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Meteorological factor contributions to the seesaw concentration pattern between PM_{2.5} and O₃ in Shanghai

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With the implementation of various stringent emission reduction measures since 2013 in China, significant declines in fine particle (PM_{2.5}) concentrations have occurred nationwide. However, China has suffered from increasing levels of ozone pollution in eastern urban areas. Many studies focus on the chemical interaction between PM_{2.5} and O₃, but the meteorological mechanisms of the seesaw variation pattern between them are still unclear. Taking the megacity Shanghai (SH) as an example, we explored the meteorological causes of two types of PM_{2.5}-O₃ concentration variation seesaw events, i.e., high PM_{2.5} concentration with low O₃ concentration (Type-one) events and low-PMhigh-O₃ (Type-two) seesaw events. The backward trajectories of the 144 Type-one events are divided into three clusters. Among the three clusters of Type-one seesaw events, the boundary layer height decreases by 20.53%-53.58%, and the wind speed decreases by 17.99%-28.29%, which is unfavorable for the diffusion of local air pollutants and contributes to the accumulation of PM_{2.5}. Additionally, a backward air mass with a high content of PM_{2.5} plays an important role in the Type-one events, especially in the case of cluster one. In terms of cluster two, the increase in cloud cover, decrease in solar radiation and increase in relative humidity also promote the hygroscopic growth of aerosols and suppress the production of O₃. As for cluster three, higher cloud cover and relative humidity contribute to the seesaw pattern of PM and O₃. The 64 Type-two seesaw events are divided into two clusters. O₃-rich air masses from the nearby east sea surface and remote northeast China increase the local O₃ of SH. Moreover, in cluster one, high boundary layer depth and wet deposition contribute to the decrease in PM_{2.5} concentration. In cluster two, a obvious decrease in cloud cover and increase in solar radiation are also favorable for the photochemical production of ozone. The results will provide suggestions for the government to use to take measures to improve the air quality of SHs.

KEYWORDS

PM_{2.5}, O₃, meteorological mechanisms, Shanghai, backward trajectory

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Introduction

In recent years, cities in eastern China have been facing severe air pollution. PM2.5 (particles with aerodynamic diameters less than 2.5 µm) and ozone concentrations often exceed air quality standards (Ministry of Ecology and Environment, 2013). Air pollution is one of the main environmental problems in urban areas, particularly in megacities, such as Shanghai (SH). Exposure to high levels of PM2.5 has adverse effects on human health and ecosystem productivity (Yue et al., 2017; Wang et al., 2019a). Thus, in September 2013, the State Council issued the Action Plan on the Prevention and Control of Air Pollution, known as the Clean Air Actions (CAA), to aggressively control anthropogenic emissions. The Yangtze River Delta region was required to reduce its PM2.5 concentrations by 20% within 5 years (Ministry of Ecology and Environment, 2013). Starting that year, PM2.5 data from a nationwide monitoring network of approximately 1000 sites became available from the China National Environmental Monitoring Center (CNEMC) of the Ministry of Ecology and Environment of China (MEE). Afterward, in a three-year action plan to fight air pollution (State Council of the People's Republic of China, 2018), sulfur dioxide, nitrogen oxides and PM2.5 were further restricted. With joint efforts, the average annual PM2.5 concentration in China decreased by 44% in 2019 compared with 2013 (Zhong et al., 2021). However, at the same time, the surface ozone concentration in China continues to rise (Liu et al., 2018; Dang and Liao, 2019, 2012-2017), which partially offsets the health benefits brought by the decrease in PM_{2.5} concentration.

Stage II of the CAA plan began in 2018 (State Council of the People's Republic of China, 2018), and new emission control measures were implemented for surface ozone.

Given the seesaw variation between PM2.5 and O3, i.e., the decrease in O₃ concentrations with the increase in PM_{2.5} contents, many studies have been conducted to investigate the chemical mechanisms (Shao et al., 2022; Wu et al., 2022). The interaction between PM2.5 and O3 is closely related and very complicated. NOx and VOCs (volatile organic compounds) in the atmosphere are able to generate ozone under light conditions in photochemical reactions (Carrillo-Torres et al., 2017). In this process, NOx, SO2 and VOCs are also oxidized to form secondary PM_{2.5} (Zhu et al., 2015). At nighttime and in the immediate vicinity of large NO emissions, ozone concentrations are depressed through the process of NOx titration. The result is the net conversion of O₃ to NO₂ (Gillani and Pleim, 1996). In regions with high NOx emissions, O3 formation can be VOClimited (Sillman et al., 1990; Kleinman et al., 2003). As a result of CAA, NOx decreased obviously (Silver et al., 2018, 2015-2017). Thus, due to the nonlinear relationship between O₃ production and NOx (Simon et al., 2015), the concentration of O₃ in VOClimited areas in eastern China increases to some extent. Through heterogeneous chemical reactions, PM_{2.5} can indirectly affect O₃. In winter on the North China Plain, formaldehyde is produced by the oxidation of reactive VOCs. The fast production of ozone is driven by HOx radicals from the photolysis of formaldehyde, overcoming radical titration from the decreased NOx emissions (Li et al., 2021). Research has demonstrated an ~40% decrease in PM_{2.5}, slowing the aerosol sink of hydroperoxyl (HO2) radicals



