- 
- Jürgen Schnelle-Kreis, Pamela A. Dominutti, Jean-Luc Jaffrezo, Sönke Szidat, Yang Chen, Junji Cao, Urs Baltensperger, Gaëlle Uzu, Kaspar R.
- 8 Daellenbach, Imad El Haddad\*, André S. H. Prévôt\*

# Supplementary Information **Local incomplete combustion emissions define the PM2.5 oxidative potential in Northern India** Deepika Bhattu\*, Sachchida Nand Tripathi\*, Himadri Sekhar Bhowmik, Vaios Moschos, Chuan Ping Lee, Martin Rauber, Gary Salazar, Gülcin Abbaszade, Tianqu Cui, Jay G. Slowik, Pawan Vats, Suneeti Mishra, Vipul Lalchandani, Rangu Satish, Pragati Rai, Roberto Casotto, Anna Tobler, Varun Kumar, Yufang Hao, Lu Qi, Peeyush Khare, Manousos Ioannis Manousakas, Qiyuan Wang, Yuemei Han, Jie Tian, Sophie Darfeuil, Mari Cruz Minguillon, Christoph Hueglin, Sébastien Conil, Neeraj Rastogi, Atul Kumar Srivastava, Dilip Ganguly, Sasa Bjelic, Francesco Canonaco,

9 \*Corresponding author. Email: [dbhattu@iitj.ac.in](mailto:dbhattu@iitj.ac.in) (D.B.); [snt@iitk.ac.in](mailto:snt@iitk.ac.in) (S.N.T.); [imad.el-haddad@psi.ch](mailto:imad.el-haddad@psi.ch) (I.E.H.); [andre.prevot@psi.ch](mailto:andre.prevot@psi.ch) (A. S. H. P.)

## **This PDF file includes:**

- 1. Supplementary Methods (Method 1-5)
- 2. Supplementary Figures (Figure 1-12)
- 3. Supplementary Tables (Table 1-8)
- 4. Supplementary References

#### 

## **Supplementary Method 1: Sampling site details**

 The Delhi urban roadside site experiences high wind speeds and less variable RH conditions as compared to the urban background and sub-urban industrial site. The next sub-urban sampling site (IITK) is ~600 km to the South-East. It is also a residential campus in proximity of roads with 35 moderate traffic  $(\sim 800 \text{ m})$  and upwind of a coal power plant (Panki). More details on the sampling sites are provided in Bhowmik et al.<sup>1</sup> and 36 Tripathi et al.<sup>2</sup>.

 New Delhi is situated 160 km south of the Himalayas and near the Thar Desert. It is surrounded by the adjoining state of Haryana on three sides (North, West and South) and Uttar Pradesh on the other side. During winter, New Delhi experiences prevailing northwesterly winds originating in the Himalayas and crossing northern plain whereas during summer, south-westerly winds prevail (Supplementary Fig. 9). To understand the effect of local sources and regional transport of the pollutants in the capital city, 5 sites were selected. Two sites were in Delhi (North-west: Indian Institute of Tropical Meteorology Delhi (IITMD) representative of urban background, South-west: Indian Institute of Technology Delhi (IITD) as urban roadside). Two sites encompassing Delhi were in the north-west direction (Chaudhary Charan Singh Haryana Agricultural University (CCSHAU) Hisar, Haryana) and south-east direction (Manav Rachna International University, Faridabad (MRIUF)) as upwind rural background site and a sub-urban industrial site, respectively. The last downwind site in south-east direction is a sub-urban Kanpur site (Indian Institute of Technology Kanpur (IITK).

 The Delhi urban background site is in central urban part of Delhi, surrounded by forest areas and residential colonies whereas Delhi urban roadside site is an urban residential site surrounded by roads with heavy traffic (<100 m away) and industries located in the north-west and east direction.

