Co-Editor Decision: Publish subject to minor revisions (review by editor) (26 Jan 2019) by Luisa Molina

Comments to the Author:

Dear Authors,

I am pleased to accept your manuscript for publication after considering an additional comment from one of the referees, "In principle, all factors identified in the PMF analysis need to be independent of each other. However in the manuscript the 3rd factor is a derivative of other 4 factors. A more detailed clarification and/or explanation is needed."

I would appreciate if you could respond to this comment. Best regards, Luisa Molina

Response: Thanks very much for the editor's kind decision. Following your suggestion, we changed the 3rd factor of "secondary particle formation" to "secondary source", which should be independent from other primary sources identified in the manuscript. The Figure 4 and 5 were also modified accordingly.

Impacts of short-term mitigation measures on PM_{2.5} and radiative effects: case study at a regional background site near Beijing, China

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19 Abstract. Measurements at a background site near Beijing showed that pollution controls implemented

20 during the 19th National Congress of the Communist Party of China (NCCPC) were effective in reducing

21 PM_{2.5}. Mass concentrations of PM_{2.5} and its major chemical components were 20.6–43.1% lower during

22 the NCCPC-control period compared with a non-control period, and differences were greater on days

23 with stable meteorological conditions. A receptor model showed that PM_{2.5} from traffic-related emissions,

24 biomass burning, industrial processes, and mineral dust was 38.5-77.8% lower during the NCCPC-

 $25 \qquad \text{control versus non-control period, but differences in PM_{2.5} from coal burning were small, and secondary}$

26 source was higher during the control period. During one pollution episode in the non-control period,

27 secondary <u>source</u> dominated, and the WRF-Chem model showed that the Beijing-Tianjin-Hebei (BTH)

region contributed 73.6% of PM_{2.5} mass. A second pollution episode was linked to biomass burning, and

BTH contributed 46.9% of PM_{2.5} mass. Calculations based on IMPROVE algorithms showed that organic
 matter was the largest contributor to light extinction during the non-control period whereas NH₄NO₃ was

the main contributor during the NCCPC. The Tropospheric Ultraviolet and Visible radiation model

32 showed that the average direct radiative forcing (DRF) values at the Earth's surface were -14.0 and -19.3

33 W m⁻² during the NCCPC-control and non-control periods, respectively, and the DRF for the individual

34 PM_{2.5} components were 22.7–46.7% lower during the NCCPC. The information and dataset from this

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37 study will be useful for developing air pollution control strategies in the BTH region and for 38 understanding associated aerosol radiative effects.

39 1 Introduction

High loadings of fine particulate matter (PM_{2.5}, particulate matter with an aerodynamic diameter ≤ 2.5 40 41 μm) cause air quality to deteriorate (Pui et al., 2014; Tao et al., 2017), reduce atmospheric visibility 42 (Watson, et al., 2002; Cao et al., 2012), and adversely affect human health (Feng et al., 2016; Xie et al., 43 2016). Moreover, PM2.5 can directly and indirectly affect climate and ecosystems (Lecoeur et al., 2014; Tie et al., 2016). With the rapid increases in economic growth, industrialization, and urbanization in the 44 past two decades, Beijing has experienced serious PM2.5 pollution, especially in winter (e.g., Zhang et al., 45 2013; Elser et al., 2016; Wang et al., 2016a; Zhong et al., 2018). Since the Chinese government 46 promulgated the National Ambient Air Quality Standards for PM2.5 in 2012 (NAAQS, GB3095-2012), a 47 48 series of emission control strategies have been implemented in Beijing and surrounding areas to alleviate the serious air pollution problems. These measures include installing desulphurization systems in coal-49 50 fired power plants, banning high-emission motor vehicles, and promoting natural gas as an alternative to the China Environmental State Bulletin 51 coal in rural areas. According to (www.zhb.gov.cn/hjzl/zghjzkgb/lnzghjzkgb, in Chinese), the annual levels of PM_{2.5} during 2013–2016 in 52 Beijing showed a decreasing trend (r = 0.98 and slope = -5.3 µg m⁻³ year⁻¹), but there were still 45.9% of 53 days in 2016 that suffered from varying degrees of pollution. 54 55 Identifying the causes of air pollution in Beijing is challenging because the chemical composition of PM2.5

56 is variable and complex, and the particles originate from a variety of sources and processes. For example,

Elser et al. (2016) reported that organic aerosol (OA) was the largest contributor to PM_{2.5} mass during extreme haze periods in Beijing, and the primary aerosol from coal combustion (46.8%) was the dominant contributor to OA, followed by the oxygenated OA (25.0%) and biomass burning OA (13.8%). In addition, Zheng et al. (2016) found that organic matter (OM) was the most abundant component (18–60%) in PM_{2.5}, but its relative contribution usually decreased as pollution levels rose while those of secondary inorganic species (e.g., sulfate and nitrate) increased.

In recent years, the Chinese government has taken temporary control measures to ensure good air quality
 during some important conferences and festivals held in Beijing, including the 2008 Summer Olympic
 Games, the 2014 Asia-Pacific Economic Cooperation (APEC) summit, and the 2015 Victory Day parade
 (VDP). These actions provide valuable opportunities for evaluating the effectiveness of emission controls

on air pollution, and the information gathered during the control periods should useful for making policy
decisions. Numerous studies have demonstrated that temporary aggressive control measures were
effective in reducing primary pollutants and secondary aerosol formation in Beijing (e.g., Wang et al.,
2010; Guo et al., 2013; Li et al., 2015; Tao et al., 2016; Xu et al., 2017).

71 Air pollution in Beijing is not only influenced by local emissions and the regional transport of pollutants but also by meteorological conditions (e.g., Li and Han, 2016; Bei et al., 2017). In this regard, Zhong et 72 al. (2018) concluded that heavy pollution episodes in Beijing can be generally divided into two phases (1) 73 a transport stage, which is characterized by increases in pollutants mainly transported from the south of 74 Beijing and (2) an accumulation stage, during which there is dramatic growth in PM2.5 loadings due to 75 76 stagnant meteorological conditions. Moreover, several studies showed that the emission controls put in place during important events were effective in decreasing aerosol concentrations, but meteorological 77 78 conditions also played an important role in determining aerosol loadings (Gao et al., 2011; Liang et al., 79 2017). For example, Liang et al. (2017) found that meteorological conditions and emission control measures had comparable impacts on PM2.5 loadings in Beijing during the 2014 APEC (30% versus 28%, 80 respectively) and the 2015 VDP (38% versus 25%). 81

The existing studies on the effects of temporary air pollution controls in Beijing have not covered mid-82 autumn when meteorological conditions are typically complex and variable. Indeed, Zhang et al. (2018) 83 reported that two weather patterns common in October caused heavy pollution episodes in Beijing. One 84 episode was linked a Siberian high-pressure system and a uniform high-pressure field while the second 85 86 was associated with a cold front and a low-pressure system. For this study, measurements were made at a regional background site in the Beijing-Tianjin-Hebei (BTH) region to investigate the changes of PM2.5 87 during the 19th National Congress of the Communist Party of China (NCCPC), which was held in Beijing 88 from 18-24 October. Temporary control measures were implemented in Beijing and neighboring areas; 89 these included restrictions on the number of vehicles, prohibition of construction activities, and 90 restrictions on factories and industrial production. The primary objectives of this study were to (1) 91 investigate the effectiveness of emission control measures on PM2.5 and the associated changes in its 92 chemical composition; (2) determine the contributions of emission sources to PM_{2.5} mass during the 93 NCCPC-control and non-control periods; and (3) evaluate the impacts of reductions of PM2.5 on aerosol 94 direct radiative forcing (DRF) at the Earth's surface. The study produced a valuable dataset and the results 95 96 provide insights into how controls on air pollution can affect Beijing.

