



Source Apportionment of Fine Aerosol at an Urban Site of Beijing using a Chemical Mass Balance Model

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21 Abstract

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Fine particles were sampled from 9th November to 11th December 2016 and 22nd May 23 to 24th June 2017 as part of the Atmospheric Pollution and Human Health in a Chinese 24 25 megacity (APHH-China) field campaigns in urban Beijing, China. Inorganic ions, trace 26 elements, OC, EC, and organic compounds including biomarkers, hopanes, PAHs, n-27 alkanes and fatty acids, were determined for source apportionment in this study. 28 Carbonaceous components contributed on average 47.2% and 35.2% of total reconstructed PM2.5 during the winter and summer campaigns, respectively. Secondary 29 inorganic ions (sulfate, nitrate, ammonium; SNA) accounted for 35.0% and 45.2% of 30 31 total PM_{2.5} in winter and summer. Other components including inorganic ions (K⁺, Na⁺, 32 Cl^{-}), geological minerals, and trace metals only contributed 13.2% and 12.4% of $PM_{2.5}$ during the winter and summer campaigns. Fine OC was explained by seven primary 33 sources (industrial/residential coal burning, biomass burning, gasoline/diesel vehicles, 34 35 cooking and vegetative detritus) based on a chemical mass balance (CMB) receptor 36 model. It explained an average of 75.7% and 56.1% of fine OC in winter and summer, 37 respectively. Other (unexplained) OC was compared with the secondary OC (SOC) estimated by the EC-tracer method, with correlation coefficients (R²) of 0.58 and 0.73, 38 and slopes of 1.16 and 0.80 in winter and summer, respectively. This suggests that the 39 unexplained OC by CMB was mostly associated with SOC, PM_{2.5} apportioned by CMB 40 41 showed that the SNA and secondary organic matter were the highest two contributors to PM_{2.5}. After these, coal combustion and biomass burning were also significant 42 sources of PM2.5 in winter. The CMB results were also compared with results from 43 Positive Matrix Factorization (PMF) analysis of co-located Aerosol Mass Spectrometer 44 (AMS) data. The CMB was found to resolve more primary OA sources than AMS-PMF 45 but the latter apportioned more secondary OA sources. The AMS-PMF results for major 46 47 components, such as coal combustion OC and oxidized OC correlated well with the 48 results from CMB. However, discrepancies and poor agreements were found for other OC sources, such as biomass burning and cooking, some of which were not identified 49 50 in AMS-PMF factors.





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52 Keywords: PM_{2.5}, Beijing, mass closure, CMB, AMS-PMF, source apportionment





53 1 Introduction

54 Beijing is the capital of China and a hotspot of particulate matter pollution. It has been experiencing severe PM_{2.5} (particulate matter with an aerodynamic diameter of 55 <2.5 µm) pollution in recent decades, as a result of rapid urbanization and 56 57 industrialization, and increasing energy consumption (Wang et al., 2009). High PM_{2.5} pollution from Beijing could have significant impact on human health (Song et al., 58 2006a; Li et al., 2013). A case study in Beijing revealed that a 10 μ g m⁻³ increase of 59 ambient PM_{2.5} concentration will correspondingly increase 0.78%, 0.85% and 0.75% 60 of the daily mortality of the circulatory diseases, cardiovascular diseases and 61 cerebrovascular diseases, respectively (Dong et al., 2013). Furthermore, PM_{2.5} causes 62 63 visibility deterioration in Beijing. A better understanding of the sources of $PM_{2.5}$ in 64 Beijing is essential to provide scientific evidence to control the PM_{2.5} pollution.

Many studies have identified the possible sources of fine particulate matter in Beijing 65 using various methods (Zheng et al., 2005; Song et al., 2006a; Song et al., 2006b; Li et 66 al., 2015; Zhang et al., 2013; Yu and Wang, 2013). Song et al. (2006a) applied 67 two eigenvector models, principal component analysis/absolute principal component 68 scores (PCA/APCS) and UNMIX to study the sources of PM_{2.5} in Beijing. Some studies 69 70 used elemental tracers to do source apportionment of PM2.5 by applying positive matrix factorization (PMF) (Song et al., 2006b; Li et al., 2015; Zhang et al., 2013; Yu and 71 Wang, 2013). This approach has some underlying issues: firstly, PMF requires a 72 relatively large sample size; and a "best" solution of achieved factors is subjective 73 (Ulbrich et al., 2009); secondly, many important PM2.5 emission sources do not have a 74 unique elemental composition. Hence, an elemental tracer-based method cannot 75 distinguish sources such as cooking or vehicle exhaust, as they emit mainly 76 carbonaceous compounds (Wang et al., 2009). Generally, organic matter (OM) is 77 comprised of primary organic matter (POM) and secondary organic matter (SOM). 78 POM is directly emitted and SOM is formed through chemical oxidation of volatile 79 80 organic compounds (VOCs) (Yang et al., 2016). OM was the largest contributor to PM_{2.5} mass, which accounted for 30%-60% of PM_{2.5} (Song et al., 2007; He et al., 2001; 81 82 Huang et al., 2014), and can contribute up to 90% of submicron PM mass (Zhou et al., 2018). Furthermore, many organic tracers are more specific to particular sources, 83 84 making them more suitable to identify and quantify different source contributions to 85 carbonaceous aerosols and PM_{2.5}.

A few studies have also applied a Chemical Mass balance (CMB) model for source 86 apportionment of PM in Beijing. For example, Zheng et al. (2005) investigated sources 87 88 of PM_{2.5} in Beijing, but the source profiles they used were mainly derived in the United States, which were less representative of the local sources. Liu et al. (2016) and Guo et 89 al. (2013) apportioned the sources of PM_{2.5} in a typical haze episode in winter 2013 in 90 Beijing during the Olympic Games period in summer 2008, respectively. Wang et al. 91 (2009) apportioned the sources of $PM_{2.5}$ in both winter and summer. A major challenge 92 of the CMB model is that it cannot quantify the contributions of secondary organic 93 94 aerosol and unknown sources, which are often lumped as "unexplained OC".

In this study, PM_{2.5} samples were collected in an urban site of Beijing during winter and summer 2016-2017. OC, EC, PAHs, alkanes, hopanes, fatty acids and monosaccharide anhydrides were determined. To ensure that the source profiles used in





98 the CMB model are representative, we mainly selected those studies which had been 99 based in China: straw burning (Zhang et al., 2007b), wood burning (Wang et al., 2009), gasoline and diesel vehicles (Cai et al., 2017), industrial and residential coal combustion 100 (Zhang et al., 2008), and cooking (Zhao et al., 2015). Source contributions of organic 101 102 carbon were examined and quantified applying the CMB model based on the source 103 profiles mentioned above. The results of this study are discussed and compared with 104 the results from Aerosol Mass Spectrometer-PMF analysis (AMS-PMF) (Ulbrich et al., 105 2009; Elser et al., 2016) to improve our understanding of different sources of PM_{2.5}, especially for secondary organic aerosols. 106

107 2 Methodology

108 2.1 Aerosol sampling

109 $PM_{2.5}$ was collected at an urban sampling site (116.39E, 39.98N) - the Institute of Atmospheric Physics (IAP) of the Chinese Academy of Sciences in Beijing, China from 110 9th November to 11th December 2016 and 22nd May to 24th June 2017, as part of the 111 Atmospheric Pollution and Human Health in a Chinese megacity (APHH-China) field 112 campaigns (Shi et al., 2019). The sampling site (Fig. 1) is located in the middle between 113 the North 3rd Ring Road and North 4th Ring Road and approximately 200 m from a 114 major highway. Hence, it is subject to many local sources, such as traffic, cooking, etc. 115 The location of a rural site in Beijing - Pinggu during the APHH-China campaigns is 116 also shown in Fig. 1. Other information regarding the sampling site is described 117 elsewhere (Shi et al., 2019). 118



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- 120 Figure 1. Locations of the sampling sites in Beijing (IAP urban site: Institute of Atmospheric Physics
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of the Chinese Academy of Sciences; Pinggu - rural site) (source: © Google Maps).