FIGURE 1

The detailed geographical distribution of 9 observational stations in the Shanghai region. The points in the figure represent the latitude and longitude of the Shanghai air quality monitoring station. Air quality monitoring station locations can be obtained from the website of Ministry of Ecology and Environment of the People's Republic of China (http://106.37.208.233:20035). Capital letters in the figure represent abbreviations of site names.

and thus stimulating ozone production at 0.6–1 ppbv a⁻¹ in Beijing–Tianjin–Hebei (Li et al., 2019). Using the WRF-CMAQ model, Liu and Wang also uncovered a similar and substantial effect of HO2 uptake on increases in O₃ levels due to changes in PM concentrations (Liu and Wang, 2020a). Although many studies have been performed to evaluate the quantitative contributions of photolysis rates and heterogeneous chemical reactions to the production of O₃, their relative importance is still unclear (He and Carmichael, 1999; Sillman, 1999).

The absorption and scattering of shortwave radiation by $PM_{2.5}$ can attenuate incident solar shortwave radiation and change the radiation balance in the atmosphere (Charlson et al., 1992; Bond et al., 2013). The attenuation of short-wave radiation will reduce the photolysis rate of O₃ and precursors (J $[O_3^{1}D]$ and J $[NO_2]$, etc.), which will decrease the photochemical reaction intensity, and then reduce the net chemical production of O₃ (Jacobson, 1998; Li et al., 2011; Deng et al., 2012). Wang et al. (2019b) analyzed the photolysis rate data in Beijing from 2012 to 2015, and found that aerosols caused J[NO₂] and J[O₃¹D] near the ground in Beijing to decrease by 24%–30% and 27%–33%, respectively, which led to a decrease of photochemical generation rate in the local ozone in summer by approximately 25%.

Previous studies on the interaction between $PM_{2.5}$ and O_3 have mostly focused on chemical mechanisms. However, meteorological conditions also have great effects on the formation, transportation and deposition processes of $PM_{2.5}$ and O_3 and their precursors (Finlayson-Pitts and Pitts, 1986; Cox and Chu, 1996; Xu et al., 1996; Mu and Zhang, 2014; Zhang et al., 2014; Hu et al., 2015; Yin et al., 2017; Zhang, 2017; Wang et al., 2021). For example, by altering the chemical reaction rates directly and the biogenic emissions of VOCs indirectly, higher temperatures can enhance ozone formation in most instances (Guenther et al., 2006; Lu et al., 2019b). The response of $PM_{2.5}$ concentrations to temperature was largely the result of

competing changes in sulfate and nitrate concentrations with a smaller role played by organics (Dawson et al., 2007). Increasing relative humidity will increase the water content of fine particles, thus increasing the PM2.5 concentration (Dawson et al., 2007). Water vapor is able to react with the excited oxygen atom O(1D) and generate OH radicals, which eventually lead to a reduction in the reaction between $O(^{1}D)$ and O₂ (Johnson et al., 1999). Thus, the increase in water vapor can lead to a decrease in surface ozone concentrations (Camalier et al., 2007; He et al., 2017).Pollutant distributions are strongly affected by the wind field (Chen et al., 2020). PM_{2.5} concentrations typically decrease by an order of magnitude between polluted regions and the diluting background air, whereas for ozone concentrations may actually increase or decrease, affected by upwind regions (Jacob and Winner, 2009; Gu et al., 2020b). Precipitation decreases both PM2.5 and O₃ concentrations via wet removal (Shan et al., 2008). An increase in the planetary boundary layer height can decrease pollutant levels via dilution of primary pollutants into a larger volume of air (Su et al., 2018)