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99 2 Materials and methods

100 2.1 Sampling site

Intensive measurements were made from 12 October to 4 November 2017 at the Xianghe Atmospheric 101 102 Observatory (39.75° N, 116.96° E; 36 m above sea level) to investigate how the characteristics of PM_{2.5} and the associated radiative effects were affected by the controls put in place during the NCCPC. Xianghe 103 104 is a small county with 0.33 million residents, and it is located in a major plain-like area ~50 km southeast from Beijing and ~70 km north from Tianjin (Figure 1). The sampling site is surrounded by residential 105 areas and farmland, and it is ~5 km west of Xianghe city center. This regional aerosol background site is 106 influenced by mixed emission sources in the BTH region. A more detailed description of the site may be 107 found in Ran et al. (2016). 108

109 2.2 Measurements

110 2.2.1 Offline measurements

PM_{2.5} samples were collected on 47 mm quartz-fiber filters (QM/A; GE Healthcare, Chicago, IL, USA) 111 and Teflon® filters (Whatman Limited, Maidstone, UK) using two parallel mini-volume samplers 112 (Airmetrics, Oregon, USA) that operated at a flow rate of 5 L min⁻¹. The duration of sampling was 24 h, 113 and the sampling interval was from 09:00 local time to 09:00 the next day. To minimize the evaporation 114 of volatile materials, the samples were stored in a refrigerator at -4 °C before the chemical analyses. The 115 quartz-fiber filters were used for determinations of water-soluble inorganic ions and carbonaceous species 116 while the Teflon® filters were used for inorganic elemental analyses. The PM2.5 mass on each sample 117 filter was determined gravimetrically using a Sartorius MC5 electronic microbalance with $\pm 1 \mu g$ 118 119 sensitivity (Sartorius, Göttingen, Germany). For the mass determinations, the filters were equilibrated 120 under controlled temperature (20-23 °C) and relative humidity (35-45%) before the measurements were made. Field blanks (a blank quartz-fiber filter and a blank Teflon® filter) were collected and analysed to 121 account for possible background effects. 122

Water-soluble inorganic ions, including F⁻, Cl⁻, NO₃⁻, SO₄²⁻, Na⁺, K⁺, Mg²⁺, Ca²⁺, and NH₄⁺ were measured with the use of a Dionex 600 ion chromatograph (IC, Dionex Corp., Sunnyvale, CA, USA). The four anions of interest were separated using an ASII-HC column (Dionex Corp.) and 20 mM potassium hydroxide as the eluent. The five cations were separated using a CS12A column (Dionex) and an eluent of 20 mM methane sulfonic acid. More detailed description of the IC analyses may be found in Zhang et al. (2011). Carbonaceous species, including organic carbon (OC) and elemental carbon (EC)

were determined using a Desert Research Institute (DRI) Model 2001 thermal/optical carbon analyzer 129 (Atmoslytic Inc., Calabasa, CA, USA) following the Interagency Monitoring of Protected Visual 130 131 Environments (IMPROVE A) protocol (Chow et al., 2007). A standard sucrose solution was used to establish a standard carbon curve before the analytical runs. Replicate analyses were performed at a rate 132 133 of one sample for every ten samples, and the repeatability was found to be < 15% for OC and < 10% for EC. More information of the OC and EC measurement procedures may be found in Cao et al. (2003). 134 Thirteen elements were determined by energy-dispersive X-ray fluorescence (ED-XRF) spectrometry 135 (Epsilon 5 ED-XRF, PANalytical B.V., Netherlands), and these elements include Al, Si, K, Ca, Ti, Cr, 136 Mn, Fe, Cu, Zn, As, Br, and Pb. The analytical accuracy for ED-XRF measurements was determined with 137 138 a NIST Standard Reference Material 2783 (National Institute of Standards and Technology, Gaithersburg,

139 MD, USA). A more detailed description of the ED-XRF methods may be found in Xu et al. (2012).

140 2.2.2 Online measurements

The aerosol optical properties were determined using a Photoacoustic Extinctiometer (PAX, Droplet 141 Measurement Technologies, Boulder, CO, USA) at a wavelength of 532 nm. The PAX measured light 142 143 scattering (bscat) and absorption (babs) coefficients (in Mm⁻¹) simultaneously using a built-in wide-angle integrating reciprocal nephelometer and a photoacoustic technique, respectively. Before and during the 144 sampling, the PAX b_{scat} and b_{abs} were calibrated using ammonium sulfate and fullerene soot particles, 145 respectively, which were generated with an atomizer (Model 9302, TSI Inc., Shoreview, MN, USA). 146 Detailed calibration procedures have been described in Wang et al. (2018a; 2018b). For this study, the 147 148 PAX was fitted with a PM2.5 cutoff inlet, and the sampled particles were dried by a Nafion® dryer (MD-700-24S-1; Perma Pure, LLC., Lakewood, NJ, USA). The time resolution of the data logger was set to 1 149 150 minute.

One-minute average mixing ratios of NOx (NO + NO₂), O₃, and SO₂ were measured using a Model 42*i* gas-phase chemiluminescence NOx analyzer (Thermo Fisher Scientific, Inc., Waltham, MA, USA), a Model 49*i* photometric ozone analyzer (Thermo Fisher Scientific, Inc.), and a Model 43*i* pulsed UV fluorescence analyzer (Thermo Fisher Scientific, Inc.), respectively. Standard reference NO, O₃, and SO₂ gases were used to calibrate the NOx, O₃, and SO₂ analyzers, respectively, before and during the campaign. All the online data were averaged to 24 h and matched to the duration of the filter sampling.

157 2.2.3 Complementary data

Wind speed (WS) and relative humidity (RH) were measured with the use of an automatic weather station 158 159 installed at the Xianghe Atmospheric Observatory. Surface weather charts for East Asia were obtained from the Korea Meteorological Administration. The three-day backward in time trajectories and mixed 160 layer heights (MLHs) were calculated using the Hybrid Single-Particle Lagrangian Integrated Trajectory 161 (HYSPLIT) model (Draxler and Rolph, 2003), which was developed by the National Oceanic and 162 Atmospheric Administration (NOAA). The aerosol optical depth (AOD) was measured using a 163 164 sunphotometer (Cimel Electronique, Paris, France), and those data were obtained from the Aerosol Robotic Network data archive (http://aeronet.gsfc.nasa.gov). Fire counts were obtained from the 165 Moderate Resolution Imaging Spectroradiometer (MODIS) instruments on the Aqua and Terra satellites 166 167 (https://firms.modaps.eosdis.nasa.gov/map).

168 2.3 Data analysis methods

169 2.3.1 Chemical mass closure

170 The chemically reconstructed PM_{2.5} mass was calculated as the sum of OM, EC, SO₄²⁻, NO₃⁻, NH₄⁺, Cl⁻,

fine soil, and trace elements. A factor of 1.6 was used to convert OC to OM (OM = $1.6 \times OC$) to account

172 for those unmeasured atoms in organic materials based on the results of Xu et al. (2015). The mass

173 concentration of fine soil was calculated by summing the masses of Al, Si, K, Ca, Ti, Mn, and Fe oxides

using the following equation (Cheung et al., 2011):

175 [Fine soil] = $[Al_2O_3] + [SiO_2] + [K_2O] + [CaO] + [TiO_2] + [MnO_2] + [Fe_2O_3]$

 $176 = 1.89 \times [Al] + 2.14 \times [Si] + 1.21 \times [K] + 1.4 \times [Ca] + 1.67 \times [Ti] + 1.58 \times [Mn] + 1.43 \times [Fe](1)$

The mass concentration of trace elements was calculated as the sum of measured elements that were not used in the calculation of fine soil:

179 [Trace elements] = [Cr] + [Cu] + [Zn] + [As] + [Br] + [Pb]

As shown in Figure S1, the reconstructed $PM_{2.5}$ mass was strongly correlated (r = 0.98) with the

(2)

181 gravimetrically determined values, and this attests to the validity of the chemical reconstruction method.

