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Levels and sources of hourly PM$_{2.5}$-related elements during the control period of the COVID-19 pandemic at a rural site between Beijing and Tianjin

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HIGHLIGHTS

- The effects of COVID-19 and its control were studied for a rural site in Xianghe.
- PM$_{2.5}$-related elements were hourly measured during our study.
- Dust, coal combustion and industrial sources decreased during the control period.
- Beyond expectation, vehicle emissions increased during the control period.
- Lowest reductions of industrial emission occurred for air masses from northeast.

**GRAPHICAL ABSTRACT**

ABSTRACT

To control the spread of the novel coronavirus disease 2019 (COVID-19) in China, many anthropogenic activities were reduced and even closed on the national scale. To study the impact of this reduction and closing down, hourly concentrations of PM$_{2.5}$-related elements were measured at a rural site before (12–25 January 2020), during (26 January–9 February 2020) and after (22 March–2 April 2020) the control period when all people remained socially isolated in their homes and could not return to economic zones for work. Nine major sources were identified by the positive matrix factorization model, including firework burning, coal combustion, vehicle emissions, dust, Cr industry, oil combustion, Se industry, Zn smelter, and iron and steel industry. Before the control period, K, Fe, Ca, Zn, Ba and Cu were the main elements, and fireworks burning, Zn smelter and vehicle emissions provided the highest contributions to the total element mass with 55%, 12.1% and 10.3%, respectively. During the control period, K, Fe, Ba, Cu and Zn were the dominating elements, and fireworks burning and vehicle emissions contributed 55% and 27% of the total element mass. After the control period, Fe, K, Ca, Zn and Ba were the main elements, and dust and iron and steel industry were responsible for 56% and 21% of the total element mass. The increased contribution from vehicle emissions during the control period could be attributed to our sampling site being near a town hospital and the fact that the vehicle activities were not restricted. The source apportionment results were also related to air mass backward trajectories. The largest reductions of dust, coal
1. Introduction

The recent outbreak of the coronavirus disease 2019 (COVID-19) became a global pandemic impacting over 200 countries with over 7 million confirmed cases until 10 June 2020 (WHO, 2020). As the first country fighting COVID-19, China has implemented a series of control measures, including stay-at-home orders, closing non-essential businesses, and regional lockdown, etc. On 23 January 2020, the government of Wuhan announced a shutdown and other cities of Hubei announced a lockdown on 24 January 2020, which included actions such as quarantining, traffic restrictions, and factory closures. On 25 January 2020, all provinces in mainland China except Tibet launched the highest level of emergency response. Additionally, most provincial governments in China allowed non-essential businesses to resume work after 10 February (Zhang et al., 2020), while Wuhan began to cancel control measures and resume the production and common living order on 8 April (Hubei COVID-19 Headquarter, 2020).

During the control period of the pandemic, a series of measures was undertaken to reduce gatherings of people. Most transportation was prohibited and almost all avoidable outdoor anthropogenic activities stopped all around the country. The Spring Festival holiday was extended beyond 10 February, and only essential enterprises involving people’s immediate needs (e.g., health care, providing food) were allowed to operate. All non-essential factories were shut down, and all schools, entertainment venues and restaurants were closed. This lockdown enforced restrictions and self-quarantine measures resulted in a dramatically reduced number of vehicles on the road and a near total reduction in factory production. The flow of commercial trucks and buses in the Beijing-Tianjin-Hebei region and its surrounding areas decreased by 77% and 39%, respectively, during the control period (Y. Wang et al., 2020). P. Wang et al. (2020) found that decreased anthropogenic emissions contributed to a decrease of PM$_{2.5}$ mass concentrations due to suspension of transportation and industry. The variations of air pollution during the control period provide an opportunity for better understanding the air quality improvement when significant restrictions in emissions from many sources are implemented.