39 High volume PM<sub>2.5</sub> samplers (Thermo at the Delhi urban roadside site and Tisch at the other 4 sites with a flowrate of 1.13 m<sup>3</sup> min<sup>-1</sup>) were used to 40 collect 893 samples (865 samples and 28 field blanks) from January to May 2018 on quartz fiber filters (8" x 10", pre-combusted (~350 °C) Pallflex® at Delhi urban roadside site and Whatman QMA at the other 4 sites). The samplers were installed at the roof top of the Agromet Observatory, Department of Agricultural Meteorology, CCSHAU (29.10◦N, 75.46◦E; ~215 m amsl) Hisar (~5 m above ground), main building of IITM Delhi (~15 m above ground; 28.63◦N, 77.167◦E; ~220 m amsl), Centre for Atmospheric Science Building (~15 m above ground) at IIT Delhi (28.54◦N, 77.19◦E; ~218 m amsl), C Block of MRIUF (~7 m above ground; 28.45◦N, 77.28◦E ~278 m amsl) and Centre for Environmental Science and Engineering (CESE), IIT Kanpur (~8 m above ground; 26.51◦N, 80.23◦E ~142 m amsl). Filters were wrapped in aluminium foils, sealed in a 46 zipped plastic bag and stored at -18 to -20 °C at all sampling sites. Filters were shipped to IIT Kanpur once every two months in a thermally 47 insulated box filled with dry ice to avoid loss of semi-volatile species. These filters were again stored at -20 °C on receipt and later kept at room 48 temperature (20 °C) for 30 minutes before cutting the whole filter area into two halves. This procedure was performed in a temperature and RH- controlled clean room and one half of all the filters was wrapped in a new aluminium foil, sealed in a zipped plastic bag, and packed with dry ice for shipment to PSI, Switzerland. A total of 330 filter samples (including field blanks and repeats), referred to as "2018 filters" in the following text, were selected from: (1) every  $6<sup>th</sup>$  sampling day (day and night filters until mid-March and 1 filter from every third day for the rest of the 52 period), and (2) additional periods of interest identified for analysis, along with the days immediately prior and posterior to such periods. PM<sub>2.5</sub> mass was reconstructed based on all the measured components (organics, inorganic ions, and elements as their oxides).

 Further, sub-urban industrial site is a residential campus near the main road with both heavy- and light-duty vehicles and surrounded by industries as well as sewage treatment plants in the north direction. The north-west rural background upwind site is an Agromet Observatory situated in the agricultural research farm at CCSHAU, Hisar.

55 We compared TOC analyzer-measured WSOC and WSOC derived from AMS [i.e., total AMS-OA/ (OM/OC)bulk\*(<sup>15</sup>NO + <sup>15</sup>NO<sub>2</sub>)], where (<sup>15</sup>NO  $+ 15NO_2$ ) was used to correct for any changes in the AMS signal intensity. It was found that the periods with high concentrations of dominant 57 cations ( $Ca^{2+}$ , Mg<sup>2+</sup> and K<sup>+</sup>) have undergone recombination with labelled ions, yielding compounds that are not readily vaporized, thus resulting in negative bias for the AMS-derived WSOC.

## **Supplementary Method 2: Filter collection, storage, and transportation**

#### **Supplementary Method 3: WSOC measurement vs estimation**

#### **Supplementary Method 4: Aerosol instrumentation and measurements**

**A) AMS and EESI-ToF measurements:** The AMS was set up to provide average (40 seconds) mass spectra in the mass to charge (*m/z*) range of

12- 440 (V-mode). The acquisition rate for the EESI mass spectra for up to *m/z* 960 was 1 Hz. Overall, a total of 394 samples (including 28 field