182 The slope of 0.86 indicates that our measured chemical species accounted for most of the PM_{2.5} mass.

183 The difference between the reconstructed and measured PM_{2.5} mass was defined as "others".

184 2.3.2 Receptor model source apportionment

Positive matrix factorization (PMF) has been widely used in source apportionment studies in the past two decades (e.g., Cao et al., 2012; Xiao et al., 2014; Tao et al., 2014; Huang et al., 2017). The principles of PMF are described in detail elsewhere (Paatero and Tapper, 2006). Briefly, PMF is a bilinear factor model that decomposes an initial chemically-speciated dataset into a factor contribution matrix G_{ik} ($i \times k$ dimensions) and a factor profile matrix F_{kj} ($k \times j$ dimensions) and then iteratively minimizes the object function Q:

191
$$X_{ij} = \sum_{k=1}^{p} G_{ik} F_{kj} + E_{ij}$$
 (3)

192
$$Q = \sum_{i=1}^{m} \sum_{j=1}^{n} \left(\frac{\varepsilon_{ij}}{\sigma_{ij}} \right)^2$$
(4)

where X_{ij} is the concentration of the *j*th species measured in the *i*th sample; E_{ij} is the model residual; and σ_{ij} represents the uncertainty.

In this study, the PMF Model version 5.0 (PMF 5.0) from US Environmental Protection Agency (EPA) 195 196 (Norris et al., 2014) was employed to identify the PM2.5 sources. Four to nine factors were extracted to determine the optimal number of factors with random starting points. When the values of scaled residuals 197 198 for all chemical species varied between -3 and +3 and a small Q_{true}/Q_{expect} was obtained, the base run was considered to be stable. Further, bootstrap analysis (BS), displacement analysis (DISP), and bootstrap-199 displacement analysis (BS-DISP) were applied to assess the variability and stability of the results. A more 200 detailed description of the methods for the determination of uncertainties in PMF solutions can be found 201 in Norris et al. (2014). 202

203 2.3.3 Regional chemical dynamical model

The Weather Research and Forecasting model coupled to chemistry model (WRF-Chem) is a 3-D online-204 coupled meteorology and chemistry model, and it was used to simulate the formation processes that led 205 to high PM2.5 loadings after the NCCPC. The WRF-Chem uses meteorological information, including 206 clouds, boundary layer, temperature, and winds; pollutant emissions; chemical transformation; transport 207 (e.g., advection, convection, and diffusion); photolysis and radiation; dry and wet deposition; and aerosol 208 209 interactions. A detailed description of the WRF-Chem model may be found in Li et al. (2011a; 2011b; 2012). A grid of 280×160 cells covering China with a horizontal resolution of 0.25° was used for the 210 simulation, which also included twenty-eight vertical layers from the Earth's surface up to 50 hPa. Seven 211 layers below 1 km were used to ensure a high vertical resolution near ground-level. The meteorological 212

initial and boundary conditions were retrieved from the National Centers for Environmental Prediction
(NCEP) reanalysis dataset, and the chemical initial and boundary conditions were obtained from the 6 h

215 output of the Model for Ozone and Related chemical Tracers (MOZART, Emmons et al., 2010).

216 In this study, the mean bias (MB), root mean square error (RMSE), and index of agreement (IOA) were

used to evaluate the performance of WRF-Chem simulation. The IOA is representative of the relative
difference between the predicted and measured values, and it varies from 0 to 1, with 1 indicating perfect
performance of the model prediction. These parameters were calculated using the following equations (Li
et al., 2011a):

221
$$MB = \frac{1}{N} \sum_{i=1}^{N} (P_i - O_i)$$
(5)

222 RMSE =
$$\left[\frac{1}{N}\sum_{i=1}^{N}(P_i - O_i)^2\right]^{\frac{1}{2}}$$
 (6)

223
$$IOA = 1 - \frac{\sum_{i=1}^{N} (P_i - O_i)^2}{\sum_{i=1}^{N} (|P_i - P_{ave}| + |O_i - O_{ave}|)^2}$$
(7)

where P_i and P_{ave} represent each predicted $PM_{2.5}$ mass concentration and the average value, respectively; O_i and O_{ave} are the observed $PM_{2.5}$ mass concentrations and the average value, respectively; and *N* is representative of the total number of predictions used for comparison.

227 2.3.4 Calculations of chemical b_{scat} and b_{abs}

To determine the contributions of individual $PM_{2.5}$ chemical species to particles' optical properties, b_{scat} and b_{abs} were reconstructed based on the major chemical composition of the $PM_{2.5}$ using the revised IMPROVE equations as follows (Pitchford et al., 2007):

231
$$b_{scat} \approx 2.2 \times f_{s}(RH) \times [(NH_{4})_{2}SO_{4}]_{small} + 4.8 \times f_{L}(RH) \times [(NH_{4})_{2}SO_{4}]_{Large} + 2.4 \times f_{s}(RH) \times [($$

232
$$[NH_4NO_3]_{Small} + 5.1 \times f_L(RH) \times [NH_4NO_3]_{Large} + 2.8 \times [OM]_{Small} + 6.1 \times [OM]_{Large} + 1 \times 10^{-10}$$

233 [Fine soil] (8)

234
$$[X]_{Large} = \frac{[X]^2}{20 \ \mu g \ m^{-3}}, for [X] < 20 \ \mu g \ m^{-3}$$
 (9)

235
$$[X]_{Large} = [X], for [X] \ge 20 \,\mu g \, m^{-3}$$
 (10)

236
$$[X]_{Small} = [X] - [X]_{Large}$$
 (11)

237 where the mass concentrations of ammonium sulfate ([(NH₄)₂SO₄]) and ammonium nitrate ([NH₄NO₃])

238 $\,$ were estimated by multiplying the concentrations of ${\rm SO_4^{2-}}$ and ${\rm NO_3^-}$ by factors of 1.375 and 1.29,

respectively (Tao et al., 2014); f(RH) is the water growth for the small (S) and large (L) modes of (NH₄)₂SO₄ and NH₄NO₃ in PM_{2.5}; and [X] represents the PM_{2.5} composition as used in Eq. (8). This analysis is based on the assumption that the particles were externally mixed. More detailed information concerning the IMPROVE algorithms may be found in Pitchford et al. (2007).

A second assumption for this part of the study was that there was negligible absorption by brown carbon in the visible region (Yang et al., 2009), and on this basis, the b_{abs} can be determined from the EC mass concentration using linear regression (Eq. 12). As shown in Figure S2, the derived slope (a) and intercept (b) for the regression model were 10.8 m² g⁻¹ and -4.7, respectively.