To date, some studies have found that strict quarantine measures to control COVID-19 improved China’s air quality in the short term (P. Wang et al., 2020; Y. Wang et al., 2020; Xu et al., 2020; Zhang et al., 2020). Compared with before the control period, the concentrations of NO$_2$, CO, PM$_{2.5}$, PM$_{10}$ and SO$_2$ decreased during the control period, while the concentrations of O$_3$ increased. However, the variations and contributions of emission sources during this period are still rarely reported (Li et al., 2020; Zheng et al., 2020), in particular for rural areas. Trace elements can act as markers for specific source categories of PM$_{2.5}$ (Cui et al., 2019; Y. Liu et al., 2019). For example, arsenic (As) is used to indicate coal combustion, Cr is a tracer of industry (Cui et al., 2019). To study the variations of the PM$_{2.5}$ sources, hourly PM$_{2.5}$-related elements were measured at a rural site between Beijing and Tianjin in this study. The impact of regional transportation on PM$_{2.5}$ was analyzed by combining the backward trajectories with the results of source apportionment. This study provides useful information for other countries to assess the impact of COVID-19 on the environment and can be used as a reference for future research on improving air quality.

2. Methodology

2.1. Sampling site

The measurement period is from 12 January to 2 April 2020 at the Xianghe Atmospheric Integrated Observatory (39.75°N, 116.96°E) at a height of 8 m. Xianghe is a small county with 0.33 million residents, and it is located in a major plainlike area 50 km southeast of Beijing and 70 km north of Tianjin (Fig. 1). As shown in Fig. 1(b), a town hospital and a road are located at approximately 100 m to the northeast and east of the sampling site, respectively.

2.2. Instrumentation

The hourly PM$_{2.5}$ mass concentration was measured using a synchronized hybrid ambient real-time particulate monitor (model 5030,
Thermo-Fisher, USA) in rural Xianghe. The hourly-mean PM$_{2.5}$ mass concentrations in "2 + 26" cities in Beijing-Tianjin-Hebei and its surrounding areas, including Beijing, Tianjin and seven cities (Jinan, Dezhou, Liaocheng, Bengzhou, Zibo, Jining, Heze) in the Shandong Province, eight cities (Langfang, Tangshan, Shijiazhuang, Handan, Baoding, Cangzhou, Hengshui, Xingtai) in the Hebei Province, four cities (Taiyuan, Zhangji, Jincheng, Yangquan) in the Shanxi Province and seven cities (Zhengzhou, Anyang, Kaifeng, Jiaozuo, Xining, Hebi and Puyang) in the Henan Province, were downloaded from the China National Environmental Monitoring Center (CNEMC) (available at http://106.37.208.233:20035/). The locations of the "2 + 26" cities and Xianghe are shown in Fig. S1. Elements in PM$_{2.5}$ were measured with 1-h time resolution by an Xact 625 Ambient Metals Monitor (Cooper Environmental Services (CES), Beaverton, OR, USA). This instrument is a sampling and analyzing X-ray fluoresce spectrometer designed for the semi-continuous measurement of elements in aerosols; it can measure 23 elements, i.e., K, Ca, V, Cr, Cu, Zn, Ga, As, Mn, Fe, Co, Ni, Se, Ag, Cd, Sn, Sb, Au, Hg, Ba, Ti, Pb, and Pd. More information on the instrument, as used by us, is given in Cui et al. (2019). For Co, Sn, Sb, and Ti, more than 85% of the data were below the method determination limit (MDL) and these elements were therefore excluded from the further analysis. More than 70% of the hourly concentrations of Ga, Ag, Cd and Au were below 3 × MDL and these were also excluded in this study. V was retained as it is an important tracer of oil combustion (Liu et al., 2018; Wang et al., 2018). The Xact 625 has already been used in several studies (Chang et al., 2018; Gao et al., 2016; Phillips-Smith et al., 2017; Tremper et al., 2018). Due to malfunction of the instrument, data from the period of 11 February to 21 March are not available. The details about quality assurance (QA)/quality control (QC) can be found in the Supporting Information.

