Meteorological characteristics of extreme ozone pollution events in China and their future predictions

Yang Yang[1*], Yang Zhou[2], Hailong Wang[3], Mengyun Li[1], Huimin Li[1], Pinya Wang[1], Xu Yue[1], Ke Li[1], Jia Zhu[1], Hong Liao[1]

[1]Joint International Research Laboratory of Climate and Environment Change, Jiangsu Key Laboratory of Atmospheric Environment Monitoring and Pollution Control, Jiangsu Collaborative Innovation Center of Atmospheric Environment and Equipment Technology, School of Environmental Science and Engineering, Nanjing University of Information Science and Technology, Nanjing, Jiangsu, China

[2]Shanghai Baoshan Meteorology Bureau, Shanghai, China

[3]Atmospheric Sciences and Global Change Division, Pacific Northwest National Laboratory, Richland, WA, USA

*Correspondence to yang.yang@nuist.edu.cn

Abstract

Ozone (O_3) has become one of the most concerning air pollutants in China in recent decades. In this study, based on surface observations, reanalysis data, global atmospheric chemistry model simulations and multi-model future predictions, meteorological characteristics conducive to extreme O_3 pollution in various regions of China are investigated, and their historical changes and future trends are analyzed. During the most severe O_3 polluted months, the chemical production of O_3 is enhanced under the hot and dry conditions over the North China Plain (NCP) in June 2018 and Yangtze River Delta (YRD) in July 2017, while the regional transport is the main reason causing the severe O_3 pollution over Sichuan Basin (SCB) in July 2015 and Pearl River Delta (PRD) in September 2019. Over the last four decades, the frequencies of high temperature and low relative humidity conditions increased in 2000-2019 relative to 1980-1999, indicating that O_3 pollution in both NCP and YRD became more frequent under the historical climate change. In SCB and PRD, the occurrence of atmospheric circulation patterns similar to those during the most polluted months increased, together with the more frequent hot and dry conditions, contributing to the increases in severe O_3 pollution in SCB and PRD during 1980–2019. In the future (by 2100), the frequencies of months with anomalous high temperature show stronger increasing trends in the high forcing scenario (SSP5-8.5) compared to the sustainable scenario (SSP1-2.6) in China. It suggests that high anthropogenic forcing will not only lead to slow economic growth and climate warming, but also likely result in environmental pollution issues.

1. Introduction

Tropospheric ozone (O_3), one major air pollutant, is formed in photochemical reactions of nitrogen oxides (NO_x) and volatile organic compounds (VOCs) when exposed to sunlight (Finlayson-Pitts and Pitts, 1997; Silman, 1999). Enhanced O_3 pollution harms ecosystems and human health (Fleming et al., 2018; Maji et al., 2019) by reducing crop yields (Ainsworth et al., 2012; Mills et al., 2018) and aggravating cardiopulmonary disease (Ebi and McGregor, 2008; Liu et al. 2018). In recent years, near-surface ozone concentrations in many regions of China have been increasing considerably (Verstraeten et al., 2015; Cheng et al., 2019; Zhang et al., 2020, Li et al., 2019; Lu et al., 2018; Silver et al., 2018; Yin et al., 2019, Lu et al., 2020). Lu et al. (2020) revealed that the daily maximum of 8-h average O_3 concentration (MDA8-O_3) in China increased by 2.4 ppb per year (5.0% relative to the average) during April–September over 2013–2019.

In addition to emissions, O_3 concentrations are influenced by meteorological factors such as temperature, relative humidity, solar radiation, and winds (Mott et al., 2005; Fu and Tian, 2019; Gong and Liao, 2019; Li et al., 2019, 2020; Le et al., 2020; Zhao et al., 2020). Typically, strong solar radiation, high surface air temperatures, and low relative humidity are conducive to photochemical production of O_3, causing a raise of O_3 concentration (Peterson and Flowers, 1977; Xu, et al., 2011; Coates et al., 2016; Li et al., 2020; Dang et al., 2021). Wind speed is negatively correlated with surface O_3 because low wind speed facilitates the accumulation of O_3 upon production (Zhang et al., 2015; Wang et al., 2017; Liu and Wang, 2020). Han et al. (2020) explored the impacts of various meteorological factors on the daily variation of summer surface O_3 in eastern China based on a multiple linear regression method and suggested that relative humidity is the primary factor affecting O_3 concentration in central and south parts of eastern China, while temperature is the most important factor governing O_3 concentration in north of eastern China. Gong and Liao (2019) reported that the meteorological characteristics of O_3 pollution events in North China during 2014–2017 were the high daily maximum temperature, low relative humidity, abnormal southerly winds and high pressure at 500 hPa. These findings emphasize that meteorological factors play a crucial role in regulating O_3 pollution in China.

Atmospheric circulation patterns affect O_3 concentrations over China through changing meteorological factors (Yang et al, 2014, 2022; Zhao and Wang, 2017; Shu et al., 2019; Dong et al., 2020; Zhou et al., 2022). Zhao and Wang (2017) examined the influence of the Western Pacific Subtropical high (WPSH) on O_3 over eastern China based on observations and reanalysis data from 2014 to 2016. They found that stronger WPSH enhanced the moisture transport to southern China, which was detrimental to the photochemical reaction of O_3, leading to a decrease in surface O_3 concentration in southern China, whereas O_3 concentrations in northern China increased under the stronger WPSH related to the dry and hot conditions favoring O_3 production. On the basis of observational O_3 data and ERA5 reanalysis data during 2014–2018, Dong et al. (2019) analyzed the impact of synoptic patterns on summertime O_3 pollution in the North China Plain and revealed that the most severe O_3 pollution weather pattern is associated with anomalous southwesterly winds, which carry dry, warm air from inland southern China to the North China Plain and favor the chemical production of O_3. Zhou et al. (2022) explored the impacts of Asian summer monsoon on the interannual variation of O_3 concentrations based on surface measurements and GEOS-Chem model simulations. They showed that the East Asian summer monsoon strength was positively correlated with O_3 concentration in south-central China and South Asian summer monsoon has complex effects on O_3 pollution in China, mainly through changing transboundary transport related to large-scale circulations.

As mentioned above, many previous studies have examined the meteorological characteristics of O_3 pollution in China. However, they focused on O_3 pollution over limited regions in China in each study (e.g., the North China Plain, southern China). These studies only examined the meteorological characteristics of O_3 pollution in a short time period due to the lack of observational data and did not consider the historical and future trends of these meteorological factors. In this study, the meteorological characteristics conducive to the most severe O_3 pollution in several polluted areas of China, including the North China Plain (NCP), Yangtze River Delta (YRD), Sichuan Basin (SCB), and Pearl River Delta (PRD), are respectively investigated based on the observed surface O_3 concentrations, reanalysis data, and GEOS-Chem model simulations. Besides, the contributions from various chemical and physical processes inducing regional O_3 pollution are quantified using an integrated process rate (IPR) analysis method. The historical changes in these meteorological factors favoring the most severe O_3 pollution over 1980–2019 are provided. Moreover, variations in future meteorological patterns during 2021–2100 leading to severe O_3 pollution in China are presented under the sustainable and high forcing scenarios according to the multi-model data from the Coupled Model Intercomparison Project Phase 6 (CMIP6).

2. Methods

2.1 Surface ozone observations and meteorological reanalysis

Hourly surface O_3 concentrations are obtained from the Ministry of Ecology and Environment (MEE) of China. The observational network was established in 2013 with 450 monitoring sites and increased to 1,500 monitoring sites by 2019, covering about 360 cities in China. MDA8-O_3 are calculated based on hourly O_3 concentrations from April-September during 2013 to 2020. In this study, O_3 pollution days are defined as the days when MDA8-O_3 exceeds 160 μg m^{-3} according to the China National Ambient Air Quality Standard (GB3095-2012).