However, most of the existing studies focused on the meteorological condition contributions to the individual pollution of $PM_{2.5}$ or O_3 (He et al., 2017; Wang et al., 2018; Lu et al., 2019a; Zhai et al., 2019; Li et al., 2020; Yin et al., 2020). In terms of the synergetic seesaw pattern between $PM_{2.5}$ and O_3 , its meteorological mechanism has been poorly understood. Individual meteorological factors will have the same or adverse effects on the concentration of $PM_{2.5}$ and O_3 . What is the net contribution of meteorological conditions to the seesaw concentrations pattern between $PM_{2.5}$ and O_3 . This study will take the seesaw concentration events of $PM_{2.5}$ and O_3 in Shanghai as a case to reveal their meteorological causes, which will benefit the prediction of future synergetic pollution of $PM_{2.5}$ and O_3 .



Data and method

Air pollution dataset

Hourly surface observations of O_3 , $PM_{2.5}$ and nitrogen dioxide (NO_2) were all obtained from the CNEMC network (http://106.37. 208.233:20035/, last access: 28 December 2020). As part of the CAA, the network started in 2013 with 496 sites in 74 major cities across the country, growing to more than 2500 sites in 366 cities by 2020. At each monitoring site, the concentration of O_3 was measured using ultraviolet absorption spectrometry and differential optical absorption spectroscopy. $PM_{2.5}$ mass concentrations are measured using the micro-oscillating balance

method and/or the β absorption method. NO₂ concentrations are measured by the molybdenum converter method, which is known to have positive interferences from NO₂ oxidation products (Ministry of Ecology and Environment, 2012). Instrumental operation, maintenance, data assurance, and quality control were conducted based on the most recent revisions of China's environmental protection standards (Ministry of Ecology and Environment, 2013). In terms of time, all the data we used are from 13 May 2014 to 7 November 2020.

Air quality data were measured at 9 sites in the Shanghai region, and each site has distinct representation for different influences from urban air pollutant emissions. Figure 1 shows the locations of the nine sites, i.e., Putuo Station, No.15 Factory



The seasonal distribution (A), precipitation distribution (B) and annual distribution (C) of seesaw events in Shanghai during 2013–2020. If the hourly precipitation of any grid exceeds 0.5 mm, it is considered that there is an obvious wet cleaning process in this study.





Station, Hongkou Station, Xuhui Shanghai Normal University Station, Yangpu Sipiao Station, Jing'an Monitoring Station, Pudong Chuansha Station, Pudong New Area Monitoring Station and Pudong Zhangjiang Station. As the observation system is being developed and gradually improved, the duration of available data varies from station to station. The concentration unit of air quality data provided by the official website of MEE is micrograms per cubic meter (μ g·m⁻³). Before 31 August 2018, the concentration was under the standard condition (273 K, 1013 hPa). From 1 September 2018 and on, the reference gas concentration was changed to 298 K and 1013 hPa, and the $PM_{2.5}$ concentration was changed to the measured local environmental conditions. Additionally, we removed invalid values and abnormal values from the hourly CNEMC data due to instrument calibration issues.

Meteorological dataset

The hourly dataset of the fifth-generation European Centre for Medium-Range Weather Forecasts (ECMWF ERA5) atmospheric reanalysis dataset with a resolution of 0.25° was



used to describe the meteorological characteristics in this research (https://cds.climate.copernicus.eu/cdsapp#!/dataset/reanalysis-era5-pressure-levels?tab=overview, https://cds.climate.copernicus.eu/cdsapp□!/dataset/reanalysis-era5-

single-levels?tab=overview, last access: 18 May 2022). The U and V components of 10 m wind, 2 m air temperature, relative humidity, boundary layer height (BLH), mean surface downward longwave radiation flux (Longwave), mean surface downward shortwave radiation flux (Shortwave), total precipitation, fraction of cloud cover, surface net solar radiation (SSR), and the vertical distribution of vertical velocity, geopotential, and ozone mass mixing ratio are

involved in this study. In order to find out the statistical difference between meteorological variable values of five clusters, t-test is used to examine significance. The t-test is a popular statistical tool used to test differences between the means of two groups, or the difference between one group's mean and a standard value. Running t-tests help us to understand whether the differences are statistically significant (Box, 1987).

Additional information on ozone in ERA5 is that the heterogeneous ozone chemistry is updated and the interaction between ozone and the atmosphere is one-way. Ozone is advected by the atmospheric flow. However, the



Transport pathways of air masses in 5 clusters with (A,B) PM_{2.5} mean concentration and (C,D) O₃ mean concentration (ppb). The background PM_{2.5} concentration is the annual average from 2014 to 2020, sourced from MERRA-2. The background ozone concentration is the annual average from 2014 to 2020, sourced from MERRA-2.