PM_{2.5} samples were collected on pre-baked (450°C for 6h) large quartz filters (Pallflex, 8×10 inch) by Hi-Vol air sampler (Tisch, USA) at a flow rate of 1.1 m³ min⁻ 1. A Medium-Vol air sampler (Thermo Scientific Partisol 2025i) was also deployed at the same location to collect PM_{2.5} samples simultaneously on 47 mm PTFE filters at a flow rate of 15.0 L min⁻¹. Field blanks were also collected with the pump turned off during the sampling campaign. Before and after sampling, all filters were put in a





balance room and equilibrated at a constant temperature and relative humidity (RH) for 128 129 24h prior to any gravimetric measurements, which were 22°C and 30% RH for summer samples, 21°C and 33% RH for winter samples. PM_{2.5} mass was determined through 130 the weighing of PTFE filters using a microbalance (Sartorius model MC5, precision: 1 131 132 μ g). After that, filters were wrapped separately with aluminum foil and stored at under 133 -20° C in darkness until analysis. The large quartz filters were analyzed for OC, EC, 134 organic compounds and ion species, while small PTFE filters were used for the 135 determination of PM2.5 mass and metals. Online PM2.5 were determined by the TEOM FDMS 1405-DF instrument at IAP with filter equilibrating and weighing conditions 136 comparable with the Federal Reference Method (RH: 30-40%; temperature; 20-23°C). 137

138 2.2 Chemical Analysis

139 **2.2.1 OC and EC**

140 A 1.5 cm² punch from each large quartz filter sample was taken for organic carbon (OC) and elemental carbon (EC) measurements by a thermal/optical carbon analyzer (model 141 RT-4, Sunset Laboratory Inc., USA) based on the EUSAAR2 (European Supersites for 142 Atmospheric Aerosol Research) transmittance protocol. The uncertainties from 143 144 duplicate analyses of filters were <10%. Replicate analyses were conducted once every 145 ten samples. All sample results were corrected by the values obtained from field blanks, which were 0.40 and 0.01 µg m⁻³ for OC and EC, respectively. Details of the OC/EC 146 147 measurement method can be found elsewhere (Paraskevopoulou et al., 2014). The instrumental limits of detection of OC and EC in this study were estimated to be 0.03 148 and 0.05 µg m⁻³, respectively. 149

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151 2.2.2 Organic compounds

Organic tracers, including 11 n-alkanes (C24-C34), 2 hopanes (17a (H) -22, 29, 30-152 153 Trisnorhopane, 17b (H), 21a (H) -Norhopane), 17 PAHs (retene, phenanthrene, anthracene, fluoranthene, pyrene, benz(a)anthracene, chrysene, benzo(b)fluoranthene, 154 benzo(k)fluoranthene, benzo(e)pyrene, benzo(a)pyrene, perylene, Indeno(1,2,3-155 cd)pyrene, dibenz(a,h)anthracene, benzo(ghi)perylene, coronene, picene), 3 156 157 anhydrosugars (levoglucosan, mannosan, galactosan), 2 fatty acids (palmitic acid, stearic acid) and cholesterol in the $PM_{2.5}$ samples were determined in this study. A 158 portion of the filters were extracted 3 times with dichloromethane/methanol (HPLC 159 160 grade, v/v: 2:1) under ultrasonication for 10 minutes. The extracts were then filtered 161 and concentrated using a rotary evaporator under vacuum, and blown down to dryness with pure nitrogen gas. 50 µL of N.O-bis-(trimethylsilyl)trifluoroacetamide (BSTFA) 162 with 1% trimethylsilyl chloride and 10 µL of pyridine were then added to the extracts, 163 164 which were left reacting at 70 °C for 3 h to derivatize -COOH to TMS esters and -OH to TMS ethers. After cooling to room temperature, the derivatives were diluted with 165 140 µL of internal standards (C13 n-alkane, 1.43 ng µL⁻¹) in n-hexane prior to GC-MS 166 analysis. The final solutions were analyzed by a gas chromatography mass spectrometry 167 system (GC/MS, Agilent 7890A GC plus 5975C mass-selective detector) fitted with a 168 DB-5MS column (30 m \times 0.25 mm \times 0.25 μ m). The GC temperature program and MS 169 170 detection details were reported in Li et al. (2018). Individual compounds were identified through the comparison of mass spectra with those of authentic standards or literature 171 172 data (Fu et al., 2016). Recoveries for these compounds were in a range of 70-100%,





- 173 which were obtained by spiking standards to pre-baked blank quartz filters followed by
- the same extraction and derivatization procedures. Field blank filters were analyzed the
- 175 same way as samples for quality assurance, but no target compounds were detected.

176 **2.2.3 Inorganic components**

Major inorganic ions including Na⁺, K⁺, NH₄⁺, Cl⁻, NO₃⁻ and SO₄²⁻ were determined by 177 using an ion chromatograph (IC, Dionex, Sunnyvale, CA, USA), the detection limits 178 (DLs) of them were 0.032, 0.010, 0.011, 0.076, 0.138, 0.240 and 0.142 μ g m⁻³ 179 respectively. The analytical uncertainty was less than 5% for all inorganic ions. An 180 intercomparison study showed that our IC analysis of the above-mentioned ions agreed 181 well with those of the other laboratories (Xu et al., 2020). Trace metal including Al 182 (DLs in μg m⁻³, 0.221), Si (0.040), Ca (0.034), Ti (0.003) and Fe (0.044) were 183 184 determined by X-ray fluorescence spectrometer (XRF). Other elements including V, Cr, 185 Co, Mn, Ni, Cu, Zn, As, Sr, Cd, Sb, Ba and Pb were analyzed by Inductively-coupled 186 plasma-mass spectrometer (ICP-MS), the detection limits of them were 1.32, 0.25, 0.04, 0.06, 2.05, 1.25, 1.22, 1.74, 0.02, 0.03, 0.11, 0.06 and 0.04 ng m⁻³, respectively. Mass 187 concentrations of all inorganic ions and elements in this study were corrected for the 188 189 field blank values, and the methods were quality assured with standard reference 190 materials.

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192 **2.3 Chemical Mass Closure (CMC) Method**

A Chemical Mass Closure analysis was carried out, which includes secondary inorganic
ions (sulfate, nitrate, ammonium; SNA), sodium, potassium and chloride salts,
geological minerals, trace elements, organic matter (OM), EC and bound water in
reconstructed PM_{2.5}. Geological minerals were calculated applying the equation (Eq. 1)
(Chow et al., 2015):

198 Geological minerals = 2.2Al + 2.49Si + 1.63Ca + 1.94Ti + 2.42Fe (1)

199 Trace elements were the sum of all analysed elements excluding Al, Si, Ca, Ti and Fe. The average OM/OC ratios of organic aerosols (OA) from AMS elemental analysis 200 were applied to calculate OM, which were 1.75 ± 0.16 and 2.00 ± 0.19 in winter and 201 summer, respectively. Based on the concentrations of inorganic ions and gas-phase NH₃, 202 particle bound water was calculated by ISORROPIA II model (available 203 204 at http://isorropia.eas.gatech.edu) in forward mode and thermodynamically metastable 205 phase state (Fountoukis and Nenes, 2007). Two sets of calculations were done for online and offline data, differing at the temperature and relative humidity as specified above. 206

207 2.4 Chemical Mass Balance (CMB) model

A receptor model, namely the chemical mass balance model (US EPA CMB8.2), was applied in this study to apportion the sources of OC. It utilizes a linear least squares solution and both uncertainties in source profiles and ambient measurements were taken into consideration in this model. The essential criteria in this model were met to ensure reliable fitting results. For instance, in all samples, R^2 were >0.80 (mostly >0.9), Chi² were <2, Tstat values were mostly greater than 2 except the source of vegetative detritus, and C/M ratios (ratio of calculated to measured concentration) for all fitting





species were in range of 0.8-1.2 in this study. The source profiles applied here were 215 216 mostly from local studies in China to better represent the source characteristics, including straw burning (wheat, corn, rice straw burning) (Zhang et al., 2007b), wood 217 burning (Wang et al., 2009), gasoline and diesel vehicles (including motorcycles, light-218 219 and heavy-duty gasoline and diesel vehicles) (Cai et al., 2017), industrial and residential 220 coal combustion (including anthracite, sub-bituminite, bituminite, and brown coal) 221 (Zhang et al., 2008), and cooking (Zhao et al., 2015), except vegetative detritus (Rogge 222 et al., 1993; Wang et al., 2009). Fitting species should be stable during the transport from sources to receptor site and can represent the chemical characteristics of the 223 224 sources (Wang et al., 2009). The selected fitting species were EC, levoglucosan, 225 palmitic acid, stearic acid, fluoranthene, phenanthrene, retene, benz(a)anthracene, 226 chrysene, benzo(b)fluoranthene, benzo(k)fluoranthene, benzo[ghi]perylene, picene, 227 17a (H) -22, 29, 30-trisnorhopane, 17b (H), 21a (H) -norhopane and n-alkanes (C24-228 C33), the concentrations of which are provided in Table 1.