Additionally, the concentrations of SO$_2$ and NOx (NOx = NO + NO$_2$) were measured by a pulsed-fluorescence analyzer (model 43i, Thermo, USA) and a chemiluminescence analyzer (model 42i, Thermo, USA), respectively. Wind speed (WS) and wind direction (WD) were measured with an automatic weather station.

2.3. Positive matrix factorization (PMF) model for source apportionment

The PMF model is a widely used receptor model to resolve pollution sources and quantify the source contributions to the particulate matter (Gao et al., 2016; Wang et al., 2019; Zíková et al., 2016). The US EPA PMF5.0 model was employed to quantify the mass contributions and the source profiles of the elements in PM$_{2.5}$ in this study (Liu et al., 2018). The error fraction in the PMF analysis was set to 10% for the elements (Khan et al., 2016; Li et al., 2017). The detailed information about the calculation of the uncertainties of the measured elements can be found in the Supporting information. The uncertainties of the results from the PMF analysis were evaluated by displacement (DISP) and bootstrap (BS) (Paatero et al., 2014).

2.4. Air mass backward trajectory analysis

The 48-h backward trajectories arriving at the study site at a height of 500 m were calculated every hour using the Global Data Assimilation System meteorological data set (ftp://arlftp.arlhq.noaa.gov/pub/archives/gdas1). Cluster analysis was performed using the GIS-based software named TrajSat (Wang et al., 2009).

3. Results and discussion

3.1. Concentrations of PM$_{2.5}$ mass during the study period

The time-series of hourly PM$_{2.5}$ mass concentrations in Xianghe and Beijing, Tianjin, Tangshan and Langfang in Hebei province during the study period are shown in Fig. 2. The PM$_{2.5}$ mass shows a similar variation in Xianghe and Beijing, Tianjin, Langfang, and Tangshan, indicating that the pollution of PM$_{2.5}$ is regional (Table S1). The hourly PM$_{2.5}$ mass concentrations in Xianghe varied from 2.2 μg/m$^3$ to 450 μg/m$^3$, with an average value of 76 ± 80 μg/m$^3$. The average values of the PM$_{2.5}$ mass concentrations (μg/m$^3$) in Xianghe and "2 + 26" cities in Beijing-Tianjin-Hebei and its surrounding areas during the study period are given in Table 1. The average concentrations of the PM$_{2.5}$ mass in Xianghe are higher than in Beijing, Tianjin and most cities in the Hebei, Shanxi, Shandong and Henan provinces, while they are lower than in Shijiazhuang, Handan, Baoding, Xingtai in the Hebei province and in Anyang, Kaifeng and Puyang in the Henan province.

From 10 February 2020 on, except for Hubei, some industrial activities and businesses began to resume work in some provinces, including Beijing, Tianjin and Hebei. The news reported that the percentage of work resumption of large industrial enterprises in Beijing reached about 90% on 16 March 2020 (China News, 2020). According to governmental intervention, the daily transportation index provided by the Baidu migration dataset (Baiduqianxi, 2020) and conditions of resuming work of some industries and businesses in Beijing, Tianjin and Langfang, three periods were selected to better illustrate the source emission changes: before the control period of the pandemic (P1: 12–25 January 2020), during the control period of the pandemic (P2: 26 January–9 February 2020) and after the control period of the pandemic (P3: 22 March–2 April 2020). It is worth noting that the average...
PM$_{2.5}$ mass concentration in Xianghe was higher during P2 than during P1, which is similar to the variations in Langfang and Beijing (Table 1). In contrast, for all other cities of Table 1 the concentration of the PM$_{2.5}$ mass was lower during P2 than during P1. The different variations of the PM$_{2.5}$ mass concentrations during P1 and P2 can be attributed to meteorological conditions and increased secondary formation (Huang et al., 2020; P. Wang et al., 2020).

