anthropogenic driver, studies also addressed the roles of meteorological factors, biogenic VOCs (BVOCs) emissions and transboundary transport of pollutants on O_3 enhancement in China (Monks et al., 2015; Lu et al., 2019; Cao et al., 2022; Wang et al., 2022; Weng et al., 2022; Jiang et al., 2023). Meteorological factors, including temperature, humidity, wind, etc., influence the chemical reactions associated with O_3 production and elimination, and the transportation of O_3 precursors (Gong and Liao, 2019). The changed meteorology was estimated to enhance the summer MDA8 O_3 concentration by 1.4 ppb yr^{-1} during 2013–2019 in the North China Plain (NCP), nearly half of the overall O_3 growth of 3.3 ppb yr^{-1} (Li et al., 2020a). BVOCs refer to VOCs emitted from terrestrial ecosystems and possess high reactivity in atmospheric chemical processes, mainly including isoprene, monoterpenes and sesquiterpene (Wu et al., 2020a). Cao et al. (2022) reported that BVOCs emissions in summer 2018 enhanced 8.6 ppb MDA8 O_3 averaged over China with the highest contribution over 30 ppb in southern China.

Moreover, O_3 and its precursors could be transported over long distance, and transboundary foreign anthropogenic emissions were estimated to contribute 2–11 ppb to near surface O_3 in China (Ni et al., 2018; Han et al., 2019).

In the context of future global change, substantial but uncertain changes will occur in economy, climate and land cover. According to the Sixth Assessment Report of Intergovernmental Panel on Climate Change (IPCC AR6 report), the global surface temperature will increase 0.2–3.7 °C till 2100 under different scenarios compared to 2015 (IPCC, 2021, 2022). As a result, unfavorable meteorological extremes, such as high temperature extremes and ecological drought events, will be more frequent and intense (Hong et al., 2019; Porter and Heald, 2019; IPCC, 2021), leading to the deterioration of air quality named as climate penalty. To conquer the climate change and resulting air quality deterioration, a series of measures will be implemented, such as accelerating the transition to clean energy, upgrading industrial production technologies and strengthening pollution control measures. Attributable to these changes, annual mean surface O_3 in East Asia was projected to change by −13.9–6.1 ppb till 2100 compared to 2015 (IPCC, 2021). However, limited by the coarse resolution of earth system model and the lack of consideration of the regional measures for reducing air pollution and carbon emissions, global estimation is insufficient for understanding how O_3 pollution in China will respond to the complex future change.

There have been some studies on how the above-mentioned changes will affect future O_3 level in China. Hong et al. (2019) reported that the 1-hour maximum O_3 in April to September will be enhanced by 2–8 ppb within large areas of China under RCP4.5 (representative concentration pathways 4.5, van Vuuren et al., 2011) scenario from the 2010s to 2050s. Under high-forcing scenarios, Li et al. (2023) projected the climate-driven O_3 concentration in the 2100s and found that O_3 concentration in southeast China would increase 5–20 % compared to the 2020s by a machine learning method. A warming climate should enhance the O_3 level, given the increasing frequency of atmospheric stagnation and heat waves (Hong et al., 2019; Wang et al., 2022; Gao et al., 2023; Li et al., 2023). The effect of anthropogenic emission change on China's O_3 level has been estimated by studies under different scenarios. Zhu and Liao (2016) applied global emission estimates under RCPs, and found that the maximum growth of annual mean O_3 would be 6–12 ppb during 2000–2050 under different scenarios. Using the Dynamic Projection model for Emissions in China (DPEC) that better includes local information of energy transition and emission controls (Cheng et al., 2021b), Xu et al. (2022) reported that the joint impact of climate change and emission reduction would reduce the annual MDA8 O_3 concentration to 63.0 μg m^{-3} under ambitious scenario of carbon neutrality. Biogenic emission change is another factor influencing future O_3 (Chen et al., 2009; Andersson and Engardt, 2010; Harper and Unger, 2018; Wang et al., 2020). Liu et al. (2019) predicted a 24 % growth of BVOCs emissions driven by climate change under RCP8.5 from 2015 to the 2050s, resulting in a variation of daily 1-h maximum O_3 concentration ranging from −10.0 to 19.7 ppb across different regions in China.

Limitations exist in current studies, which prevent comprehensive assessment and understanding of the joint impacts of future changes of multiple factors on China's O_3 pollution. Firstly, the above estimations mainly focused on the influence of future changes on summertime or annual average O_3 concentration. As China's O_3 pollution has been reported to spread into spring and fall, it is of great importance to separate the impacts on warm (April to September, the six months with heaviest O_3 pollution for most part of China, Liu et al., 2023a) and non-warm season O_3 (October to March), considering the diverse air pollution sources and O_3 formation sensitivity to precursors for different seasons (Li et al., 2021; Wang et al., 2023). Recent studies have suggested diverse effects of future emission change on O_3 evolution for difference seasons in China (Hou et al. 2023; Liu et al. 2023b). In addition, the rising frequency of extreme weathers and declining anthropogenic emissions will further influence the possibility of extreme O_3 events, which has been scarcely discussed.

Secondly, to restrain global warming, China has made a national commitment to achieving "carbon neutrality" by 2060 (Shi et al., 2021), and accordingly launched a series of energy and climate action plans to reduce greenhouse gas emissions. These actions will also cause substantial reductions in air pollutant emissions, but have not been fully included in existing predictions of global emissions (Tong et al., 2020; Cheng et al., 2021b). Large bias will then be caused in the simulation of anthropogenic-induced future changes of air quality, with a less realistic estimate of local emission path (Cheng et al., 2021a). Due to probably faster decline of emissions in China but slower in surrounding countries in the future, the contributions of transboundary emissions on China's O_3 can be greatly changed and has not yet been fully considered (Hou et al. (, 2023). Thirdly, BVOCs emissions will not only be affected by meteorological factors but also by land use and land cover change (Penuelas and Staudt, 2010; Szogs et al., 2017; Wang et al., 2021a). Future land management will change due to socio-economic development and necessary actions as climate change response, and the changed shares of forest, cropland and grassland will alter the magnitude and distribution of BVOCs emissions and thereby affect O_3 concentration (Hurtt et al., 2020; Liao et al., 2020; Liu et al., 2022). Finally, the existing evaluations were conducted separately for individual influencing factors, with diverse methods and data. The interactions between different factors were seldom included in existing analyses, and the relative contributions of multiple factors were difficult to be evaluated or compared. Relevant studies have been conducted in developed countries (Gonzalez-Abraham et al., 2015), and are still lack in China.

In this study, we evaluate the complex influence of future changes of multiple factors on surface O_3 concentration in China within a uniform framework. The evaluation is conducted from the perspectives of seasonal, regional and extreme events of O_3 pollution. Four factors are included in the analyses, i.e., meteorological conditions, local anthropogenic emissions, BVOCs emissions, and anthropogenic emissions from surrounding foreign countries. The analyses are conducted based on a series of sensitivity experiments in numerical modelling of future air quality, and up-to-date input data from multiple sources are utilized in the model (see details in next section). We provide a comprehensive perspective on the spatiotemporal change of China's O_3 pollution till the 2060s, under a moderate way SSP2 of Shared Socioeconomic Pathways (SSPs, Riahi et al., 2017) and a midrange mitigation scenario RCP4.5, a scenario at the middle of the socio-economic developing way with radiative forcing at 4.5 W m^{-2} nominally by 2100 (Meinshausen et al., 2020). The outcomes highlight the regional and seasonal heterogeneity of O_3 pollution risks driven by complex future change of multiple factors, and support strategy design of O_3 pollution alleviation with specific principles, targets and action pathways.

2 Data and Methods

2.1 Main framework and research domain

The simulation framework incorporates the Weather Research and Forecasting model (WRF, version 3.7.1) to the generate hourly meteorological fields, the Model of Emissions of Gases and Aerosols from Nature (MEGAN, version 2.1) to calculate gridded BVOCs emissions, and the Community Multiscale Air Quality model (CMAQ, version 5.2) to simulate O_3 concentration. BVOCs emission calculations and air quality simulations are driven by meteorological fields of 2018–2022 (the 2020s, representing the current situation) and 2058–2062 (the 2060s, representing the future situation). All simulation results are averaged over a period of five years to mitigate the influence of interannual variability of meteorology. The modelling domain, same for WRF, MEGAN and CMAQ, covers East Asia, most areas of South Asia and Central Asia, and part of Southeast Asia and North Asia (Figure 1). It applies the Lambert Conformal Conic projection centered at (110° E, 34° N), and the horizontal resolution is 27 km×27 km, with 303×203 grids. The target area, Chinese mainland, includes 31 provincial-level administrative regions (excluding Hong Kong, Macao and Taiwan). Eight geographical regions are defined, and locations of the three regions with dense population and relatively heavy air pollution are also shown in Figure 1, namely BTH (Beijing-Tianjin-Hebei), YRD (Yangtze River Delta) and PRD (Pearl River Delta).

2.2 Data sources and processing methods

We use the bias-corrected RCP4.5 output of the National Center for Atmospheric Research's Community Earth System Model (NCAR CESM) as initial and boundary conditions for WRF (Monaghan et al., 2014). A ten-year dynamic downscaling simulation for 2018–2022 and 2058–2062 is conducted. Note we do not utilize the real-time reanalysis data to drive the simulation of the 2020s, in order to minimize the systematic error between the simulation driven by real meteorological conditions (for current simulations) and climate projection (for future simulations),

The BVOCs emissions are basically determined by meteorology and vegetation. The meteorological conditions are supplied by WRF. The vegetation data, including leaf area index (LAI), plant functional types (PFTs) and emission factors (EFs) of each PFT, are determined for 2020 and 2060. Gridded LAI data for 2020 are obtained from Global Land Surface Satellite product (Liang et al., 2021), and those for 2060 under SSP2-45 scenario are downscaled from the daily CESM2 output of Coupled Model Intercomparison Project Phase 6 (CMIP6). PFTs data for 2020 are derived from MCD12C1 product of Moderate-Resolution Imaging Spectroradiometer (MODIS) dataset and mapped to the 16 types required for MEGAN following Liao et al. (2020). The PFTs data for 2060 in China are obtained from Liao et al. (2020) under SSP2-45 scenario, with other regions maintaining those of 2020. EFs for each PFT are taken from Guenther et al. (2012).