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230 2.5 Positive Matrix Factorization analysis of data obtained from Aerosol Mass

231 Spectrometer (AMS-PMF)

232 An Aerodyne AMS with a PM_1 aerodynamic lens was deployed on the roof of the neighboring building- the Tower branch of IAP for real-time measurements of non-233 refractory (NR) chemical species from 15th November to 11th December 2016 and 22nd 234 May to 24th June 2017. The detailed information of the sampling sites is given 235 elsewhere (Xu et al., 2019b). The submicron particles were dried and sampled into the 236 237 AMS at a flow of ~0.1 L min⁻¹. NR-PM₁ can be quickly vaporized by the 600 $^{\circ}$ C 238 tungsten vaporizer and then the NR-PM₁ species including organics, Cl⁻, NO₃⁻, SO₄²⁻ and NH_4^+ were measured by AMS in mass sensitive V mode (Sun et al., 2020). Details 239 240 of AMS data analysis, including the analysis of organic aerosol (OA) mass spectra can be found elsewhere (Xu et al., 2019b). While the source apportionment of fine OC in 241 this study was conducted by using an offline chemical speciation dataset and source 242 profiles, the source apportionment of organics in NR-PM₁ was carried out by applying 243 PMF to the high-resolution mass spectra of OA. The procedures of the pretreatment of 244 245 spectral data and error matrices can be found elsewhere (Ulbrich et al., 2009). The PMF analysis resulted in an optimal solution of 2 primary factors in summer: traffic-related 246 hydrocarbon-like OA (HOA) and cooking OA (COA) and 3 secondary factors of 247 oxygenated OA (OOA): OOA1, OOA2, OOA3. In winter, 3 primary factors were 248 identified: coal combustion OA (CCOA), COA, biomass burning OA (BBOA), and 3 249 secondary factors: oxidized primary OA (OPOA), less-oxidized OA (LOOOA), and 250 251 more-oxidized OA (MOOOA). It is noted that the data were missing during the period 09th - 14th November 2016 due to the malfunction of the AMS. 252

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254 3 Results and discussion

255 **3.1 Characteristics of PM2.5 and Carbonaceous Compounds**

Mean concentrations of PM_{2.5}, OC, EC and organic tracers during wintertime (9th November to 11th December 2016) and summertime (22nd May to 24th June 2017) at the





IAP site are summarized in Table 1 and Fig. S1. The average $PM_{2.5}$ concentration was 94.8±64.4 µg m⁻³ during the whole winter sampling campaign. The winter sampling period was divided as haze (daily $PM_{2.5} > 75 \mu g m^{-3}$) and non-haze days (<75 µg m⁻³). The average daily $PM_{2.5}$ was 136.7±49.8 and 36.7±23.5 µg m⁻³ on haze and non-haze days, respectively. Daily $PM_{2.5}$ in the summer sampling period was 30.2±14.8 µg m⁻³, comparable with that on winter non-haze days.

OC concentrations ranged between 3.9-48.8 µg m⁻³ (mean: 21.5 µg m⁻³) and 1.8-12.7 264 μg m⁻³ (mean: 6.4 μg m⁻³) during winter and summer, respectively. They are 265 comparable with the OC concentrations in winter $(23.7 \ \mu g \ m^{-3})$ and summer $(3.78 \ \mu g \ m^{-3})$ 266 m⁻³) in Tianjin, China during an almost simultaneous sampling period (Fan et al., 2020), 267 but much lower than the OC concentration (17.1 µg m⁻³) in summer 2007 in Beijing 268 269 (Yang et al., 2016). The average OC concentration during haze days $(29.4\pm9.2 \ \mu g \ m^3)$ was approximately three times that of non-haze days $(10.7\pm6.2 \ \mu g \ m^{-3})$ during winter. 270 The average EC concentration during winter was $3.5\pm2.0 \ \mu g \ m^{-3}$; its concentration was 271 4.6±1.3 µg m⁻³ on haze days, approximately 2.4 times that on winter non-haze days 272 $(1.9\pm1.6 \,\mu\text{g m}^{-3})$ and 5 times that $(0.9\pm0.4 \,\mu\text{g m}^{-3})$ during the summer sampling period. 273 The OC and EC concentrations in this study were comparable with the OC (27.9 ± 23.4 274 μ g m⁻³) and EC (6.6 ± 5.1 μ g m⁻³) concentrations in winter Beijing in 2016 (Qi et al., 275 2018), but much lower than those in an urban area of Beijing during winter (OC and 276 EC: 36.7 ± 19.4 and $15.2\pm11.1 \ \mu g \ m^{-3}$) and summer ($10.7\pm3.6 \ and \ 5.7\pm2.9 \ \mu g \ m^{-3}$) in 277 2002 (Dan et al., 2004). 278

On average, OC and EC concentrations in winter were 3.3 and 3.9 times those in summer. Additionally, OC and EC were well-correlated in this study, with R^2 values of 0.85 and 0.63 during winter and summer, respectively, suggesting similar sources of carbonaceous aerosols, especially in winter. Less correlated OC and EC in summer could be a result of SOC formation. SOC in this study was estimated and is discussed in section 3.3.7.

G 18/ -3	Winter		W. (21)	6 (24)	
Compounds ^a / ng m ⁻³	Haze ^d (n=18)	Non-haze ^c (n=13)	- Winter (n=31)	Summer (n=34)	
PM _{2.5} (µg m ⁻³)	136.7±49.8 (80.5-239.9)b	36.7±23.5 (10.3-72)	94.8±64.4 (10.3-239.9)	30.2±14.8 (12.2-78.8)	
OC (µg m ⁻³)	29.4±9.2 (13.7-48.8)	10.7±6.2 (3.9-21.5)	21.5±12.3 (3.9-48.8)	6.4±2.3 (1.8-12.7)	
EC (µg m-3)	4.6±1.3 (1.6-6.6)	1.9±1.6 (0.3-5.2)	3.5±2.0 (0.3-6.6)	0.9±0.4 (0.2-1.7)	
SOC ^c (µg m ⁻³)	10.3±5.7 (2.9-24.6)	2.9±1.4 (0.0-5.5)	7.2±5.7 (0.0-24.6)	2.3±1.4 (0.0-6.0)	
Levoglucosan	348.2±148.0 (83.1-512.5)	195.0±163.7 (19.1-539.5)	278.5±171.4 (19.1-539.5)	26.1±28.3 (2.9-172.2)	
Palmitic acid	376.2±234.9 (44.5-1089.6)	278±280.6 (33.8-1137.2)	335±255.3 (33.8-1137.2)	25.2±11.9 (9.4-68)	
Stearic acid	207.1±181.4 (23-846.7)	163.6±228.1 (17.3-903.2)	188.8±199.8 (17.3-903.2)	16.0±7.2 (5.6-36.4)	
Phenanthrene	8.6±6.1 (1.8-19)	5.6±6.1 (1-24.8)	7.3±6.2 (1-24.8)	0.7±0.7 (0-3.8)	
Fluoranthene	25.1±19.6 (4.2-76.2)	16.1±21.3 (4.2-84.3)	21.3±20.5 (4.2-84.3)	0.4±0.2 (0-0.9)	
Retene	16±14.9 (2-52.2)	11.1±12.1 (0.5-45.5)	13.9±13.8 (0.5-52.2)	0±0 (0-0.1)	
Benz(a)anthracene	21.5±16.5 (0.3-62.7)	10.8±9.3 (1.4-30.5)	17±14.8 (0.3-62.7)	0.2±0.1 (0-0.5)	
Chrysene	22.6±14.1 (3.7-47.3)	13.6±15.6 (0.1-59.5)	18.8±15.2 (0.1-59.5)	0.2±0.1 (0-0.3)	
Benzo(b)fluoranthene	52.6±29 (10.7-98)	28.1±31 (2.4-113.6)	42.3±31.8 (2.4-113.6)	0.7±0.5 (0-2)	
Benzo(k)fluoranthene	12.2±8 (0-25.3)	6.7±6.8 (0-23.7)	9.9±7.9 (0-25.3)	0.2±0.1 (0-0.4)	
Picene	0.8±0.8 (0-2.6)	0.3±0.5 (0-1.3)	0.6±0.7 (0-2.6)	0±0 (0-0)	
Benzo(ghi)perylene	7.0±4.7 (0-13.6)	4.0±4.1 (0-14.0)	5.6±4.6 (0-14.0)	0±0.1 (0-0.3)	
17a (H) -22, 29, 30- Trisnorhopane	2.7±1.6 (0.6-6.7)	1.6±1.5 (0.3-6)	2.2±1.6 (0.3-6.7)	0±0.1 (0-0.4)	
17b (H), 21a (H) - Norhopane	3.1±1.6 (0.9-6.6)	1.8±1.8 (0.3-7.3)	2.6±1.8 (0.3-7.3)	0±0 (0-0.2)	

285 **Table 1.** Summary of measured concentrations at IAP site in winter and summer.





