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Characteristics of chemical composition and source apportionment of PM_{2.5} during a regional haze episode in the yangtze river delta, china

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The source of PM_{2.5} varies at different stages of urban haze pollution. In addition, there is obvious regional transport of pollutants between urban agglomerations. PM_{2.5} and its major chemical compositions in a regional haze episode were measured continuously from 16 to 27 November 2018 in Nanjing, China. The types of primary sources resolved by principal component analysis (PCA) and positive matrix factorization (PMF) were similar, and the result of PMF was more refined. The average contribution of each source by PMF was: secondary nitrate (64.01%), secondary sulfate (11.62%), incomplete combustion (4.49%), sea salt (8.61%), biomass burning (6.90%), and crustal dust (4.37%). In different haze stages, the distribution characteristics of air pollutants differed. The concentrations of SO_4^{2-} , NO_3^{-} , NH_4^+ , and black carbon were the highest in the haze developing stage, which was 2.0, 3.1, 3.0, and 2.4 times, respectively, higher than that under clean conditions. The increment of NO₃⁻ dominated the development of haze, and the proportion of NO3⁻ from haze generation to development increased by 4.05%. The concentration contributions of secondary nitrate, sea salt, and biomass burning were highest in haze development, secondary sulfate was highest in haze generation, and incomplete combustion was highest in haze dissipation, which was 3.5, 1.8, 3.3, 1.7, and 9.5 times higher than the clean stage, respectively. In the haze episode, the contribution of crustal dust was lower than in the clean stage. Potential source contribution function (PSCF) and concentration weighted trajectory (CWT) revealed that the major source area of air pollutants in Nanjing came from the southeast, and the northwest was the major impact area.

KEYWORDS

haze, PM2.5, chemical composition, Source apportionment, the yangtze river delta

1 Introduction

In China, haze pollution has become a hot topic of common concern among the public, government, and researchers (Wang Q. et al., 2009; Peplow, 2014; Fu and Chen, 2017; Shen et al., 2022). $PM_{2.5}$ has become as much a concern for people's daily life as temperature, wind, and other meteorological variables (Zhang et al., 2013; Ding and Liu, 2014; Maji et al., 2018). The most intuitive effect of the increases of fine particle concentrations is the atmospheric visibility degradation (Chang et al., 2009; Zhang et al., 2010), affecting radiation and climate (Ramanathan et al., 2001; Andreae et al., 2005; Lohmann and Feichter, 2005), and harmful health effects in human (Steinle et al., 2015).

Studies on the analysis and quantitative assessment of $PM_{2.5}$ source have been conducted in China, Song et al. (2006) employed the PCA model to discover five sources of $PM_{2.5}$ in Beijing, where the major chemical contributors to PM2.5 mass were secondary sulfate and nitrate, mixed sources of coal

combustion and biomass burning, industrial emissions, traffic, and dust. Compared with the PCA model, the PMF model can effectively reduce the deviation caused by the "mixing factor" (Gao et al., 2018; Han et al., 2016). The regional transport of air pollutants also play a key role on the pollutant concentrations in a specific area (Liu et al., 2016). Sun et al. (2015) used the potential source contribution function (PSCF) model to find that Beijing's high potential pollution source area located in its southwest. Using concentration weighted trajectory (CWT) model, Jing et al. (2019) found that the episodes of high BC concentrations in Nanjing during the autumn and winter were caused by the long-range transport from the southwestern region. The CAMx model extended by particulate matter source spportionment technology (PSAT) tool was used for tracking and accounting for particulate matter concentrations that originate or are formed as a result of emissions form longrange transportation. With CAMx-PSAT, Wu et al. (2013) found that the primary sources of particulate pollution in the Pearl





River Delta region were regional transmission (42%) and vehicle emissions (21%).

The Yangtze River Delta (YRD) is one of the fastest-growing regions in East Asia. The rapid development of urbanization and industrialization has resulted in the increasingly severe pollution problem and health risk. (Kang et al., 2013; Liu et al., 2013; Wang et al., 2016). Using the CAMx model, Li et al. (2015) found that the main sources of haze pollution in the YRD region are industrial processing, combustion emissions, vehicle emissions, and fugitive dust. Both local emission and regional transportation can cause heavy haze pollution in YRD, which usually depends on meteorological conditions (Wang et al., 2013). Long-range transportation of pollutants was more conducive to the growth of organic carbon therein, while particles mixed with more chloride and nitrate/sulfate (Hu et al., 2018).