Anthropogenic emissions for Chinese mainland are obtained from the Multi-resolution Emission Inventory for China (MEIC, http://meicmodel.org.cn/?page_id=560) for 2020, and DPEC version 1.1 under SSP2-45 incorporating the best available end-of-pipe pollution control technologies for 2060. The annual total eEmissions by country/region outside Chinese mainland are obtained from CMIP6 dataset under SSP2-45 scenario (O'neill et al., 2016; Gidden et al., 2019~~).~~), and t~~T~~hese emissions are downscaled into gridded monthly data for CMAQ simulation, based on the spatial and temporal distributions of emissions ~~outside Chinese mainland are assumed the same as those~~ in MIX Asian emission inventory (Li et al., 2017). The speciation profiles of NMVOCs are taken from MIX as well. Supplementary Figure S1 shows the emissions of two main precursors of O_3 by year and region. The NO_x and NMVOCs emissions for Chinese mainland were estimated to decline 58 % and 51 % from 2020 to 2060, respectively, much faster than those of surrounding areas within the modelling domain (8 % and 14 % respectively). In particular, the NO_x emissions would decline 57–62 % for the three developed regions BTH, YRD and PRD, while the reductions of anthropogenic NMVOCs would vary a lot among regions (36 %, 49 % and 60 % for BTH, YRD and PRD, respectively).

Carbon Bond 2005 (CB05, Yarwood et al., 2005) is adopted as the gas-phase chemical mechanism and the sixth-generation CMAQ aerosol module AERO6 (Appel et al., 2013) as aerosol chemistry mechanism. The initial and boundary conditions are set by default clean air conditions in CMAQ, and the first 10 days for each year are determined as the spin-up period to minimize the effects of initial and boundary conditions.

2.3 Simulation cases

Six cases of CMAQ simulations are conducted to investigate the impacts of future change of the four factors on O_3 concentration in China (Table 1). Cases 1 and 2 represent the current (2020s) and future (2060s) baseline, respectively, and the difference between them indicates the joint effect of the future changes of multiple factors. Each of Cases 3–6 applies the prediction for 2060s for one specific factor but keeps the remaining factors at current condition (2020s). Thus, the difference between each of those four cases and Case 1 indicates the impact of individual factor, including meteorological conditions (Case 3), domestic anthropogenic emissions

(Case 4), BVOCs emissions (Case 5) and anthropogenic emissions of surrounding countries (countries other than Chinese mainland within the modeling domain, Case 6). Surrounding areas refer to areas within the modelling domain but excluding Chinese mainland. Each case contains a five-year (2018–2022 or 2058–2062) WRF-MEGAN-CMAQ simulation driven by the varying meteorological conditions for individual years, and the five-year average of simulated O_3 concentrations is adopted for further analyses.

2.4 Model performance

To evaluate the model performance, we conduct a comparative analysis between simulations and observations for meteorological factors and O_3 concentrations, as well as an intercomparison for BVOCs estimates between different studies.

We first examine the capability of downscaled CESM climate projections in capturing the meteorological conditions of the 2020s. We applied the meteorological data from the National Climate Data Center (NCDC, archived at https://quotsoft.net/air) in 2020, and the statistical metrics are presented in Supplementary Table S1. The modeled temperature at 2 m (T2) is in good spatiotemporal agreement with the observations, with the correlation coefficient (R) of 0.96 and index of agreement (IOA) of 0.98. The relative humidity (RH) is also well predicted with R and IOA at 0.78 and 0.88, respectively. The model shows an overestimation on the wind speed by 1.41 m s^{-1}, which is also reported by Hu et al., (2022). The correlation coefficients of wind speed and direction are higher than 0.5. Overall, the modeled meteorological fields have basically captured the conditions in China and are appropriate for subsequent MEGAN and CMAQ simulations.

For BVOCs emissions, we compare our estimates for the 2020s with previous studies, as summarized in Supplementary Table S2. The total BVOCs, isoprene and terpenes emissions in this study are estimated at 33.55, 21.08 and 3.30 Tg yr^{-1}, respectively, and are comparable to other studies. In particular, our estimate is larger than others except for Li et al. (2020b) for isoprene, while smaller than others except for Wu et al. (2020a) for terpenes. The differences between studies might result from the diverse strategies of mapping PFTs from the original satellite products and the difference between downscaled climate conditions and the real meteorological fields.

We apply the observed MDA8 O_3 concentration data from the national network of China Ministry of Ecology and Environment (archived at https://quotsoft.net/air) to evaluate CMAQ performance. As shown in Supplementary Figure S2, the simulation could capture the spatiotemporal distribution of surface MDA8 O_3 concentration for the whole country and specific O_3 pollution hot spots, e.g., BTH and eastern Sichuan province with their surrounding areas. The statistical metrics of the comparisons between the simulated and observed monthly average MDA8 O_3 concentration of 2020s are summarize in Supplementary Table S1. The normalized mean biases (NMB) are calculated at 14.12 % and 10.90 % for warm and non-warm season, and R values at 0.71 and 0.32, respectively. Even with a slight overestimation, the reliability of our simulation is comparable to most previous studies in China, with a better performance in the warm season (Hu et al., 2016; Lu et al., 2019; Gao et al., 2020; Yang and Zhao, 2023).

~~In addition, w~~We evaluate the interannual varia~~tion~~bility within each of the five-year simulation~~s~~s, based on ~~using~~ the coefficient of variation (CV), the ratio of standard deviation ~~and~~ to mean of the simulated O_3 concentration~~s~~. As ~~is~~ shown in Supplementary Table S3, the CVs are generally below 5 % in most cases, indicating relatively small interannual variability ~~that~~ in ~~the simulated~~ O_3 concentration~~s exhibited~~ simulation~~ relatively low interannual variability~~. Th~~is suggest~~e results thus justify the representativeness of the five-year averages for present and future scenarios.~~s that the concentrations are stable and closely centered around the mean, with minimal fluctuation across the five-year simulations.~~

3 Results and Discussions

3.1 Future change of meteorology and BVOCs emissions

The downscaled changes in the meteorological factors from the 2020s to 2060s (SSP2-45 scenario) are shown in Figure 2, including temperature, RH and wind speed (WS). The changes are analyzed separately for April–September (warm season) and October–March (non-warm season). For the warm season, daily maximum temperature at 2 m (T-max) will increase across China with an average change of 1.0 °C, and the minimum and maximum changes are found in Tibetan Plateau at 0.1 °C and in Heilongjiang province at 2.1 °C, respectively. The RH will decrease slightly by −0.6 % for the whole country, with the changes for most areas within the range between −3 % and 0 % except for some areas of Northwestern China, Southwestern China, and Tibetan Plateau (see the region definitions in Figure 1). The growing T-max and declining RH will enhance the photochemical production of O_3 and BVOCs emissions. For the non-warm season, the national average growth of T-max will be smaller at 0.2 °C and some areas in Northeastern, Northern and Eastern China will even experience a decline ranging from −1.8 to 0 °C. The RH will change diversely across the country, ranging from −6.0 to 6.3 %. Very limited change in WS will occur, ranging from −0.1 to 0.2 m s^{-1} in most areas of the country. Generally, the decreasing wind speed in future East Asia could be attributed to weakened atmospheric circulation (Coumou et al., 2018; Deng et al., 2021). The increasing wind speed in non-warm season might result from the temperature and pressure gradients between the land and adjacent oceans (Yao et al., 2019; Wu et al., 2020b). The spatial distribution of downscaled future meteorological field changes is generally in agreement with those predicted by Hong et al. (2019) and Hu et al. (2022). Some discrepancies in temperature and wind speed change of non-warm season between studies result from the different choices of base year and parameterization schemes of WRF.

Table 2 shows China's BVOCs emissions of the 2020s and 2060s (SSP2-45 scenario) estimated with MEGAN, as well as the BVOCs emission intensity (emissions per unit area) for the three developed regions. The emissions will increase from 33.6 Tg yr^{-1} for the 2020s to 43.4 Tg yr^{-1} for the 2060s. The growth rates in BTH, YRD and PRD are predicted to be ~~22~~21.4 %, 23.9 % and 23.0 %, respectively, smaller than that for the whole country (29.2 %). The spatial distributions of BVOCs emissions for the 2020s and the changes from the 2020s to 2060s, are shown in Supplementary Figure S3. Areas all over China will experience the growth of BVOCs emissions, and it will be more prominent in areas with high vegetation coverage (e.g., Southern and Southwestern China) rather than urban areas. The growth of BVOCs emissions will enhance the contribution of natural sources to O_3 formation, especially along with declining anthropogenic emissions in the future (Penuelas and Llusia, 2003; Riahi et al., 2017; Gao et al., 2022).

3.2 Response of surface O_3 concentration to combined future changes

Figure 3 illustrates the spatial distributions of MDA8 O_3 concentrations for the warm and non-warm seasons of the 2020s and 2060s (SSP2-45 scenario), as well as the differences between the two periods. Briefly, future changes of the four factors under SSP2-45 are estimated to jointly reduce MDA8 O_3 by 7.7 and 1.1 ppb in the warm and non-warm season, respectively, while the O_3 responses to future changes will differ by region.

In the warm season of the 2020s (Figure 3a), the nationwide average MDA8 O_3 concentration is simulated at 57.3 ppb, and those of BTH, YRD and PRD are 73.7, 68.7 and 52.3 ppb, respectively. Hot spots of O_3 pollution, with average MDA8 O_3 over 75 ppb, are mainly located in Northern China and Sichuan province. The pattern is predicted to persist into the 2060s (Figure 3b), with a decline in both the severity and size of highly polluted regions. The nationwide MDA8 O_3 concentration will decline 13.4 % to 49.6 ppb, and that in most areas of China will be within the range of 37.5–67.5 ppb. The highest concentration will be lower than 75 ppb for the two hotspots of Northern China and Sichuan. BTH will remain as the most O_3-polluted area in warm season, with the O_3 concentration at 63.9 ppb (13.3 % smaller than the 2020s), while that of YRD and PRD will decrease to 53.9 (21.5 %) and 39.8 ppb

(23.9 %), respectively. O_3 concentration in the developed regions will decline faster than or roughly the same as that for the whole country. The reductions in MDA8 O_3 from 2020s to 2060s will be 10–20 ppb for Northern, Eastern, Central and Southern China and 0–10 ppb for Northeastern and Northwestern China as well as the Tibetan Plateau (Figure 3c). Notably, some areas in Sichuan are expected to experience a substantial decline of MDA8 O_3 over 20 ppb.

O_3 concentration of the non-warm season is simulated to be much lower than that of the warm season. The 2020s average MDA8 O_3 is 48.4 ppb, ranging from 30.0 to 67.5 ppb in most areas of China (Figure 3d). Different from the warm season in which highest concentration is found for Northern China and Sichuan, the Southern and Southwestern parts of China suffer the highest O_3 level for the non-warm season. A general west-to-east and south-to-north gradient is found for MDA8 O_3, with the lowest concentration found in Northern and Northeastern China. The concentrations in BTH and YRD are simulated at 33.8 and 45.1 ppb, respectively, much lower than that of PRD (58.9 ppb). Relatively high temperature during even the non-warm season is expected to expand the O_3 pollution period in Southern China. Resulting from complex change of multiple factors, the national average MDA8 O_3 concentration in the non-warm season of 2060s will decrease slightly to 47.3 ppb under SSP2-45, and that in most regions will be within the range of 37.5–52.5 ppb except for some areas in Northeastern China and Tibetan Plateau (Figure 3e). The MDA8 O_3 concentrations of the three developed regions will become closer at 39.3, 48.4 and 51.6 ppb for BTH, YRD and PRD, respectively. As illustrated in Figure 3f, MDA8 O_3 is predicted to increase in BTH and YRD and the surrounding areas, with the growth mostly ranging 0–15 ppb. In other areas (especially in Southern China), the concentration will decrease in the non-warm season by −15 to −5 ppb. As a result of the increased O_3 in the less polluted Eastern and Northern China and decreased O_3 in the more polluted Southwestern and Southern parts, the 2060s regional disparity in the non-warm season O_3 pollution will get smaller compared to the 2020s (Figure 3d and 3e).