The meteorological fields are taken from the European Centre for Medium-Range Weather Forecasts (ECMWF) ERA5 monthly reanalysis dataset during 1980–2020, with a horizontal resolution of 0.25° × 0.25°. To explore the meteorological characteristics that are conducive to O_3 pollution, sea level pressure (SLP), geopotential height (GPH) at 500 hPa, wind fields at 850 hPa and 500 hPa, temperature at 2m (T2m) and surface relative humidity (RH) are adopted, which can have significant impacts on O_3 variations in China (Jiang et al., 2020; Dong et al., 2020; Le et al., 2020).

2.2 GEOS-Chem model simulations

O_3 concentrations and the related chemical and physical processes causing O_3 variations over 1981–2020 are simulated in the global atmospheric chemistry model GEOS-Chem (version V12.9.3), driven by the Modern-Era Retrospective analysis for Research and Application, Version 2 (MERRA-2). Simulations are performed on 47 vertical layers from surface to 0.01 hPa, and a horizontal grid of 2° latitude × 2.5° longitude. GEOS-Chem model incorporates a fully coupled O_3-NO_x-hydrocarbon-aerosol chemical mechanism (Pye et al., 2009; Mao et al., 2013; Sherwen et al., 2016). Boundary-layer mixing uses a non-local scheme

(Lin and McElroy, 2010), and stratospheric O_3 chemistry employs the linearized O_3 parameterization (LINOZ) (McLinden et al., 2000).

Global anthropogenic emissions driving the simulations are from the Community Emissions Data System (CEDS, Hoesly et al., 2018) and biomass burning emissions are from the Global Fire Emissions Database, Edition 4 (GFED4, Van der Werf et al., 2017). VOCs emissions from biogenic sources are provided offline by the Model of Emissions of Gases and Aerosols from Nature version 2.1 (MEGAN V2.1, Guenther et al., 2012). Lightning and soil emissions are specified in the model (Hudman et al., 2012; Ott et al., 2010). Anthropogenic emissions in China are updated with the Multi-resolution Emission Inventory (MEIC), a localized emission dataset for China. Anthropogenic, biomass burning, biogenic and other natural emissions are kept at 2017 levels during the simulations, so as to eliminate the influence of emission changes on the interannual variation and trends of O_3. Simulated O_3 distributions with the same configuration in GEOS-Chem have been extensively evaluated in many studies, and the model has been reported to capture O_3 concentrations well in China (e.g., Li et al., 2019; Lu et al., 2019; Ni et al., 2018).

2.3 CMIP6 multi-model simulations

The multi-model simulations from historical and the Scenario Model Intercomparison Project (ScenarioMIP) in CMIP6 are used to analyze the historical variations and future trends of meteorological conditions conducive to the most severe O_3 pollution. Two different future scenarios of the Shared Socioeconomic Path (SSPs) are applied, including the sustainable scenario (SSP1-2.6) and the high forcing scenario (SSP5-8.5). Totally simulations from 13 models (ACCESS-CM2, ACCESS-ESM1-5, CAS-ESM2-0, CMCC-CM2-SR5, CMCC-ESM2, FGOALS-f3-L, FGOALS-g3, GFDL-ESM4, INM-CM4-8, INM-CM5-0, IPSL-CM6A-LR, MPI-ESM1-2-HR, MPI-ESM1-2-LR) are analyzed in this study.

3. Results

3.1 Meteorological characteristics conductive to regional ozone pollution

To investigate the relationship between meteorological conditions and regional O_3 pollution in China, the frequencies of O_3 pollution days from April to October during 2013–2020 are calculated for Beijing, Shanghai, Chengdu and Guangzhou, representing the typical four polluted regions in China (i.e., NCP, YRD, SCB and PRD) (Figure 1). Observational data show the highest frequencies of O_3 pollution days in June 2018, July 2017 and September 2019 in Beijing, Shanghai and Guangzhou, with pollution days up to 22, 20 and 19 days per month, respectively. The top three highest frequencies of O_3 pollution days in Chengdu are in July 2016, July 2015 and July 2018 (16, 15 and 15 days per month, respectively). Variations in O_3 concentration in the real world are driven by changes in both meteorological factors and emissions. With fixed emissions, the positive anomalies of near-surface O_3 concentrations over NCP, YRD and PRD during their most polluted months can also be reproduced by the GEOS-Chem model (Figure 2), suggesting that the O_3 pollutions during the most polluted months over NCP, YRD and PRD are likely attributable to the anomalies of meteorological conditions. In the top three O_3 polluted months in Chengdu, only in July 2015 the higher concentrations than the long-term averages can be captured by the simulations with fixed emissions. Therefore, in this study, we focus on the meteorological characteristics in June 2018, July 2017, July 2015 and September 2019, that were conducive to the most severe O_3 pollution over NCP, YRD, SCB and PRD, respectively.

When O_3 pollution was the most severe over NCP in June 2018, an anomalous high pressure occurred at 500 hPa over NCP (Fig. 3b), relative to the 40-year climatological averages from 1980 to 2019, leading to positive T2m anomalies near the surface (Fig. 3c). Anomalous lows located over northeastern China and northwestern Pacific (Fig. 3a) and the associated anomalous northerly winds prevent the moisture moving from the ocean to NCP, causing negative RH anomalies over NCP (Fig. 3d). The meteorological conditions with the high T2m and low RH are favorable for the photochemical production of O_3. When the most severe O_3 pollution occurred in July 2017, YRD was dominated by anomalous high pressure in the lower and middle troposphere (Figs. 4a and 4b). Under the control of high pressure, the meteorological conditions (e.g., high T2m and low RH) enhance the photochemical production of O_3 (Figs. 4c and 4d). In the O_3 pollution event of SCB in July 2015, the negative T2m anomaly is not conducive to the O_3 production (Fig. 5c), although the RH was low (Fig. 5d). Meanwhile, the anomalous low over eastern China and northwestern Pacific in the middle troposphere favors regional O_3 transport from the polluted source region over eastern China to SCB (Fig. 5b) and the anomalous high over central-western China is conducive to the vertical transport of upper tropospheric O_3 down to the lower troposphere (Fig. 5a). For the PRD in September 2019, the anomalous high covering almost the entire China along with the anomalous low over East China Sea generates northerly wind anomalies in the lower troposphere over eastern China, which tend to transport polluted air from northern China and weaken the inflow of oceanic clean air (Fig. 6). The temperature increase is much more significant in the upwind regions as compared to PRD, suggesting that the strong regional transport could be the primary reason causing this severe O_3 pollution event of PRD.

3.2 Physical and chemical mechanisms leading to regional ozone pollution

To further explore the mechanisms of meteorological changes leading to the severe O_3 pollution over the four typical polluted regions in China, contributions of individual chemical and physical processes to O_3 variations are quantified based on the IPR analysis from GEOS-Chem simulations and summarized in Table 1.

Consistent with the meteorological anomalies analyzed above, high temperature and low RH meteorological conditions in NCP are conducive to the photochemical production of O_3. During the polluted month over NCP, the chemical production of tropospheric O_3 is higher than the long-term average by 2.36 Gg day^{-1}, while the horizontal transport also contributes to the increase in O_3 mass by 1.58 Gg day^{-1} (Table 1). Due to the enhanced northwesterly winds, the import of O_3 mass from the north and west of NCP was increased by 1.80 and 0.62 Tg, respectively (Table 2). In YRD, the chemical production (2.38 Gg day^{-1}) is also the dominant process that drives the O_3 concentration increase during the most severe polluted month, associated with the warm and dry conditions. Therefore, the anomalous chemical production is the major process that induced O_3 pollution in NCP and YRD during the most severe polluted months.