ERA5 prognostic ozone has no feedback on the atmosphere *via* the radiation scheme (Hersbach et al., 2020).

Since it has a relatively high temporal and spatial resolution (1 hour averages and $0.5^{\circ} \times 0.625^{\circ}$), we also obtain surface PM_{2.5} concentrations in the Modern-Era Retrospective Analysis for Research and Applications, version 2 (MERRA-2). Using fields from the 2D aer_Nx collection, the concentration of particulate matter can be computed using the following formula:

 $\label{eq:PM2.5} PM_{2.5} = DUSMASS25 + OCSMASS+ BCSMASS + SSSMASS25 + SO4SMASS* (132.14/96.06)Because the species tracer in MERRA-2 is the sulfate ion, sulfate requires a multiplication factor (https://gmao.gsfc.nasa.gov/reanalysis/MERRA-2/FAQ/). In addition to the lack of nitrate aerosols$

in MERRA-2, the bias in MERRA-2 mainly comes from the uncertainties of the emission inventory and meteorological simulation in GOES-5(Geostationary Operational Environmental Satellite-5) models (Buchard et al., 2016; He et al., 2019).

Definition of seesaw events of $\text{PM}_{2.5}$ and O_3

As shown in Figure 2, the $PM_{2.5}$ concentration and O_3 concentration in SH have obvious seasonal and diurnal seesaw variation characteristics. Figure 2 shows the climatology of

TABLE 1 Absolute anomalies and relative anomalies of PM_{2.5}, O₃ and NO₂ concentrations in each backward trajectory cluster of the two seesaw events in Shanghai. Units: $\mu g/m^3$. The anomalies are with respect to their average value of every hour in every month.

Pollutants	Cluster1	Cluster2	Cluster3
Type-one			
PM _{2.5}	+49.84	+40.45	+45.22
O ₃	-33.19	-39.49	-33.63
NO ₂	+41.06	+30.61	+22.77
Type-two			
PM _{2.5}	-33.88	-40.34	
O ₃	+25.77	+22.65	
NO ₂	-22.83	-33.87	

TABLE 2 Relative anomaly percent changes in PM_{2.5}, O₃ and NO₂ concentrations in each backward trajectory cluster of the two seesaw events in Shanghai. Units: %. The relative anomalies are with respect to the average value of every hour in every month.

Pollutants	Cluster1	Cluster2	Cluster3
Type-one			
PM _{2.5}	+107.54%	+95.92%	+105.65%
O ₃	-58.45%	-54.70%	-54.56%
NO ₂	+82.88%	+71.88%	+51.81%
Type-two			
PM _{2.5}	-67.97%	-71.09%	
O ₃	+53.18%	+64.25%	
NO_2	-46.61%	-57.64%	

diurnal and seasonal variations in PM2.5 and O3 concentrations. It shows higher PM2.5 concentrations during the morning rush hour and lower concentrations in the afternoon; while O3 shows lower concentrations at 7-8 a.m. and higher concentrations at 2-3 p.m. In terms of seasonal variation, the PM_{2.5} concentration in winter is the highest, but winter is the cleanest season for O₃ pollution in SH. The chemical and physical mechanisms of the diurnal and seasonal seesaw pattern between PM2.5 and O3 in Figure 2 have been revealed in previous studies from the perspective of the variation in emission and meteorological factors. In short, ozone has a strong correlation with high temperatures, so it peaks at midday and during mid-spring to early summer in SH (Gu et al., 2020a; Chang et al., 2021). The reason that ozone concentration of Shanghai in spring is higher than summer is the photochemical reaction of ozone affected by plum-rain in summer (Gao et al., 2017). Deep and persistent cloud cover impedes solar radiation to reach the surface of the Earth, affecting the photochemical generation of ozone Affected by human activities, household heating in winter and vehicle emissions in the morning and evening are able to increase PM_{2.5} (Xiao et al., 2015; Zhang and Cao, 2015). Outbreaks of haze

generally occur during the winter as a result of temperature inversions. During an inversion warm air settles above a layer of cool air near the surface. The lid-like warm air traps pollutants near the surface. Lower PM in afternoon can be explained by the enhanced emission for heating and relatively low the boundary layer (Zhang and Cao, 2015). However, there are still some seesaw events of $PM_{2.5}$ and O_3 under the background of diurnal and seesaw patterns shown in Figure 2, the mechanisms of which are still unclear. This study will focus on the meteorological causes of these seesaw events.