C24	26.3±15.3 (7.8-55.5)	18±19.2 (2.1-71.2)	22.5±17.4 (2.1-71.2)	1.4±0.6 (0.5-3.3)
C25	28.2±15.6 (8.5-59)	19.5±20.5 (2.3-76.2)	24.2±18.3 (2.3-76.2)	2.9±1.5 (0.5-6.5)
C26	18.9±10.2 (5.8-40.2)	13±13.1 (1.8-48.2)	16.2±11.8 (1.8-48.2)	1.6±0.7 (0.3-4.3)
C27	20.4±9.2 (6.1-37.1)	13.8±12.5 (2.2-43.5)	17.4±11.2 (2.2-43.5)	4.4±2 (0.6-11.7)
C28	10.6±4.8 (3.2-19.2)	6.9±5.7 (1.5-19.3)	8.9±5.5 (1.5-19.3)	1.4±0.6 (0.3-2.9)
C29	22.3±10.1 (5.9-39.7)	14.3±12.6 (3-39)	18.7±11.9 (3-39.7)	5.2±3.3 (0.4-20.7)
C30	6.8±2.9 (2.2-11.4)	4.5±3.1 (1-9.7)	5.7±3.2 (1-11.4)	1±0.4 (0.2-2)
C31	11.6±4.2 (3.5-17.7)	7.7±5.8 (1.2-18.7)	9.8±5.3 (1.2-18.7)	4.3±3.2 (0.4-20)
C32	6.1±2.6 (1.7-9.3)	3.9±2.6 (0.7-8.2)	5.1±2.8 (0.7-9.3)	0.9±0.4 (0.2-1.7)
C33	5.8±2.7 (1.7-11.5)	3.9±3.1 (0.9-9.6)	4.9±3 (0.9-11.5)	1.8±1.1 (0.1-6.3)
C34	2.1±2.1 (0-5.5)	1.2±1.4 (0-4)	1.7±1.8 (0-5.5)	0.3±0.3 (0-0.9)

^a The unit is ng m⁻³ for all organic compounds and µg m⁻³ for PM_{2.5}, OC, EC and SOC; ^b mean±SD
 (min-max); ^c SOC concentration was calculated by EC-tracer method; ^d Haze days: PM_{2.5}≥75 µg m⁻³; ^e Non-haze days: PM_{2.5}<75 µg m⁻³;

289 3.2 Chemical Mass Closure (CMC)

The composition of $PM_{2.5}$ applying the chemical mass closure method is plotted in Fig.2 and summarized in Table S1. Because the gravimetrically measured mass (offline $PM_{2.5}$) differs slightly from online $PM_{2.5}$ (Fig. S2), the regression analysis results between mass reconstructed using mass closure (reconstructed $PM_{2.5}$) and both measured $PM_{2.5}$ (offline $PM_{2.5}$ / online $PM_{2.5}$) were investigated and plotted in Fig. 3.

295

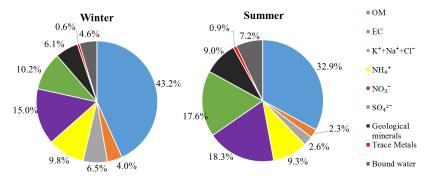




Figure 2. Chemical components of reconstructed PM_{2.5} (offline) applying mass closure method.

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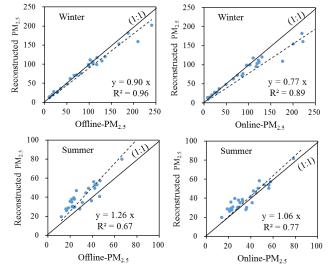




Figure 3. Regression results between reconstructed PM_{2.5} and offline/online PM_{2.5} by chemical mass
 closure method.

As shown in Fig. 3, measured offline/online PM_{2.5} generally agrees well with the 303 reconstructed PM_{2.5}. In winter, the regression results were good between reconstructed 304 305 PM_{2.5} and offline-PM_{2.5}. For online-PM_{2.5}, it was much higher than the reconstructed PM_{2.5} when the mass was over 170 µg m⁻³. After excluding the outliers (2 outliers of 306 offline-PM_{2.5} > 200 μ g m⁻³ and 4 outliers of online-PM_{2.5} > 170 μ g m⁻³), the regression 307 308 results improved with both slopes and R^2 approaching unity (Fig. S3). This could 309 indicate some uncertainties in offline and online PM2.5 measurement for heavily 310 polluted samples. During the summer campaign, the slope of the reconstructed PM_{2.5} and online-PM_{2.5} was close to 1, but that of reconstructed PM_{2.5} and offline-PM_{2.5} was 311 312 1.26. This could be due to the positive artifacts of quartz filters for chemical analyses, which can absorb more organics than PTFE filters that are used for PM weighing. The 313 datapoints were more scattered in summer, which could result from the large difference 314 in OM-OC relationships from day to day. The reconstructed inorganics (reconstructed 315 316 PM_{2.5} excluding OM) correlated well with offline-PM_{2.5}, but OM did not (Fig. S4).

During the winter campaign, the carbonaceous components (OM & EC) accounted for 47.2% of total reconstructed PM_{2.5}, followed by the secondary inorganic ions (NH₄⁺, SO₄²⁻, NO₃⁻) (35.0%). In summer, on the contrary, secondary inorganic salts represented 45.2% of PM_{2.5} mass, followed by carbonaceous components (35.2%). Bound water contributed 4.6% and 7.2% of PM_{2.5} during the winter and summer, respectively. All other components combined accounted for 13.2% and 12.4% of PM_{2.5} during the winter and summer campaigns, respectively.

324

325 **3.3** Source apportionment of fine OC in urban Beijing applying a CMB model

The CMB model resolved seven primary sources of OC in winter and summer, including vegetative detritus, straw and wood burning (biomass burning, BB), gasoline





vehicles, diesel vehicles, industrial coal combustion (Industrial CC), residential coal combustion (Residential CC) and cooking. It explained an average of 75.7% (45.3-91.3%) and 56.1% (34.3-76.3%) of fine OC in winter and summer, respectively. The averaged CMB source apportionment results in winter and summer are presented in Table 2. Daily source contribution estimates to fine OC and the relative abundance of different acurace contribution to OC in winter and summer are field.

different sources contributions to OC in winter and summer are shown in Fig. 4.

During the winter campaign, coal combustion (industrial and residential CC, 7.5 µg 334 m⁻³, 35.0% of OC) was the most significant contributor to OC, followed by Other OC 335 $(5.3 \ \mu g \ m^{-3}, 24.8\%)$, biomass $(3.8 \ \mu g \ m^{-3}, 17.6\%)$, traffic (gasoline and diesel vehicles, 336 $2.6 \,\mu g \,\mathrm{m}^{-3}$, 11.9%), cooking ($2.2 \,\mu g \,\mathrm{m}^{-3}$, 10.3%), vegetative detritus ($0.09 \,\mu g \,\mathrm{m}^{-3}$, 0.4%). 337 On winter haze days, industrial coal combustion, cooking and Other OC were 338 339 significantly higher (nearly tripled) compared to non-haze days. During the summer campaign, Other OC (2.9 µg m⁻³, 45.6%) was the most significant contributor to OC, 340 followed by coal combustion (2.0 µg m⁻³, 31.1%), cooking (0.7 µg m⁻³, 10.3%), traffic 341 $(0.4 \,\mu\text{g}\,\text{m}^{-3}, 6.1\%)$, biomass burning $(0.3 \,\mu\text{g}\,\text{m}^{-3}, 5.3\%)$, and vegetative detritus $(0.1 \,\mu\text{g}\,\text{m}^{-3}, 5.3\%)$ 342 m⁻³, 1.7%). 343

344	Table 2. Source contribution estimates (SCE, $\mu g m^{-3}$) for fine OC in urban Beijing
345	during winter and summer from the CMB model

	Winter			Summer (n=34)
Sources	Haze (n=18)	Non-haze (n=13)	Winter (n=31)	
Vegetative detritus	0.11±0.08	0.07 ± 0.08	0.09±0.08	0.11±0.08
Biomass burning	4.80±2.23	2.38 ± 2.57	3.78±2.64	0.34±0.39
Gasoline vehicles	2.35±1.27	1.59 ± 1.85	2.03±1.56	0.31±0.16
Diesel vehicles	0.83±1.43	0.14±0.33	0.54±1.15	0.08±0.16
Industrial coal combustion	7.09±4.17	1.95±1.36	4.94±4.15	1.82 ± 0.72
Residential coal combustion	3.64±3.72	1.16±0.96	2.60±3.12	0.18 ± 0.11
Cooking	3.23±2.30	0.85 ± 0.52	2.23±2.13	0.66±0.43
Other OC ^a	7.4±5.6	2.5±1.4	5.3±4.9	2.9±1.5
Calculated OC ^b	22.0±6.5	8.2±5.3	16.2±9.1	3.5±1.2
Measured OC	29.4±9.2	10.7±6.2	21.5±12.3	6.4±2.3

^a Other OC is calculated by subtracting calculated OC from measured OC;.

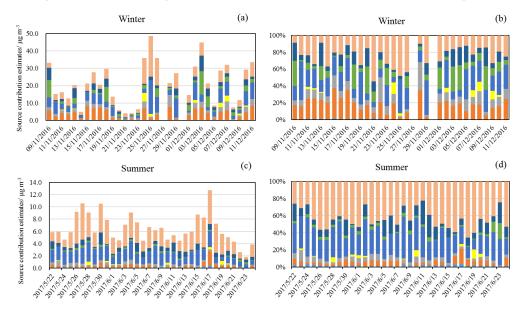
347 ^b Calculated OC is the sum of OC from all seven primary sources: vegetative detritus, biomass burning,

348 gasoline vehicles, diesel vehicles, industrial coal combustion, residential coal combustion and cooking.