 $247 b_{abs} = a \times [EC] + b (12)$

248 2.3.5 DRF calculations

The Tropospheric Ultraviolet and Visible (TUV) radiation model developed by the National Center for 249 Atmospheric Research was used to estimate the aerosol DRF for 180-730 nm at the Earth's surface. A 250 detailed description of the model may be found in Madronich (1993). Aerosol DRF is mainly controlled 251 252 by the aerosol column burden and chemical composition, and important properties include the AOD, aerosol absorption optical depth (AAOD), and single-scattering albedo (SSA = (AOD-AAOD)/AOD). 253 Based on an established relationship between the AODs measured with sunphotometer and the light 254 extinction coefficients ($b_{ext} = b_{scat} + b_{abs}$) observed with PAXs, an effective height can be retrieved which 255 makes it possible to convert the IMPROVE-based chemical bext values into the AODs or AAODs caused 256 257 by the PM2.5. There are hygroscopic effects to consider, and therefore, the dry best values measured here were modified to the wet best based on the water-growth function of particles described in Malm et al. 258 (2003). We note that the estimated chemical AODs were based on the assumption that the aerosols were 259 distributed homogeneously throughout an effective height. 260 Finally, the calculated chemical AOD and SSA for different PM2.5 composition scenarios were used in 261

the TUV model to obtain shortwave radiative fluxes. Values for the surface albedo, another factor that influences DRF, were obtained from the MOD43B3 product measured with the Moderate Resolution Imaging Spectroradiometer (https://modis-atmos.gsfc.nasa.gov/ALBEDO/index.html). The solar component in the TUV model was calculated using the δ -Eddington approximation, and the vertical profile of b_{ext} used in the model was described in Palancar and Toselli (2004). The aerosol DRF is defined as the difference between the net shortwave radiative flux with and without aerosol as follows:

268 DRF_{surface} = Flux (net)_{with aerosol, surface}-Flux (net)_{without aerosol, surface}

(13)

269 3 Results and discussion

270 3.1 Effectiveness of the control measures on reducing PM_{2.5}

We divided the study period into two phases based on the dates that the pollution control measures were 271 272 put into effect (1) the NCCPC-control period from 12 to 24 October and (2) non-control period from 25 October to 4 November. Temporal variations in the PM2.5 mass concentrations and those of the major 273 274 aerosol components during these two phases are shown in Figure 2, and a statistical summary of those data is presented in Table 1. During the NCCPC-control period, the PM2.5 mass concentrations remained 275 consistently low relative to the NAAQS II (75 μ g m⁻³), generally < 75 μ g m⁻³. In contrast, higher fine 276 particle loadings ($PM_{2.5} > 75 \ \mu g \ m^{-3}$) frequently were observed during the non-control period. On average, 277 the mass concentration of PM_{2.5} during the NCCPC-control period was $57.9 \pm 9.8 \ \mu g \ m^{-3}$, which is lower 278 by 31.2% compared with the non-control period (84.1 \pm 38.8 μ g m⁻³). Meanwhile, the PM_{2.5} mass 279 concentrations obtained from the China Environmental Monitoring Center also showed a decreasing trend 280 over most of the BTH region during the NCCPC-control period (see Figure S3). Compared with previous 281 282 events when pollution control measures were implemented in Beijing and surrounding areas, the percent decrease in PM2.5 found for the present study falls within the lower limit of the 30-50% reduction for 283 Olympic Games (Wang et al., 2009; Li et al., 2013), but it is less than the range of 40-60% for the APEC 284 (Tang et al., 2015; Tao et al., 2016; J. Wang et al., 2017) or the range of 60-70% for the VDP (Han et al., 285 2016; Liang et al., 2017; Lin et al., 2017). 286

287 As shown in Figure 2 (right panel), the chemical mass closure calculations for PM_{2.5} showed that on average OM was the largest contributor (30.4%) to PM2.5 mass during the non-control period, followed 288 289 by NO₃⁻ (16.7%), fine soil (11.2%), and EC (7.6%). In contrast, OM (24.3%) and NO₃⁻ (22.9%) dominated the PM_{2.5} mass during the NCCPC-control period, followed by SO_4^{2-} (9.8%), NH₄⁺ (9.1%), and EC (7.9%). 290 The OM mass concentration was decreased largely by 43.1% from 24.6 $\mu g \ m^{\text{-3}}$ during the non-control 291 period to 14.0 µg m⁻³ during the NCCPC-control period. For secondary water-soluble inorganic ions, the 292 average mass concentrations of NO₃ $(13.4 \ \mu g \ m^{-3})$ versus 16.9 $\ \mu g \ m^{-3})$ and NH₄⁺ (5.4 versus 6.8 $\ \mu g \ m^{-3})$ 293 were lower by 20.7% and 20.6% during the NCCPC-control period, respectively. However, SO42-294 exhibited similar loadings during the NCCPC-control (5.8 µg m⁻³) and non-control (5.3 µg m⁻³) periods. 295 This is consistent with the small differences in SO_2 concentrations for the NCCPC-control (8.5 μ g m⁻³, 296 297 Figure S4) versus the non-control (12.4 µg m⁻³, Figure S4) periods. Indeed, the low SO₂ concentrations may not have provided sufficient gaseous precursor to form substantial amounts of sulfate. The loadings 298 of EC, Cl⁻, and fine soil were lower by 25.0, 44.8, and 40.8%, respectively, when the controls were in 299

place. The variations in reductions for specific aerosol components imply differences in the effectiveness of the emission controls on the chemical species, but as discussed below, meteorological conditions probably had an influence on the loadings, too.

As shown in Figure S4, both WSs $(0.7 \pm 0.3 \text{ versus } 1.3 \pm 0.8 \text{ m s}^{-1})$ and MLHs $(304.3 \pm 60.6 \text{ versus } 373.7 \pm 217.9 \text{ m})$ were lower for the NCCPC-control period compared with the non-control period. This

indicates that horizontal and vertical dispersion were weaker during the NCCPC-control period than in the non-control period. More to the point, this shows that one needs to consider the effects of WS and MLH to fully evaluate the effectiveness of the pollution control measures. A simple and effective way to do this is to compare the concentrations of air pollutants for the two periods when atmospheric conditions were stable (Wang et al., 2015; Liang et al., 2017).

We first evaluated atmospheric stability based on relationships between PM2.5 mass concentrations and 310 WS and MLH. As shown in Figure 3, the PM_{2.5} mass concentrations exhibited a power function 311 312 relationship with WS (r = -0.65) and MLH (r = 0.77). The approach used to determine stable conditions was to find the WS and MLH values that were less than the inflection points in the PM2.5 loadings; that 313 is, where the slopes in the loadings changed from large to relatively small values. As there are no true 314 inflection points for the power functions, we used piecewise functions to represent them. As shown in 315 Figure 3, the intersections of two linear regressions can be used to represent the inflection points of the 316 317 influences of meteorological conditions on PM_{2.5} mass. Using these criteria, days with WS < 0.4 m s⁻¹ and MLH < 274 m were subjectively considered to have stable atmospheric conditions. 318

319 There were two days for the NCCPC-control period and three days for the non-control period that satisfied the stability criteria. The surface charts (Figure S5) show that the weather conditions for those selected 320 stable atmosphere days during the NCCPC-control and non-control periods were mainly controlled by 321 uniform pressure fields and weak low-pressure systems, respectively, and those conditions led to weak or 322 calm surface winds. Due to the lower WS (0.2 versus 0.3 m s⁻¹) and MLH (213 versus 244 m) during the 323 324 NCCPC-control period relative to the non-control period, the horizontal and vertical dispersion for the stable atmospheric days were slightly weaker during the NCCPC-control period. As shown in Table 1, 325 the percent differences for PM_{2.5} (43.4%), NO₃ (25.9%), OM (68.1%), EC (40.0%), and fine soil (58.7%) 326 327 were larger for the days with stable atmospheric conditions compared with those for all days. These results are a further indication that the control measures were effective in reducing pollution, but meteorology 328

329 also influenced the aerosol pollution.