Different from NCP and YRD, horizontal transport is the main process that caused O_3 pollution in SCB and PRD during the most severe months. It contributes to the rate of increase in O_3 mass by 5.10 and 6.67 Gg day^{-1}, respectively, over SCB and PRD, while other processes tend to decrease the O_3 mass (Table 1). Due to the anomalous northerly winds over SCB, more O_3 is transported into SCB from north (by 4.02 Tg), and the anomalous northeasterly winds enhance the O_3 transport from the north and east of PRD by 1.97 and 1.09 Tg, respectively, leading to the increase in O_3 concentrations over SCB and PRD during the most severe months relative to the climatological averages (Table 2). Note that, the chemical production of tropospheric O_3 decreased in SCB and PRD during the most severe months. It could have been biased by the relatively coarse model resolution in this study (2° latitude × 2.5° longitude), since that the SCB and PRD for calculating the chemical and physical processes only cover limited grid boxes. Further studies should be performed using a model with finer resolution or a nested simulation method.

3.3 Historical and future changes in the meteorological conditions

O_3 pollution has deteriorated in China during recent decades, which could be related to the changes in meteorological conditions. Time series of T2m and RH anomalies in the polluted months during the 1980–2019 and frequencies of high T2m and low RH months during 1980–1999 and 2000–2019 over the four polluted regions in China based on ERA5 reanalysis data are shown in Figure 7. Due to climate change, both the high temperature and low RH conditions in NCP, YRD, SCB and PRD all increased during the past four decades (2000-2019 versus 1980-1999). Based on the analysis showing that chemical production is the dominant process of the most severe O_3 pollution in NCP and YRD, the increases in the frequency of high temperature and low RH indicate that severe O_3 pollution in both NCP and YRD has become more frequent under the historical climate change. In SCB and PRD, the most severe O_3 pollution is more related to changes in regional transport. Similar to the analyzing method used in previous studies (Li et al., 2018; Yang et al., 2021), the SLP and 500 hPa GPH over East Asia and Western Pacific in the same month of each year similar to those during the most severe months in both SCB and PRD have increased (2000-2019 versus 1980-1999) (Figure 8), together with the more frequent hot and dry conditions (Figure 7), leading to the increases in severe O_3 pollution in SCB and PRD during 1980–2019.

Many studies have reported that future climate change will have significant influences on O_3 pollution in China through changing meteorological factors (e.g., Li et al., 2023; Wang et al., 2022). Here, the frequencies of extreme months with high T2m and low RH and the frequencies of extreme months with SLP and 500 hPa GPH that have moderate to high correlation to those in the most polluted months in the four regions of China, under the sustainable (SSP1-2.6) and high forcing (SSP5-8.5) scenarios during 2021–2100 from CMIP6 multi-model results, are presented in Figures 9 and 10, respectively. Unlike the historical changes in the meteorological conditions that caused the severe O_3 pollution through chemical production and regional transport, future variations in meteorological conditions conducive to the severe O_3 pollution are more related to the global warming process that enhances the O_3 production in China. The frequencies of months with anomalous high temperature show obvious upward trends in both SSP1-2.6 and SSP5-8.5 scenarios over the four regions, and the increasing trends in SSP5-8.5 are much more significant than in SSP1-2.6. Frequencies of low RH months show downward trends in NCP, YRD and SCB, especially under SSP5-8.5, while there is an upward trend in PRD. Note that the trends in frequencies of low RH months are much less significant than in high temperature months. The frequencies of extreme months with SLP and 500 hPa GPH that are similar to those in the most severe O_3 pollution months in the four regions do not show significant trends in the SSPs. Hence, the future climate change may aggregate O_3 pollution in China by enhancing the chemical production related to temperature increases. The O_3 pollution exacerbation is projected to be less significant in the sustainable scenario due to the moderate temperature increase than in the high forcing scenario, suggesting that the sustainable scenario is the optimal path to retaining clean air in China. High anthropogenic radiative forcing will not only lead to slow economic growth and climate warming, but also result in the environmental pollution.

4. Conclusions and Discussions

O_3 pollution harms ecosystems and human health. In recent years, near-surface O_3 concentrations in many regions of China have been increasing considerably. Base on observational O_3 data, ERA5 reanalysis data and GEOS-Chem model simulations, meteorological characteristics conducive to extreme O_3 pollution in different regions of China are investigated in this study. Contributions from various chemical and physical processes inducing O_3 pollution are quantified using the IPR analysis method. Furthermore, historical changes and future trends of meteorological conditions leading to severe O_3 pollution in China are explored based on the meteorological reanalysis and CMIP6 multi-model future predictions, which is of great implication for the mitigation and prevention of O_3 pollution over China.

In this study, June 2018, July 2017, July 2015 and September 2019 are identified as the most severe O_3 pollution months influenced by meteorological factors over NCP, YRD, SCB and PRD, respectively. Severe O_3 pollution in June 2018 over NCP and in July 2017 over YRD is mainly due to enhanced chemical production related to hot and dry conditions. The chemical production of O_3 in the most severe months over NCP and YRD are 2.36 Gg day^{-1} and 2.38 Gg day^{-1}, respectively, higher than the climatological averages. Different from NCP and YRD, regional transport is the main process leading to the high O_3 concentration in SCB and PRD during the respective severely polluted months, which contributes to the rate of increase in O_3

mass by 5.10 and 6.67 Gg day^{-1}, respectively, over SCB and PRD. During the most severe months, related to large-scale circulation patterns, anomalous northerly winds transport more O_3 into SCB from north, and anomalous northeasterly winds enhance the O_3 transport from the north and east into PRD.

Over the last four decades (2000-2019 versus 1980-1999), the frequencies of high temperature and low RH increased, indicating that O_3 pollution in both NCP and YRD has become more frequent under the historical climate change. In SCB and PRD, the occurrence of atmospheric circulation patterns similar to those during the most polluted months in both SCB and PRD has increased, together with the more frequent hot and dry conditions, leading to the increases in severe O_3 pollution in SCB and PRD during 1980–2019. In the future (by 2100), the frequencies of months with anomalous high temperature show obvious upward trends in both sustainable (SSP1-2.6) and high forcing (SSP5-8.5) scenarios over the four regions, and the increasing trends in SSP5-8.5 are much more significant than in SSP1-2.6. This suggests that high anthropogenic radiative forcing will not only lead to slow economic growth and climate warming, but also likely result in environmental pollution issues. The sustainable scenario is the optimal path to retaining clean air in China.

There are some limitations and uncertainties in this work that can be further addressed in future studies. For example, the model only captures the high O_3 concentrations in July 2015 in Chengdu among its top three polluted months. It is probably because the emissions are kept at 2017 levels during the simulations. The high O_3 anomalies in July 2016 and July 2018 are more likely influenced by the interannual changes in local precursor emissions in the background of country-level increases in O_3 concentration in recent years. However, we also can not rule out the possible inaccuracy in the model simulations to interpret severe O_3 pollution events in the SCB, which deserves further investigation with multi-model simulations. In addition, this study focuses on the most extreme O_3 pollution in several polluted areas of China. However, many other meteorological conditions can also cause O_3 pollution, although they may not be as extreme as the cases analyzed in this study, which requires comprehensive analysis for individual regions in future studies. Although the historical changes in the meteorological patterns causing severe O_3 pollution are in accordance with the elevated O_3 levels in China in the recent decade, the quantitative analysis of meteorological impacts needs full consideration of factors leading to O_3 pollution, including changes in anthropogenic and natural emissions of its precursors, O_3 chemical regime, other meteorological factors conducive to O_3 pollution, and stratosphere-troposphere exchange.