To further explore the temporal pattern of O_3 level in the future, we compare the monthly average MDA8 O_3 in the 2020s and 2060s under SSP2-45 for the whole country and three developed regions (Figure 4 and Supplementary Figure S4). For the whole country (Figure 4a), the changes of monthly average MDA8 O_3 from 2020s to 2060s are estimated to range from −3.2 to −10.7 ppb in the warm season but less prominent in the non-warm season (from −2.7 to 0.9 ppb). Along with the more reduction in summertime (June, July and August), in particular, the periods with the highest O_3 concentration will expand into spring (March) and fall (October), as presented in Supplementary Figure S4. For the three regions, a greater decline in O_3 concentration is found in the warm season while a smaller or even a growth is found in the non-warm season. For BTH (Figure 4b), the monthly MDA8 O_3 concentrations range between 24.7 and 88.4 ppb in the 2020s with a clear difference between the warm and non-warm season. This pattern will remain in the 2060s with smaller difference between months (30.6–70.2 ppb). The temporal change pattern of YRD is similar to that in BTH, with decline in the warm season and growth in the non-warm season (Figure 4c). The shift of O_3 pollution from the warm towards the non-warm season is more prominent in the PRD, the only region where O_3 concentration of all the months in 2060s is predicted to decline (Figure 4d). Different from BTH and YRD, as mentioned above, higher O_3 concentrations during spring and autumn and lower in summer (due to the abundant summertime precipitation and high humidity) are found for PRD in the 2020s (Gao et al., 2020; Han et al., 2020). With great O_3 decline in the warm season, the periods experiencing peak O_3 pollution are predicted in the non-warm season of the 2060s, predominantly between October and March (Supplementary Figure S4).

3.3 Identifying surface O_3 response to individual factors

3.3.1 Local anthropogenic emission change

Figure 5 shows the influences of changes of each individual factors (local anthropogenic emissions, meteorological conditions, BVOCs emissions, and anthropogenic emissions from surrounding countries) on the warm and non-warm season O_3 concentrations. Out of the four, the change of local anthropogenic emissions is predicted to be the most influential factor, resulting in a national average decline of 7.2 and 0.8 ppb for the warm and non-warm season, respectively (Figure 5a and 5e). In the warm season, the emission reduction will play a positive role in reducing O_3 pollution in most areas of China, and the decrease will exceed 10 ppb across Northern, Eastern, Central, Southern and part of Southwestern China. In the non-warm season, emission reduction will have contrasting effects on MDA8 O_3 levels in the north and south part of China, enhancing MDA8 O_3 by 0–15 ppb for the former while restraining it by 0–10 ppb for the latter. Especially, the emission reduction is predicted to elevate the O_3 concentration by 5.9 and 4.0 ppb for BTH and YRD respectively.

Supplementary Figure S5 shows the relative emission reductions from 2020s to 2060s by region. Under SSP2-45 scenario, the reductions of NO_x and VOCs emissions will range from 35.6 % to 63.6 % for different regions, and VOCs emission reduction will be less than that of NO_x except for PRD. As the NO_x-limited regime for O_3 formation (i.e., O_3 is more sensitive to NO_x emission change) occurs more frequently in the warm season while the VOC-limited regime more in the non-warm season, the larger decline of NO_x emissions than VOCs should be more effective in restraining the warm season O_3 pollution but has less benefit or even negative effect in the non-warm season (Sillman and He, 2002). Wintertime of NCP and YRD have been reported under the VOC-limited regime and the excessive NO_x emissions play an important role in removing O_3 by titration effect (Jin and Holloway, 2015; Li et al., 2021; Wang et al., 2021b). This may explain the MDA8 O_3 increase during the non-warm season with insufficient reduction of VOCs (35.6 % and 49.5 %) but sharp reduction of NO_x of 53.4 % and 60.3 % for NCP and YRD, respectively. Similarly, Hou et al. (2023) and Liu et al. (2023b) also predicted a growth of O_3 concentration in non-warm season over BTH and YRD under a net-zero carbon emission scenario, resulting from a weakened titration effect. Supplementary Figure S6 shows the monthly variation of O_3 and odd oxygen (O_x, O_x=O_3+NO_2, representing the real photochemical production potential of O_3 considering the titration effect) in the 2020s and 2060s. It should be noted that the growth of O_3 in the non-warm season in 2060s for BTH and YRD will be accompanied by minimal change of O_x, while the declines of O_3 and O_x will appear simultaneously in the warm season for the three regions and in non-warm season for PRD. This indicates that the growth of non-warm season O_3 in BTH and YRD should result partly from NO_x reduction and thereby weakened NO titration, as titration is a key pathway of O_3 loss when the chemical reactivity is relatively low in winter (Gao et al., 2013; Akimoto and Tanimoto, 2022). The differentiated O_3 responses to precursor reduction between YRD and PRD have also been detected during the COVID-19 breakout period. With the O_3 isopleth plots, Wang et al. (2021b) illustrated that 40–60 % reduction of NO_x and VOCs enhanced the O_3 formation in YRD under the VOC-limited regime but suppressed O_3 in PRD under the transitional regime (a regime between NO_x- and VOC-limited). Therefore, VOCs emission controls should be better addressed for O_3 pollution alleviation when it expands to non-warm season in the future.

3.3.2 Meteorological condition change

As shown in Figure 5b and 5f, the influence of meteorological change exhibits different patterns for the warm and non-warm season.–

In the warm season, meteorological change due to global warming will play a positive role on O_3 formation in most of China, with the enhancement within 0–4 ppb, but it will reduce the O_3 level in remote areas like Tibetan Plateau. The national average growth will be 0.3 ppb and that for YRD, PRD and BTH will be 1.9, 0.7, and 0.3 ppb, respectively. The response of O_3 to meteorological change is associated with some specific variables (Hong et al., 2019). For example, the great enhancement of O_3 in YRD might be attributable to a hotter, dryer and more stable atmosphere with growth of T-max (over 0.6 ℃) and decline of RH and WS (Figure 2). The result is similar to Hong et al. (2019), which reported a change of 2–8 ppb of daily 1-hour maximum O_3 concentration for the peak season from the 2010s to 2050s under RCP4.5. In addition, the declining O_3 in Tibetan Plateau and the surrounding areas might result partly from the weakened long-range transport of peroxyacetyl nitrate (PAN, the principal NO_x reservoir) from the polluted areas (Fischer et al., 2014). Driven by the elevated temperature, PAN from relatively polluted regions will undergo stronger thermal decomposition locally, thus fail to be transported far away to the remote regions to promote O_3 formation (Liu et al., 2013; Lu et al., 2019).

The influence of meteorological change on O_3 production is predicted to be much smaller for the non-warm season, with the magnitude within ±1 ppb in most areas and nationwide average at −0.2 ppb. In the three developed regions, the changes are predicted to range from −0.4 to 0.3 ppb, with little regional difference. The limited influence might be attributable to the modest change in temperature and RH in the non-warm season.

3.3.3 BVOCs and surrounding anthropogenic emission change

Compared to domestic emissions, change of BVOCs emissions and anthropogenic emissions from surrounding countries will have a less influence (within ±3 ppb) on surface O_3 concentration in China. BVOCs change tends to enhance O_3 while foreign emission change tends to restrain it in most areas (Figure ~~7~~5).

The growing BVOCs emissions due to vegetation and climate change is estimated to enhance O_3 concentration by 0–3 ppb in the most areas of China, with a larger influence of 0.6 ppb in the warm season than that of 0.3 ppb in the non-warm season across the country (Figure 5c and 5g). In the warm season, relatively large growth of O_3 concentration will occur in BTH at 2.1 ppb, and those of YRD and PRD will be 1.5 and 1.0 ppb, respectively. The abundant NO_x emissions in BTH are expected to result in a larger O_3 concentration response to BVOCs emission change than YRD and PRD, even the BVOCs emission change of BTH will be smaller than the other two regions (Table 2). The result is in agreement with other numerical simulation experiments. Liu et al. (2019) reported a prominent O_3 enhancement even with a low BVOCs emission rate under RCP8.5, in a NO_x-abundant environment. In the remote areas like Tibetan Plateau and part of Northeastern China, the increased BVOCs will remove O_3 due to the isoprene ozonolysis in low-NO_x environment (Hollaway et al., 2017; Zhu et al., 2022). In general, regions with higher O_3 pollution levels and NO_x emissions will suffer more risk of O_3 growing from rising BVOCs emissions in the future.

Most areas of China will benefit from the foreign emission change in terms of O_3 pollution alleviation (Figure 5d and 5h). An exception is Tibetan Plateau and its surrounding areas, which will be affected by the elevated emissions of NO_x and VOCs from South Asia under SSP2-45. Limited by the range of pollutant transport, greater impacts will be found for coastal and border areas and less for inland areas (Ni et al., 2018). Larger O_3 changes in the three developed regions are predicted than that of the whole country, benefitting from the precursor emission reduction in East Asia and Southeast Asia.

3.4 The relationship between the joint and separate effects of multiple factors

Figure 6 summarizes the contributions of individual factors to the total O_3 change by region and season. Due to the nonlinear response of O_3 to multi-factor changes, the aggregated contribution of the four factors does not equal to the joint contribution (i.e., there exist gaps between the difference of the 2020s and 2060s and the aggregated contribution of four factors).

The varying domestic anthropogenic emissions are predicted to dominate the change of the future O_3, with a relative contribution ranging from 75 % to 117 % for different regions and seasons. The relative contributions of the other three factors are estimated to be limited within ±25 % at national and regional level. Among different regions, YRD will be more affected by climate change with the contribution of −13 % and −12 % for the warm and non-warm season, respectively, far greater than that of

BTH and PRD (−6 % to 0 %). BTH will be more affected by BVOCs emission change than other regions in the warm season (−21 %), while YRD and PRD will be more affected in the non-warm season with the relative contributions of 17 % and −20 %, respectively. Little regional difference is found for the relative contributions of foreign emission change.