330 **3.2 Estimates of source contributions**

The mass concentrations of water-soluble inorganic ions (SO₄²⁻, NO₃⁻, NH₄⁺, K⁺, and Cl⁻), carbonaceous 331 (OC and EC), and elements (Al, Si, Ca, Ti, Cr, Mn, Fe, Cu, Zn, As, Br, and Pb) were used as data inputs 332 for the PMF 5.0 model. Through comparisons between the PMF profiles and reference profiles from 333 previous studies, the presumptive sources for the aerosol were identified as (i) coal combustion, (ii) 334 335 traffic-related emissions, (iii) secondary source, (iv) biomass burning, (v) industrial processes, and (vi) mineral dust. As shown in Figure S6, the PMF modelled PM2.5 mass concentrations were strongly 336 337 correlated with the observed values (r = 0.98, slope = 0.94), and the model-calculated concentrations for each chemical species exhibited good linearity and correlations with the measured values (r = 0.68-0.99) 338 339 (Table S1). These results show that the six identified sources were physically interpretable and accounted 340 for much of the variability in the data.

Figure 4 presents the source profiles and the average contribution of each source to PM_{2.5} mass during 341 342 the NCCPC-control and non-control periods. The first source factor was identified as coal burning 343 emissions because it was enriched with As (38.8%), Pb (32.9%), and Fe (30.3%) and had moderate loadings of Mn (26.2%), Zn (23.8%), Si (23.1%), and Ca (22.8%) (Figure 4a). Of these elements, As is a 344 well-known tracer for coal burning (Hsu et al., 2009; Y. Chen et al., 2017); Pb, Fe, Mn, and Zn (Xu et al., 345 2012; Men et al., 2018) are enriched in particles generated by this source; and Ca and Si can be 346 347 components of coal fly ash (Pipal et al., 2011). There was no significant difference in PM_{2.5} loadings contributed by this source between the NCCPC-control (8.5 µg m⁻³) and non-control (7.8 µg m⁻³) periods. 348 349 This may be because coal burning is mainly used for domestic purposes, especially heating, and the 350 control measures did not include this sector. The contribution of coal burning to PM2.5 mass in our October/November study was lower than its contribution in the BTH region in winter (~20-60 µg m⁻³) 351 (Huang et al., 2017), and that can be explained by the increased domestic usage of coal for heating 352 activities during the colder winter season. 353

The second source factor was linked to traffic-related emissions, and it was characterized by strong loadings of EC (42.1%) and Cu (40.7%) and moderate contributions of OC (29.1%), Zn (27.1%), and Br (22.2%). Previous studies have indicated that carbonaceous aerosols are components of gasoline and diesel engine exhaust (Cao et al., 2005), and therefore, EC and OC have been used as indicators for motor vehicle emissions (Chalbot et al., 2013; Khan et al., 2016a), and Br, too, may be emitted from internal combustion engines (Bukowiecki et al., 2005). Aerosol Cu and Zn are derived from other types of vehicle emissions, including those associated with lubricant and oil, brake linings, metal brake wear, and tires

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(Lin et al., 2015). Furthermore, the mass concentration of $PM_{2.5}$ from this source was strongly correlated (r = 0.72) with vehicle-related NOx concentrations (Figure S7), which further suggests the validity of this PMF-resolved source. Traffic-related emissions showed similar percent contributions to $PM_{2.5}$ mass during the NCCPC-control (14.8%) and non-control (15.4%) periods (Figure 4c), but the mass concentration was 38% lower for the NCCPC-control period (8.9 µg m⁻³) than the non-control period (14.4 µg m⁻³). This shows that the reduction in motor vehicle activity during the control period led to better air quality.

The third source factor was a clear signal of secondary particle formation because it was dominated by 369 370 high loadings of SO4²⁻ (45.4%), NO3⁻ (43.4%), and NH4⁺ (47.0%) (Zhang et al., 2013; Amil et al., 2016). 371 Thus, this factor was assigned to the secondary source. Moreover, moderate loadings of As (30.5%), Pb (27.4%), Cr (31.4%), Cu (30.7%), and EC (30.8%) also were assigned to this factor, suggesting influences 372 from coal burning and vehicle exhaust emissions. Although the concentrations of gaseous precursors, 373 374 especially SO₂ and NOx, were lower during the NCCPC-control period (Figure S4), the average mass contribution of secondary PM2.5 was larger when the controls were in effect (22.5 versus 18.3 µg m⁻³); 375 indeed, this source was the largest contributing factor (37.3% of PM2.5 mass) during the NCCPC-control 376 period. We note that the higher RH (84%) during the NCCPC-control period compared with the non-377 control period (69%) may have promoted the formation of the secondary inorganic aerosols through 378 379 aqueous reactions (Sun et al., 2014).

The fourth source factor, identified as emissions from biomass burning, was characterized by the high 380 loadings of K⁺ (59.5%) and moderate loadings of Cl⁻ (33.3%), OC (28.5%), NO₃⁻ (37.1%), SO₄²⁻ (21.1%), 381 and NH4+ (39.6%). Soluble K+ is an established tracer for biomass burning (Zhang et al., 2013; Wang et 382 al., 2016b), and Cl⁻ and OC also are emitted during biomass burning (Tao et al., 2014; Huang et al., 2017). 383 Previous studies have shown that SO2 and NO2 can be converted into sulfate and nitrate on KCl particles 384 during the transport of biomass-burning emissions (Du et al., 2011). Therefore, the abundant NO3⁻, SO4²⁻, 385 and NH4⁺ associated with this factor may be indicative of aged biomass-burning particles. As shown in 386 Figure 4c, biomass burning contributed substantially to PM2.5 mass during both the NCCPC (21.6%) and 387 non-control periods (27.3%). This is to be expected because Hebei Province is a major corn and wheat 388 producing area, and the residues of these crops commonly are used for residential cooking and heating or 389 burned in the fields (J. Chen et al., 2017). The mass concentrations of PM2.5 from this source were lower 390 during the NCCPC (13.0 µg m⁻³) than in the non-control period (25.7 µg m⁻³), and this indicates the 391 392 effectiveness of the control policy that forbade the open space biomass-burning during the NCCPC. As

the control measures did not include prohibitions on the household use of biofuels, substantial contributions of biomass burning were still evident during the NCCPC-control period.

395 The fifth source factor was identified as emissions from industrial processes because it had high loadings of Zn (41.3%), Br (38.0%), Pb (19.9%), As (19.2%), Cu (17.5%), and Mn (19.1%) (Q. Q. Wang et al., 396 2017; Sammaritano et al., 2018). This source contributed 3.6 µg m⁻³ to PM_{2.5} mass during the NCCPC-397 control period, which was lower than the non-control period (16.2 µg m⁻³) by 78%, and its percent 398 contribution to PM2.5 mass also increased correspondingly from 6.0 to 17.2% after the controls were 399 400 removed. The results showed that restrictions on industrial activities during the NCCPC-control period led to improvements in air quality. Iron and steel production are among the most important industries in 401 402 BTH region, and the iron and steel production there accounted for 28.8% of the total for China in 2016 (NBS, 2017). The sintering process in iron and steel industries produce large amounts of heavy metal 403 pollutants including Zn, Pb, and Mn (Duan and Tan, 2013). Hence, the iron and steel industries in the 404 405 BTH region were probable sources for these metals during the non-control period.