- 62 blanks, 36 sample repeats and 12 pure labelled and 13 unlabeled  $NH_4^{15}NO_3$  and  $(NH_4)_2^{34}SO_4$  standards) were measured. The details of AMS 63 operating principles, calibration procedures, and procedures for data analysis are provided elsewhere<sup>3</sup>, as well as the detailed description of the 64 EESI-ToF and its operating principles<sup>4</sup>. Briefly, the EESI used a 200 ppm NaI solution consisting of a 1:1 water to acetonitrile mixture as 65 electrospray solution (ES) and was configured for positive ion detection. The aerosol stream from the nebulizer intersected a jet spray of charged 66 droplets generated from an untreated fused silica capillary (360 µm outer diameter and 50 µm inner diameter) (BGB Analytik, AG) contained in 67 an electric field. The aerosol components soluble in the electrospray jet of charged droplets were extracted, ionized, and ejected into the gas phase 68 for detection by the LToF-MS. A potential difference of 3.0 kV relative to the MS interface was applied to the ES solution and an air pressure 69 difference of 500 mbar was used to drive the ES solution through the capillary to make ES droplets. The stability and performance of both 70 instruments were tracked by nebulizing NH<sub>4</sub>NO<sub>3</sub> and (NH<sub>4</sub>)<sub>2</sub>SO<sub>4</sub> standard solutions. Due to negligible thermal degradation and ion-induced 71 fragmentation<sup>4,5</sup>, near-molecular level (i.e., molecular ion formulae) OA composition was obtained. For AMS data analysis, Squirrel v1.59B was used for  $m/z$  calibration and baseline subtraction, and PIKA (Peak Integration by Key Analysis) v1.19B for high resolution mass spectrum analysis<sup>6</sup> 72 73 in the IGOR Pro software package 6.37. A total of 785 fragments and the isotope ions up to *m/z* 182 were fitted in HR peak fitting procedure. EESI 74 data processing was performed using Tofware version 2.5.7 (Tofwerk AG, Switzerland). Ten seconds pre-averaged data was used for high-75 resolution peak fitting. A total of 2003 ions were fitted in the selected mass transmission window between *m/z* 120 and 444.
- **B) Radiocarbon measurements:** The <sup>14</sup>C content of TC was measured using a one-step combustion protocol under pure  $O_2$  (99.9995 %) at 760 <sup>77</sup> °C for 400<sup>s<sup>7</sup></sup> using an elemental analyzer coupled with the accelerator mass spectrometer Mini Carbon Dating System (MICADAS) at the 78 Laboratory for the Analysis of Radiocarbon (LARA; University of Bern, Switzerland)<sup>8,9</sup>. The EC fraction was separated by a thermo-optical 79 OC/EC analyzer (Model 5L, Sunset Laboratory, USA) coupled online with the MICADAS<sup>10</sup>.
- 
- 80 For the determination of  ${}^{14}C$  of EC, the Swiss\_4S protocol<sup>8</sup> had been developed for samples from Switzerland or other samples with similar 81 characteristics/mass loading. However, preliminary analysis with water extraction showed that OC in the Indian PM samples was less water-82 soluble and more methanol-soluble than the Swiss samples. Therefore, the temperatures applied in the Swiss protocol for OC removal resulted in 83 high charring (i.e., a large fraction of OC got converted to EC, eventually leading to artificially high EC yields, while a large fraction of the EC 84 was lost in this OC removal step. Therefore, a modified extraction and desorption temperature protocol was developed (Bern-India\_4S) to achieve 85 efficient EC separation for these samples: we used methanol instead of water extraction for OC removal with the Sunset OC/EC analyzer and 86 reduced the temperature in step 2 from 475 °C to 425 °C. This led to efficient removal with minimized charring and acceptable EC yields.
- The <sup>14</sup>C results are expressed as fractions of modern  $(f_M)$ , i.e., the ratio of the <sup>14</sup>C/<sup>12</sup>C content of the sample related to the isotope ratio of the 88 reference year  $1950^{11}$ . The data analysis was carried out accounting for the blank correction (one field blank per site was analyzed, not relevant 89 for EC), decay of <sup>14</sup>C since the 1950s, nuclear bomb correction, OC charring (average and 1 standard deviation:  $4 \pm 3$ %; lower at the urban 90 roadside), and EC yield after OC removal (average and 1 standard deviation:  $72 \pm 9\%$ ; no significant site-to-site variability)<sup>12,13</sup>.