To better understand the regional and seasonal differences of the relative contributions of future changes to O_3 concentration, we examine the nonlinear response of O_3 to precursor change in the three developed regions. We follow Chen et al. (2021) and Schroeder et al. (2017), and conduct a fit of lognormal distribution for the relationship of modeled hourly O_3 and NO_2 concentrations, as shown in Figure 7. The data points on the left of the turning point of fitted curve suggest a NO_x-limited regime while on the right a VOC-limited regime, and data points around the turning point are under transitional regime.

The O_3-NO_2 relationship from the 2020s to 2060s will be mostly influenced by the changing domestic anthropogenic emissions, indicated by the close distributions of data points and fitted curves between "EMIS" and "2060s" in Figure 7. In the warm season, the future O_3-NO_2 relations in BTH and YRD are predicted to change greatly from a highly O_3 polluted situation with moderate NO_2 concentration to a situation with a relatively low level of NO_2 (mostly under 10 ppb) and a moderate level of O_3 (under 60 ppb). A weak VOC-limited regime appeared for the whole BTH in 2020s, ~~and it is~~ consistent with recent observation-based analysis (Chen et al. 2023; Kong et al. 2024).~~and t~~ There is big diversity within the region, including a dense area with strong VOC-limited regime and other areas with transitional or NO_x-limited regime (Figure 7a). Represented by the moving of most points from the right of the turning point to near or left of the turning point, the NO_x-limited and transitional regimes will dominate BTH in the 2060s. Compared to 2020s, the data points of 2060s are more closely distributed, indicating a reduced diversity of O_3 formation regime in the region. For YRD, most areas were under transitional or weak

VOC-limited regime in the 2020s with limited diversity within the region, and the situation in 2060s will be similar to that of BTH (Figure 7a and 7b). The shift from weak VOC-limited regime in 2020s to transitional or NO_x-limited regime in 2060s for BTH and YRD implies the influence of emission reduction on altering the sensitivity of O_3 formation to precursors. Most areas of PRD in the 2020s are under transitional or NO_x-limited regimes, and the regime will transfer to a strong NO_x-limited one in 2060s, with an almost positive correlation between NO_2 and O_3 in a low-NO_2 environment (Figure 7c). In the non-warm season, O_3 and NO_2 will remain negatively correlated for BTH and YRD till the 2060s, which suggests a persistent VOC-limited regime and explains the O_3 concentration growth along with substantial precursor emission reductions. The turning points are simulated at extremely low NO_2 concentrations of 2.0 and 1.2 ppb for BTH and YRD, respectively (Figure 7d and 7e). A big challenge still exists on effective emission controls to reduce the O_3 concentration in the non-warm season for the two regions. Differently, the O_3 formation sensitivity in most of PRD will shift from transitional regime towards a more NO_x-limited situation (Figure 7f). A simple comparison with the O_3 evolution and its precursor emission changes in developed country provided more policy implication. According to Chen et al. (2021) and the US Environmental Protection Agency, the northeastern US ~~underwen~~experienced~~t~~ rapid cross of the turning point of O_3 formation sensitivity during 1990s~~—~~–2010s, with approximately 60 % reductions in both anthropogenic NO_x and VOCs emissions. ~~However~~In BTH, the emissions ~~declines~~ are predicted to decline~~ be~~ 57 % and 36 % for NO_x and NMVOCs during 2020s–2060s ~~for BTH~~ under SSP2-45~~ scenario~~. ~~To accelerate the shift in the O_3 chemical regime for BTH,~~ Therefore, more ambitious reductions in NMVOCs will be necessary (~~—potentially~~ideally ~~even~~ double the current projected~~ efforts~~ abatement under SSP2-45), to accelerate the shift in the O_3 chemical regime for BTH.~~,~~

The fitted curves of other three factors are similar to those of the 2020s, and the change of these factors will make little difference on NO_2 concentration but will result in moderate changes on O_3 concentration within ±2 ppb. The limited changes of climate, BVOCs emissions and foreign anthropogenic emissions will not essentially alter the O_3 formation regime, but may change the O_3 production under the nearly same NO_2 concentration. Changes of individual meteorological factors are expected to easily influence the O_3 and NO_2 concentrations (Pope et al., 2015; Liu and Wang, 2020; Dewan and Lakhani, 2022). The modeled little response of NO_2 to meteorological change, except that in the non-warm season for BTH, might be attributed to the compensating effect of different variables. The limited influence of BVOCs on the O_3 formation sensitivity to precursors is consistent with Gao et al. (2022), which reported comparable empirical kinetic modelling approach (EKMA) curves with and without BVOCs emissions. The transboundary O_3 pollution results from the transport of both O_3 and its precursors (mainly associated with PAN), while NO_2 is less influenced by long-range transport due to its shorter lifetime (Ni et al., 2018; Yin et al., 2022).

The change in O_3 formation regime might partly explain the finding that the joint effect of multiple factors on restraining O_3 pollution will be larger than the aggregated effects of individual factors. Under a NO_x-limited regime, O_3 is less sensitive to changing VOCs emissions (e.g., BVOCs emissions) than that under a VOC-limited one. Therefore, the enhancement of O_3 due to BVOCs emission growth in the future will be restrained with a much lower NO_2 concentration. This indicates a co-benefit of reducing the anthropogenic emissions to restrain the potential O_3 pollution elevation due to growing BVOCs emissions (as a part of climate penalty) in the future.

3.5 Change of O_3 exceedance events over the east of China

Figure 8 shows the "O_3 exceedance events" over the east of China (mainly including Northern, Eastern, Central and Southern China) in the 2020s and 2060s, and the changes influenced by different factors. The exceedance is defined as number of days with the MDA8 O_3 exceeding the Chinese National Air Quality Standard-Grade II (160 μg m^{-3} or 81.6 ppb). The exceedance events appear mainly in the warm season

(Figure S7). Areas with frequent exceedance (over 50 days) in the 2020s were mainly located in Northern China. Much fewer exceedances are found for YRD and PRD (19.3 and 8.2 days in 2020s, respectively). In the 2060s, the O_3 exceedance events will drop significantly. The exceedance days will be fewer than 10 days for most of the country, except for some areas in BTH which will still have more than 20 exceedance days over the year.

Domestic emission abatement will be the most important factor reducing the O_3 exceedance, particularly in Northern China. The exceedance days will be cut by 45.3, 19.1 and 8.1 days for BTH, YRD and PRD, respectively, with the maximum reduction reaching 80 days within BTH and YRD. Notably, the spatial pattern of changing O_3 exceedance due to emission reduction is different from that of changing MDA8 O_3 due to emission reduction as shown in Figure 5a. Even the warm season MDA8 O_3 concentration of BTH will decline only 9.7 ppb, the O_3 exceedance events will be greatly reduced, indicating that national emission controls will be especially effective in reducing serious O_3 pollution. Climate change will mainly affect Jiangsu, Anhui, Henan and Hebei provinces, elevating the exceedance by more than 15 days in most of these areas. For YRD and PRD, climate change will elevate the exceedance by 9.5 and 3.3 days, respectively. Some areas of BTH will benefit from climate change, with the exceedance declining 0–10 days. The influences of BVOCs and foreign emission change on exceedance days are of limited regional differences, with a growth of 5 to 15 days for the former and a decline of −5 to 0 days for the latter. The exceedances elevated by BVOCs emission growth will be 6.6, 6.1 and 2.8 days for BTH, YRD and PRD with the maximum reaching 19, 18 and 12 days within the region, respectively, reflecting an unneglectable role of biogenic source change on future O_3 episodes.

4 Conclusions

We explore the response of China's surface O_3 concentration to the future changes of multiple factors under SSP2-45, based on a series of sensitivity experiments with WRF-MEGAN-CMAQ simulations. From the 2020s to 2060s, the MDA8 O_3 concentration is predicted to decline by 7.7 and 1.1 ppb in the warm and non-warm season, respectively, and the O_3 exceedances of Chinese National Air Quality Standard (Grade II) will be largely eliminated. In the warm season, MDA8 O_3 in BTH, YRD and PRD will decline by 9.7, 14.8 and 12.58 ppb, respectively, larger than the national average level. However, MDA8 O_3 will increase in BTH and YRD in the non-warm season attributed to the reduced NO_x emissions and thereby titration effect. The O_3 pollution will expand towards the non-warm season in the future, bringing new challenge for policy makers to optimize the strategy of precursor emission controls based on local conditions.

Reduction of local anthropogenic emissions is estimated to dominate the spatial distribution and magnitude of future O_3 change. Meteorological variation will lead to a change of MDA8 O_3 ranging between −1 and 4 ppb for most areas in the warm season. The influences of changing BVOCs and foreign anthropogenic emissions will be within ±3 ppb, with the former elevating O_3 while the latter reducing O_3. Especially in areas with high O_3 pollution and intense NO_x emissions, the growing BVOCs emissions will more enhance the risk of O_3 pollution. The joint effect of multiple factors on restraining O_3 pollution will be larger than the aggregated effects of individual factors, which can be partly explained by the changing O_3 formation regime. Large amount of emission reduction under SSP2-45 will reshape the O_3 formation sensitivity to precursors. In BTH and YRD, O_3 formation in the warm season is projected to shift from weak VOC-limited to transitional or NO_x-limited regime, while VOC-limited regime will still dominate in the non-warm season. In the future, O_3 will be less sensitive to BVOCs change in a low NO_x environment along with persistent emission controls, highlighting the benefit of anthropogenic emissions abatement on mitigating the climate penalty and limiting O_3 pollution.

Limitations exist in current study. Firstly, the future climate data are taken from one single model CESM, subject to bias in the assessment of meteorological influence on O_3. Secondly, some factors that will influence future O_3 level are not included in our analyses, such as the changing CH_4 concentration, increasing soil NO_x emissions and the stratosphere-troposphere exchange of O_3. For example, the hotspot of soil NO_x emissions in northern China is also the region with ~~high~~ large reduction of anthropogenic NO_x ~~reductions~~ emission but relatively small ~~reductions~~decline in O_3 concentrations. Under a warmer ~~climte~~climate, ~~The~~a growing trend of ~~future increasing~~ soil NO_x emissions ~~are~~is expected ~~to~~ for the future, and may thus present an additional challenge for ~~anthropogenic NO_x~~O_3 pollution ~~allievation~~alleviation. ~~emissions control.~~ Thirdly, there exist gaps between the downscaled and realistic conditions of meteorology for the 2020s, leading to uncertainty in the O_3 simulation. Finally, the changing O_3 formation regime is presented through the relation between O_3 and NO_2 concentrations, and the mechanism how the climate penalty will influence O_3 formation under substantial reduction of anthropogenic emissions needs to be better analyzed in future studies.

Data availability

All data in this study are available from the authors upon request.

Author contributions

JYang developed the methodology, conducted the work and wrote the draft. YZhao improved the methodology, supervised the work and revised the manuscript. YWang and LZhang contributed to the methodology and provided supports to the scientific interpretation and discussions.

Competing interests

The authors declare that they have no conflict of interest.