### 3.2. Element characteristics of the three selected periods

#### 3.2.1. P1 before the control period

The time series of the concentrations of the hourly elements in PM$_{2.5}$ during the study period is shown in Fig. 3. K, Cu, Ba, and Pb exhibited higher concentrations during the period from 24 January, 18:00 to 25 January, 12:00. 24 January is the Chinese New Year Eve, intensive fireworks burning during the Spring Festival (24 January – 9 February 2020) have great effects on the concentrations of K, Cu, Ba, and Pb, especially on the three nights of Chinese New year eve, Lunar Fifth Day, and the Lantern Festival. Major oxidizers in the fireworks are potassium compounds such as KNO$_3$, KClO$_3$, KClO$_4$, K$_2$CrO$_4$, and K$_2$Cr$_2$O$_7$ (Azhagaran et al., 2011). Ba compounds are commonly used to generate green flames and are widely used as coloring agent and stabilizer (Zhang et al., 2017). Cu can generate blue flames and is used as a catalyst (Kong et al., 2015). Pb as an oxidant (PbO$_2$) can help to achieve a steady and reproducible burning rate (Kong et al., 2015; J. Liu et al., 2019). Other elements showed a different variation; Mn, Fe, Zn, As, Se also exhibited slightly higher concentrations during the period of 15 to 19 January, which was related with higher contributions from different industries. Except for the effects of fireworks burning, some industries also affected the concentrations of Pb; Pb also showed higher concentrations during the period of 15 to 19 January. Cu also presented a higher concentration during the non-firework burning period; this could be attributed to the effects from vehicle emissions during travel before the Spring Festival (Zhu et al., 2018).

#### 3.2.2. P2 during the control period

During P2, K, Cu, Ba, and Pb still exhibited higher concentrations during the period of 26 to 29 January 2020 (Fig. 3). This could reflect aging processes after the emission of fireworks burning on the Chinese New Year Eve night (Kong et al., 2015). The occurrences of peak values of K, Cu, Ba, and Pb on 29 January, 23:00 and 8 February, 22:00 were caused by fireworks burning, which occurred on the Lunar Fifth Day and the Lantern Festival, respectively. From 30 January 2020 on, almost all elements, with the exception of V and Ni, showed a very lower concentration, indicating that the control measures during the control period of the pandemic were effective. Table 2 gives the average concentrations and associated standard deviations for the various elements during the three periods. Compared with P1, the concentrations of Ca, Cr, Mn, Fe, Zn, As, and Se also decreased during P2, which could be caused by the control measures conducted for fighting COVID-19, including social distancing, stay-at-home orders, closing non-essential businesses, and regional lockdown. 10 February 2020 was the first day of resuming work for many provinces, including Beijing, Tianjin and Hebei; as a result, the concentrations of some elements like Cu, Cr, Zn, and Pb began to increase.

#### 3.2.3. P3 after the control period

During P3, many industrial activities and businesses already resumed work. The percentage of work resumption of large industrial enterprises in the Hebei province reached 92% on 1 March 2020 (Hebei News, 2020); the percentage of work resumption of large industrial enterprises in Beijing reached about 90% on 16 March 2020 (China News, 2020); the percentage of work resumption of large industrial enterprises in Beijing reached about 90% on 16 March 2020 (China News, 2020).
3.3. Sources of elements in PM$_{2.5}$

In this study, 13 of the measured elements were used for the PMF analysis. The whole dataset including P1, P2 and P3 was used for the PMF runs. We also ran the PMF model separately for each of the three periods and compared the results with those for the whole dataset. Solutions with from five to eleven factors solutions were examined for the whole dataset; the 9-factor solution provided the most reasonable source profiles. The detailed comparison and analysis is given in the Supporting information. The relations between the predicted and measured concentrations of the elements are given in Table S2. With the exception of V, the coefficients of determination ($r^2$-values) between the measured and predicted concentrations of the species for the 9-factor PMF solution are all higher than 0.90. The uncertainties of the PMF results for the 9-factor solution were evaluated by DISP and BS. There were no swaps in the DISP runs and more than 98% of the BS runs could be mapped to base runs, indicating that the PMF analysis results were stable and reliable. The nine factors were identified as fireworks burning, coal combustion, vehicle emissions, Se industry, Cr industry, Zn smelter, oil combustion, dust, and iron and steel industry. Figure 4 presents the source profiles identified by the PMF model, while the time series of the nine factors for the measurement period are shown in Fig. S5. The polar plots and diurnal variations of the pollution source concentrations during the study period are shown in Figs. 5 and 6, respectively. The average concentrations and contributions of the nine sources during the three periods are given in Table 3.