- 91 Non-fossil fractions ( $f_{NF}$ ) were determined from their corresponding  $f_M$  values and reference values for pure non-fossil sources ( $f_{NF,ref}$ ) by  $f_{NF} = f_M$  $\gamma$  /  $f_{\text{NF,ref.}}$  For the bomb-peak correction<sup>12</sup>, based on the AMS-PMF results we considered that biomass-burning (bb) and biogenic (bio) sources 93 contribute 90% and 10%, respectively, to the  $f_{\text{NF,ref}}$ ;  $f_{\text{M,bb}}$  and  $f_{\text{M,bio}}$  values from Zotter et al.<sup>13</sup>. Based on our tests, we used the lower value of 0.07 94 for the EC yield slope from Zotter et al.<sup>13</sup> to correct the  $f_M$  values for charring and EC losses. The fraction of fossil-fuel sources was calculated by 95 *f*<sub>FF</sub> = 1 − *f*<sub>NF</sub>. The apportionment to fossil and non-fossil OC was based on mass balance calculations (OC = TC – EC) (Supplementary Fig. 10). 96 The uncertainties were determined by error propagation. The mass concentration uncertainties were assumed to be 10% for EC and 6% for OC 97 and TC (ref.<sup>12</sup>, typical values for EUSAAR2).
- 98 **C) Organic Carbon- Elemental Carbon (OC-EC):** Organic and elemental carbon was measured by a Sunset EC-OC analyzer (Model 4F, Sunset 99 Laboratory Inc. USA) using the EUSAAR-2 thermal-optical transmittance protocol<sup>14</sup>. Details of the analysis method and choice of protocol are 100 described elsewhere<sup>1</sup>.
- 101 **D) Water-soluble OC (WSOC), water soluble inorganic carbon (WSIC) and ions:** Water-soluble organic carbon was measured by a total 102 organic carbon analyzer (Model: Shimadzu-TOC-L-CPN, Shimadzu Corporation) by catalytically oxidizing water-soluble total carbon and 103 measuring the resultant  $CO_2$  using a non-dispersive infrared (NDIR) detector<sup>1</sup>. WSIC was measured from  $CO_2$  evolved after acidification of water 104 extracts. The concentrations of major ions  $(K^+, Na^+, NH_4^+, Cl^-, NO_3, SO_4^2)$  were measured by ion chromatography<sup>1</sup>.
- 105 **E) Trace elements:** 29 trace elements (Li, Mg, As, Ca, Sc, Ti, V, Cr, Mn, Fe, Co, Ni, Cu, Se, Rb, Sr, Zr, Mo, Pd, Cd, ln, Sn, Sb, Cs, Ba, Ce, Pt, 106 Tl and Pb) were measured by inductively coupled plasma mass spectrometry (ICP-MS; Thermo Scientific iCAP Q ICP-MS). Briefly, filter samples 107 were digested by an acid mix of HF, HNO<sub>3</sub> and HClO<sub>4</sub> solution. Percentage recovery was calculated compared with high purity multi-element (35 108 trace elements) standards (soluble in  $1\%$  HNO<sub>3</sub>,  $100$  ppm) (Sigma Aldrich)<sup>15</sup>.
- 109 **F) Targeted organic compounds:** A range of 81 targeted organic compounds including 15 dicarboxylic acids, 22 polycyclic aromatic 110 hydrocarbons (PAHs), 8 oxygenated polycyclic aromatic hydrocarbons (oxy-PAHs), 3 anhydrous sugars, 2 resin acids, 2 alkaloids, 10 hopanes, 111 10 n-alkanes, 7 higher n-alkanes and 2 lignin pyrolysis products were determined to support the source apportionment results. A total of 140 112 samples were measured from the CCSHAU, IITMD, IITD and MRIUF sites. Details of the methods are described in ref.<sup>16,17</sup>.
- 113 **G) PM2.5 oxidative potential:** A proxy for health-relevant exposure of PM2.5 through oxidative damages was assessed by determining the oxidative 114 potential of PM collected on filters, using three different acellular assays (dithiothreitol: DTT; 2´,7´-dichlororfluorescin: DCFH and ascorbic acid: 115 AA). Details of the methods are described elsewhere<sup>18,19</sup>. Briefly, all PM samples, including field blanks, were extracted at iso-concentration (25 116  $\mu$ g ml<sup>-1</sup>) using simulated lung fluid (Gamble solution + dipalmitoylphosphatidylcholine (DPPC) solution) and further subjected to different assays 117 in triplicate, without filtering the extracts to keep the PM in contact with the lung antioxidant or surrogate. Multi-assay measurements were