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Code and data availability. The GEOS-Chem model is available at https://zenodo.org/record/3974569#.YTD81NMzagR (last access: June 2023). O_3 observations over China can be obtained at https://quotsoft.net/air (last access: June 2023). ERA5 reanalysis data can be downloaded at https://www.ecmwf.int/en/forecasts/datasets/reanalysis-datasets/era5 (last access: June 2023). The multi-model simulations of the Coupled Model Intercomparison Project Phase 6 (CMIP6) are from https://esgf-node.llnl.gov/search/cmip6/ (last access: June 2023).

Author contribution. YY designed the research; YY and YZ performed simulations and analyzed the data. All authors including HW, LH, PW, and HL discussed the results and wrote the paper.

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

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Table 1. Anomalies in net rate of changes in tropospheric O_3 mass (Gg day^{-1}) over NCP (115°–120°E, 38°–44°N), YRD (120°–125°E, 28°–32°N), SCB (102.5°–105°E, 30°–32°N) and PRD (110°–115°E, 22°–26°N) due to physical and chemical processes in the most polluted months (June 2018, July 2017, July 2015 and September 2019, respectively) relative to the same months averaged during 1981–2019.

	Beijing	Shanghai	Chengdu	Guangzhou
Chemical reaction	2.36	2.38	-2.80	-1.52
Horizontal transport	1.58	–1.18	5.10	6.67
Diffusion and dry deposition	0.29	0.24	-0.73	-0.93

Table 2. Horizontal mass transport (Tg) of O_3 from the surface to 500 hPa over NCP (115°–120°E, 38°–44°N), YRD (120°–125°E, 28°–32°N), SCB (102.5°–105°E, 30°–32°N) and PRD (110°–115°E, 22°–26°N) areas in the severe polluted months (June 2018, July 2017, July 2015 and September 2019, respectively) and averaged over the same months of a year during 1981–2019, as well as their differences. Positive values indicate incoming fluxes and negative values indicate outgoing fluxes.

	Polluted month	Average	Anomalies
		NCP	
North	4.43	2.62	1.80
South	-2.22	-1.42	-0.81
East	-12.30	-11.31	-0.99
West	11.83	11.20	0.62
		YPD	
North	-4.13	-3.88	-0.25
South	3.58	3.20	0.37
East	-2.05	-3.90	1.85
West	2.03	4.04	-2.01
		SCB	
North	4.15	0.13	4.02
South	-2.30	0.48	-2.78
East	-1.10	-1.15	0.05
West	1.73	1.84	-0.11
		PRD	
North	2.70	0.72	1.97
South	-2.87	-0.90	-1.96
East	2.24	1.15	1.09
West	-2.32	-1.55	-0.76

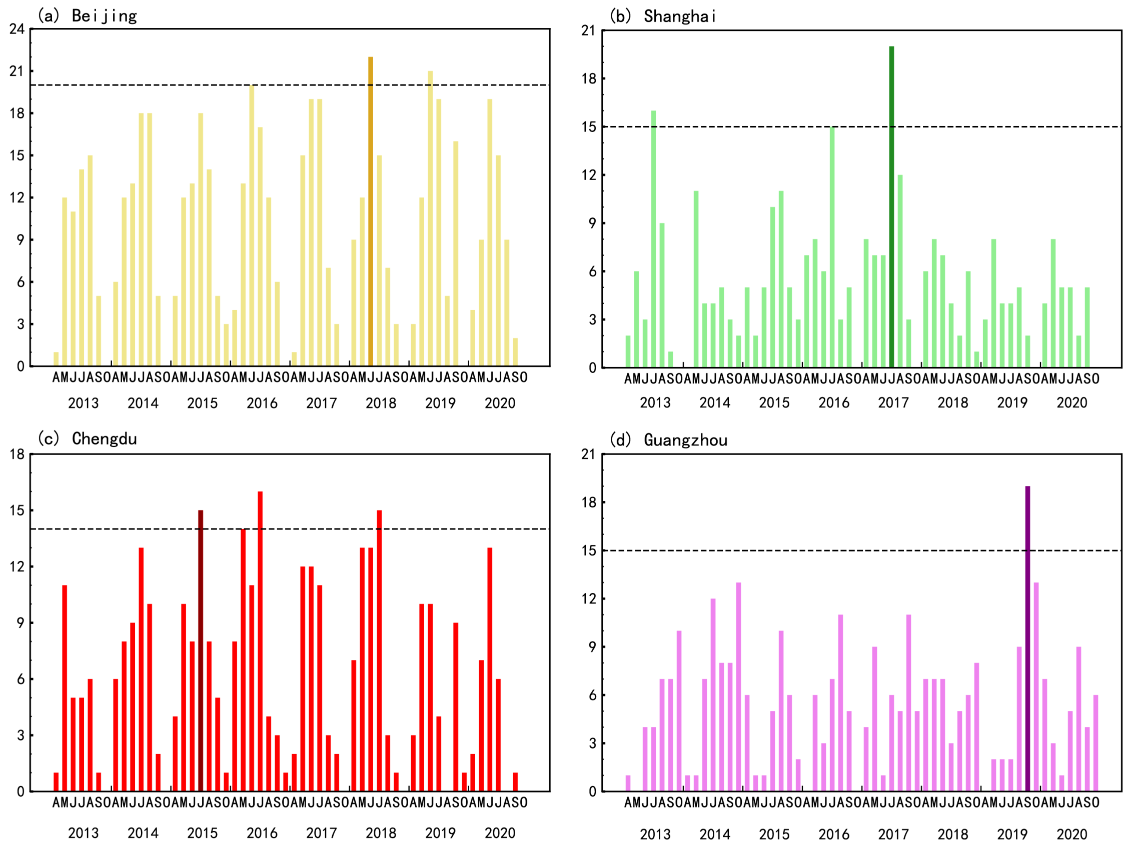

Figure 1. Time series of frequencies of severe O_3 pollution days (defined by daily maximum of 8-h average ozone (MDA8-O_3) concentration greater than 160 μg m^{-3}) in Beijing, Shanghai, Chengdu and Guangzhou (a–d) from April to October during 2013–2020. The dark-colored bars represent the most severe month (second most for Chengdu) that has the highest frequency of O_3 pollution days for the individual cities.