Factor 1 was identified as fireworks burning, because it has high loadings of K, Cu, Ba and Pb and these elements are reliable tracers of fireworks burning (Kong et al., 2015; J. Liu et al., 2019). The time series of this factor also showed three main peaks during the Chinese New Year Eve, the lunar fifth day and the Lantern Festival, respectively (Fig. S5). As seen in Fig. 5, the higher contribution of this factor occurred with WS less than 1 m/s and fireworks burning was a local source. The fireworks burning showed higher concentrations during the nighttime (20:00–5:00), which was consistent with the occurrence of the fireworks burning (Fig. 6). This factor contributed overall 40% of the total element mass in PM$_{2.5}$. During P1 and P2, fireworks burning was responsible for 55% of the total elements mass.

Factor 2 was considered as coal combustion; in this factor, arsenic (As) has the higher loading. As is reliable tracer of coal combustion (Cui et al., 2019). This factor contributed overall 21% of the total element mass. Although the concentration of SO$_2$ is also affected by the fireworks burning (Huang et al., 2012), coal combustion was somewhat positively correlated with SO$_2$ ($r^2 = 0.14$). As shown in Fig. 5, higher concentrations of coal combustion occurred with WS less than 2 m/s and higher than 4 m/s, indicating that both local emissions and regional transport played a role. The concentrations of coal combustion during the pre-control period, control and post-control periods were 104 ng/m$^3$, 40 ng/m$^3$ and 123 ng/m$^3$, respectively (Table 3). The Kruskal-Wallis ANOVA test showed that there are significant differences for coal combustion during the three periods (Kruskal-Wallis chi-squared = 57, df = 2, p-value < 0.05). With the implementation of the project “Coal to Gas” in Xianghe, coal combustion is mainly used for industry (Zhao et al., 2020). The lower concentration of coal combustion during the control period indicates that the control of coal combustion was effective.
Factor 3 was linked to vehicle emissions, and it was characterized by high loadings of Cu (34%) and Zn (6.5%). According to the emission inventories of trace elements by Zhu et al. (2018), more than 55% of the Cu and about 10% of the Zn emissions originate from brake and tire wear. Cu is mainly used in lubricants and in friction materials that constitute major contents of brake linings, while Zn is added to the tire tread to facilitate the vulcanization process (Lin et al., 2015). The concentrations of NOx are also affected by the combustion of natural gas for heating in Xianghe (Zhao et al., 2020) in addition to the contribution from vehicle emissions; a moderate correlation between vehicle emissions and NOx was found ($r^2 = 0.40$). As shown in Fig. 5(c), the high contributions of vehicle emissions occurred with wind speeds of approximately 1 m/s. Thus, vehicle emissions were significantly affected by local emissions. Traffic activities on the road near the sampling site (approximately 100 m) were the main source of vehicle emissions. The diurnal variation for vehicle emissions (Fig. 6) shows morning and evening peaks and confirms the reasonability of this source identification. This factor contributed 13.3% of the total element mass during the measurement period. Across the pre-control period, the control period, and the post-control period, the mean concentrations from vehicle emissions were 430 ng/m$^3$, 930 ng/m$^3$ and 134 ng/m$^3$, respectively ($\text{Kruskal–Wallis chi-squared} = 360, \text{df} = 2, p$-value < 0.05). On 25 January 2020, all provinces except Tibet launched the highest level of emergency responses. In Xianghe, the government also implemented
a series to control the spread of the disease on 25 January 2020. Surprisingly, the concentrations of vehicle emissions did not decrease like was the case in Beijing and Tianjin according to the transportation index (Baiduqianxi, 2020). Possible reasons for this are: (1) our sampling site was located in a rural area in Xianghe, the vehicle emissions were not restricted strictly in the rural areas; thus, vehicle emissions still had higher contributions from 26 to 28 January 2020 (Fig. S5). From 28 January 2020 on, vehicle emissions showed decreased levels and exhibited lower concentrations during the period of 1 to 8 February 2020. On 9 February 2020, the increased vehicle emissions could be ascribed to some people going back to the city for work. The traffic flow within and between cities increased significantly with the resumption of work and production (MEP, 2020). (2) A town hospital is located at approximately 100 m to the east of our sampling site; the roads to the hospital were not restricted during the control period of COVID-2019. After the control period, lower contributions of vehicle emissions were observed due to the end of the Spring travel rush and relief of the pandemic; this then reflected the general condition of traffic flow on normal days near the sampling site. The road near the sampling site is not the main road in Xianghe and the traffic flow was not heavy.