At present, there are relatively few studies on the source analysis of persistent and regional haze pollution events in the YRD based on high-resolution online measurements. Exploring the difference of sources and the pollutant transport characteristics at different stages of haze was very important for control strategy in urban agglomerations of the YRD. At different stages of urban haze pollution, the emission of atmospheric aerosols is quite different. The pollutants exchange and transportation between cities will also affect the emission variations of atmospheric aerosols. In this study, the concentration of water-soluble ions, BC, and number concentration of aerosol was observed with high-resolution online monitoring instruments during a haze pollution event in Nanjing from November 16 to 27, 2018. In this article, we will combine with receptor models (PCA, PMF, PSCF, CWT), to explore the different emission sources of PM2.5 at different stages and its impact on upstream and downstream areas on the regional haze pollution event in YRD.

2 Experimental methods

2.1 Descriptions of the measurement site

The Yangtze River Delta lies in the middle and lower Yangtze Plain of East China, which is one of the most developed regions. Nanjing is an important typical industry city in YRD, with comprehensive manufacturing and transportation hub, where the regional haze pollution breaks out frequently (Leng et al., 2016; Sun et al., 2018; Liu et al., 2019). The measurement site is installed on the top of the Meteorological Building (40 m above the ground and 62 m above sea level) of Nanjing University of Information Science and Technology (NUIST, 32.21°N, 118.72°E) (Figure 1).

2.2 Descriptions of measurement instruments

The online system MARGA (Monitor for AeRosols and Gases in Ambient air, ADI 2080) with 1-h resolution implemented the measurement of water-soluble ions and trace gases (Rumsey et al., 2014). The BC concentration was measured by the Aethalometer (AE33) with 5-min resolution. Particle size distribution is important for understanding the sources and effects of atmospheric aerosols. The aerosol number concentrations (5 nm–10 μ m) were measured with a Wide-Range Particle Spectrometer (WPS) with 5-min resolution (Wang et al., 2014). PM_{2.5} was measured using a β -ray duster meter (Model 5030 SHARP) (Zou et al., 2014). The meteorological variables (temperature, RH, wind speed, wind direction, and precipitation) were obtained from the comprehensive meteorological measurement site in NUIST, and the time resolution is 1 h.

Туре	Average	Clear	Haze
Ws m/s)	1.29 ± 0.85	1.67 ± 0.87	1.01 ± 0.73
Т (°С)	10.85 ± 2.79	10.23 ± 2.12	11.3 ± 3.12
RH (%)	80.23 ± 15.14	78.77 ± 13.55	81.28 ± 16.13
Visibility (km)	4.83 ± 4.40	7.71 ± 5.08	2.77 ± 2.18
HCl ($\mu g \cdot m^{-3}$)	0.11 ± 0.11	0.10 ± 0.11	0.12 ± 0.12
$HNO_2 \ (\mu g \cdot m^{-3})$	0.03 ± 0.10	0.01 ± 0.02	0.05 ± 0.13
$HNO_3 \ (\mu g \cdot m^{-3})$	1.09 ± 0.53	0.78 ± 0.37	1.31 ± 0.52
$SO_2 (\mu g \cdot m^{-3})$	6.25 ± 7.99	3.97 ± 6.81	7.87 ± 8.38
NH ₃ (μg·m ⁻³)	9.42 ± 6.83	6.26 ± 2.26	11.68 ± 8.01
Cl^{-} (µg·m ⁻³)	1.91 ± 1.43	1.27 ± 1.07	2.36 ± 1.49
NO_{3}^{-} (µg·m ⁻³)	28.31 ± 18.14	14.59 ± 9.25	38.1 ± 16.51
SO4 ²⁻ (µg·m ⁻³)	12.24 ± 6.76	8.07 ± 5.17	15.21 ± 6.16
Na^+ ($\mu g \cdot m^{-3}$)	0.12 ± 0.13	0.05 ± 0.09	0.17 ± 0.13
NH_4^+ (µg·m ⁻³)	12.73 ± 7.92	6.54 ± 4.69	17.14 ± 6.71
K^{+} (µg·m ⁻³)	0.30 ± 0.40	0.11 ± 0.34	0.43 ± 0.39
Mg^{2+} (µg·m ⁻³)	0.04 ± 0.13	0.03 ± 0.06	0.05 ± 0.17
Ca^{2+} (µg·m ⁻³)	0.31 ± 0.32	0.39 ± 0.35	0.25 ± 0.29
BC ($\mu g \cdot m^{-3}$)	5.22 ± 3.14	2.76 ± 1.47	6.97 ± 2.82
$PM_{2.5} \ (\mu g \cdot m^{-3})$	74.98 ± 42.99	40.79 ± 21.54	99.37 ± 37.38
dN (cm ⁻³)	$22,404.07 \pm 14,023.60$	$17,106.05 \pm 7,522.13$	26,188.38 ± 16,219.73
dS ($\mu m^2 \cdot cm^{-3}$)	858.59 ± 398.78	526.13 ± 244.20	1,065.31 ± 330.71