The sixth source factor was obviously mineral dust because it had high loadings of Al (55.9%), Si (55.7%), 406 Ca (52.6%), and Ti (36.7%) (Zhang et al., 2013; Tao et al., 2014; Kuang et al., 2015). This factor 407 contributed 3.8 μ g m⁻³ (6.3% of PM_{2.5} mass) during the NCCPC-control period and 11.5 μ g m⁻³ (12.3%) 408 to PM_{2.5} mass in the non-control period. Possible sources for the mineral dust include (i) natural dust, 409 which contains crustal Al, Si, and Ti (Milando et al., 2016), (ii) construction dust, which includes Ca (Liu 410 et al., 2017), and (iii) road dust, which is characterized by traffic-related species, such as Cu, Zn, Br, and 411 412 EC (Khan et al., 2016b; Zong et al., 2016). Here, the mineral dust factor did not contain any notable contributions from the traffic-related species. Thus, this factor can be explained by the natural and 413 construction dusts. As shown in Figure S8, WS was positively correlated (r = 0.75) with the PM_{2.5} mass 414 from mineral dust. To reduce the effects of wind speed on crustal dust resuspension, we compared the 415 days with low winds (< 1 m s⁻¹) during the sampling periods, and only three sampling days were excluded 416 417 from the analysis. This comparison showed that the mass concentration of PM2.5 from mineral dust was 60.0% lower in the NCCPC-control period (3.8 µg m⁻³) compared with the non-control period (9.5 µg m⁻³) 418 419 ³). This was a strong indication that restrictions on construction activities during the NCCPC-period were 420 effective in reducing the mineral dust component of PM2.5, but as noted above, this was not a large component of the PM2.5 mass. 421

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424 3.3 Pollution episodes after the NCCPC-control period

As shown in Figure 2 (left panel), two pollution episodes occurred after the NCCPC-control period (PE1: 425 426 25-27 October and PE2: 31 October-1 November); the average PM2.5 mass concentrations in PE1 and PE2 were 117.5 and 124.5 µg m⁻³, respectively. For PE1, secondary source was the dominant contributor 427 to the fine particle population, accounting for 54.6% of PM_{2.5} mass (Figure 5a), and the secondary species 428 429 that showed the largest contribution to PM2.5 mass was NO3 (26.8%) (Figure 5b). The mass concentration of NO₃-increased from $< 10 \,\mu g \,\mathrm{m}^3$ before PE1 to $> 25 \,\mu g \,\mathrm{m}^3$ during the episode (Figure 2). Molar ratios 430 431 of NO₃⁻ to NO₂ (NOR = $n-NO_3^{-1}(n-NO_2 + n-NO_3^{-1})$) were calculated to investigate nitrogen partitioning between the particulate and gas phases (Zhang et al., 2011). As shown in Figure 6a, the mass concentration 432 of $PM_{2.5}$ increased with NOR (r = 0.65) throughout the entire campaign, which indicates that nitrate 433 434 formation was involved in the high PM_{2.5} loadings. The NORs ranged from 0.32 to 0.71 during the PE1, and those values were significantly different (t-test, p < 0.01) from the ratios before (0.23–0.29) or after 435 436 PE1 (0.03–0.10), thus reflecting stronger nitrate formation during the pollution period. Furthermore, NOR 437 exhibited an exponential increase with RH (r = 0.80, Figure 6b), and the higher RHs (91–93%) during the PE1 may have led to greater aqueous nitrate production relative to the periods before (80-86%) or after 438 (33–57%) the first pollution episode. 439

The second largest contributor to $PM_{2.5}$ mass during PE1 was OM, which accounted for 22.9% of the fine aerosol mass. A widely used EC-tracer method (Lim and Turpin, 2002) was used to estimate the primary and secondary OA (POA and SOA). For this, the lowest 10% percentile of the measured OC/EC ratios was used as a measure of the primary OC/EC ratio (Zheng et al., 2015). The estimated mass concentrations of POA and SOA were 17.2 and 9.7 µg m⁻³ during the PE1, which amounted to 63.9 and 36.1% of the OM mass, respectively.

Photochemical oxidation and aqueous reactions are two of the major mechanisms that lead to the 446 formation SOA (Hallquist et al., 2009), and we evaluated the roles of these chemical reactions by 447 investigating trends in the EC-scaled concentrations of SOA (SOA/EC). We note that normalizing the 448 449 data in this way eliminates the impacts of different dilution/mixing conditions on the SOA loadings (Zheng et al., 2015). As shown in Figure 6c, the SOA/EC ratios increased (r = 0.65) with Ox (NO₂ + O₃), 450 which is a proxy for atmospheric aging caused by photochemical reactions (Canonaco et al., 2015), and 451 the EC-scaled concentrations showed a weaker correlation with RH (r = -0.32) (Figure 6d). These results 452 indicate that photochemical reactions rather than aqueous phase oxidation were the major pathways for 453

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SOA formation. Thus, the small contribution of SOA to PM_{2.5} during the PE1 may have been due to low
 photochemical activity during that episode.

459 In contrast to the first pollution episode, OM (31.8%) was the most abundant PM_{2.5} species during PE2, and that was followed by NO_3^- (19.2%) (Figure 5b). The mass concentration of K⁺ increased substantially, 460 from 0.1 μ g m⁻³ before PE2 to 1.7 μ g m⁻³ during the event, indicating a strengthening influence of 461 biomass-burning emissions. Indeed, the results of PMF showed that biomass burning was the largest 462 contributor to PM_{2.5} mass during the PE2, accounting for 36.0% of the total (Figure 5a). Furthermore, the 463 72-h back trajectories showed that air masses sampled during the PE2 either originated from or passed 464 over areas with fires in Inner Mongolia and Shanxi Province (see Figure 7), and this can explain the 465 apparent impacts from biomass burning emission. Moreover, SOA contributed an estimated 47.7% of the 466 OM mass, and that is a strong indication that secondary organics were a major component of the pollution. 467 The mass concentration of SOA was 19.0 μ g m⁻³ during the PE2, and that was higher than in 9.7 μ g m⁻³ 468 469 during the PE1. As the oxidizing conditions-as indicated by Ox-were similar for both pollution episodes (78.0 µg m⁻³ in PE1 versus 86.7 µg m⁻³ in PE2) (Figure S4), the larger SOA during the PE2 can 470 best be explained by SOA that formed from gaseous biomass-burning emissions during transport. 471

472 3.4 Meteorological considerations

Previous studies have shown that meteorological conditions play an important role in the accumulation 473 474 of pollution in the BTH region (Bei et al., 2017). Surface weather charts (Figure 8) were used to analyze the synoptic conditions during the two pollution episodes, and the WRF-Chem model was applied to 475 476 simulate the formation of PM2.5 (Figure 9). As shown in Figure S9, the predicted PM2.5 and its major 477 chemical components exhibited trends roughly similar to the observed values. The calculated MB and RMSE for PM_{2.5} were -6.8 and 32.8 μ g m⁻³, and the IOA was 0.75, indicating that the formation of PM_{2.5} 478 during the two pollution episodes was reasonably well captured by the WRF-Chem model even though 479 the predicted average PM_{2.5} mass concentration was lower than the observed value. The most probable 480 reason for this is that uncertainties associated with the complex meteorological fields can affect the 481 transport, diffusion, and removal of air pollutants in the atmosphere (Bei et al., 2012). Additionally, 482 discrepancies in the emission inventories for PM2.5 for different years may have contributed to the 483 differences in modelled versus measured values. 484

485 On 22 October, that is, before PE1, a weak cold high-pressure system in Siberia moved southward (Figure

486 8), and the BTH region was under the influence of a cold high-pressure system; conditions such as those

487 tend to keep pollutants at low levels. After the passage of the low-pressure system, the BTH region was

under the control of a weak high-pressure system from 24 to 25 October, and that led to a convergence of southerly airflow in the BTH region. Those meteorological conditions were favourable for the gradual accumulation of pollutants (Figure 9). For example, as shown in Figure S4, the NOx concentrations increased from 71.6 μ g m⁻³ on 22 October to 147.6 μ g m⁻³ on 25 October, and that increase provided a supply of gaseous precursors that can explain the observed large loadings of aerosol nitrate.