Factor 4 was identified as dust because it has high loading of Ca, Fe and Ba (Gao et al., 2016). Higher concentrations of dust came from the southern and northwestern areas of the sampling site with WS higher than 4 m/s. The average concentrations of dust during P1, P2 and P3 were 410 ng/m³, 45 ng/m³ and 2600 ng/m³, respectively (Kruskal-
Wallace et al., 2013. Amato et al. (2010) also reported a PMF factor with high concentrations of Zn, Pb and Mn and attributed it to industrial emissions, whereas a similar factor was attributed more to local workshops and ateliers in the city of Zurich (Richard et al., 2011). Higher concentrations of this factor were found for both low and high WS, thus both local and regional transportation contributed to it. Across the three periods, the source contributions were 510 ng/m³, 220 ng/m³ and 300 ng/m³, respectively (Kruskal-Wallis chi-squared = 175, df = 2, p-value < 0.05).

Factor 6 was identified as Zn smelter. According to the emission inventory of the Se in the Beijing-Tianjin-Hebei region in 2012, non-metallic mineral manufacturing accounts for 52% of the Se emission (Zhu et al., 2018). Glass production contributes the largest fraction of the Se emission (about 51%) because of the widespread application of selenium powder as decolorizing agent in the glass production process (Zhu et al., 2018). Hebei province is the largest producer of plate glass in China, accounting for about 19.9% of the national output (Zhu et al., 2018). As shown in Fig. 5, higher concentrations of this factor occurred with WS at 1–4 m/s for winds from western and southeastern areas of the sampling site. The contributions of the Se industry during the three periods (pre-control, control, post-control) were 135 ng/m³, 77 ng/m³, and 71 ng/m³, respectively (Kruskal-Wallis chi-squared = 17.5, df = 2, p-value < 0.05).

Factor 7 has high loadings of V and Ni. These two elements are good indicators of oil combustion sources and thus this factor was identified as oil combustion. V and Ni are abundant in crude oil and primarily retained in fuel oil-related industrial products (i.e., marine heavy oil, industrial heavy oil, asphalt, and petroleum coke) during the oil refinery process (Liu et al., 2018). Heavy oil and petroleum coke are widely used as power fuels in petrochemical plants, marine heavy oil, small power plants, and construction industries. The concentrations of oil combustion during P1, P2 and P3 are 47 ng/m³, 71 ng/m³, and 72 ng/m³, respectively (Kruskal-Wallis chi-squared = 15.6, df = 2, p-value < 0.05).