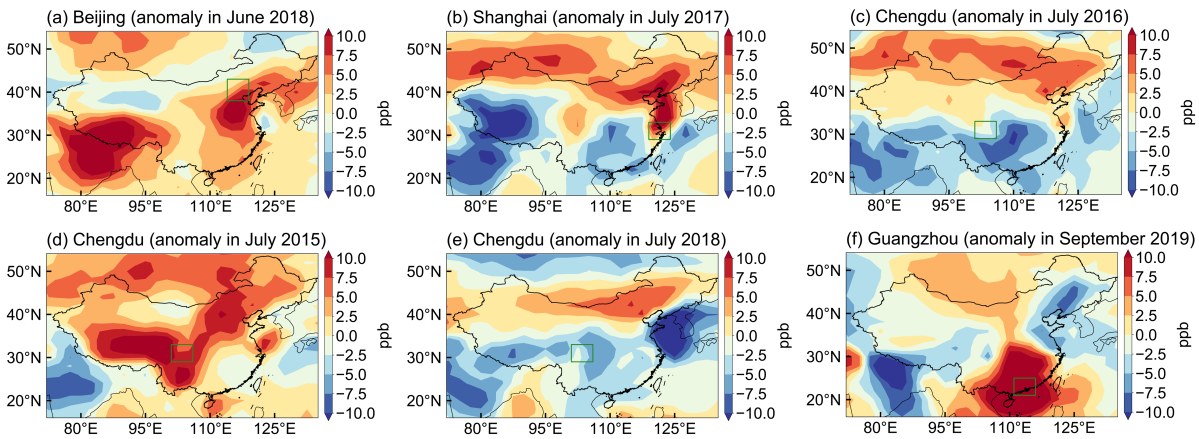

Figure 2. Spatial distribution of monthly O_3 concentration anomalies (part per billion, ppb) in June 2018 (a), July 2017 (b), July 2016 (c), July 2015 (d), July 2018 (e) and September 2019 (f) relative to 40-year (1980–2019) monthly average for June (a), July (b, c, d, e) and September (f), simulated in the GEOS-Chem model. The green boxes mark NCP (a), YRD (b), SCB (c, d, e) and PRD (f).

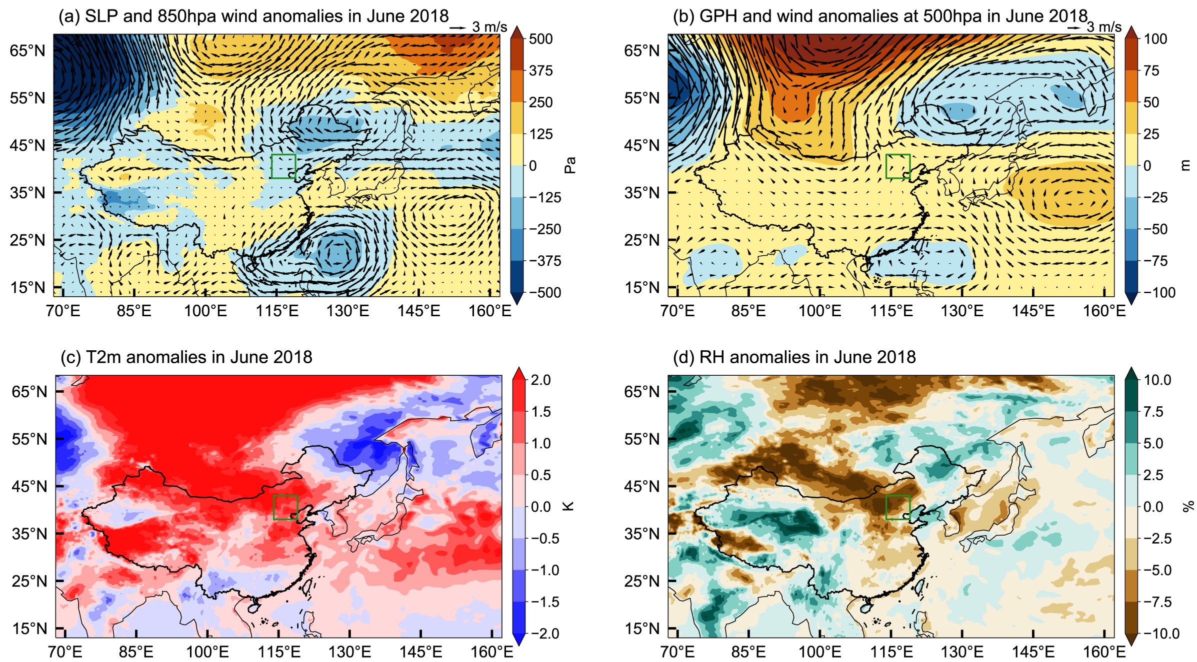

Figure 3. Anomalies in sea level pressure (SLP, Pa, shaded) and 850 hPa winds (m s^{-1}, vector) (a), geopotential height (GPH, m, shaded) and winds at 500 hPa (m s^{-1}, vector) (b), 2-meter air temperature (T2m, K) (c) and surface relative humidity (RH, %) (d) in June 2018 relative to the 40-year (1980–2019) monthly average for June. The green boxes mark NCP.

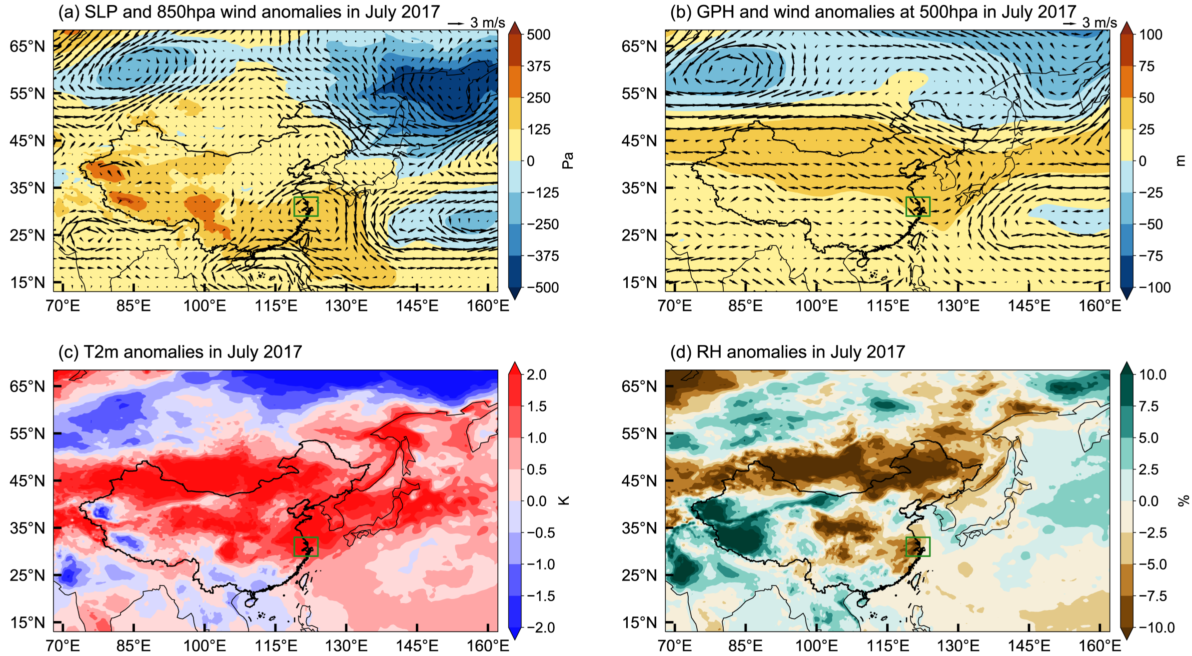

Figure 4. Same as Figure 3 but for the monthly anomalies in July 2017. The green boxes mark YRD.