TABLE 1 Summary of average meteorological variables and pollutants during the study period.

TABLE 2 Rotated component matrix of the principal component analysis during the measurement period.

Chemical composition	Component (KMO = 0.64)		
	PC1	PC2	PC3
Cl-	0.43	0.75	0.30
NO ₃ ⁻	0.93	0.15	-0.03
SO4 ²⁻	0.85	0.15	0.14
Na ⁺	0.48	0.80	0.18
$\mathrm{NH_4^+}$	0.97	0.19	0.02
K^+	0.46	0.76	-0.07
Mg ²⁺	-0.04	0.03	0.95
Ca ²⁺	-0.31	0.68	-0.35
BC	0.75	0.32	-0.05
Characteristic values	4.83	1.47	1.13
Contribution rates (%)	53.72	16.30	12.50

2.3 Principal component analysis

A preliminary source analysis of the water-soluble ions was carried out by principal component analysis (PCA, SPSS version 19.0, SPSS Inc. 1988). PCA is a statistical procedure that is used to reduce the dimensionality. It uses an orthogonal transformation to convert a set of observations of possibly correlated variables into a set of values of linearly uncorrelated variables called principal components (Song et al., 2006).

2.4 Positive matrix factorization

Positive matrix factorization is a multivariate factorization tool that can decompose the sample concentration matrix of the sample into two non-negative constant matrices factor contribution matrix and factor profile matrix, which was used successfully at the US Environmental Protection Agency for the chemometric evaluation and modelling of environmental data sets. Compared to other methods it offers some advantage that consent to better resolve the problem under analysis (Shi et al., 2017).

2.5 Potential source contribution function

The PSCF developed by Ashbaugh et al. (1985) has been widely used in many studies (Lucey et al., 2001; Wang Y. Q. et al., 2009; Zhang et al., 2019). It is calculated based on the spatial grid, which is defined as the ratio of the pollution trajectories. Detailed principles can be found in Ji et al. (2018).



2.6 Concentration weighted trajectory

The CWT model was established by Seibert et al. (1994). In order to avoid the laminations of the PSCF method whereby some grid cells could have the same PSCF value when sample concentrations could either be slightly higher or much higher than the criterion required as a result, it could be much more difficult to distinguish moderate sources from strong ones. Hence CWT method, with a weight concentration for each grid, was used with the aiming of producing a geographical overview of emission source areas within the study region (Wang et al., 2006; Shen et al., 2021).

3 Results and discussion

3.1 Overview of the measurements

The Yangtze River Delta region was mainly controlled by the continental cold high-pressure system during the study period. The weather situation was relative stability under strong high-pressure, and affected by the middle branch trough. As the YRD located in front of the trough, the water vapor transport from the southwest made the humidity higher, which was conducive to the accumulation and development of pollutants. As shown in Figure 2, haze pollution occurred in YRD around 20–22 and 24–27. During the haze period, the PM_{2.5} concentration in YRD changed greatly, and the PM_{2.5} maximum concentrations in Xuzhou, Hefei, Wuhu, Nanjing, Suzhou, and Shanghai were 241, 162, 175, 191, 162, and 125 µg m⁻³, respectively.