To eliminate the influence of seasonal variation, diurnal variation and interannual variation, the original data are classified according to different years, months and hours, and the mean, first quartile and third quartile are calculated. The original data of PM2.5 and O3 are compared with the corresponding quartile. If the hourly concentration of PM2.5 is greater than the corresponding third quartile, and O3 is less than the corresponding first quartile, we call it High-PM-Low-O3 seesaw hour (Type-one for short). Similarly, if the hourly concentration of O3 is greater than the corresponding third quartile, and PM2.5 is less than the corresponding first quartile, it is defined as Low-PM-High-O₃ hour (Type-two). The quartiles are not fixed, but change with the specific year, month and hour. The above three influencing factors, together with the original data of pollutant concentration, any change in four variables will affect the quartiles. That is, the relatively high/ low air pollutant concentrations are used to define the seesaw cases instead of their absolute values. To ensure the spatial and temporal continuity of the two types of cases, a seesaw event was defined as at least 12 consecutive seesaw hours occurring at more than half of the available observation stations in Shanghai.

HYSPLIT model and clustering analysis

Backward trajectory analysis essentially follows a parcel of air backward in hourly time steps for a specified length of time. The Hybrid–Single Particle Lagrangian Integrated Trajectory (HYSPLIT) model (Draxler et al., 2009) developed by the National Oceanic and Atmospheric Administration (NOAA) was used in this study to identify potential source regions of $PM_{2.5}$ or O_3 for a specific city of SH and capture the horizontal movement of the air masses from the source region. The model uses internal terrain following sigma coordinates and meteorological element fields are interpolated linearly to the corresponding coordinates.

To show the origins of air masses arriving at SH, 48-hour backward trajectories were computed every 6 h (at 00:00, 06:00, 12: 00, and 18:00 UT) for the years 2014–2020 for 100 m above ground over SH using the HYSPLIT- 4 model (Draxler et al., 2009.). Meteorological archive data used for the HYSPLIT model are from NCEP (National Centers for Environmental Prediction), at a horizontal resolution of 2.5° (latitude) $\times 2.5^{\circ}$ (longitude).



FIGURE 7

Anomalies of boundary layer height in 5 clusters. The solid blue line in the background indicates the provincial boundaries of China. Different subtitles represent different clusters. The black dots indicate that the grids are above the 99% confidence level (p < 0.01). The solid red box marks the location of Shanghai. Unit: m.



FIGURE 8

Anomalies of relative humidity in 5 clusters. The solid blue line in the background indicates the provincial boundaries of China. Different subtitles represent different clusters. The black dots indicate that the grids are above the 99% confidence level (p < 0.01). The solid red box marks the location of Shanghai. Unit: %.

TABLE 3 Meteorological factor percent changes in 5 clusters. The bold font indicates an important influencing factor. Units: % Data source: ERA5-1000 hPa & Surface. The relative anomalies are with respect to the average value of every hour in every month. Bold fonts indicate the main factors influencing the specific seesaw cluster.

Variable	Cluster1	Cluster2	Cluster3
Type-one			
Temperature	-3.27%	+11.18%	-0.27%
BLH	-53.58%	-48.13%	-20.53%
Shortwave	+2.91%	-20.96%	-9.67%
Longwave	-3.15%	+6.75%	+0.95%
SSR	+1.78%	-15.70%	-6.48%
Precipitation	-32.76%	+32.25%	-33.55%
Windspeed	-28.29%	-39.11%	-17.99%
Cloud Cover	+2.62%	+24.92%	+8.69%
Relative Humidity	-9.89%	+12.78%	+5.80%
Type-two			
Temperature	+21.75%	-22.80%	
BLH	+45.63%	+100.02%	
Shortwave	-18.87%	+6.27%	
Longwave	+7.08%	+0.84%	
SSR	-13.94%	+3.90%	
Precipitation	+136.85%	-40.58%	
Windspeed	+38.35%	+38.88%	
Cloud Cover	+37.10%	-51.40%	
Relative Humidity	+13.22%	+1.60%	

TrajStat software is used to classify the backward trajectories arriving at SH to better determine the transportation of $PM_{2.5}$ and O_3 . The trajectories were assigned to distinct clusters according to their moving speed and direction using Ward's hierarchical method based on the Euclidean distance between all pairs of trajectories (Sirois and Bottenheim, 1995). Major transport pathways leading to elevated or dropped $PM_{2.5}$ concentrations and ozone concentrations were identified by combining trajectories with the corresponding mean concentrations of the pollutant.