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FIGURE CAPTIONS

Figure 1 The modelling domain and geographical definitions (denoted by colors) of this study. Boundaries of the three regions, including BTH (Beijing-Tianjin-Hebei), YRD (Yangtze River Delta) and PRD (Pearl River Delta), are marked by dark grey lines.

Figure 2 Projected changes of the ~~strongly~~ ozone-related meteorological ~~elements~~factors, including daily maximum temperature at 2 m (T-max, a and d), relative humidity (RH, b and e) and wind speed (WS, c and f), from the 2020s to 2060s. Panels (a-c) represent those of the warm season, and panels (d-f) represent those of non-warm season.

Figure 3 Simulation and projection of seasonal average MDA8 O_3 in the 2020s (Case1, a and b) and 2060s (Case2, d and e), and the changes over this period (Case2−Case1, c and f). Panels (a-c) represent those of the warm season, and panels (d-f) represent those of non-warm season. Regional mean concentrations across China (CHN), BTH, YRD and PRD are inset.

Figure 4 Simulation and projection of monthly average MDA8 O_3 in the 2020s and 2060s across CHN (a), BTH (b), YRD (c) and PRD (d).

Figure 5 Projected changes of MDA8 O_3 from the 2020s to 2060s attributed to anthropogenic emissions from local sources (Case3−Case1, a and e), meteorological conditions (Case4−Case1, b and f), BVOCs emissions (Case5−Case1, c and g) and anthropogenic emissions from surrounding countries (Case6−Case1, d and h). Panels (a-d) represent those of the warm season, and panels (e-h) represent those of non-warm season. Regional mean changes across CHN, BTH, YRD and PRD are inset.

Figure 6 The relationships between the separate MDA8 O_3 changes attributed to the four factors (denoted by the name of Case3–6) and the total changes from the 2020s to 2060s over China and the three regions. Panels (a-d) represent those of the warm season, and panels (e-h) represent those of non-warm season. The relative contributions of the four factors to the total influence of future change are shown in the light grey box.

Figure 7 The relationships between simulated hourly NO_2 and O_3 concentrations with the lognormal fits for different regions and seasons. The colored circles, representing different cases, come from the seasonal average concentrations for each grid in the target region. The specific circles with black border represent the regional average situation, and the turning points of every fitted curve are marked by the “+” sign. The density plots of the 2020s and 2060s are inset.

Figure 8 Projected annual O_3 exceedance over the east of China in the 2020s and 2060s, and the exceedance changes when the four factors at 2060s level. Regional mean changes across CHN, BTH, YRD and PRD are inset.

TABLES

Table 1. List of simulation cases to investigate the impact of future change upon surface O_3 in China, with sensitivity experiments from the perspectives of four main influencing factors.

Case number	Case name	China's local emissions	Meteorological conditions	BVOCs emissions	Surrounding emissions
Case1	2020s	2020	2018-2022	2018-2022	2020
Case2	2060s	2060	2058-2062	2058-2062	2060
Case3	EMIS	2060	2018-2022	2018-2022	2020
Case4	CLIM	2020	2058-2062	2018-2022	2020
Case5	BVOC	2020	2018-2022	2058-2062	2020
Case6	SURR	2020	2018-2022	2018-2022	2060

Table 2. The BVOCs estimation over China and emission intensity in BTH, YRD and PRD of the 2020s and 2060s, as well as the corresponding growth rates over this period.

	China	BTH	YRD	PRD
	Emissions (Tg)	Emission intensity (Gg grid^{-1})		
2020s	33.6	1.4	4.6	8.7
2060s	43.4	1.7	5.7	10.7
Growth rate	29.2 %	21.4 %	23.9 %	23.0 %

FIGURES

Figure 1

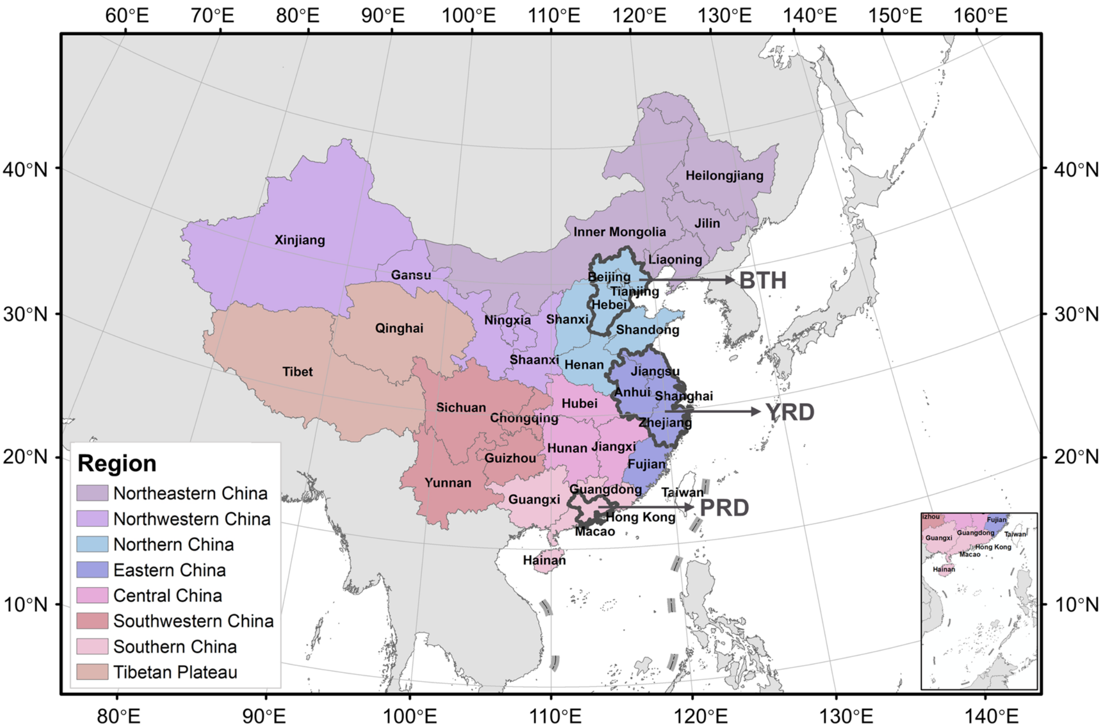
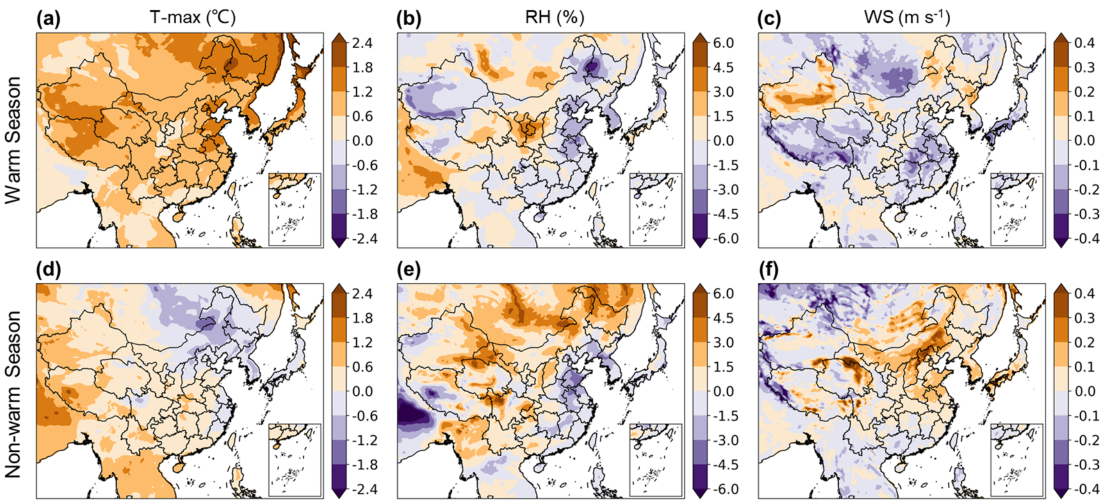
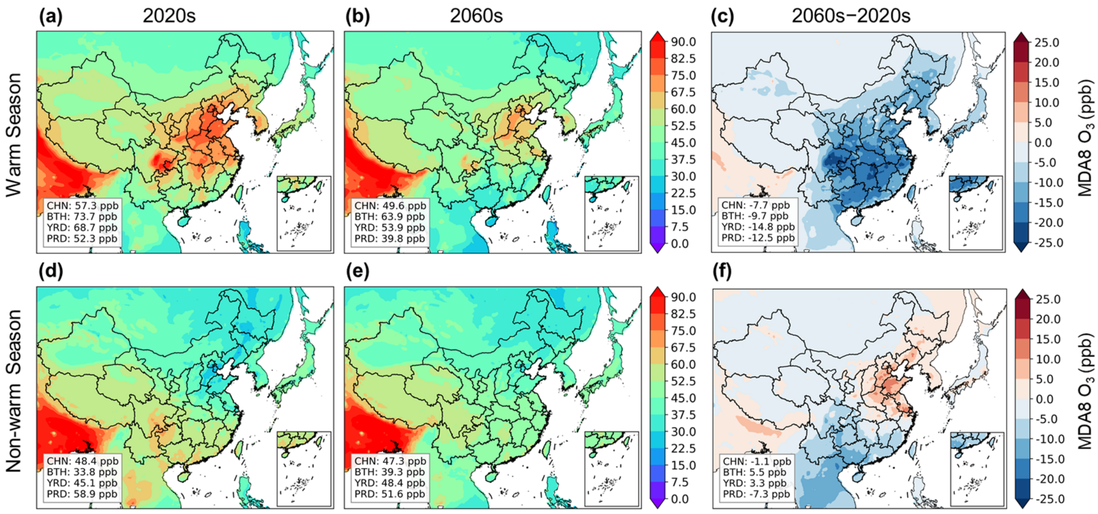


Figure 2



(a)
T-max (℃)
(b)
RH (%)
(c)
WS (m s-1)
Warm Season
(d)
(e)
(f)
Non-warm Season
2.4
1.8
1.2
0.6
0.0
-0.6
-1.2
-1.8
-2.4
6.0
4.5
3.0
1.5
0.0
-1.5
-3.0
-4.5
-6.0
0.4
0.3
0.2
0.1
0.0
-0.1
-0.2
-0.3
-0.4

Figure 3


(a)
2020s
(b)
2060s
(c)
2060s−2020s
Warm Season
Non-warm Season
(d)
(e)
(f)
CHN: 57.3 ppb
BTH: 73.7 ppb
YRD: 68.7 ppb
PRD: 52.3 ppb
CHN: 49.6 ppb
BTH: 63.9 ppb
YRD: 53.9 ppb
PRD: 39.8 ppb
CHN: -7.7 ppb
BTH: -9.7 ppb
YRD: -14.8 ppb
PRD: -12.5 ppb
CHN: 48.4 ppb
BTH: 33.8 ppb
YRD: 45.1 ppb
PRD: 58.9 ppb
CHN: 47.3 ppb
BTH: 39.3 ppb
YRD: 48.4 ppb
PRD: 51.6 ppb
CHN: -1.1 ppb
BTH: 5.5 ppb
YRD: 3.3 ppb
PRD: -7.3 ppb
90.0
82.5
75.0
67.5
60.0
52.5
45.0
37.5
30.0
22.5
15.0
7.5
0.0
25.0
20.0
15.0
10.0
5.0
0.0
-5.0
-10.0
-15.0
-20.0
-25.0
MDA8 O_3 (ppb)