Concerning the locations of six cities, Figure 1 also showed it was the regional haze pollution occurred in the central and northern regions of YRD, while the concentration of $PM_{2.5}$ decreased from north to south. Although the emissions and locations were different in the six major cities, the trends of $PM_{2.5}$ variation of them are similar. The correlation between Nanjing and other five cities on $PM_{2.5}$ concentrations was 0.43–0.83 (*t*-test, p = 0.01), so Nanjing selected as the typical city for study is reasonable.



According to Table 1, the average of wind speed was $1.2 \pm 0.8 \text{ m s}^{-1}$ during the study period in Nanjing. The wind speed on clean days was slightly higher than that on haze days. The relative humidity (RH) was high and the average value was $80.23 \pm 15.14\%$, which was helpful to pollutants accumulation. During the early measurement period, the light rain occurred in Nanjing from November 18 to 19, the rainfall intensity was about 0.4 mm h⁻¹. HCl is mainly produced by industrial emissions, and which concentrations variation between clean days and haze days is not obvious. The concentration of SNA (SO₄²⁻, NO₃⁻ and NH₄⁺) is closely related to its gaseous precursors (HNO₂, HNO₃, SO₂, and NH₃) (Chen et al., 2015), and the latter concentrations on haze days were about 1.7–5 times higher than that in clean days.

The average PM_{2.5} value over Nanjing was 74.98 ± 42.99 µg m⁻³ (Table 1), slightly lower than the 75 µg m⁻³ limitation of Grade-II regulation (Ambient Air Quality Standards of China, GB 3095–2012). Overall, water-soluble ions and BC contributed 81.57% of PM_{2.5}, the mass concentration of inorganic components showed a strong positive correlation with PM_{2.5} (R = 0.94, p = 0.01). The correlation between anion and cation in water-soluble ions was also very high ($R^2 = 0.99$, p = 0.01). SNA was still the most important component of PM_{2.5} (up to 71.06%), whose order of proportion was NO₃⁻ (37.76%), NH₄⁺ (16.96%), and SO₄²⁻ (16.32%). NO₃⁻/SO₄²⁻ = 2.31 (>1), so indicated that the vehicle emission occupied main contribution (Yao et al., 2002). The contributions of Cl⁻, Ca²⁺, K⁺, Na⁺, and Mg²⁺ were fewer, totaling 3.57%.

The average concentration of BC on haze days was 6.97 \pm $2.82 \ \mu g \ m^{-3}$, which was 2.5 times that of clean days, and the BC concentration showed an ascending trend during the measurement period. The average aerosol number concentration was 17,106.05 \pm 7,522.125 cm⁻³ on clean days, while on haze days it increased to 1.5 times. The dominant size in PM distribution was 25-55 nm. The average aerosol surface area concentration on haze days was 1,065.31 \pm 330.71 μm^2 cm^{-3}, increased to 2.0 times on the clean days, and the dominant particle size in particle surface area distribution was around 170 and 450 nm.

3.2 Source apportionment of PM_{2.5} by PCA

As shown in Table 2, PCA method was reliable, since KMO = 0.64 (>0.6) (Kaiser-Meyer-Olkin test). All eigenvalues of three principal factors groups were greater than one that resolved 82.52% sources of PM2.5 mass concentrations.

Factor 1 resolved a mixture of secondary sources and incomplete combustion accounting for 53.72% of the total variance. It was strongly loaded with SO_4^{2-} , NO_3^{--} and NH_4^+ (all around 0.9), suggesting probably contribution from secondary sources (Taghvaee et al., 2018). The partial correlation coefficient of BC also reached 0.75, indicating that the above process was accompanied by incomplete combustion (Rehman et al., 2011).



Factor 2 was identified as a mixture of sea salt and biomass burning because of high factor loadings by Cl⁻, Na^{+,} and K⁺, which accounted for 16.30% of the total variance. Cl⁻ is the identification component of sea salt (Liu et al., 2017), K⁺ is the identification component of biomass burning (Santoso et al., 2008), and Na⁺ can be produced in both processes.

Factor 3 was explained by crustal dust and accounted for 12.50% of the total variance. It was mainly loaded with Mg^{2+} (0.95), Mg^{2+} and Ca^{2+} abundant elements in the crust, both of which are considered to be the identification components of crustal dust in related studies (Almeida et al., 2005). So, in this haze pollution event, the inorganic components of PM_{2.5} primarily controlled by secondary sources, incomplete combustion, sea salt, biomass burning, and crustal dust.