Results

Occurrence frequency of the seesaw events

Figure 3A shows the seasonal distribution of the two types of $PM_{2.5}$ and O_3 seesaw events during the 2013–2020 period. Obviously, the frequency of Type-one seesaw events is higher than Type-two in total. There were 144 cases in Type-one and 64 cases in Type-two. In terms of seasons, most of the two types of seesaw events occur in winter. Both seesaw events show the lowest frequency in summer, which is approximately 22.73% and 10% of that in winter for Type-one and Type-two, respectively.

Considering the wet deposition, the physical mechanisms of the seesaw events are supposed to be different from those on dry days. We reclassified the seesaw events with the occurrence of precipitation in Figure 3B. In summer, more than half of the seesaw events occurred on rainy days, which is different from other seasons. The sample size is 5 for both dry and rainy Type-one events in summer, but all the Typetwo events occur on rainy days. The positive correlation between PM2.5 and O3 in summer results in the lowest occurring frequency in event summer. High O3 concentrations in a strong oxidative air condition promoted the formation of secondary particles in summer. During plum-rain in summer, under most circumstances PM2.5 concentrations and O3 concentrations decrease at the same time due to precipitation (Jia et al., 2017; Zhu et al., 2019). The interannual variation in the seesaw event frequency is shown in Figure 3C.

Atmospheric circulation anomalies of the seesaw events

The roles of meteorology on pollutants are complex and varying. To further explore the meteorological causes behind the seesaw incidents, we calculated the anomalies of meteorological variables under the seesaw events in different seasons, including geopotential height (Z) and U/V components at 925 hPa, 850 hPa and 500 hPa. Figure 4 indicates the seasonal average of Z&UV anomalies at 925 hPa for Type-one seesaw events (i.e., high PM2,5 with low O3 concentrations). It shows almost the same atmospheric circulation pattern in the four seasons, with a cyclonic anomaly over North China and the Northeast China region, which brings northwest or west winds to SH. For the case of Type-two seesaw events in Figure 5 (i.e., high O₃ with low PM_{2.5} concentrations), the circulation anomalies are also the same in spring, autumn and winter. SH is located at the edge of the west positive and east negative pressure, and the prevailing wind of SH comes from the northeast. In terms of the summertime, the sample size of Type-two is small, which leads to the vague distribution of its circulation anomaly pattern. The atmospheric circulation anomalies at 850 hPa and 500 hPa also show the same pattern for each type of seesaw event (figures not shown here). According to Figures 4, 5, there is almost no difference in the seasonal circulation anomalies, therefore, we investigate the mechanisms of seesaw events without considering their seasonal difference hereafter.

Regional transport contributes to the seesaw events

Although it shows a consistent seasonal circulation anomaly distribution for each type of seesaw events in Figures 4, 5, it



Precipitation intensities in 5 clusters of Shanghai seesaw events. No rain: the rate of fall varying between a trace and 0.5 mm per hour. Light rain: the rate of fall varying between 2.5 mm and 7.6 mm per hour (American Meteorological Society).



indicates an average pattern of atmospheric circulation. To reveal the mechanism of the specific seesaw event, the 48 h backward trajectory of each seesaw events is calculated and clustered, which can provide the information on the origins and transport pathways of air masses. As shown in Figure 6, cluster analysis yielded a total of three air-mass clusters for the 144 Type-one seesaw events, and two air-mass clusters for the 64 Type-two events.

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For the case of Type-one events, the three kinds of air masses in SH come from the eastern ocean, north China and southwest of SH. The frequencies of these three kinds of air masses are almost the same. There are 56 samples in C1, 36 samples in C2 and 52 samples in C3. Tables 1, 2 summarize the anomalies of $PM_{2.5}$, O₃ and NO₂ concentrations in each backward trajectory cluster. For Type-one seesaw events, the magnitudes of $PM_{2.5}$ anomalies are higher in cluster-one (C1 for short) (+107%) and cluster-three (C3) (+105%) than cluster-two (+95%), which may be due to the higher $PM_{2.5}$ background concentrations transported from the north and southwest upwind regions of SH (as shown in Figure 6). Moreover, the NO₂ concentration increased by 83% and 52% in C1 and C3 in the Type-one events, respectively, which can also imply the regional transportation of aerosols and precursors.