349







■ Vegetative detritus ■ Biomass burning ■ Gasoline vehicles ■ Diesel vehicles ■ Industrial CC ■ Residential CC ■ Cooking ■ Other OC

Figure 4. Daily source contribution estimates to fine OC in (a) winter and (c) summer

351 and their relative abundance in winter (b) and summer (d)





352 3.3.1 Industrial and residential coal combustion

In China, a large amount of coal is used in thermal power plant, industries, urban and 353 354 rural houses in northern China, especially during the heating period (mid-November to mid-March) (Huang et al., 2017; Yu et al., 2019). But urban household coal use 355 experienced a remarkable drop of 58% during 2005-2015, which is much higher than 356 that of rural household coal use (5% of decrease) (Zhao et al., 2018). In this study, coal 357 358 combustion is the single largest contributor to primary OC in both winter and summer. In addition, industrial CC was a more significant source of OC than residential CC in 359 urban Beijing. On average, coal combustion related OC was 7.5±5.0 µg m⁻³ (34.5±9.8% 360 of OC) in winter, which was more than 3 times of that in summer - $2.0\pm0.8 \ \mu g \ m^{-3}$ 361 362 (32.3±10.2% of OC), but the percentage contribution is similar. A similar seasonal trend 363 was also found in other studies in Beijing (Zheng et al., 2005; Wang et al., 2009), but the relative contribution of coal combustion was much lower than in this study. 364 Industrial CC derived OC was 4.94±4.15 and 1.82±0.72 µg m⁻³ in winter and summer, 365 respectively. Residential CC derived OC was 2.60 ± 3.12 and 0.18 ± 0.11 µg m⁻³ in winter 366 and summer, respectively. Residential CC was much higher in winter compared to that 367 368 in summer. On haze days, industrial CC and residential CC derived OC were 3.6 and 3.1 times that on non-haze days, respectively, indicating an important contribution to 369 haze formation from industrial CC. 370

371 Chloride has been considered as a tracer for coal combustion (Chen et al., 2014). The time series of OC from coal combustion (OC-CC) and Cl⁻ during winter and summer 372 of Beijing are shown in Fig. 5. OC-CC and Cl⁻ exihibited similar trends in both seasons. 373 The correlation coefficient (R²) between OC-CC and Cl⁻ during winter was 0.62 but 374 375 there is no significant correlation between the two during the summer campaign while. 376 This is probably related to the semi-volatility of ammonium chloride, which is liable to evaporate in summer (Pio and Harrison, 1987). A similar phenomenon has been 377 observed in Delhi (Pant et al., 2015). 378

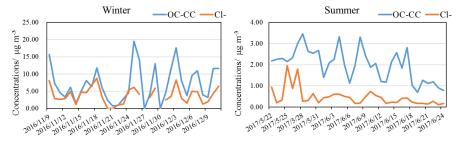




Figure 5. Time series of OC from coal combustion (OC-CC) and Cl⁻ in winter and summer in Beijing

381 382

383 3.3.2 Biomass burning

Biomass burning (BB), including straw and wood burning, is an important source of atmospheric fine OC, which ranked as the second highest primary source of OC, after industrial coal combustion during the winter campaign, and third highest during the summer campaign after industrial CC and cooking. As shown in Fig. 4, the relative





abundance of BB derived-OC during the winter campaign is much higher than the 388 summer campaign. BB-derived OC from the CMB results was 3.78±2.64 µg m⁻³ and 389 $0.34\pm0.39 \ \mu g \ m^{-3}$ in winter and summer, contributing 17.6% and 5.3% of OC in these 390 two seasons, respectively. These results are lower than those in 2005-2007 Beijing 391 when BB accounted for 26% and 11% of OC in winter and summer, respectively (Wang 392 393 et al., 2009). The BB-derived OC on winter haze days $(4.80\pm2.23 \ \mu g \ m^{-3})$ was 394 approximately double that of non-haze days (2.38±2.57 µg m⁻³), accounting for 16.3% 395 and 22.2% of OC on haze and non-haze days, respectively.

Levoglucosan is widely used as a key tracer for biomass burning emissions (Bhattarai 396 et al., 2019; Cheng et al., 2013; Xu et al., 2019a). Based on a levoglucosan to OC ratio 397 398 of 8.2 % (Zhang et al., 2007a; Fan et al., 2020), the BB-derived OC was $3.40\pm2.09 \,\mu g$ 399 m^{-3} and $0.32\pm0.35 \,\mu g m^{-3}$ during the winter and summer campaigns, respectively. These results are comparable to BB-derived OC from the CMB in this study. The estimated 400 BB-derived OC concentration are also comparable with the BB-derived OC during the 401 same sampling periods in Tianjin (Fan et al., 2020), but higher than those at IAP in 402 403 2013-2014 (Kang et al., 2018).. Both of the studies applied the levoglucosan/OC ratio method to estimate the BB-derived OC although the actual ratio in Beijing air may be 404 very different to 8.2%. The heavily elevated OC concentration in winter compared to 405 summer could be a result of increased biomass burning activities for house heating and 406 cooking in Beijing in addition to the unfavorable dispersion conditions under stagnant 407 weather conditions in the winter. 408

In summer, the total OC concentration was highest on 17th June. The sudden rise of OC on this day was attributed to the enhanced biomass burning activities, which led to the highest level of BB-derived OC and highest BBOC to OC abundance. The levoglucosan concentration on this day was also the highest in summer, which reached 172 ng m⁻³.

414 **3.3.3 Gasoline and diesel vehicles**

OC and EC are the key components of traffic emissions (gasoline vehicles & diesel 415 engines) (Chen et al., 2014; Chuang et al., 2016). Traffic related OC, as represented by 416 the total sum of OC from gasoline and diesel vehicles, was 2.4 ± 2.3 and $0.39\pm0.22 \mu g$ 417 418 m⁻³, and contributed 12.1±7.8% and 6.1±3.3% of OC in winter and summer, respectively. These results are lower than the contribution of vehicle emissions to OC 419 420 (13-20%) in Beijing during 2005 and 2006 (Wang et al., 2009), suggesting traffic emissions may be a less significant contributor to fine OC in the atmosphere in Beijing 421 in 2016/2017. By multipling by OM/OC factors of 2.39 and 1.47 in winter and summer, 422 423 respectively, as mentioned in section 2.3, traffic related organic aerosol contributed 424 $8.2\pm6.5\%$ and $2.3\pm1.7\%$ of PM_{2.5} in winter and summer, respectively. The summer 425 result was comparable with the vehicular emissions contribution to $PM_{2.5}$ (2.1%) in 426 summer in Beijing, but higher than that in winter (1.5%) in Beijing estimated by using a PMF model (Yu et al., 2019). Gasoline vehicles dominanted the traffic emissions; 427 gasoline vehicle-derived OC was 2.03±1.56 and 0.31±0.16 µg m⁻³ in winter and 428 429 summer, respectively, which are approximately four times than that in winter 430 $(0.54\pm1.15 \ \mu g \ m^{-3})$ and summer $(0.08\pm0.16 \ \mu g \ m^{-3})$ attributed to diesel vehicles. On 431 haze days, gasoline- and diesel-derived OC were 2.35±1.27 and 0.83±1.43 µg m⁻³, 432 respectively, much higher than gasoline- $(1.59\pm1.85 \ \mu g \ m^{-3})$ and diesel-derived (0.14±0.33 µg m⁻³) OC on non-haze days. Even though diesel vehicles played a less 433





434 important role in OC emissions, diesel-derived OC on haze days increased by around 6
 435 times above that of non-haze days, and such an increase was much higher than for
 436 gasoline, suggesting a potentially important role of diesel emissions on haze formation.

437 **3.3.4 Cooking**

Cooking is expected to be an important contributor of fine OC in densily populated 438 Beijing, which has a population of over 21 million. The cooking source profile was 439 selected from a study which was carried out in the urban area of another Chinese 440 megacity- Guangzhou, which includes fatty acids, sterols, monosaccharide anhydrides, 441 alkanes and PAHs in particles from the Chinese residential cooking (Zhao et al., 2015). 442 The resultant cooking related OC concentrations were $2.23\pm2.13 \,\mu g \,m^{-3}$ and 0.66 ± 0.43 443 µg m⁻³ in winter and summer, respectively, and both accounted for about 10% to total 444 OC. Cooking OC was 3.23±2.30 µg m⁻³ on winter haze days, around four times higher 445 than that on non-haze days $(0.85\pm0.52 \ \mu g \ m^{-3})$. 446

447 **3.3.5 Vegetative detritus**

Vegetative detritus made a minor contribution to fine particle mass. Its concentration 448 was $0.09\pm0.08 \,\mu\text{g m}^{-3}$ (0.4%) and $0.11\pm0.08 \,\mu\text{g m}^{-3}$ (1.7%) of OC during the winter and 449 450 summer campaigns, respectively. These contributions are comparable with that in winter (0.5%), but higher than that in summer (0.3%) in urban Beijing during 2006-451 2007 (Wang et al., 2009). These results are also higher than the plant debris-derived 452 OC in Tianjin in winter 2016 ($0.02 \,\mu g \, m^{-3}$) and summer 2017 ($0.01 \,\mu g \, m^{-3}$), which were 453 calculated based on the relationship of glucose and plant debris and a OM/OC ratio of 454 455 1.93 (Fan et al., 2020).