1103 **Figure 4**

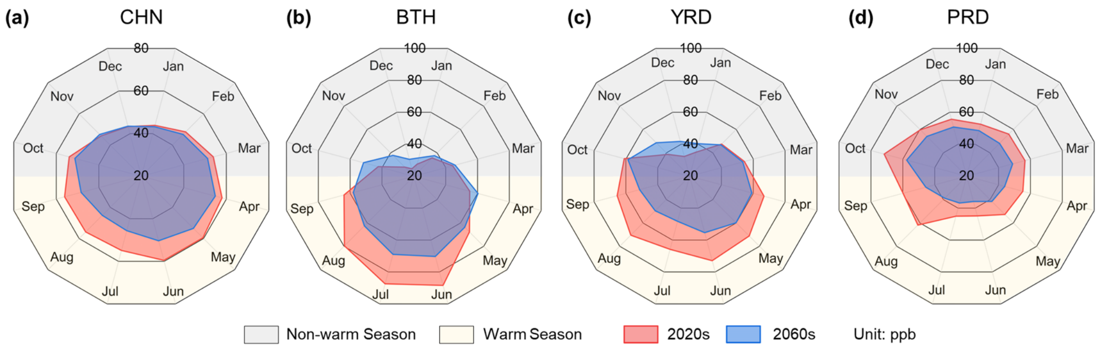


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Figure 5

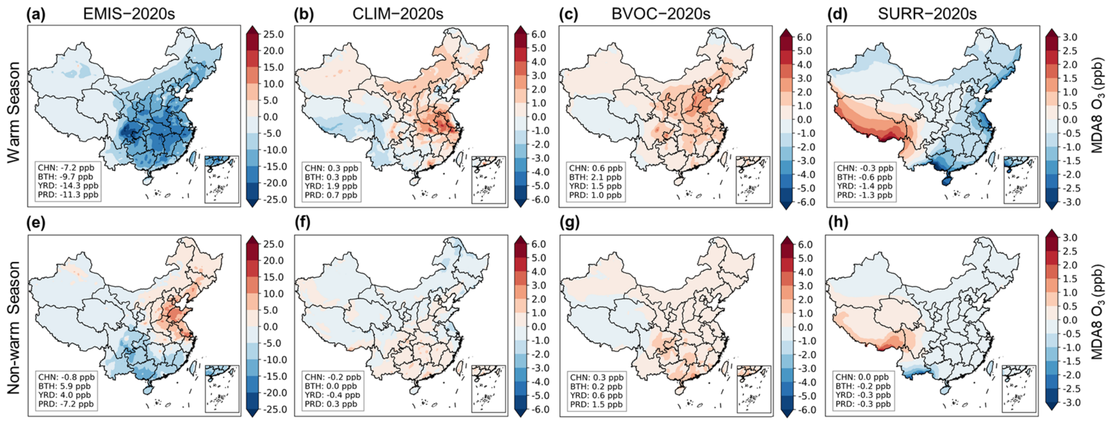

(a)
EMIS−2020s
(b)
CLIM−2020s
(c)
BVOC−2020s
(d)
SURR−2020s
Warm Season
Non-warm Season
CHN: -7.2 ppb
BTH: -9.7 ppb
YRD: -14.3 ppb
PRD: -11.3 ppb
CHN: 0.3 ppb
BTH: 0.3 ppb
YRD: 1.9 ppb
PRD: 0.7 ppb
CHN: 0.6 ppb
BTH: 2.1 ppb
YRD: 1.5 ppb
PRD: 1.0 ppb
CHN: -0.3 ppb
BTH: -0.6 ppb
YRD: -1.4 ppb
PRD: -1.3 ppb
(e)
(f)
(g)
(h)
CHN: -0.8 ppb
BTH: 5.9 ppb
YRD: 4.0 ppb
PRD: -7.2 ppb
CHN: -0.2 ppb
BTH: 0.0 ppb
YRD: -0.4 ppb
PRD: 0.3 ppb
CHN: 0.3 ppb
BTH: 0.2 ppb
YRD: 0.6 ppb
PRD: 1.5 ppb
CHN: 0.0 ppb
BTH: -0.2 ppb
YRD: -0.3 ppb
PRD: -0.3 ppb
25.0 20.0 15.0 10.0 5.0 0.0 -5.0 -10.0 -15.0 -20.0 -25.0
6.0 5.0 4.0 3.0 2.0 1.0 0.0 -1.0 -2.0 -3.0 -4.0 -5.0 -6.0
3.0 2.5 2.0 1.5 1.0 0.5 0.0 -0.5 -1.0 -1.5 -2.0 -2.5 -3.0
MDA8 O_3 (ppb)

Figure 6

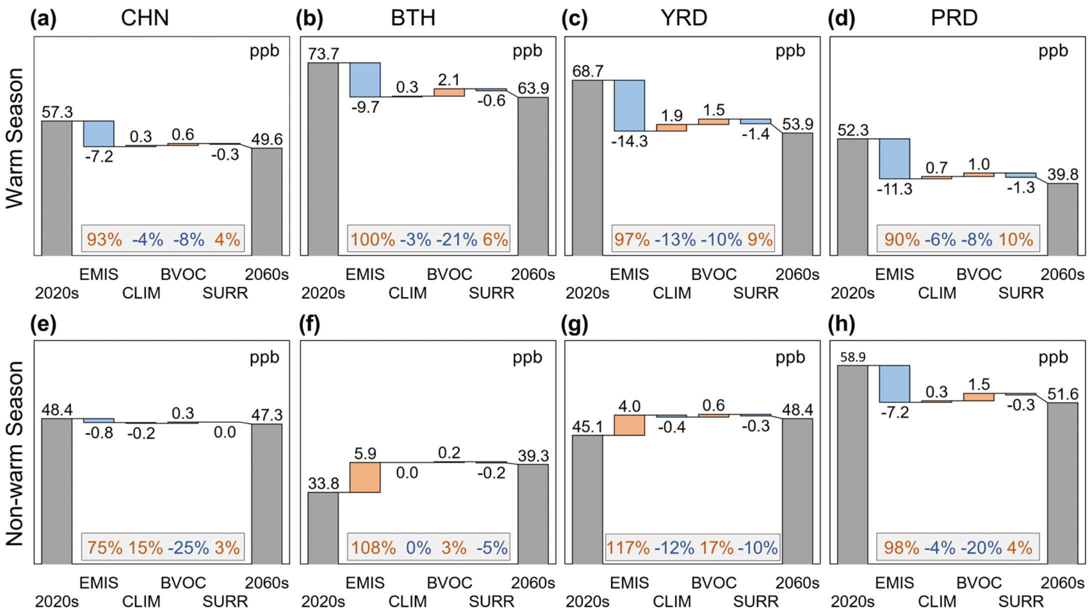

(a)
CHN
(b)
BTH
(c)
YRD
(d)
PRD
Warm Season
Non-warm Season
ppb
57.3
-7.2
0.3
0.6
-0.3
49.6
93% -4% -8% 4%
73.7
-9.7
0.3
2.1
-0.6
63.9
100% -3% -21% 6%
68.7
-14.3
1.9
1.5
-1.4
53.9
97% -13% -10% 9%
52.3
-11.3
0.7
1.0
-1.3
39.8
90% -6% -8% 10%
(e)
48.4
-0.8
-0.2
0.3
0.0
47.3
75% 15% -25% 3%
(f)
33.8
5.9
0.0
0.2
-0.2
39.3
108% 0% 3% -5%
(g)
45.1
4.0
-0.4
0.6
-0.3
48.4
117% -12% 17% -10%
(h)
58.9
-7.2
0.3
1.5
-0.3
51.6
98% -4% -20% 4%
2020s
EMIS
CLIM
BVOC
SURR
2060s

Figure 7

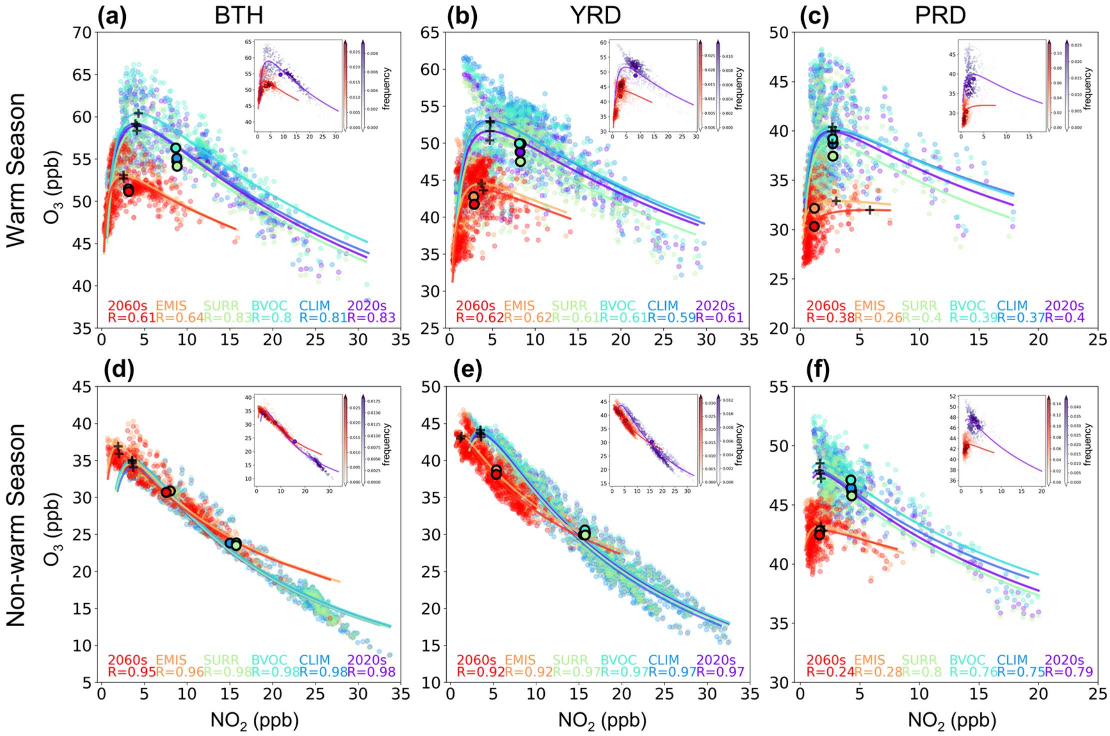
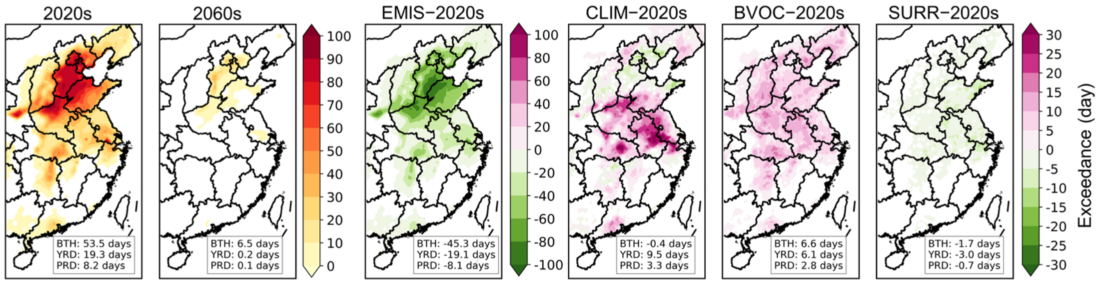