In terms of Type-two seesaw events, the air mass in SH mainly comes from the nearby east sea surface (61% of events, 39 samples) with minor influence from the remote northeast China (39% of events, 25 samples) Compared to the average pollutant concentrations, the PM2.5 concentrations of the two types of clusters decreased by 67.97% and 71.07%, respectively. Correspondingly, the ozone concentration increased by 53.18% and 64.25%, respectively. Based on the climatology of O3 and $\text{PM}_{2.5}$ in Figure 6, it shows higher O_3 concentrations and lower PM_{2.5} contents in the east sea surface. The easterly wind brings O3-rich and low PM2.5 air masses to SH, and leads to high O3 and low PM_{2.5} pollution for the two clusters of Type-two events. Several studies have found that ozone tends to be continuously produced in the downwind regions of major cities (Kleinman et al., 2003; Tie et al., 2009, 2013). O3 formation usually occurs under strong VOC-limited conditions in coastal megacities, which inhibits O3 chemical production. In the downwind

region, O_3 formation shifts to NOx-limited conditions gradually on the contrary to the sea surface, which results in pronounced O_3 production (Kley, 1997). O_3 precursors emitted from a coastal city are advected to the ocean by land breezes and produce O_3 over the ocean, the O_3 -rich air mass can be advected back to the city by sea breezes.

Meteorological cause for the seesaw events

The meteorological influence on the pollutant varied by region and could be comparable to or even more significant than the impact of changes in anthropogenic emissions (Liu and Wang, 2020b). The relative anomalies of air temperature, BLH, shortwave, longwave, SSR, precipitation, 10 m wind, cloud cover and relative humidity are involved to reveal the meteorological effects on seesaw events. Figures 7, 8 show the spatial distribution of BLH and RH relative anomalies in the five clusters. Table 3 summarizes the regional mean relative anomalies of the above nine meteorological factors of SH in the five clusters. For the case of C1 in Type-one events, it shows lower than normal BLH around SH and the East Ocean, with a significant negative BLH relative anomaly of 199.41 m (-54%). The BLH indicates the vertical diffusion ability of the local atmosphere, which has been demonstrated to have a negative relationship with ambient PM_{2.5} concentrations (Wang and Wang, 2014; Wang and Wang, 2016). The lower PM_{2.5} in C1 of Type-one events can be attributed to the repression of vertical diffusion to some extent. Considering that the wind speed drops by 0.93 m/s (-28%), stable and stagnant circulation may also increase the concentration of pollutants.

For C2 in Type-one, the height of the boundary layer decrease by 48.13% and the relative humidity increased by 12.78%, which is the most significant among the three clusters (Figure 8). Figure 9 shows the occurrence frequency of precipitation during the five clusters. Due to the positive precipitation anomaly (Table 3) and easterly backward trajectory (Figure 6), there is a positive relative humidity anomaly in C2 of Type-one events, which is beneficial for the hygroscopic growth of aerosols and leads to an increase in PM2.5 concentration (Wang et al., 2019b; Won et al., 2021). The scavenging process and physical removal mechanism of PM are determined by different precipitation intensities (Andronache, 2003; Chate et al., 2003; Wang et al., 2010). According to the definition of the American Meteorological Society, precipitation can be classified as light rain, mild rain and heavy rain (American Meteorological Society American Meteorological Society, 2022). Using this definition, Figure 9 shows the distribution of no rain, light rain and mild rain under the five clusters. The precipitation in C2 is primarily light rain, and the deposition ability is weak compared to moistening the air mass. The stable air mass also contributes to the accumulation of surface air pollutants. In addition, based on the regional mean meteorological factor anomalies in Table 3, the significant increase in cloud cover (+25%) and decrease in shortwave radiation (-21%) on the rainy days also caused the weakening of ozone photochemical production, which led to the decrease in surface ozone concentration in Shanghai.

For C3 in Type-one, the magnitudes of the regional mean meteorological anomalies are not as obvious as those in C1 and C2. However, the obvious increase in relative humidity (+5.8%) and decrease in solar radiation (-9.7%) can be considered a beneficial meteorological background for the increase of PM_{2.5} and decrease of O₃, respectively. Notably, the C3 air mass comes from the North China Plain where the concentration of fine particles is generally high (Figure 6). This may indicate that the high concentration of PM_{2.5} in C3 is related to transport. In brief, it is the comprehensive influence of transport, shortwave radiation and relative humidity resulting in C3 (Figure 10).