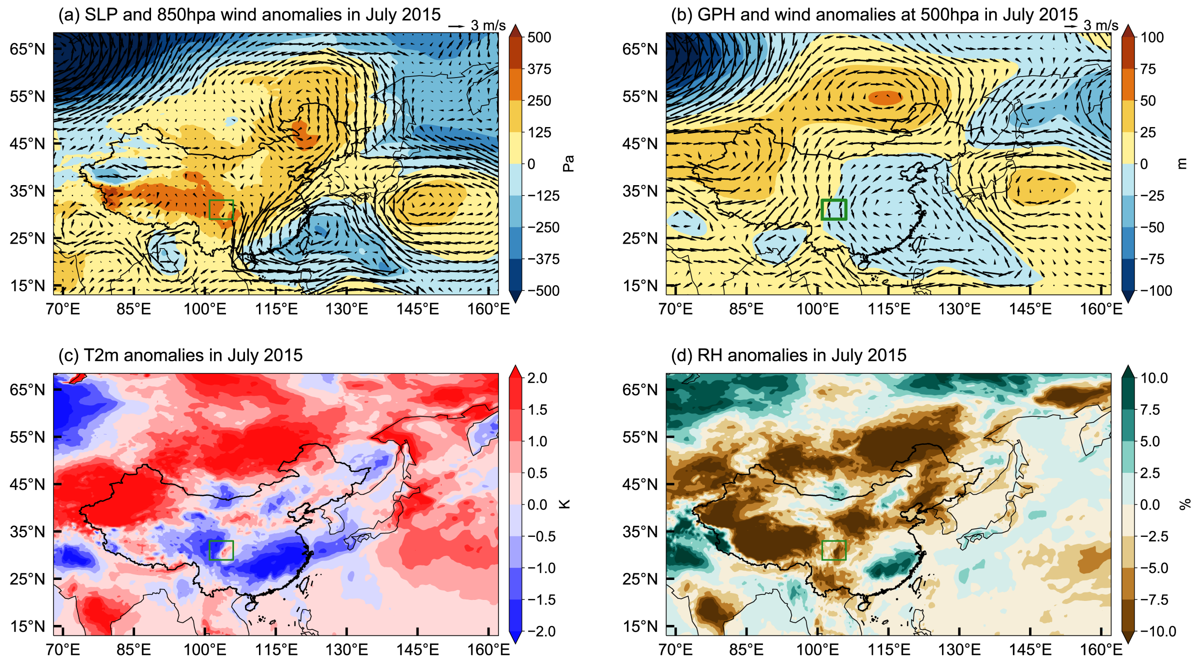

Figure 5. Same as Figure 3 but for the monthly anomalies in July 2015. The green boxes mark SCB.

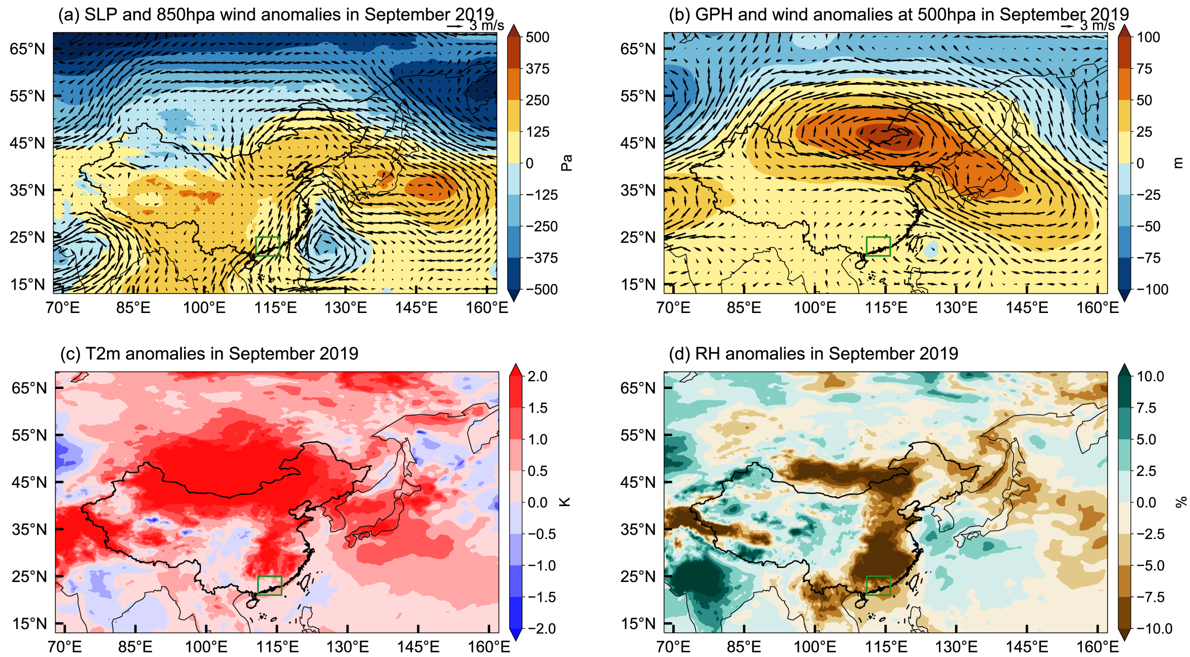

Figure 6. Same as Figure 3 but for the monthly anomalies in September 2019. The green boxes mark PRD.