456 **3.3.6 Other OC**

The Other OC was calculated by subtracting the calculated OC (the sum of OC from 457 seven main sources) from measured OC concentrations. As shown in Table S2, there 458 are four major source categories of OC in Beijing based on the Multi-resolution 459 Emission Inventory for China (MEIC), which include power, industry, residential and 460 transportation (Zheng et al., 2018). In the "industry" category, industrial coal 461 combustion has been resolved by the CMB model. The local emissions of OC from 462 463 industrial coal in Beijing were zero (shown in Table S2), and hence, the resolved POC 464 from industrial coal combustion in Beijing should be regionally-transported. The MEIC data also show a small industrial oil combustion source. Since the tracers for this are 465 likely to be the same as those for petroleum-derived road traffic emissions in CMB, this 466 may result in a small overestimation of the latter source. For the industrial processes 467 related OC which have not been resolved by the CMB model, the annual average OC 468 emissions in Beijing were 1161 and 1083 tonnes in 2016 and 2017 respectively, which 469 accounted for 7.7% and 9.0% of the total OC emissions (POC). Therefore, the 470 471 contribution from industrial processes to the total OC in the atmosphere (POC+SOC) 472 was considered relatively small. The Other OC in this study is likely to be a mixture of predominantly SOC and a small portion of POC from sources such as industrial 473 processes. 474





The Other OC was 5.3±4.9 and 2.9±1.5 µg m⁻³ in winter and summer, respectively, 475 contributing 24.8% and 43.9% of total measured OC. This is in good agreement with 476 the Other OC estimated by CMB in another study in urban Beijing, for which Other OC 477 contributed 22% and 44% of OC in winter and summer, respectively (Wang et al., 2009). 478 479 SOC/OC in summer was more than 10% higher than that in summer 2008 in Beijing 480 estimated using a tracer yield method, with the SOC derived from specific VOC 481 precursors (toluene, isoprene, α-pinene and β-caryophyllene) accounting for 32.5% of 482 OC (Guo et al., 2012).

Even though the Other OC concentration was lower in summer, its relative abundance was higher than that in winter, suggesting relatively higher efficiency of SOA formation in summer due to more active photochemical processes under higher temperature and strong radiation. The Other OC on winter haze days was $7.4\pm5.6 \ \mu g$ m⁻³, approximately 3 times of that on non-haze days ($2.5\pm1.4 \ \mu g \ m^{-3}$). Other OC is also compared with the SOC estimated by EC-tracer method below.

489 **3.3.7 SOC calculated based on the EC-tracer method**

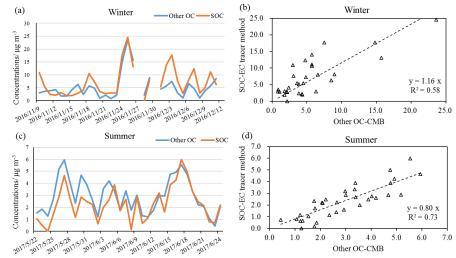
EC is a primary pollutant, while OC can originate from both primary sources and form 490 491 in the atmosphere from gaseous precursors, namely primary organic carbon (POC) and SOC, respectively (Xu et al., 2018). The OC/EC ratios can be used to estimate the 492 primary and secondary carbonaceous aerosol contributions. Usually, OC/EC ratios > 493 2.0 or 2.2 have been applied to identify and estimate SOA (Liu et al., 2017). In this 494 study, all samples were observed with higher OC/EC ratios (>2.2). SOC in this study 495 496 was estimated using the equation below, assuming EC comes 100% from primary sources and the OC/EC ratio in primary sources is relatively constant (Turpin and 497 Huntzicker, 1995; Castro et al., 1999): 498

499
$$\operatorname{SOC}_i = \operatorname{OC}_i - \operatorname{EC}_i \times (\operatorname{OC/EC})_{\operatorname{pri}}$$
 (4)

where SOC_i, OC_i and EC_i are the ambient concentrations of secondary organic 500 501 carbon, organic carbon and elemental carbon of sample i, respectively. (OC/EC)_{pri} is 502 the OC/EC ratio in primary aerosols. It is difficult to accurately determining the ratio 503 of (OC/EC)_{pri} for a given area. (OC/EC)_{pri} varies with the contributions of different 504 sources and can also be influenced by meteorological conditions (Dan et al., 2004). In this work, (OC/EC)pri was determined based on the lowest 5% of measured OC/EC 505 506 ratios for the winter and summer campaigns, respectively (Pio et al., 2011). The average SOC concentrations during summer and winter were calculated and are shown in Table 507 1. Daily concentrations of Other OC estimated by CMB and SOC estimated by the EC-508 tracer method in winter and summer are plotted in Fig. 6, as well as their correlation 509 510 relationship.







511 512 Figure 6. Time series of mean values for Other OC estimated by CMB and SOC
513 estimated by the EC-tracer method in winter (a) and summer (c); Correlation
514 relationship between Other OC estimated by CMB and SOC estimated by the EC-tracer
515 method in winter (b) and summer (d).

The average SOC concentrations in winter and summer are presented in Table1. The 516 average SOC concentration during winter was 7.2±5.7 µg m⁻³, accounted for 517 36.6±15.9% of total OC. The average SOC concentration during summer was one third 518 of that in winter, which was $2.3\pm1.4 \ \mu g \ m^{-3}$, accounting for $36.2\pm16.0\%$ of total OC. 519 The mean SOC concentrations during winter haze and non-haze periods were 10.3±5.7 520 μ g m⁻³ and 2.9±1.4 μ g m⁻³, contributing to 34.0±12.0% and 40.5±20.4% of OC during 521 haze and non-haze episodes, respectively. As shown in Fig. 6, the SOC estimated by 522 the EC tracer method followed a similar trend to the Other OC calculated by the CMB 523 model. They were well-correlated in both seasons with R^2 of 0.58 and 0.73 in winter 524 525 and summer samples, respectively and gradients of 1.16 and 0.80. This suggests that the estimates of Other OC calculated from the CMB outputs were reasonable and 526 mainly represented the secondary organic aerosol. 527

528 3.4 Comparison with the source apportionment results in rural Beijing

The OC source apportionment results in this study are also compared with those in another study conducted at a rural site of Beijing - Pinggu during APHH-Beijing campaigns (Wu et al., 2020). CMB was run based on the results from high-time resolution PM_{2.5} samples that were collected in Pinggu during the same sampling period, but not on identical days. The comparison results are presented in Table 3.

As shown in Table 3, slightly more OC was explained by CMB at the urban site (75.7%) than the rural site (69.1%) during winter, but less OC was explained at the urban site (56.1%) than the rural site (63.4%) during summer. As at the urban site, biomass burning and coal combustion are important primary sources in rural Beijing. Diesel contributed more to OC at the rural site, while cooking contributed more at the urban site. The rural site also had a larger contribution from vegetative detritus to OC than the urban site. The source contribution estimates from biomass burning at the rural





site was approximately 2 and 4 times that at the urban site during winter and summer. 541 In winter, biomass burning contributed a similar percentage of OC at both sites. A 542 higher percentage of OC from biomass burning was found at the rural site than the 543 urban site in summer, possibly because of use of biomass for cooking. For traffic 544 545 emitted OC, gasoline exceeded diesel at the urban site, while the rural site by contrast 546 has a larger diesel contribution. Industrial CC emitted OC is higher at the urban site 547 during winter, but lower in summer compared to the rural site. The source contribution estimates of residential CC at the urban site is only half that of the rural site in both 548 seasons, and its relative contribution to OC was also lower at the urban site. Coal is 549 widely used for cooking and heating at the villages around the rural site at the time of 550 551 observations. Cooking accounted for over 10% of OC at the urban site, but less than 5% 552 at the rural site, which is plausible as the urban site is more densely populated.

Table 3. Comparison of the source contribution estimates (SCE in μ g m⁻³ (%OC)) at IAP with those at a rural site in Beijing- Pinggu

	IAP (Urban) (This study)		Pinggu (Rural)	
	Winter (31 days)	Summer (34 days)	Winter (14 days)	Summer (6 days)
OC	21.5±12.3	6.4±2.3	36.5±29.3	10.7±4.9
OC explained	75.7±11.0%	56.1±11.3%	69.1±7.1%	63.4±12.6%
Vegetative detritus	0.1±0.1 (0.5±0.4%)	0.1±0.1 (1.7±0.8%)	1.5±3.0 (2.8±3.4%)	0.3±0.3 (2.1±1.4%)
Biomass burning	3.8±2.6 (17.4±8.7%)	0.3±0.4 (4.8±3.4%)	6.8±5.6 (18.1±3.4%)	1.1±0.6 (10.7±2.6%)
Gasoline	2.0±1.6 (10.2±6.6%)	0.3±0.2 (4.9±2.2%)	1.0±0.9 (3.4±1.6%)	0.1±0.0 (1.3±0.6%)
Diesel	0.5±1.2 (1.9±3.7%)	0.1±0.2 (1.2±2.5%)	6.2±6.0 (13.7±6.0%)	0.6±0.3 (6.2±4.8%)
Industrial CC	4.9±4.1 (22.0±11.2%)	1.8±0.7 (29.0±9.0%)	3.2±2.6 (10.2±5.7%)	3.8±2.5 (34.1±11.0%)
Residential CC	2.6±3.1 (12.5±10.2%)	0.2±0.1 (3.3±3.5%)	5.7±4.3 (19.0±12.4%)	0.4±0.2 (4.2±1.8%)
Cooking	2.2±2.1 (10.6±7.3%)	0.7±0.4 (11.1±7.1%)	0.5±0.5 (2.0±2.3%)	0.5±0.4 (4.9±3.9%)
Other OC	5.3±4.9 (24.8±12.1%)	2.9 ± 1.5 (43.9±11.4%)	11.7±10.4 (30.9±7.1%)	3.9±2.3 (36.6±12.6%)