118 performed to obtain a comprehensive picture of the effects of different reactive oxygen species (ROS) generation mechanisms: particle-bound 119 ROS (DCFH) and catalytic generation via redox-active constituents (AA and DTT). The DTT assay method measures the electron transfer ability 120 of PM, thereby producing ROS. It mimics interaction of surrogates (nicotinamide adenine dinucleotide: NADH and nicotinamide adenine 121 dinucleotide phosphate-oxidase: NADPH) with PM<sup>18</sup> and is sensitive to both soluble metals and organic carbon (HULIS, quinones and water-122 soluble organics)<sup>20</sup>, and, as a thiol-based reagent, provides a balanced answer to a wide range of pollutants. The AA assay is a simplified version 123 of the respiratory tract lining fluid (RTLF) assay which also measures the electron-transfer ability of PM from AA to  $O<sub>2</sub>^{18}$ . It is sensitive to redox-124 active transition metals (Fe and Cu) and mainly specific organic aerosol sources (biomass burning). DCFH, on the other hand, is a fluorescent 125 probe sensitive to peroxides  $(H_2O_2)$  and organic peroxides). It measures particle-bound ROS but is also known to be sensitive to reactive nitrogen 126  $species<sup>21</sup>$ .

#### 127

#### 128 **Supplementary Method 5: Data analysis**

129 **A) 2D hierarchical clustering on AMS-derived factor profiles:** For each AMS-derived factor, a specific cluster of closely associated ions was 130 observed and is discussed below.

135 GC-MS. The signal intensity of these representative ions was 10 times higher than other unique and abundant ions like *m/z* 126 {C7H10O2}, *m/z* 136 73  $\{C_2H_3NO_2\}$ ,  $m/z$  61  $\{C_2H_5O_2\}$ ,  $m/z$  60  $\{CH_2NO_2\}$  and  $m/z$  57  $\{C_3H_5O\}$ .

137 COOA was identified as a highly oxidized factor with  $m/z$  28 (CO<sup>+</sup>) and  $m/z$  44 (CO<sub>2</sub><sup>+</sup>) as representative ions.

131 CPOA was identified by the relatively high abundance of CHN ions signals ( $m/z$  95 {C<sub>5</sub>H<sub>7</sub>N<sub>2</sub>},  $m/z$  96 {C<sub>5</sub>H<sub>8</sub>N<sub>2</sub>},  $m/z$  108 {C<sub>6</sub>H<sub>8</sub>N<sub>2</sub>},  $m/z$  109 132  ${C_6H_9N_2}$ ,  $m/z$  110  ${C_6H_{10}N_2}$ ,  $m/z$  122  ${C_8H_{12}N}$ , and  $m/z$  123  ${C_8H_{13}N}$ ) and contributions by some hydrocarbons (HC) and CHO ions  $(C_xH_yO_1)$ 133 and  $C_xH_yO_3$ ), and a few CHNO ions.

134 BBOA had high abundance of *m/z* 60 {C2H4O2} and *m/z* 73 {C3H5O2}. It correlated (Pearson's *r*=0.75; *n*=140) with levoglucosan measured by

138 UOOA was identified by  $m/z$  43  $\{C_2H_3O\}$  and other unique ions with formula  $C_xH_yO_1$  and  $C_xH_yO_2$ , including  $m/z$  58  $\{C_3H_6O\}$ ,  $m/z$  71  $\{C_4H_7O\}$ , 139 *m/z* 85 {C4H5O2}, *m/z* 83 {C5H7O}, *m/z* 95 {C6H7O}, *m/z* 59 {C3H7O}, *m/z* 99 {C5H7O2}, *m/z* 97 {C6H9O}, *m/z* 111 {C6H7O2}.

140

141 **B) HOA estimation:** HOA estimation was performed using the fossil and non-fossil fractions of EC and levoglucosan by employing two 142 different approaches.

143 a) In approach 1, the following steps were used:

144 i) Correlation of  $EC_{nf}$  (y-axis) and levoglucosan (x-axis): A moderate correlation (Pearson's  $r = 0.7$ ) was obtained (*n*=44; selected samples from

145 upwind rural background, Delhi urban background and urban roadside) with a slope of  $1.08 \pm 0.17$  and an intercept of  $0.6 \pm 0.2$ .  $EC_{nf,estimated}$ 

158 ii) We obtained a slope (1.5  $\pm$  0.1; n=38; Pearson's  $r = 0.8$ ) of *HOC* vs.  $EC_f$ , with  $EC_f$  derived from radiocarbon (<sup>14</sup>C) analysis and HOC 159 obtained as follows

146 was calculated using the slope of 1.08 for the remaining 140 samples from the same sites that had the levoglucosan concentration measured.