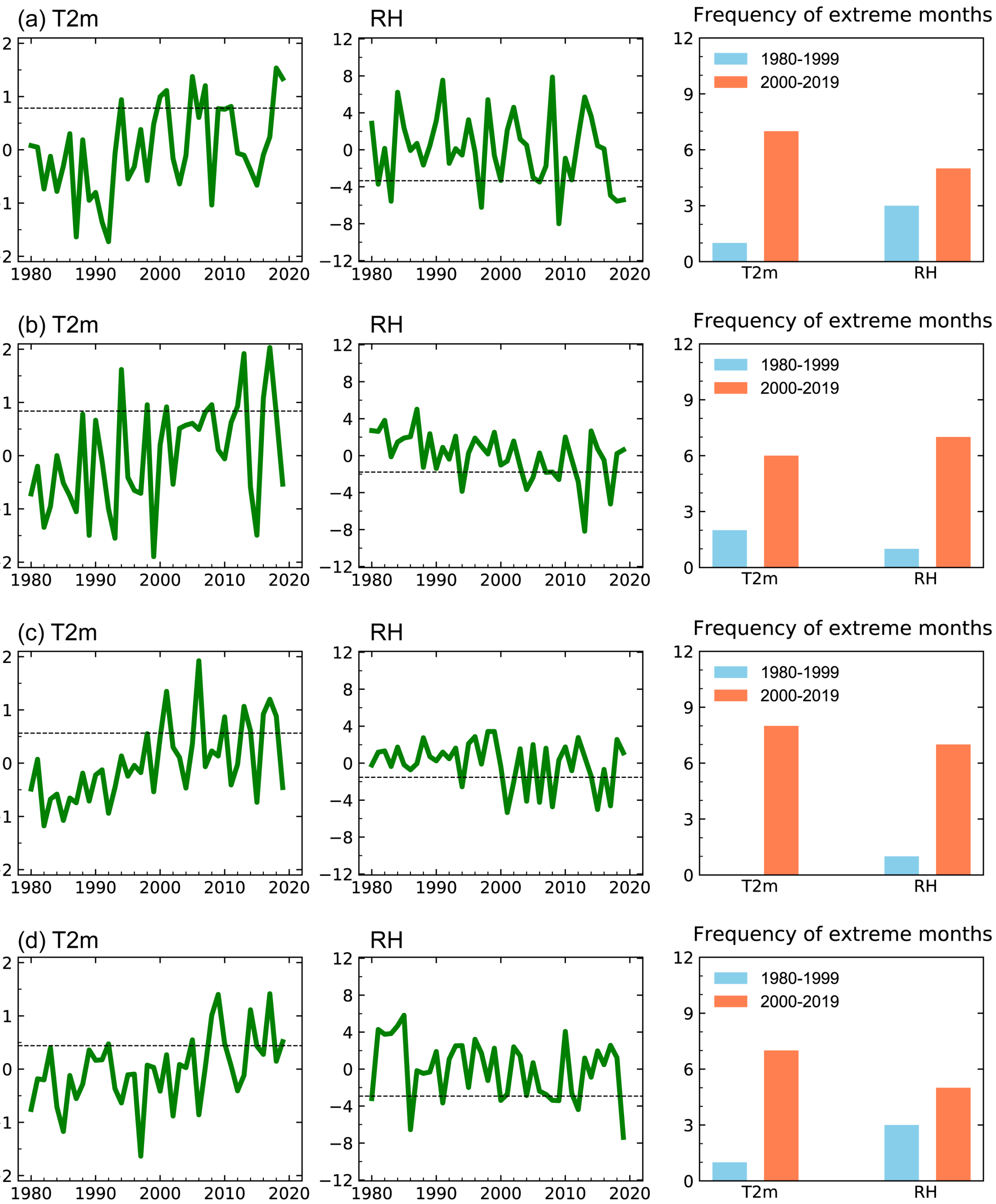

Figure 7. Time series of anomalies of T2m (K, left) and surface RH (%, middle) over (a) NCP (115°–120°E, 38°–44°N), (b) YRD (120°–125°E, 28°–32°N), (c) SCB (102.5°–105°E, 30°–32°N) and (d) PRD (110°–115°E, 22°–26°N) in the most polluted months during 1980–2019. The dotted lines mark the 80th percentile of the distributions for T2m and 20th percentile for RH. The bar charts (right) represent the frequency of T2m above the 80th percentile and RH anomalies below the 20th percentile during 1980–1999 (blue) and 2000–2019 (orange).

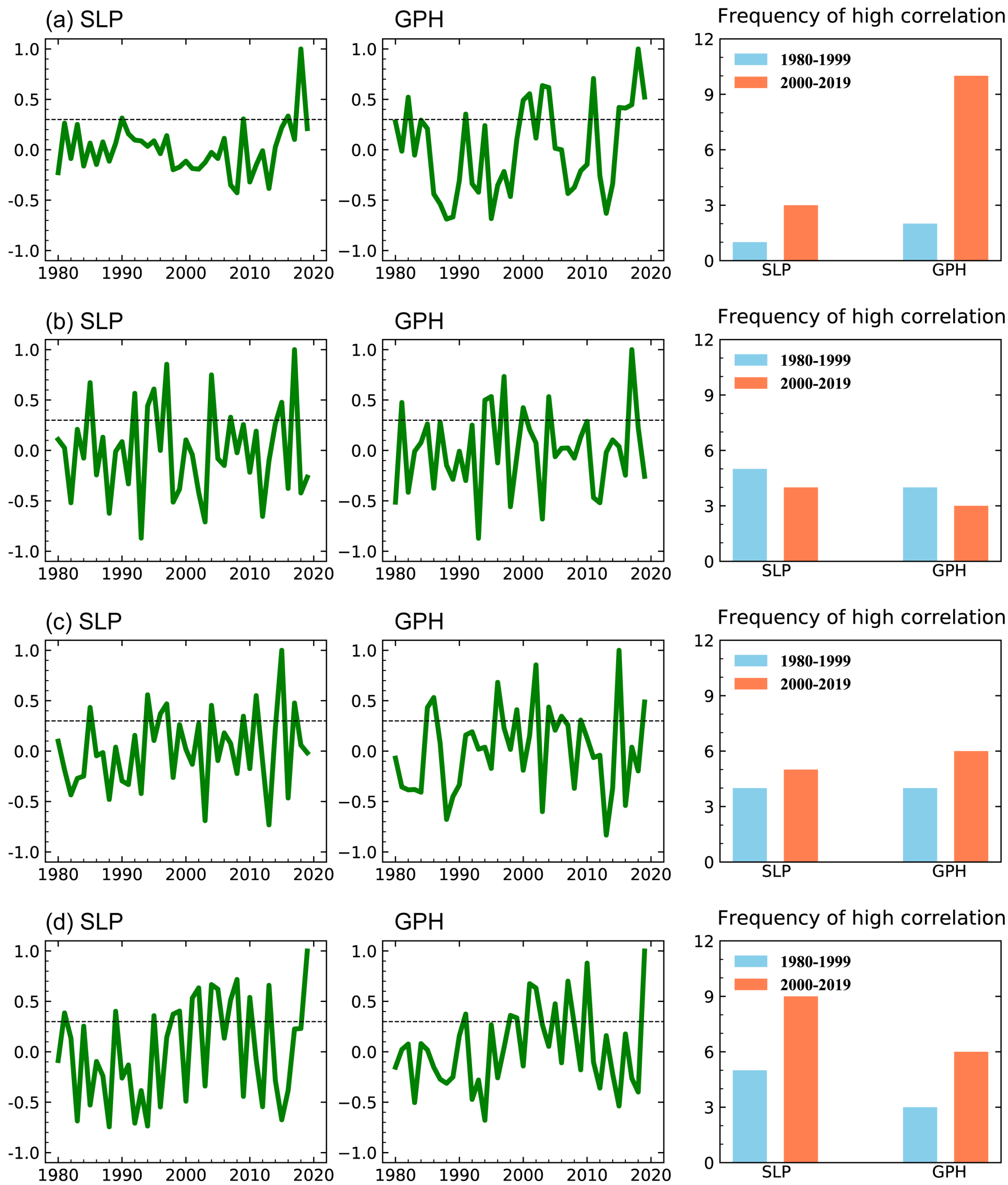

Figure 8. Time series of spatial correlation in SLP (left) and 500 hPa GPH (middle) anomalies over East Asia and Western Pacific (EAWP, 90°–160°E, 20°–60°N) in June 2018 (a), July 2017 (b), July 2015 (c) and September 2019 (d) and those in the same targeted month of each year during 1980–2019. The dotted lines mark the correlation coefficient of +0.3, which is used as a threshold to define "moderate to high correlation". The bar chart (right) represents the frequency of SLP and 500 hPa GPH anomalies in the same months during 1980–1999 (blue) and 2000–2019 (orange) that have moderate to high correlation (>0.3) with those in June 2018, July 2017, July 2015 and September 2019.