On 28 October, the first day of PE1, cold air piled up in the BTH region, and the high-pressure system gradually strengthened. The weather in the BTH region at that time was characterized by cloudiness, high RH, and low surface WSs. Those conditions promoted the accumulation of pollutants (Figure 9), and the WRF-Chem simulation indicated that the BTH region contributed 73.6% of PM_{2.5} mass during PE1. On 29 October, the cold high-pressure system moved towards the south, and northerly winds increased. Those meteorological conditions presumably led to a dilution of the air pollutants, and as a result, lower PM_{2.5} loadings were observed in the BTH region (Figure 9).

From 31 October–1 November (PE2), the BTH region was again dominated by a weak high-pressure system, and a convergence of northerly airflow was caused by the high-pressure system and a trailing low-pressure front. Local pollutants from the BTH region would have accumulated under those conditions, but as discussed above, the loadings of $PM_{2.5}$ also can be affected by the long-range transport processes. Indeed, the WRF-Chem simulation indicated that the BTH region contributed 46.9% to $PM_{2.5}$ mass, similar to the import of fine particles from other regions (53.1%). After 2 November, the cold highpressure system began to move southward, the winds strengthened, and the air quality gradually improved.

507 **3.5 Impacts of PM_{2.5} emission reduction on aerosol radiative effects**

508 The aerosol DRF refers to the change in the energy balance caused by the scattering and absorption of radiant energy by aerosols. As shown in Figure S10, the reconstructed chemical bscat correlated strongly 509 (r = 0.91) with the observed b_{scat} values; the slope of the linear regression was 0.90. This result indicates 510 that the IMPROVE-based method provided a good estimation of the chemical b_{scat}; nonetheless, it is likely 511 that more locally-measured mass scattering efficiencies for each chemical species could reduce the 512 underestimates of measured values. Moreover, a significant (p < 0.01) relationship between the measured 513 babs and EC mass (Figure S2) validates the use of EC mass loadings in Eq. 12 to estimate the chemical 514 babs. The contributions of each measured PM2.5 component to the chemical bext were calculated based on 515 Eq. 8, and on average, OM was the largest contributor (43.5%) to the chemical bext during the non-control 516 period (Figure 10a), followed by NH₄NO₃ (32.4%), EC (14.3%), (NH₄)₂SO₄ (7.6%), and fine soil (2.2%). 517 In contrast, during the NCCPC-control period, NH4NO3 was the largest contributor to the chemical bext, 518

amounting to 36.7% of bext, and it was followed by OM (33.3%), EC (16.2%), (NH₄)₂SO₄ (11.9%), and 519 fine soil (1.9%). The contributions of the various PM2.5 components to bext were different compared with 520 previous studies of the pollution controls for the Olympics and APEC. For example, Li et al. (2013) 521 reported that (NH₄)₂SO₄ (41%) had the largest contribution to b_{ext} during the Olympics, followed by 522 523 NH4NO3 (23%), OM (17%), and EC (9%); Zhou et al. (2017) found that OM (49%) was the largest contributor to best during the APEC summit, followed by NH4NO3 (19%), (NH4)2SO4 (13%), and EC 524 (12%). These differences may be attributed to variable efficiencies of the controls for the specific fine 525 particle species and to variations in RH among studies, the latter of which can influence sulfate and nitrate 526 formation. 527

As shown in Figure S11, the AODs measured with a sunphotometer were well correlated with the best 528 under ambient conditions; the slope (effective height) of the regression was 708 m and r = 0.78. Based on 529 the average effective height, the estimated chemical AOD (AOD = $708 \times b_{ext} \times 10^{-6}$) and SSA contributed 530 531 by each major component in PM2.5 were entered into the TUV model to calculate the DRF at the Earth's surface. The estimated average DRF ranged from -33.2 to -3.4 W m⁻², with an arithmetic mean ± standard 532 deviation of -16.5 ± 6.7 W m⁻² for the campaign. The average DRF for our study is similar to the -13.7533 W m⁻² calculated for photosynthetically active radiation at Xianghe, China in autumn using the Santa 534 Barbara DISORT Atmospheric Radiative Transfer model (SBDART) (Xia et al., 2007a). Further 535 comparisons with previous estimates of DRFs in China at ultraviolet and visible wavelengths show that 536 the average value from our study is similar to that at the rural site of Taihu (-17.8 W m⁻², Xia et al., 2007b), 537 538 but it was less negative than at the suburban or urban sites of Linan (-73.5 W m⁻², Xu et al., 2003), Nanjing (-39.4 W m⁻², Zhuang et al., 2014), or Xi'an (-100.5 W m⁻², Wang et al., 2016b). The more negative DRF 539 values correspond with high aerosol loadings during those studies. 540

The estimated average DRF during the NCCPC-control period was -14.0 ± 3.0 W m⁻², which was less 541 negative than the value during the non-control period $(-19.3 \pm 8.6 \text{ W m}^{-2})$ (Figure 10b), and this is 542 consistent with lower PM2.5 mass loadings during the NCCPC-control period. Even though the DRF 543 values were as high as -24.7 and -28.2 W m⁻² during PE1 and PE2, respectively, the percent reduction in 544 DRF during the NCCPC-control period versus the non-control period (26.3%) was smaller than the value 545 during the APEC-control study (61.3%, Zhou et al., 2017). Figure 10b also indicates that EC was 546 responsible for the largest (most negative) DRF effects at the surface during the non-control period: the 547 EC DRF value of -13.4 W m⁻² was followed by OM (-3.0 W m⁻²), NH₄NO₃ (-2.2 W m⁻²), (NH₄)₂SO₄ (-548 0.5 W m⁻²), and fine soil (-0.15 W m⁻²). The high EC DRF may have been due in part to EC particles 549

internally mixed with other materials because mixing can amplify light absorption and thereby increase DRF. The lower aerosol loadings during the NCCPC-control period can explain why the DRF values for EC, NH₄NO₃, OM, and fine soil in the uncontrolled period were smaller in magnitude -10.1, -1.7, -1.6, and -0.09 W m⁻², respectively, than in the non-control period; these were equivalent to decreases of 24.6, 22.7, 46.7, and 40.0%. These results suggest that the short-term mitigation measures implemented during the NCCPC reduced the cooling effects of PM_{2.5} at the surface in Beijing.