(a) BTH
(b) YRD
(c) PRD
(d)
(e)
(f)
Warm Season
Non-warm Season
O3 (ppb)
NO2 (ppb)
frequency
2060s EMIS SURR BVOC CLIM 2020s
R=0.61 R=0.64 R=0.83 R=0.8 R=0.81 R=0.83
R=0.62 R=0.62 R=0.61 R=0.61 R=0.59 R=0.61
R=0.38 R=0.26 R=0.4 R=0.39 R=0.37 R=0.4
R=0.95 R=0.96 R=0.98 R=0.98 R=0.98 R=0.98
R=0.92 R=0.92 R=0.97 R=0.97 R=0.97 R=0.97
R=0.24 R=0.28 R=0.8 R=0.76 R=0.75 R=0.79

Figure 8



2020s
2060s
EMIS−2020s
CLIM−2020s
BVOC−2020s
SURR−2020s
BTH: 53.5 days
YRD: 19.3 days
PRD: 8.2 days
BTH: 6.5 days
YRD: 0.2 days
PRD: 0.1 days
BTH: -45.3 days
YRD: -19.1 days
PRD: -8.1 days
BTH: -0.4 days
YRD: 9.5 days
PRD: 3.3 days
BTH: 6.6 days
YRD: 6.1 days
PRD: 2.8 days
BTH: -1.7 days
YRD: -3.0 days
PRD: -0.7 days
100
90
80
70
60
50
40
30
20
10
0
100
80
60
40
20
0
-20
-40
-60
-80
-100
30
25
20
15
10
5
0
-5
-10
-15
-20
-25
-30
Exceedance (day)

Supplement of

Investigating the response of China's surface ozone concentration to the future changes of multiple factors

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Table list

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Tables

Table S1 Statistical metrics for comparison between the observed and simulated monthly average meteorological variables and MDA8 O_3 in 2020. OBS is mean of observation, SIM is mean of simulation, and Bias is mean bias between SIM and OBS. R, NMB, NME and IOA refers to the correlation coefficient, normalized mean bias, normalized mean error and index of agreement between SIM and OBS, respectively.

Variables		OBS	SIM	Bias	R	NMB	NME	IOA
T2 (°C)		13.21	12.53	−0.69	0.96	−5.20 %	17.65 %	0.98
WS (m s^{-1})		2.60	4.02	1.41	0.51	54.21 %	60.04 %	0.57
WD (°)		175.75	174.78	−0.97	0.51	−0.55 %	18.40 %	0.72
RH (%)		65.25	66.17	0.92	0.78	1.41 %	12.78 %	0.88
MDA8 O_3 (ppb)	Warm season	57.45	~~8.11~~65.56	8.11	0.71	14.12 %	16.33 %	0.74
	Non-warm season	38.67	~~4.21~~42.88	4.21	0.32	10.90 %	25.48 %	0.49

Table S2 Comparison of BVOCs estimates between this study and previous estimations.

Species	Period	Annual emission (Tg)	Reference
Total BVOCs	2020s	33.55	This study
	2015–2019	29.28±0.91	Ma et al. (2021)
	2015–2019	31.42±0.95	Ma et al. (2021)
	2008–2018	54.60	Li et al. (2020)
	2001–2016	34.27	Wang et al. (2021)
	2017	23.54	Wu et al. (2020)
Isoprene	2020s	21.08	This study
	2015–2019	13.88±0.57	Ma et al. (2021)
	2015–2019	14.29±0.54	Ma et al. (2021)
	2008–2018	29.30	Li et al. (2020)
	2001–2016	15.94	Wang et al. (2021)
	2017	13.30	Wu et al. (2020)
Terpenes	2020s	3.30	This study
	2015–2019	5.28±0.12	Ma et al. (2021)
	2015–2019	4.77±0.11	Ma et al. (2021)
	2017	3.09	Wu et al. (2020)

Table S3 Coefficients of variation (CVs) for simulated O_3 concentrations within five-year simulations for the whole country (CHN) and selected developed regions (BTH, YRD, and PRD).

Simulation	Warm Season				Non-warm Season			
	CHN	BTH	YRD	PRD	CHN	BTH	YRD	PRD
2020s	2.6 %	5.7 %	1.8 %	4.0 %	1.4 %	0.9 %	2.7 %	3.8 %
2060s	1.5 %	4.0 %	2.2 %	5.6 %	0.8 %	1.3 %	2.1 %	3.4 %
CLIM	1.9 %	5.9 %	2.5 %	6.5 %	1.1 %	1.6 %	4.1 %	3.8 %
EMIS	2.2 %	3.3 %	2.2 %	3.7 %	1.4 %	1.4 %	1.5 %	2.5 %
BVOC	2.6 %	5.0 %	2.0 %	4.3 %	1.4 %	1.0 %	2.3 %	3.4 %
SURR	2.7 %	5.8 %	1.7 %	4.1 %	1.5 %	0.9 %	2.7 %	3.9 %

Figures

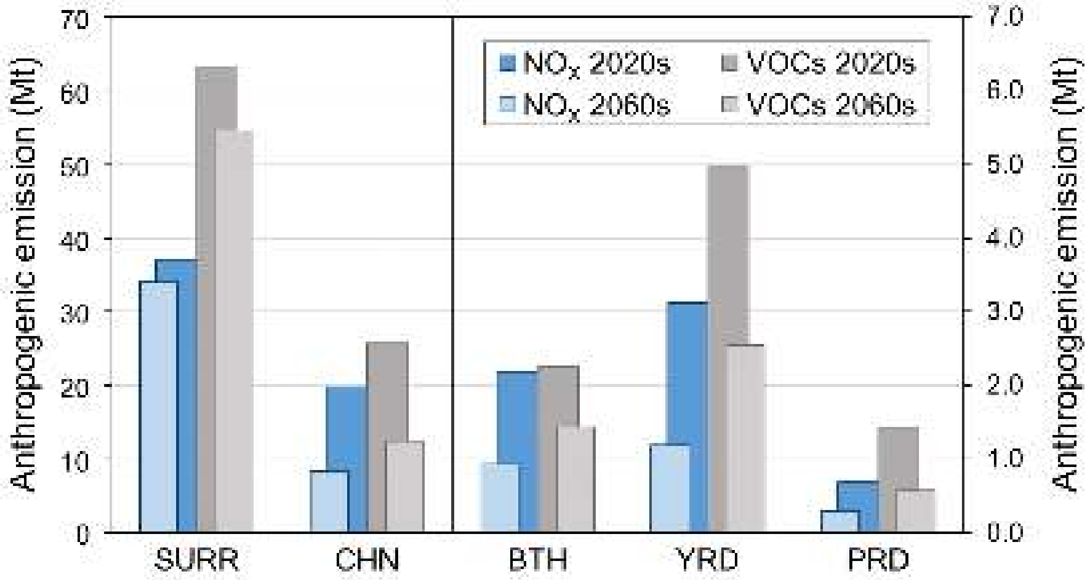


Figure S1 The NO_x and NMVOCs emissions in 2020 and 2060 for the surrounding areas within the modelling domain but excluding Chinese mainland (SURR), Chinese mainland (CHN), as well as the three regions. Data illustrated are obtained from MEIC for 2020 and DPEC for 2060 within Chinese mainland, and SSP dataset for the rest.

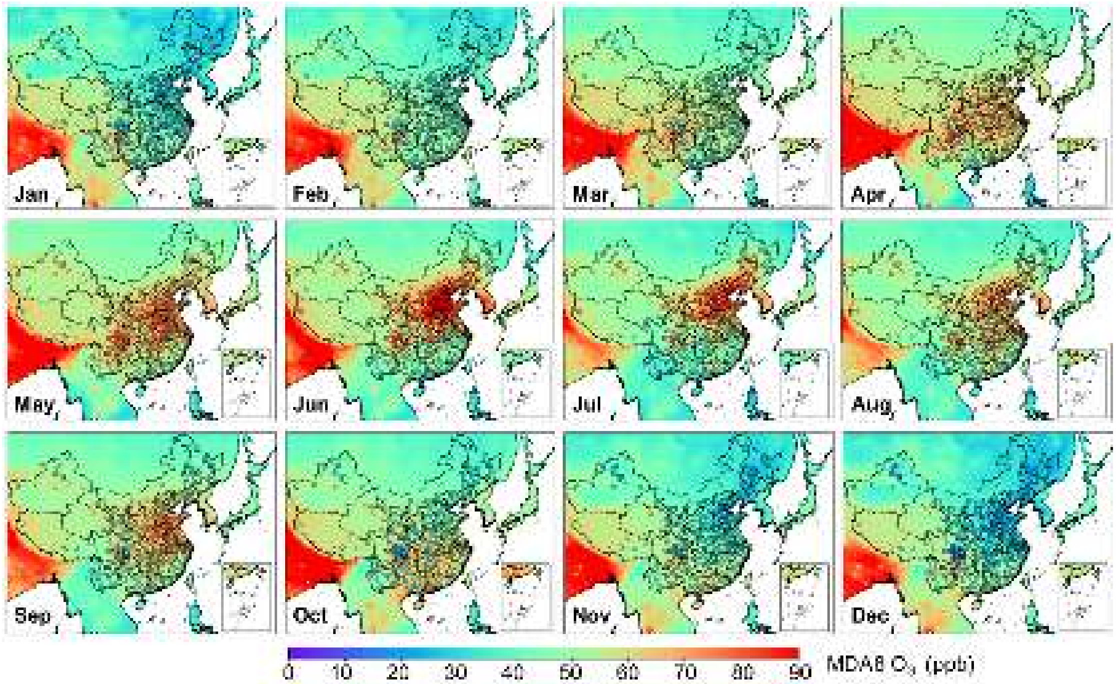


Figure S2 Spatial distribution of the observation (circles) and simulation (shaded) of monthly average MDA8 O_3 in 2020 across China.