Factor 8 was identified as Cr-related industry emissions, with a high loading of Cr (92%). Cr is widely used in electroplating (additive metal surface treatment) and leather (as a chrome tanning agent used in the tanning process) industries (Liu et al., 2018). Specific metal processing industries have also been reported to be associated with Cr emissions (Queiroz et al., 2007). The higher concentrations of the Cr industry originated from the southeastern areas of the sampling site with WS at 1–6 m/s (Fig. 5). The concentrations of Cr industry during the pre-control period (P1), control (P2) and post-control periods (P3) were 12.0 ng/m³, 4.5 ng/m³ and 19.9 ng/m³, respectively (Kruskal-Wallis chi-squared = 127, df = 2, p-value < 0.05).

Factor 9 was identified as iron and steel industry because of high loadings for Mn, Fe and Zn. Iron and steel production are among the most important industries in the Beijing-Tianjin-Hebei region (National Bureau of Statistics, 2017). The sintering process in iron and steel industries produces large amounts of heavy metals: Zn, Mn and Fe (Duan and Tan, 2013). According to the polar plot of this factor (Fig. 5), higher concentrations occurred for winds from the southeastern area of the sampling site with WS higher than 4 m/s. There are a lot of iron and steel plants in southeastern areas of the sampling site, like Tianjin and Tangshan (Zhu et al., 2018). The concentrations of iron and steel industry during the pre-control period, control and post-control periods were 240 ng/m³, 147 ng/m³ and 970 ng/m³, respectively (Table 3) (Kruskal-Wallis chi-squared = 360, df = 2, p-value < 0.05).

As can be deduced from the concentration data in Table 3, compared with P1, the dust concentration during P2 decreased by 89%, followed by decreases for the Cr industry (62%), coal combustion (62%), Zn smelter (58%), Se industry (43%) and iron and steel industry (40%). In contrast, the concentrations from vehicle emissions and oil combustion increased by 115% and 53%, respectively.

3.4. Air mass backward trajectories during the three periods

During the study period (P1, P2 and P3), the air masses were classified into five clusters (Fig. 7(a)). Source contributions (ng/m³) of eight factors for each trajectory cluster during the three periods (P1, P2 and P3) are also shown in Fig. 7(b). The air masses for Cluster 1 (C1) originated in Mongolia, passed over Xilingole League in Inner Mongolia, Chengde in Hebei; this cluster made up for 26.5% of the air masses. The air masses for Clusters 2 (C2) and 4 (C4) came from northwest areas of the sampling site; these clusters accounted for 16.7% and 24.6%, respectively, of the air masses. The transport distance was longer for C2 than for C4; the air masses for C2 originated from Siberia, while those for C4 started from Mongolia; the air masses of both clusters passed Ulanqab in Inner Mongolia and Zhangjiakou in the Hebei province before reaching Beijing. For Cluster 3 (C3), before reaching Beijing, the air masses passed Chengde and Tangshan in the Hebei province, and Tianjin. Tangshan and Tianjin are industrial cities in China. The contributions from coal combustion, Cr industry, Se industry, Zn smelter as well as from iron and steel industry were relatively larger for C3 than for the other clusters (Fig. 7(b)). The air masses for Cluster 5 (C5) came from southwestern areas of the sampling site; they successively passed Suqian and Xuzhou in the Jiangsu province, Jining and Liaocheng in the Shandong province, and Hengshui in the Hebei province. The passed areas of C5 are characterized by higher emissions from industrial activities (Zhu et al., 2018). It is worth noting that C5 only occurred during P3 after the control period (22 March–2 April 2020); its air mass trajectories accounted for 8.7% of all trajectories. During P3, in addition to C3, C5 also had higher contributions from industrial activities, which is consistent with air mass transport from the industrial regions in the southwestern Hebei province.