555 **3.5** Comparison with source apportionment results from AMS-PMF

Results from AMS-PMF were compared with the CMB source apportionment results 556 to investigate the consistency and potential uncertainties of both methods, and also to 557 provide supplemental source apportionment results. Six factors in non-refractory (NR)-558 PM_1 from the AMS were identified based on the mass spectra measured in winter at 559 560 IAP by applying a PMF model, including coal combustion OA (CCOA), cooking OA (COA), biomass burning OA (BBOA) and 3 secondary factors of oxidized primary OA 561 (OPOA), less-oxidized OA (LOOOA), and more-oxidized OA (MOOOA). In summer, 562 the PMF analysis resulted in 5 factors including 2 primary factors of hydrocarbon-like 563 564 OA (HOA), cooking OA (COA) and 3 secondary factors of oxygenated OA (OOA): 565 OOA1, OOA2, OOA3. In order to compare with the source apportionment results of OC in this study from the CMB model, the OA concentrations from the AMS-PMF 566 were converted to OC based on various OA/OC ratios measured in Beijing: 1.35 for 567 CCOA/CCOC (coal combustion organic carbon), 1.31 for HOA/HOC (hydrocarbon-568 like organic carbon) (Sun et al., 2016), 1.38 for COA/COC (cooking organic carbon), 569 570 1.58 for BBOA/BBOC (biomass burning organic carbon) (Xu et al., 2019b), and 1.78 for OOA/OOC (Huang et al., 2010). The concentrations of OA and corresponding OC 571 from AMS-PMF analysis are presented in Table 4. 572

Table 4. Source contributions of OA and OC (µg m⁻³) from AMS-PMF results in urban
 Beijing during winter and summer

Winter/µg m-3

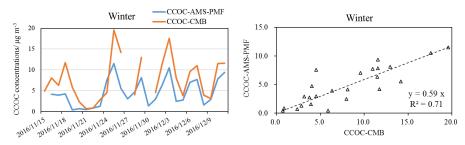




CCOA	6.2±4.4	CCOC	4.6±3.3
COA	5.9±4.1	COC	4.3±3.0
BBOA	6.5±5.8	BBOC	4.1±3.7
OPOA	4.6±2.1	OPOC	2.6±1.2
LOOOA	5.2±5.2	LOOOC	2.9±2.9
MOOOA	8.1±7.0	MOOOC	4.6±4.0
OOA ^a	18.0±13.2	OOC ^d	10.1±7.4
OM^b	36.7±24.0		
Summer/µg m ⁻³			
HOA	0.7±0.4	HOC	0.5±0.3
COA	1.8±1.0	COC	1.3±0.7
OOA1	3.3±1.4	OOC1	$1.9{\pm}0.8$
OOA2	2.4±2.4	OOC2	1.4±1.3
OOA3	1.9±1.1	OOC3	1.1±0.6
OOA ^c	7.6±3.7	OOC	4.3±2.1
OM	10.1±3.9		

^a OOA=OPOA+LOOOA+MOOOA; ^b OM is organics measured by AMS; ^c OOA=OOA1+OOA2+OOA3; 575 576 ^d OOC=OOC1+OOC2+OOC3

The CCOA factor was mainly characterized by m/z of 44, 73 and 115 (Sun et al., 577 2016). In winter, CCOA was 6.2±4.4 µg m⁻³, contributing 16.9% of OM. CCOC was 578 $4.6\pm3.3 \,\mu\text{g m}^{-3}$, which was much lower than the estimated coal combustion OC (7.5±5.0 579 μ g m⁻³, industrial and residential coal combustion OC) by CMB. The time series of coal 580 combustion related OC (CCOC) estimated by CMB and CCOC from AMS-PMF 581 analysis in Fig. 7 showed a similar trend with relatively good correlation of $R^2 = 0.71$, 582 but coal combustion estimated by CMB was consistently higher than by AMS-PMF, 583 584 probably because AMS-PMF only resolved the sources of NR-PM_{1.0}, and some coal combustion particles are larger (Xu et al., 2011). The correlation coefficients (R²) of 585 CCOC from AMS-PMF with Cl⁻ and NR-Cl⁻ were 0.49 and 0.65, respectively in the 586 587 winter data.



588 589

Figure 7. Time series and correlation of coal combustion related OC (CCOC) estimated by CMB and CCOC from AMS-PMF analysis 590

BBOA from the AMS data in winter was 6.5±5.8 µg m⁻³, contributing 17.7% of OM. 591 This BBOA factor included a high proportion of m/z 60 and 73, which are typical 592 fragments of anhydrous sugars like levoglucosan (Srivastava et al., 2019). BBOC was 593 594 $4.1\pm3.7 \,\mu\text{g}\,\text{m}^{-3}$, which was very close to the estimated BBOC ($3.78\pm2.64 \,\mu\text{g}\,\text{m}^{-3}$, 17.6%595 of OC) by CMB in this study.

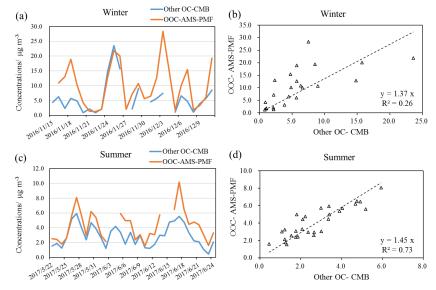




596 COA is as a common factor identified in both winter and summer AMS-PMF results. 597 It is characterized by high m/z of 55 and 57 in the mass spectrum (Sun et al., 2016). 598 COA was 5.9 ± 4.1 and $1.8\pm1.0 \ \mu g \ m^{-3}$ in winter and summer, respectively, contributing 599 16.1% and 17.8% of OM. COC was 4.3 ± 3.0 and $1.3\pm0.7 \ \mu g \ m^{-3}$ in winter and summer, 600 respectively, which were almost 2 times of those for winter ($2.23\pm2.13 \ \mu g \ m^{-3}$) and 601 summer ($0.66\pm0.43 \ \mu g \ m^{-3}$) in the CMB results.

HOA was 0.7±0.4 µg m⁻³ in summer, accounting for 6.9% of OM. HOA is usually 602 identified based on the high contribution of aliphatic hydrocarbons in this factor, 603 particularly m/z of 27, 41, 55, 57, 69 and 71 (Aiken et al., 2009). This result is lower 604 than that (17% of OM) in rural Beijing during summer 2015 (Hua et al., 2018). HOC 605 was $0.5\pm0.3 \ \mu g \ m^{-3}$ in summer, which is higher than the traffic (gasoline+diesel) 606 emitted OC (0.4±0.2 µg m⁻³) from the CMB model. No obvious correlation was 607 observed between HOC with nitrate and traffic emitted OC from the CMB model during 608 summer. 609

AMS OOA concentrations (the sum of all oxidized OA) were 18.0±13.2 and 7.6±3.7 610 611 ug m⁻³ in winter and summer, respectively, accounting for 49.0% and 75.2% of OM. 612 The derived OOC concentrations in winter and summer were 10.1 ± 7.4 and 4.3 ± 2.1 µg m⁻³ in winter and summer, respectively, higher than the Other OC estimated by CMB 613 in winter $(5.3\pm4.9 \ \mu g \ m^{-3})$ and summer $(2.9\pm1.5 \ \mu g \ m^{-3})$ in this study. The time series 614 and correlation of Other OC estimated by CMB results and OOC from AMS-PMF 615 results is plotted in Fig. 8. A similar temporal trend was found between them, especially 616 in summer, which was also observed with a better correlation ($R^2=0.73$). 617



618

Figure 8. Time series of mean values for Other OC estimated by CMB, and OOC estimated by AMS-PMF in winter (a) and summer (c); Correlation relationship between Other OC estimated by CMB and OOC estimated by AMS-PMF in winter (b) and summer (d).





In summary, CMB is able to resolve almost all major known primary OA sources, but 623 AMS-PMF can resolve more secondary OA sources. The AMS-PMF results for major 624 components, such as CCOC and OOC agreed well with the results from CMB in the 625 winter. However, discrepancies or poor agreement was found for other sources, such as 626 627 BBOA and COA, although the temporal features were very similar. Furthermore, AMS-PMF did not identify certain sources, probably due to their relatively small contribution 628 629 to particle mass. Overall, CMB and AMS-PMF offered complementary data to resolve 630 both primary and secondary sources.