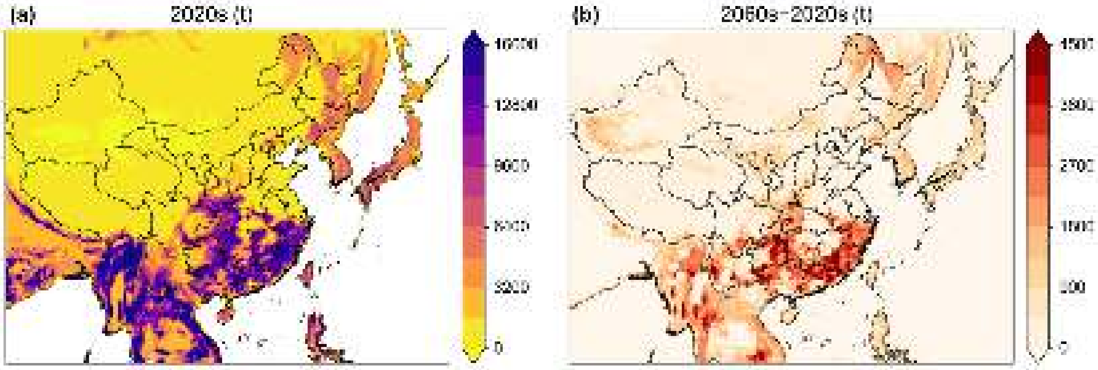
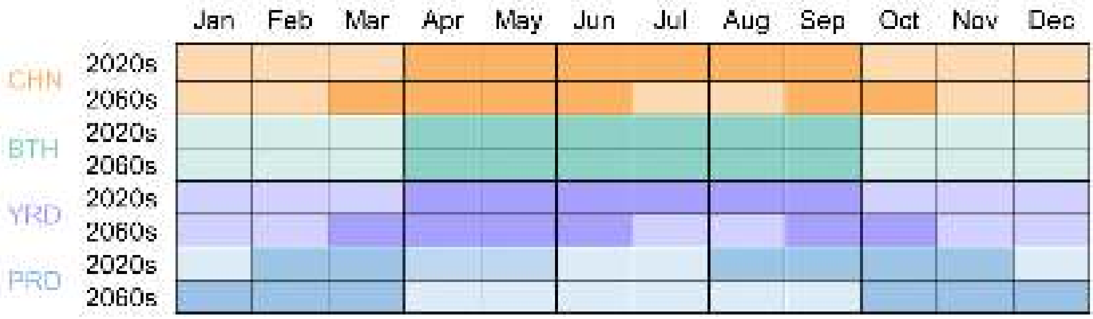


Figure S3 The spatial distribution of BVOCs emissions of the 2020s and the difference between the 2060s and 2020s.



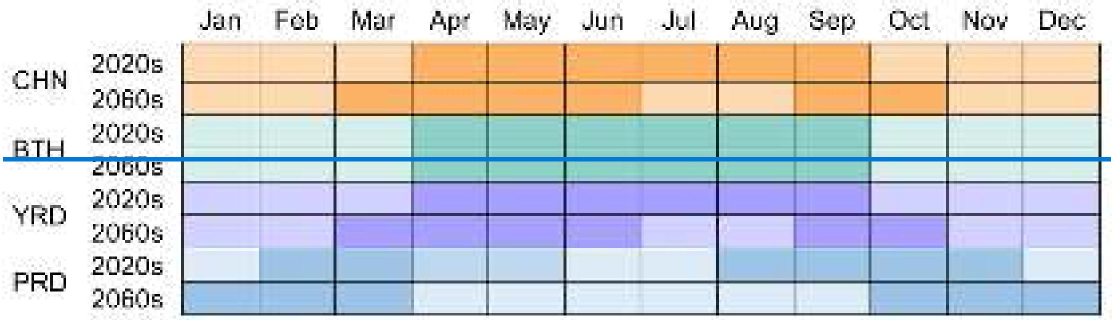


Figure S4 The distribution of O_3 seasons across China and the three regions, and the top six months with the highest levels of O_3 pollution are covered by darker color. Note that there is discrepancy between simulation and observation in PRD for 2020s. February and March are simulated as two of the six most polluted months, while the observation indicates April and May.

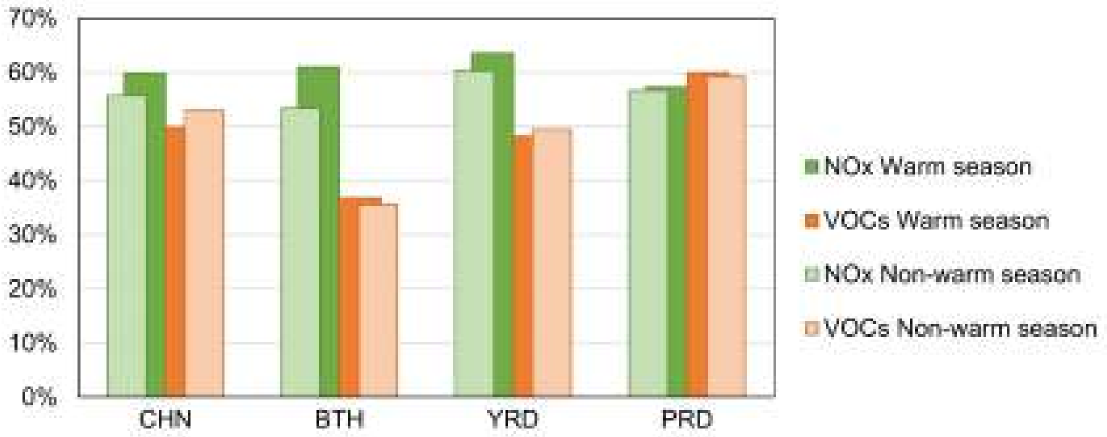


Figure S5 The emission reduction rates for the two precursors during 2020 and 2060 over China and the three regions.

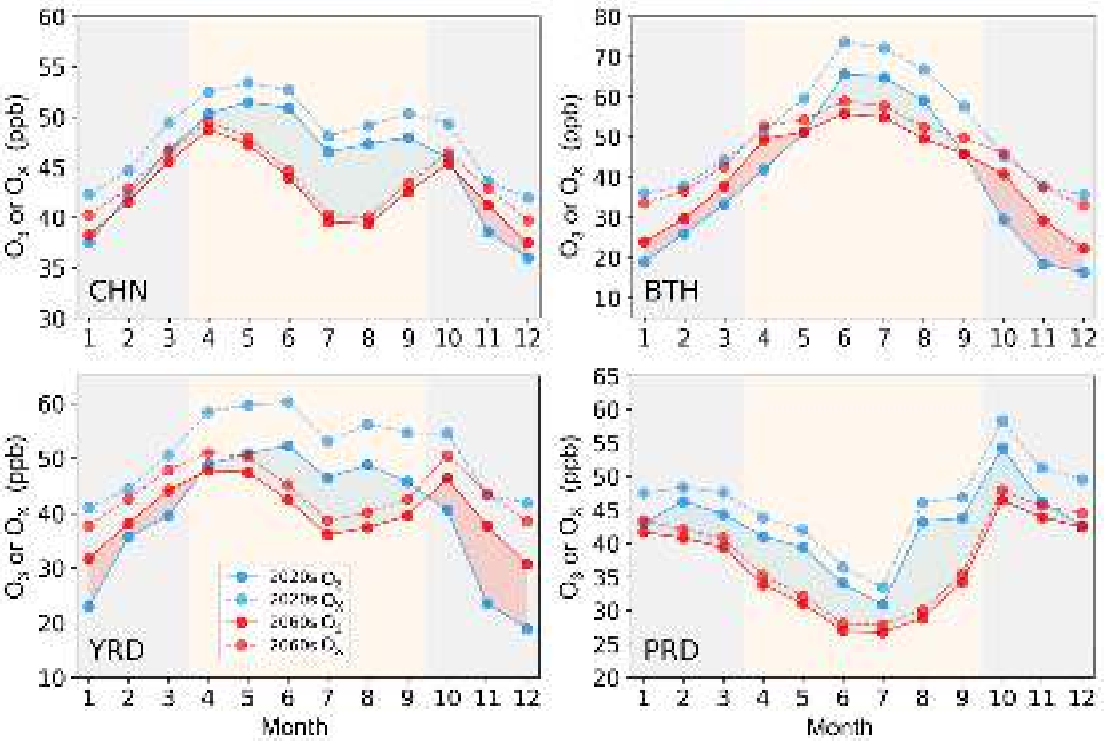


Figure S6 Simulation and projection of hourly O_3 and O_x in the 2020s and 2060s over China and the three regions.

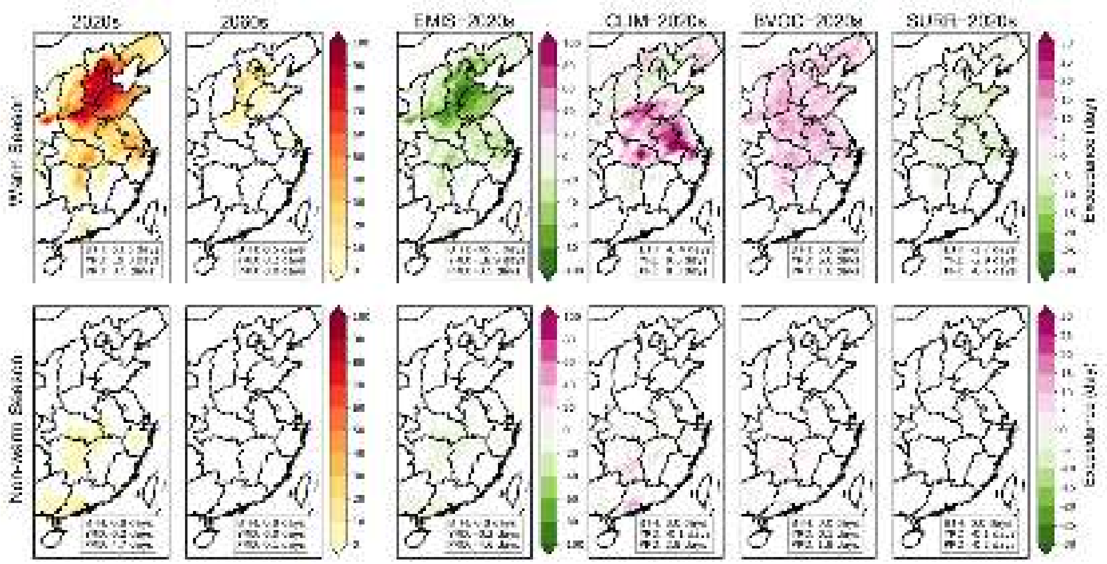


Figure S7 Projected seasonal O_3 exceedance over the east of China in the 2020s and 2060s, and the exceedance changes when the four factors at 2060s level.

References

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