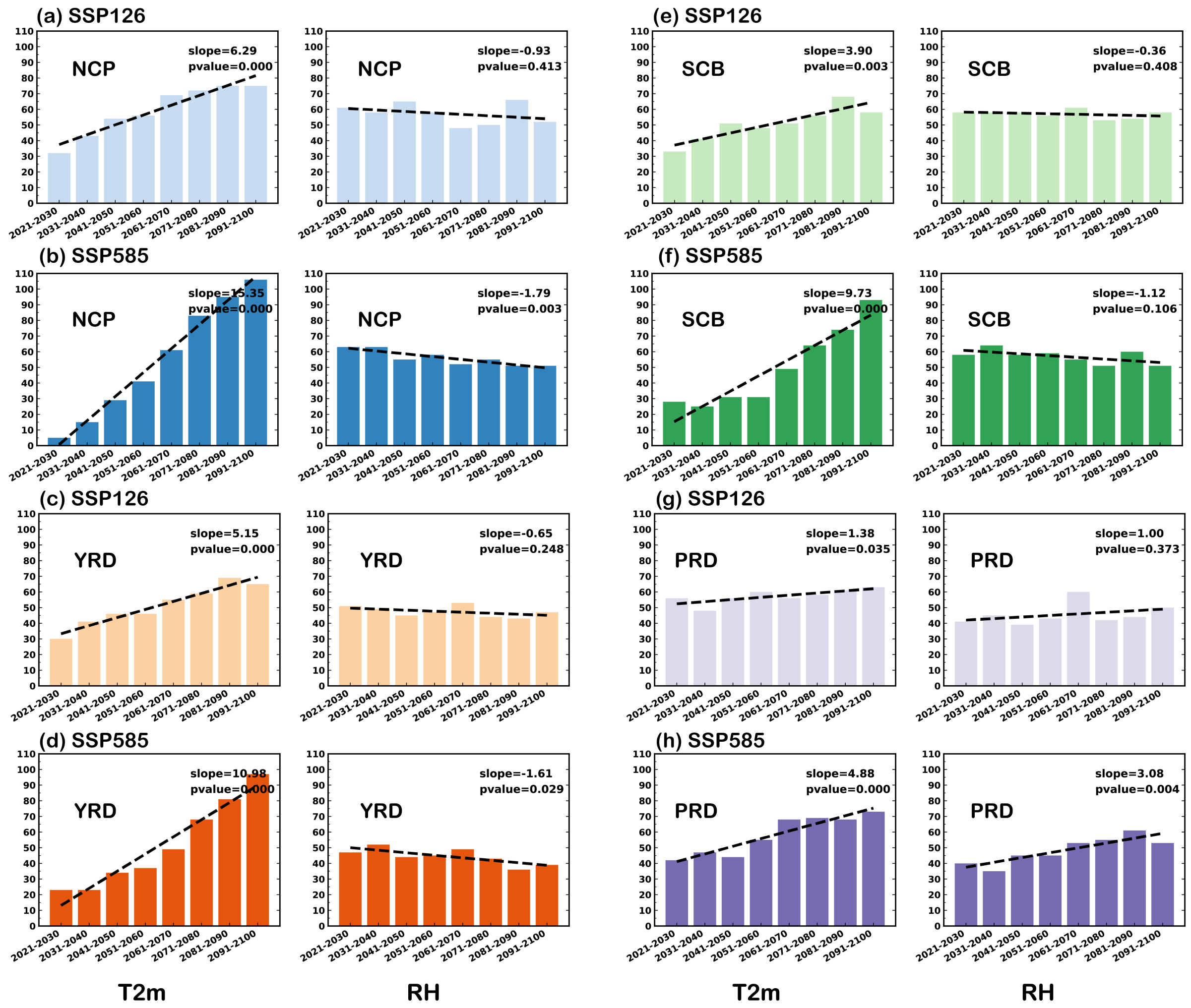

Figure 9. Frequencies of extreme months with T2m or RH anomalies exceeding the 80th percentile or below the 20th percentile of the distributions over NCP (115°–120°E, 38°–44°N) (a, b), YRD (120°–125°E, 28°–32°N) (c, d), SCB (102.5°–105°E, 30°–32°N) (e, f) and PRD (110°–115°E, 22°–26°N) (g, h) in each 10-year interval during 2021–2100 under two SSPs future scenarios of 13 CMIP6 models. The two SSPs are SSP1-2.6 and SSP5-8.5. The slope and P values of the linear regression during 2021–2100 are shown in the upper right of each panel.

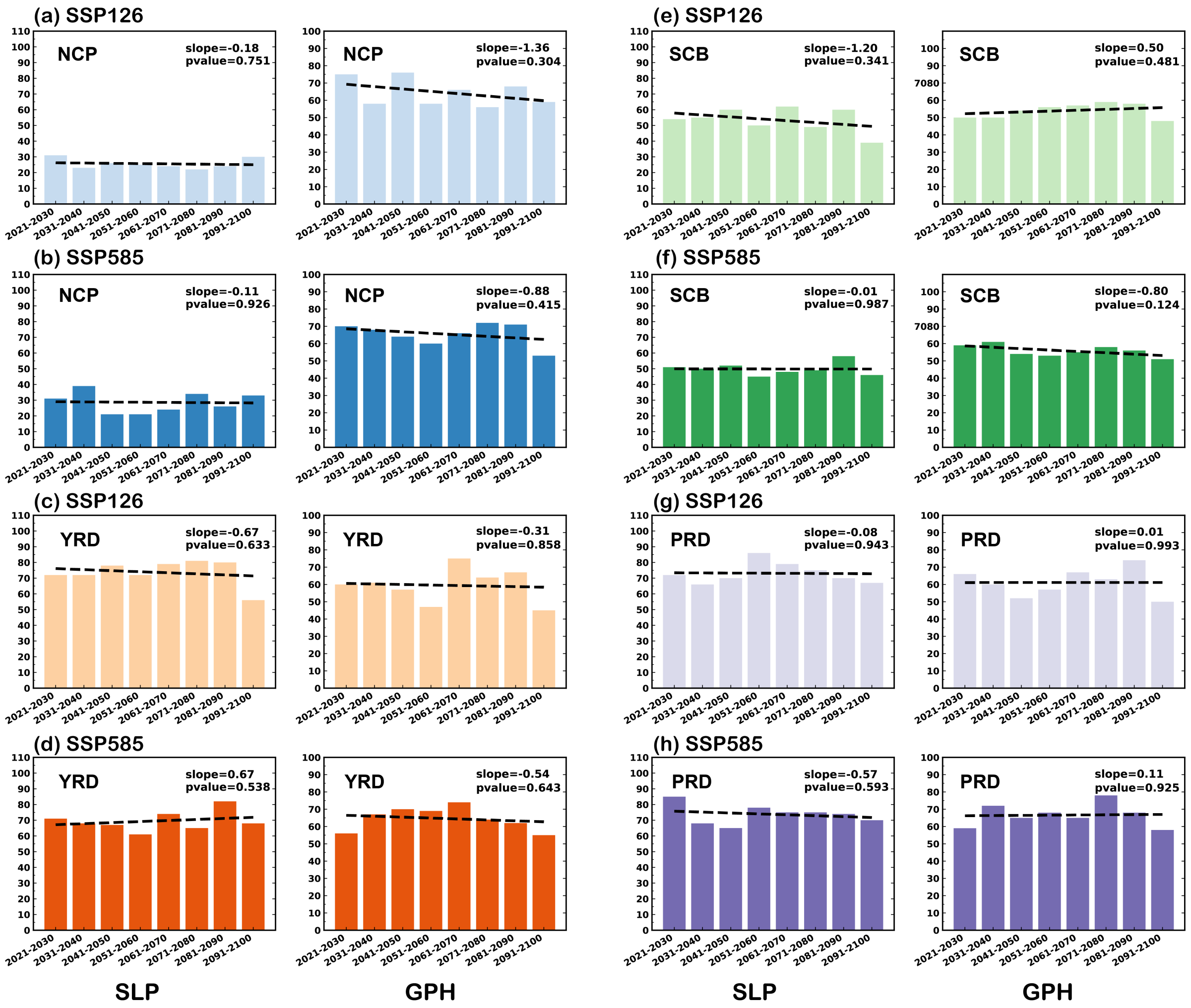

Figure 10. Frequencies of extreme months with SLP and 500 hPa GPH that have moderate to high correlation (>0.3) to those in June 2018 (a, b), July 2017 (c, d), July 2015 (e, f) and September 2019 (g, h) in each 10-year interval during 2021–2100 under two SSPs future scenarios of 13 CMIP6 models. The two SSPs are SSP1-2.6 and SSP5-8.5. The slope and P values of the linear regression during 2021–2100 are shown in the upper right of each panel. The linear trends of SLP and GPH in each model grid were removed before the correlation coefficient is calculated.