631 **3.6 Source contributions to PM_{2.5} from the CMB model**

The source contributions to $PM_{2.5}$ were calculated by multiplication of the fine OC 632 source estimates from CMB by the ratios of fine OC to PM_{2.5} mass (Table S3), which 633 634 were obtained from the same source profiles used for the OC apportionment by CMB 635 (Zhang et al., 2007b; Wang et al., 2009; Cai et al., 2017; Zhang et al., 2008). For cooking, 636 an OM/OC ratio of 1.4 was applied (Zhao et al., 2007). For vegetative detritus, OM/OC ratio of 2.1 was applied (Bae et al., 2006b). The OM/OC ratios for oxygenated OA were 637 in the range of 1.85-2.3 (Zhang et al., 2005; Aiken et al., 2008), and the OM/OC ratio 638 was 2.17 in secondary organic aerosols of $PM_{2.5}$ (Bae et al., 2006a). Therefore, an 639 640 OM/OC ratio of 2.2 is applied in this study to convert the Other OC to OM. Instead of 641 OC/PM2.5, applying an OM/OC ratio for the calculation may result in an 642 underestimation of PM_{2.5} source contributions, because sources like cooking and 643 vegetative detritus can also emit inorganic pollutants. However, cooking emissions are 644 mostly organic and the contribution from vegetative detritus to PM_{2.5} is very small, their 645 effects on source contribution estimation here are considered negligible. The daily 646 $PM_{2.5}$ contribution estimates are provided in Fig. 9. The seasonal average source 647 contributions and their relative abundance in reconstructed PM2.5 are summarized in Table S4. 648

As shown in Table S4, PM_{2.5} mass was well explained by those sources which 649 accounted for 91.9 \pm 24.1% and 99.0 \pm 19.1% of online PM_{2.5} in winter and summer, 650 respectively. In the summer, the offline PM_{25} is lower than online observations. Thus, 651 the CMB-based source contributions are more than offline $PM_{2.5}$ mass (121.7±26.6%). 652 On average, the source contributions in winter ranked as SNA (30.5 μ g m⁻³, 34.1% of 653 reconstructed PM2.5 and hereafter), coal combustion (industrial & residential CC; 17.7 654 μg m⁻³, 21.4%), Other OM (14.6 μg m⁻³, 14.8%), biomass burning (8.9 μg m⁻³, 11.0%), 655 gasoline & diesel (5.6 µg m⁻³, 7.5%), geological minerals (5.3 µg m⁻³, 7.0%), cooking 656 $(3.1 \ \mu g \ m^{-3}, 3.9\%)$ and vegetative detritus $(0.2 \ \mu g \ m^{-3}, 0.3\%)$; in summer these ranked 657 as SNA (17.7 µg m⁻³, 48.5%), other OM (8.0 µg m⁻³, 18.3%), coal combustion (4.7 µg 658 m⁻³, 14.6%), geological minerals (3.5 µg m⁻³, 10.4%), cooking (0.9 µg m⁻³, 2.8%), 659 gasoline & diesel (0.8 µg m⁻³, 2.6%), biomass burning (0.8 µg m⁻³, 2.1%) and vegetative 660 detritus (0.2 µg m⁻³, 0.7%). 661

662Zheng et al. (2005) investigated the seasonal trends of PM2.5 source contributions in663Beijing during 2000 applying a CMB model. In winter (January), the contributions from664coal combustion, biomass burning, diesel & gasoline, vegetative detritus to PM2.5 were6659.55 μ g m⁻³ (16% of PM2.5 and hereafter), 5.8 μ g m⁻³ (9%), 3.85 μ g m⁻³, 0.33 μ g m⁻³,666respectively. Contributions from gasoline, diesel, coal combustion and biomass burning667were enhanced in Beijing during winter in 2016 compared to 2000, while the668contribution from vegetative detritus basically remained similar. In summer (July) 2000,





669 coal combustion contributed 2% of $PM_{2.5}$ (2.39 µg m⁻³), much less than that in summer 670 2016 of this study. The contribution from diesel & gasoline (7.78 µg m⁻³, Zheng et al., 671 2005) was approximately 10 times of that in 2016 (0.8 µg m⁻³). Similarly, contributions 672 from vegetative detritus and biomass burning were small and insignificant.

Zhou et al. (2017) estimated that coal combustion contributions in winter and 673 summer of Beijing-Tianjin-Hebei area in 2013 were 15.9 µg m⁻³ and 2.1 µg m⁻³, 674 respectively, which are comparable with those in this study. These results are also 675 comparable with the PMF-resolved coal and oil combustion in Beijing during winter 676 (17.4 µg m⁻³) and summer (2.2 µg m⁻³) in 2010 (Yu et al., 2013). SNA contributed 52.7 677 and 26.4 µg m⁻³ of PM_{2.5} during winter (January) and summer (July), respectively (Yu 678 et al., 2013), which are much higher than those in this study. It is noteworthy that a 679 680 severe haze pollution event occurred during January 2013, which was characterized by high concentrations of sulfate and nitrate in several studies (Zhou et al., 2017; Han et 681 al., 2016). The contribution from biomass burning in winter is consistent (8.5 μ g m⁻³) 682 with this study (8.9 μ g m⁻³), but higher in summer (2.6 μ g m⁻³) (0.8 μ g m⁻³). The 683 cooking source contributed 4.8 and 1.3 µg m⁻³ in PM_{2.5} during winter and summer 2013, 684 respectively, which is also comparable with this study. 685

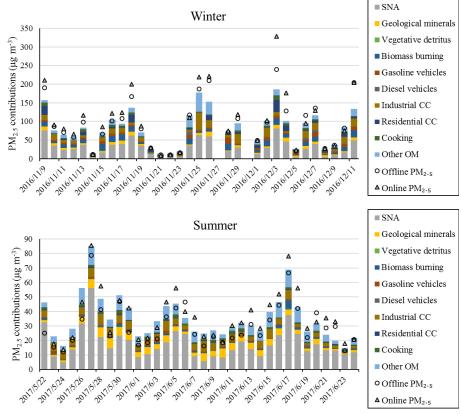




Figure 9. Daily PM_{2.5} source contribution estimates from the CMB model

688





689 4 Summary

Carbonaceous aerosols contributed approximately 59% and 41% of reconstructed PM2.5 690 691 in winter and summer at the urban IAP site in Beijing. The OC and EC concentrations were comparable with more recent studies (Fan et al., 2020; Qi et al., 2018), but lower 692 than those before 2013 (Yang et al., 2016; Dan et al., 2004), suggesting the 693 effectiveness of air pollution control measures since 2013 (Vu et al., 2019; Zhang et al., 694 695 2019). CMB modelling showed that in the winter 2016, the top three primary contributors to PM_{2.5}-OC were coal combustion (35%), biomass burning (17%), and 696 traffic (12%); these were in the same order with that at the rural site during the same 697 study period: coal combustion (29%), biomass burning (18%), and traffic (17%) (Wu 698 699 et al., 2020). In the summer 2017, the top three primary contributors to $PM_{2.5}$ -OC were 700 coal combustion (32%), cooking (11%), and traffic (6%); these were different to that at the rural site during the same study period: coal combustion (38%), biomass burning 701 (11%), and traffic (7%) (Wu et al., 2020). The Other OC, which was well-correlated 702 703 $(R^2: 0.6 \sim 0.7; slope: 0.8 \sim 1.2)$ with the secondary OC (SOC) estimated based on the ECtracer method, accounted for 25% and 44% of OC at urban site and 31% and 37% of 704 OC at rural site during winter and summer, respectively. Although the annual average 705 PM_{2.5} levels in Beijing reduced from 88 µg m⁻³ in year 2013 to 58 µg m⁻³ in year 2017 706 (Vu et al., 2019), and the deweathered concentration of PM₁ decreased by -38% in 2017 707 708 comparing to 2007 (Zhang et al., 2020), our CMB modelling results indicate that the coal combustion and biomass burning still remained the dominant primary OC sources 709 710 in 2016 winter and 2017 summer, with road traffic ranked as the third highest. Cooking was more significant than biomass burning at the urban site during summer. Compared 711 to other CMB studies in Beijing, it revealed an increase of the contributions from coal 712 combustion, biomass burning and traffic to PM2.5 in winter 2016 compared to winter 713 714 2000, while those in this study remained similar compared to winter 2013. Sulfate, nitrate and ammonium concentrations were significantly lower in this study compared 715 to 2013 (Zheng et al., 2005; Zhou et al., 2017). It is however notable that there is a 716 717 broad consistency in the findings of the CMB studies, whereas the more numerous 718 studies which have used PMF come to rather diverse conclusions (Srivastava et al., 719 2020).

720

721 *Data availability.* The data in this article are available from the corresponding authors 722 upon request.

723

Author contributions. JX did the CMB modelling and drafted the paper with the help
 of ZS, RMH and all co-authors. DL, TVV conducted the laboratory analysis of organics
 and inorganics, respectively. XW, YZ provided the CMB source profiles. YS provided
 the AMS-PMF data.

728

729 *Competing interests.* The authors have no conflict of interests.

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