556 4 Conclusions

We investigated the effects of pollution controls put in place during the 19th NCCPC on the chemical 557 558 composition of PM2.5 and aerosol radiative effects at the Earth's surface. The average mass concentration of PM_{2.5} during the NCCPC-control period was $57.9 \pm 9.8 \ \mu g \ m^{-3}$, which was 31.2% lower relative to the 559 non-control period (84.1 \pm 38.8 μ g m⁻³). The major chemical species, that is, OM, NO₃⁻, NH₄⁺, EC, and 560 561 fine soil were lower by 43.1, 20.7, 20.6, 25.0, and 40.8% during the NCCPC-control period, respectively compared with samples taken after the controls were removed. Comparisons for only those days with 562 563 stable meteorological conditions showed that the control versus non-control differences in $PM_{2.5}$ (43.4%), NO3⁻ (25.9%), OM (68.1%), EC (40.0%), and fine soil (58.7%) were larger compared with those for all 564 days. Overall, these results indicate that control measures were effective in reducing fine particle pollution. 565 566 Results of a PMF receptor model showed that biomass burning (27.3%) was the largest contributor to 567 PM_{2.5} mass during the non-control period, followed by secondary <u>source</u> (19.5%), industrial processes 568 (17.2%), traffic-related emissions (15.4%), mineral dust (12.3%), and coal burning (8.3%). In contrast, 569 secondary source (37.3%) was the largest contributor to PM2.5 mass during the NCCPC-control period, followed by biomass burning (21.6%), traffic-related emissions (14.8%), coal burning (14.1%), mineral 570 571 dust (6.3%), and industrial processes (6.0%). The mass concentrations of $PM_{2.5}$ contributed by traffic-572 related emissions, biomass burning, industrial processes, and mineral dust all were lower during the NCCPC-control period compared with the non-control period. However, there was no significant 573 574 difference in PM2.5 mass from coal burning between these two periods, and a larger PM2.5 mass 575 concentration of secondary source was found for the NCCPC-control period. There were two pollution episodes (PE1: 25-27 October and PE2: 31 October-1 November) that occurred 576 577 after the NCCPC, and the average PM2.5 mass concentrations during those events (117.5 µg m⁻³ for PE1 and 124.5 µg m⁻³ for PE2) were more than double those when the controls were in place. For PE1, 578 secondary source was the most important source for fine particles, accounting for 54.6% of PM2.5 mass. 579

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Aerosol NO₃⁻ showed the largest contribution to PM_{2.5} mass (26.8%), and the high RH during PE1 likely promoted aqueous reactions involving nitrate. In contrast, OM (31.8%) was the most abundant species in PM_{2.5} during the PE2, and the PMF indicated that biomass burning was the largest source, accounting for 36.0% of the PM_{2.5} mass. The WRF-Chem simulation showed that the BTH region contributed 73.6% and 46.9% of PM_{2.5} mass during the PE1 and PE2, respectively. Calculations based on methods developed for the IMPROVE program indicated that OM was the largest

contributor (43.5%) to the chemical best during the non-control period, followed by NH4NO3 (32.4%), EC 592 (14.3%), (NH₄)₂SO₄ (7.6%), and fine soil (2.2%). During the NCCPC-control period, NH₄NO₃ accounted 593 for 36.7% of bext, and that was followed by OM (33.3%), EC (16.2%), (NH4)2SO4 (11.9%), and fine soil 594 595 (1.9%). The TUV model showed that the estimated average DRF (-14.0 ± 3.0 W m⁻²) at the surface during the NCCPC-control period was 27.5% less negative than in the non-control period (-19.3 \pm 8.6 W m²), 596 597 and this is consistent with the lower PM2.5 loadings during the NCCPC-control period. Furthermore, EC 598 had the largest (most negative) influence on DRF at the surface during the non-control period; the EC DRF value of -13.4 W m⁻² was followed by OM (-3.0 W m⁻²), NH₄NO₃ (-2.2 W m⁻²), (NH₄)₂SO₄ (-0.5 W 599 m⁻²), and fine soil (-0.15 W m⁻²). The DRF values caused by EC, NH₄NO₃, OM, and fine soil when the 600 controls were in place were lower by -10.1, -1.7, -1.6, and -0.09 W m⁻², respectively, compared with the 601 non-control period, and the corresponding percent reductions were 24.6, 22.7, 46.7, and 40.0%. The 602 603 results suggest that the short-term mitigation measures during the NCCPC-control period were effective in reducing fine particle pollution and those actions also had radiative effects sufficient to affect surface 604 605 temperature.

606 <u>Author contribution</u>

607	W and JC designed the research. QW, SL, WD, YW, JT, YZ, and MW carried out the meas	urements.
600	W and NL number and the englysis and OW unsteaded and All the outhous commented on the	h a <i>m a</i> m a m
608	w and NL performed the analysis, and Ow wrote the paper. All the authors commented on the	ne paper.

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913	Table 1 Summary of $PM_{2.5}$ and its major chemical components at Xianghe during the 19th
914	National Congress of the Communist Party of China (NCCPC)-control and non-control periods.

Components	Grand average (µg m ⁻³)	Control period (µg m ⁻³)	Non-control period (µg m ⁻³)	Change ratio ^a (%)
PM _{2.5}	70.0	57.9 (63.7) ^b	84.1 (112.6)	31.2 (43.4)
NO ₃ -	15.0	13.4 (18.0)	16.9 (24.3)	20.7 (25.9)
SO4 ²⁻	5.6	5.8 (7.6)	5.3 (6.6)	-9.4 (-15.2)
$\mathrm{NH_{4}^{+}}$	6.0	5.4 (8.6)	6.8 (9.7)	20.6 (11.3)
Cl-	2.2	1.6 (1.5)	2.9 (3.4)	44.8 (55.9)
Organic matter	18.9	14.0 (9.5)	24.6 (29.8)	43.1 (68.1)
Elemental carbon	5.2	4.5 (4.5)	6.0 (7.5)	25.0 (40.0)
Trace elements	1.8	1.4 (1.2)	2.3 (3.0)	39.1 (60.0)
Fine soil	5.5	4.2 (2.6)	7.1 (6.3)	40.8 (58.7)

^a([Non-control period]-[NCCPC-control period])/[Non-control period]. ^bValues in parentheses show the results for days with stable meteorological conditions (wind speed < 0.4 m s^{-1} and mixed layer height < 274 m).

Figure Captions

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920 Figure 1. Location of the Xianghe sampling site. The map was drawn using the ArcGIS.

- Figure 2. (left) Daily variations in the contributions of chemical species to PM_{2.5} mass during the
 campaign and (right) average contributions of chemical species during the 19th National Congress
 of the Communist Party of China (NCCPC)-control and non-control periods. PE1 and PE2 represent
 two pollution episodes.
- Figure 3. Scatter plots showing the relationships between PM_{2.5} mass concentrations and (a) wind speed
 and (b) mixed layer height.
- Figure 4. (a) Source profiles for the six sources identified using the positive matrix factorization model
 version 5.0, (b) the mass concentrations of PM_{2.5} contributed by each source, and (c) the average
 source contribution of each source to the PM_{2.5} mass.
- Figure 5. Average source contributions of (a) each positive matrix factorization source factor and (b)
 chemical species to the PM_{2.5} mass during two pollution episodes (PE1 and PE2).
- Figure 6. Correlations for (a) PM_{2.5} mass concentrations versus molar ratios of NO₃⁻ and NO₂ (NOR), (b)
 NOR versus relative humidity (RH), (c) the ratio of secondary organic aerosol to elemental carbon
 (SOA/EC) ratios versus Ox (O₃ + NO₂), and (d) SOA/EC versus RH for all samples from the
 campaign.
- Figure 7. Three-day backward in time air mass trajectories (BT) arriving at 150 m above ground every
 hour from 31 October to 1 November 2017. The orange points represent fire counts that were derived
 from Moderate Resolution Imaging Spectroradiometer observations.
- Figure 8. Surface weather charts for 08:00 (local time) over East Asia from 22 October to 2 November
 2017. The black triangles represent Xianghe.
- Figure 9. Daily average PM_{2.5} concentrations (μg m⁻³) simulated for the Beijing-Tianjin-Hebei region
 and surrounding areas from 25 October to 2 November 2017. The Weather Research and Forecasting
 model coupled to chemistry (WRF-Chem) model was used for the simulation.
- Figure 10. Average values of (a) light extinction coefficients (including light scattering and absorption)
 and (b) direct radiative forcing (DRF) at the surface contributed by each PM_{2.5} chemical composition
 during the 19th National Congress of the Communist Party of China (NCCPC)-control and non control periods.

948



951 Figure 1.









Figure 4.



Figure 5.







968 Figure 6.





Figure 7.





Figure 8.









Figure 10.