3.3 Source apportionment of PM_{2.5} by PMF

3.3.1 PMF model results

The PMF model resolved the six sources as the PCA method proceeded: secondary nitrate, secondary sulfate, incomplete combustion, sea salt, biomass burning, and crustal dust. (Figure 3).

Source 1 was identified as secondary nitrate, it mainly consisted of NO₃⁻ (85.66%) and NH₄⁺ (66.87%), these species could indicate secondary nitrate. The contribution of source one to inorganic components was up to 64.01%. The correlation coefficient between the time series of secondary nitrate contribution and PM_{2.5} was 0.91 (p < 0.01), indicating that secondary nitrate was a major source of PM_{2.5}. Source two was secondary sulfate with an average contribution of 11.62%, of which the characteristic value of SO₄²⁻ is the highest (41.15%),



and the concentration contribution of $\rm NH_4^+$ in this source was also high, which was the second generation of sulfate process. Source three was incomplete combustion with an average contribution of 4.49%. The concentration contribution of BC is the highest (47.09%), which is mainly from various incomplete combustion processes. The above three types of sources correspond to factor 1 analyzed by PCA. The total contribution of these three types of sources accounts for up to 80.12%, which was 26.40% higher than factor 1 of PCA.

Source 4 was sea salt with an average contribution of 8.61%, of which the characteristic value of Mg^{2+} (67.07%) and Cl^{-} (59.69%) had high loadings and contained a certain amount of Na⁺ (24.73%). It is the common sea salt in the coastal area, which is in line with the actual situation in Nanjing. Source five was relevant with biomass burning, with an average contribution of 6.90%, which was identified by higher loadings of K⁺ and Na⁺. They were related to combustion activities. The above two types of sources correspond to factor 2 analyzed by PCA. The total contribution of these two types of sources accounts for up to 15.51%, which was 0.79% lower than factor 2 of PCA.

Source 6 was crustal dust, with high loading of Ca^{2+} . The average contribution of the crustal dust was 4.37%, corresponding to factor 3 analyzed by PCA, and 8.13% lower than that.

The comparison of PCA and PMF results showed that the types of main sources resolved by the two models were similar, both delivered the source of $PM_{2.5}$. The PMF result was more refined than that of PCA: the secondary nitrate and secondary sulfate was treated independently, the uncertainty of the mixing factor was reduced, and the average contribution of each source was obtained.

3.3.2 Source contributions at different stages

Based on the Grade-II limit (75 μ g m⁻³) of Ambient Air Quality Standards of China and visibility data, the PM_{2.5} concentration >75 μ g m⁻³ was defined as a haze process. PM_{2.5} and visibility changed from good to poor and reached 75 μ g m⁻³ < PM_{2.5} h concentration <115 μ g m⁻³ was classified as a haze generation process. When PM_{2.5} h concentration was> 115 μ g m⁻³, it was divided into the haze development process.



 $PM_{2.5}$ and visibility have improved from poor, and the $PM_{2.5}$ h concentration of $<\!115\,\mu g\,m^{-3}$ and finally less than 75 $\mu g\,m^{-3}$ was defined in the haze dissipation process. According to the above classification criteria, in this study, haze process was divided into five processes: clean, precipitation, haze generation, haze development, and haze dissipation.

As shown in Figure 4, the concentrations of NO₃⁻, NH₄^{+,} and SO₄²⁻ were increasing along with the haze developing, and decreased during the dissipation stage. Their concentrations in the haze development stage were 52.69, 23.00, and 18.34 μ g/m³, which were 3.1, 3.0, and 2.0 times than that in the clean condition. From the perspective of the proportion, the proportion of NO₃⁻ increased by 4.05% from the generation to the development, the proportion of SO₄²⁻ decreased by 4.54%, and the proportion of NH₄⁺ increased by only 0.22%. It indicated that the increase of NO₃⁻ dominated the development of haze pollution.