As described in the previous section, regional transport is undoubtedly the main influencing factor of the Type-two seesaw events. The trajectory in Figure 6 also shows that the air masses originate from areas with higher ozone and lower PM concentrations. Apart from that, other meteorological factors also make sense. In C1 of Type-two, the simple size of the dryday case is small, and only the rainy cases are involved. Apparently, it shows a obvious positive anomaly of precipitation (Table 3) and relative humidity (Figure 8). Compared with the case of C2 in Type-one, the frequency of moderate rain in C1 of Type-two is much higher. The effects of wet deposition are more pronounced than the hygroscopic growth of aerosols in a humid environment, which results in a decrease in $PM_{2.5}$ (Jacob and Winner, 2009).

In C2 of Type-two events, a 51% decrease in cloud cover along with a 6.27% increase in shortwave radiation contributes to the photochemical formation of ozone. Moreover, the increase in solar radiation prompts the development of boundary layer, which results in the positive BLH anomaly by 403.28 m (+100%) and the improvement in atmospheric vertical diffusion conditions. The surface wind speed, which is regularly considered the metric of horizontal diffusion ability, increases by 39% in C2 of Type-two events. The significant improvement of vertical and horizontal diffusion conditions led to a decrease in the PM_{2.5} concentration (Wang et al., 2016, Wang et al., 2018; Wang and Zhang, 2020).

Conclusions and discussion

This study explored the impact of meteorological factors on the seesaw concentration pattern between $PM_{2.5}$ and O_3 in Shanghai from 2014 to 2020. On the basis of the variation in $PM_{2.5}$ and O_3 concentrations, we divided the seesaw events into two types, Type-one (High-PM-Low-O₃) and Type-two (Low-PM- High-O₃). The trajectories of the seesaw events were assigned to distinct clusters

according to their moving speed and direction using the HYSPLIT model. The specific number of clusters is given as follows: three clusters in Type-one and two clusters in Type-two.

In Type-one seesaw events, the boundary layer height in three clusters obviously decreased, C1 (-53.58%) and C2 (-48.13%). Obviously, worse diffusion conditions lead to an increase in the PM_{2.5} concentration. The PM_{2.5} concentration in C1 upwind direction is higher than that in SH, while the ozone concentration is lower. The transport of C1 upwind also contributes. In C2, the relative humidity increased by 12.78% which contributed to the hygroscopic growth of PM_{2.5} and a decrease in O₃ production. Less SSR (-20.96%) and higher cloud cover (+24.92%) in C2 were not conducive to the formation of ozone photochemical reactions. The same conditions were found in C3, but it was less obvious than C2, +8.69% for cloud cover and -9.67% for SSR. Thus, transportation and increased relative humidity were also factors that cannot be ignored in C3.

The cause of another seesaw event is not as complicated as that of Type-one events. Figure 6 and Tables 1, 3 show that the moist oceanic air mass with rich O_3 from the northeast sea enters Shanghai and reduces the concentration of $PM_{2.5}$ at the same time in two clusters of Type-two events. Transport is undoubtedly the main influencing factor in both clusters of Type-two events. Adequate precipitation also contributes to the wet removal of fine particles in C1. A doubled BLH (+100.02%) was beneficial to the diffusion of $PM_{2.5}$ in C2. A 6.27% increase in SSR contributed to the photochemical production of O_3 under clear sky conditions.

Except for the production of ozone from photochemical reactions occurring within the troposphere, the presence of ozone in the troposphere is understood to arise from another basic process. Tropospheric-stratospheric exchange can cause the transport of stratospheric air, rich in ozone, into the troposphere (Langford, 1999). Ni et al. suggested that stratospheric ozone intrusion acts as an additional source of the near-surface tropospheric ozone concentration, which deteriorates O_3 pollution in China (Ni et al., 2019).

Therefore, we also made a vertical profile of the ozone mass mixing ratio in autumn (Figure 11). The scavenging process of precipitation can be clearly seen in this figure. Note that when the seesaw events occurred, the near-surface ozone concentration increased, but the ozone concentration in the upper levels decreased. During the precipitation, strong convective weather with obvious downward flow may have affected stratospheric-tropospheric exchange of air masses and ozone concentrations. Therefore, it is also suspected that the near-surface tropospheric ozone concentration is closely related to transport from the upper layer or stratosphere. However, there are currently no further data supporting this hypothesis. Besides autumn, the transport of ozone in the upper and lower layers is not obvious in other seasons.

Data availability statement

The original contributions presented in the study are included in the article/Supplementary Material, further inquiries can be directed to the corresponding author.

Author contributions

XW and YS designed the research. YS performed the analyses and wrote the paper. Both authors contributed to the final version of the paper.

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Conflict of interest

The authors declare that the research was conducted in the absence of any commercial or financial relationships that could be construed as a potential conflict of interest.

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