We examined the difference of contributions of eight sources during the three periods (P1, P2 and P3) for different clusters. Compared to P1, coal combustion showed almost 100% decrease for C1, C2 and C4 during P2, while the concentration of coal combustion for C3 only decreased by 24.9%. The lower reduction of coal combustion for C3 could be ascribed to no shutdown of some coal combustion related industries; note that lower reductions of some industrial sources were also observed for C3. Due to the stop of many construction activities during P2, the contributions of dust decreased by 86%, 79%, 94% and 92% for C1, C2, C3 and C4, respectively. Other industrial activities also showed a decrease during P2. The contributions of the Zn smelter decreased by 81%, 84%, 22% and 95% for C1, C2, C3 and C4, respectively. The Cr industry contributions decreased by 47%, 41%, 65% and 77% for C1, C2, C3 and C4, respectively. The contributions from the Se industry decreased by 55%, 47%, 41% and 64% for C1, C2, C3 and C4, respectively. The iron and steel industry contributions decreased by 71%, 34%, 19.1% and 87% for C1, C2, C3 and C4, respectively. In contrast, the concentrations from the dust, coal combustion and industrial sources, the concentrations of oil combustion and vehicle emissions increased during P2. The concentrations of oil combustion increased by 44%, 85%, 45% and 34% for C1, C2, C3 and C4 during P2 when compared with P1. Although the concentrations of vehicle emissions increased in all the clusters during the control period, they were mainly affected by local emissions; the reasons were discussed in Section 3.2. The least decreased contributions of coal combustion and industrial activities (Zn smelter, Se industry, iron and steel industry) during P2 were noted for C3 when the air masses passed over Tangshan and Tianjin.
Compared with P2, the concentrations of coal combustion in the different air mass clusters were higher during P3 after the control period, indicating the resumption of some coal combustion related industrial activities. However, as shown in Fig. 7(b), the concentrations of coal combustion during P3 in C1, C3 and C4 were lower than those before the control period. This is reasonable due to the stop of heating activities from 15 March 2020 on. Higher contributions of dust during P3 in all clusters can be ascribed to frequent occurrence of sand storms in spring and the resumption of large construction activities. The contributions of industrial sources (Cr industry, Se industry, Zn smelter, iron and steel industry) in different air masses during P3 were slightly higher than during P2, which also reflects the resumption of those industrial activities. The concentrations of some industrial sources like Zn smelter in different clusters were still lower during P3 than those during P1, which indicated that these industrial enterprises did not work yet at their normal operational levels. Vehicle emissions were lower during P3 when compared with those during P1 and P2; this is ascribed to the end of Spring travel rush and the relief of the pandemic.

4. Conclusion

Hourly concentrations of PM$_{2.5}$ mass and PM$_{2.5}$ related elements were measured from 12 January to 2 April 2020 at a rural site in Xianghe between Beijing and Tianjin. The following conclusions can be drawn:

1. The average concentrations of the PM$_{2.5}$ mass during the study period in Xianghe were higher than those in most “2 + 26” cities in Beijing-Tianjin-Hebei and its surrounding areas. Similar to Beijing and Langfang, the PM$_{2.5}$ mass concentrations were higher during the control period when compared with before the control period.

2. During the three selected periods, fireworks burning had stronger effects on the concentrations of K, Cu, Ba and Pb before and during the control period of the COVID-19 pandemic due to the Chinese Spring Festival.

3. Nine types of sources contributing to the total elements in PM$_{2.5}$ were identified by PMF; they were fireworks burning, coal
This study provided detailed variations of some primary PM$_{2.5}$ sources during the control period of the COVID-19 pandemic based on the measurement of elements. In a further study, combining other high-time resolution PM$_{2.5}$ chemical composition datasets (OC/EC, water-soluble-ions, etc.) could lead to a better understanding of the variations of the PM$_{2.5}$ sources (both primary and secondary sources).

CRediT authorship contribution statement

Yang Cui: Investigation, Formal analysis, Writing - original draft.

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Declaration of competing interest

The authors declare that they have no known competing financial interests or personal relationships that could have appeared to influence the work reported in this paper.

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Appendix A. Supplementary data

Supplementary data to this article can be found online at https://doi.org/10.1016/j.scitotenv.2020.148040.

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