The concentrations of BC, Cl^- , K^+ , and Na_+ were 8.83, 2.92, 0.61, and 0.21 µg/m³, keeping highest values during the haze development stage, which were 2.4, 2.1, 3.6, and 2.6 times than that in the clean stage. Due to the higher humidity and less dust in the haze, the concentration of Ca^{2+} in the whole haze episode showed lower level. The concentration and proportion of Mg^{2+} were highest in the haze generation stage. The proportion of K^+ and BC continued to rise during the generation, development, and dissipation stages of haze, which implied that biomass burning may contribute to haze formation.

Figure 4 showed that the precipitation process can effectively remove pollutants. Compared with the clean process, the concentration of SNA had decreased by 50.44, 51.43, and 59.72%, respectively. Cl⁻ had the highest proportion in the precipitation process (6.74%), which was related to the source



of sea salt. It was worth mentioning that the concentration and proportion of Ca^{2+} were significantly higher than others in the precipitation process, which may be due to rainwater being sucked into the inlet of the instrument. According to previous studies (Zhang et al., 2015), the Ca^{2+} content in precipitation in Nanjing is high.

The mass concentration and proportion of each source of $PM_{2.5}$ varied at different haze stages. The source of $PM_{2.5}$ was the primary secondary nitrate, which accounted for 56.56–73.23%. The secondary nitrate concentration in the haze development stage was the highest with 79.47 µg/m³, 3.5 times than that in clean stage. The concentration of secondary sulfate was the highest value 11.43 µg/m³ in the haze generation, which was 1.7 times than that of the clean stage. In the haze's development and dissipation process, the concentration of the secondary sulfate decreased first and then increased.

With the generation, development, and dissipation of haze, the concentration and proportion of incomplete combustion

kept increasing. The concentration during the dissipation phase was $9.56 \,\mu\text{g/m}^3$, which was $9.5 \,\text{times}$ than that in the clean process. Both sea salt and biomass burning had the highest concentrations during the haze development stage, 7.66 and $7.17 \,\mu\text{g/m}^3$, respectively, which were 1.8 and 3.3 times than that in the clean process. The concentration and proportion of crustal dust in the different stages of haze were all lower than that in the clean stage.

During the precipitation phase, the mass concentration and proportion of different sources have changed significantly. The concentration of secondary sulfate had decreased by 65.41% compared to the clean phase. The sea salt contribution in the precipitation process had reached the largest, up to 20.36%. Biomass combustion had the lowest contribution (2.04%) in the precipitation process. The concentration and proportion of crustal dust in the precipitation process were significantly larger than in other stages, which may be related to the measurement error of Ca^{2+} .

3.4 Impact of surrounding cities on nanjing

Figure 1 showed that the cluster of air masses in Nanjing during the study period can be divided into three directions: west (central Anhui province, 41.67%), northeast (northern Jiangsu province and the sea, 31.25%), and southeast (southern Jiangsu province, 27.08%). As shown in Figure 5, High PSCF values (>0.7) of PM_{2.5} loading was distributed in three pathways, which agreed with the cluster of air mass trajectories. The largest potential source area was located in the Suzhou-Wuxi-Changzhou Metropolitan area (SWC), southern Jiangsu, indicating that the source of PM_{2.5} was primarily from urban areas.

The potential source area of BC and aerosol number concentration (dN) were slightly different from $PM_{2.5}$, and the effect from the maritime pathway was low. The main source areas of BC were distributed over Central Anhui, northern Anhui, and SWC, and that of dN originated from central Anhui, SWC, and Shanghai. The potential source area distribution of SNA was similar to the three-pathway distribution of $PM_{2.5}$. The red zone size of all components showed remarkable difference, that demonstrating each component was controlled by the local emissions characteristic.

The results of the potential source area identified by CWT model were very similar to production of PSCF model (Figure 6). The total source area obtained by CWT was significantly larger than that of PSCF, and the distribution of the potential source area got smoother. The reason for these phenomena suggested that once the trajectory related pollutant concentrations at the measurement site was higher than the pollutant threshold value in PSCF, all points on the trajectory were treated as polluting points, so a thinner linear distribution of the source area would stand out. The CWT model can avoid this problem, there was no specific pollution threshold, the impact of each trajectory on the observation points is equal, and the distribution of source areas with different levels of contribution can be quantified.

As shown in Figure 6, the highest CWT values for PM 2.5 originating from SWC appeared, and another strong source area located in the northeast of Nanjing. The distribution of the strong source areas of different components in PM $_{2.5}$ was different. The largest source area of BC existed in the west of Nanjing, while dN was located in the southeast (SWC and Shanghai). The area of strong potential source area of SO₄²⁻ was very large, which meant all of the Nanjing surrounding areas provided contributions. As the primary components of PM_{2.5}, the strong potential source area of NO₃⁻ and NH₄⁺ and PM_{2.5} showed the most similar.

3.5 Impact of nanjing on surrounding cities

To explore the impact of Nanjing pollutant transport on its surrounding regions, PSCF (Figure 7) and CWT (Figure 8)

models were performed using the 24 h forward trajectory data and pollutant concentration data. The PSCF results showed the main impact area of $PM_{2.5}$ existed in central Anhui and southern Jiangsu, and the strongest impact area of BC was located in the southeast of Nanjing (SWC). High PSCF values of dN were found in northern Anhui. The distribution of the SNA was similar, located in the north and central regions of Jiangsu and Anhui.

The impact area produced by CWT was larger than that of PSCF, but the main scope with high CWT values was similar. The strong impact area contributed by $PM_{2.5}$ was mainly from the west to the southeast around Nanjing. The main impact areas of BC were located in Jiangsu Province (Huaian and SWC). The impact area of dN was smaller than that of other pollutants and mainly distributed in the north of Anhui, northwest of Nanjing. The distribution of the SNA's strong impact range was similar to $PM_{2.5}$, mainly from northwest to east in the vicinity of Nanjing.

Overall, the air masses transport from Nanjing, on the farthest northward possibly impact on the Yellow Sea region, while on the farthest southward may affect the northern region of Taiwan. The most important impact area was located in the southeast and northwest areas around Nanjing (SWC and north central Anhui).

4 Conclusion

To explore the source variations of PM2.5 at different stages on regional haze pollution event and the regional pollutant transport in urban agglomeration, receptor models (PCA, PMF, PSCF, and CWT) were employed to analyze the sources of haze in Nanjing on November 16-27, 2018. PCA showed that the main sources of PM2.5 were a mixture of secondary sources and incomplete combustion, a mixture of sea salt and biomass burning, and crustal dust. PMF results were more refined than PCA. Secondary nitrate and secondary sulfate were separated in the secondary sources, and the average contribution from each source was obtained: secondary nitrate (64.01%), secondary sulfate (11.62%), incomplete combustion (4.49%), sea salt (8.61%), biomass burning (6.90%) and crustal dust (4.37%). According to the different stages of the haze, the measurement was divided into clean, precipitation, haze generation, haze development, and haze dissipation. In the haze development stage, the concentrations of SO42-, NO3-, NH_{4}^{+} , and BC were 18.34, 52.69, 23.00, 8.83 μ g/m³, respectively, occupied the highest value during the whole pollution event, which were 2.0, 3.1, 3.0, and 2.4 times higher than the clean stage. The increase of NO3- dominated the development of haze, and the proportion of NO3- from haze generation to development increased by 4.05%. The concentration contributions of secondary nitrate, sea salt, and biomass burning were highest in haze development, secondary sulfate was highest in haze generation, and incomplete combustion was highest in haze dissipation, which was 3.5,

1.8, 3.3, 1.7, and 9.5 times higher than that in the clean stage. The crustal dust contribution in different haze stages was all lower than that in the clean stage. The potential source areas of $PM_{2.5}$ in Nanjing were distributed in three pathways: west, northeast, and southeast (central Anhui, northern Jiangsu province, and the sea, SWC). The main potential source areas of SNA, BC, and dN were in SWC, southeast of Nanjing. The impact area of air pollutants in Nanjing can reach Taiwan and Shandong, and the most important impact area was in the southeast and northwest areas around Nanjing.

Data availability statement

The raw data supporting the conclusion of this article will be made available by the authors, without undue reservation.

Author contributions

Conceptualization: AL and YD Methodology: Writing–Original Draft: AL, YD, XK, and TZ Methodology, Software: AL, YD, QH, XK, and MY Writing–Review and Editing: YD Data curation, Resources: YC and SS. Funding acquisition, Supervision: TZ Project administration: TZ.

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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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