147 Later,  $EC_{f,estimated}$  was estimated using Eq. S1 as follows.

148 
$$
EC_{f,estimated} = Sunset EC_{measured} - EC_{nf,estimated}
$$
 Eq. S1

Further, uncertainties ( $\sigma_{EC_{nf,estimated}}$  &  $\sigma_{EC_{f,estimated}}$ ) were estimated using Eq. S2 and Eq. S4.

150 
$$
\sigma_{EC_{nf,estimated}} = \sqrt[2]{\left(\frac{\sigma_{slope}}{slope}\right)^2 + \left(\frac{\sigma_{levo}}{C_{levo}}\right)^2}
$$
 Eq. S2

$$
151
$$

152 
$$
\sigma_{levoglucosan} = \left\{ \frac{5}{6} * MDL_{levo}, \text{ if } C_{levo} < MDL_{levo}^{22,23} \right\}
$$
 Eq. S3

153

- 177 Finally, approach 1 and 2 were compared to evaluate their compatibility to estimate HOA for the sites where either  $EC_f$  (sub-urban industrial site 178 in Delhi and downwind Kanpur site outside Delhi) or levoglucosan concentrations were not available. However, an  $R^2$  of 0.96 and slope of 0.98 179 while comparing estimated HOA suggested the suitability of approach 2 in the absence of specific source markers i.e., levoglucosan.
- 180

154 Otherwise, 
$$
\sqrt{(U_{levo} * C_{levo})^2 + (MDL_{levo})^2}
$$

156 where  $MDL_{levo}$  is the minimum detection limit of the instrument for levoglucosan (0.03 ng m<sup>-3</sup>) and  $U_{levo}$  is the analytical uncertainty (10%).

$$
\sigma_{EC_{f,estimated}} = \sqrt[2]{\sigma_{EC_{nf,estimated}}^2 + {\sigma_{EC}}^2}
$$
 Eq. S4

$$
HOC = \frac{HOA_{AMSDelhi\;urban\,roadside}/ACSM_{Delhi\;urban\,background}}{\left(\frac{OM}{OC}=1.25\right)_{HOA}} \qquad \qquad \mathbf{Eq. S5}
$$

161 iii)  $HOC_{estimated}$  for the remaining 140 samples was obtained by multiplying  $EC_{f,estimated}$  computed in step (i) with the slope of 1.5 obtained 162 in step (ii) (Eq.6).  $HOA_{estimated}$  was further computed using the OM:OC ratio of 1.25 again and uncertainty was estimated using Eq. S7.  $163$   $163$   $163$   $164$   $165$   $169$ 

187 Where *a, b, c* and *d* are the fitting coefficients. The fossil fraction of each factor was computed by subtracting their respective non-fossil fraction from 1. These values represent an average over both the cold and warm period which might have a certain seasonality. The uncertainties were estimated by performing 1000 bootstrap runs. A similar approach was used to calculate the fossil fractions of the factors by replacing *OCnf* with  $OC_f$ -  $HOC_{est}$  (Eq. S11), based on the assumption that  $HOC_{est}$  is completely fossil and assumed to have similar relative ionization efficiency (RIE) as other fractions of OC.

164

165 
$$
\sigma_{HOA_{estimated}} = \sqrt[2]{\sigma_{EC_{f,estimated}}^2 + (\frac{\sigma_{slope\,(HOC\,vs\,EC_f)}}{slope\,(HOC\,vs\,EC_f)})^2}
$$
 Eq. S7

166

167 In approach 2, the following steps were used:

168 i) The average  $EC_f/EC_{nf}$  ratio of 2.4  $\pm$  0.8 (at upwind rural background and Delhi urban sites) was calculated

$$
EC = EC_f + EC_{nf} = 2.4 * EC_{nf} + EC_{nf} = 3.4 * EC_{nf}
$$
 Eq. S8

- **9Solution**  $EC_{nf} = 0.3 * EC$  **Eq. S9**
- 172 ii)  $EC_{nf,estimated}$  was computed for the remaining sites/samples using a factor of 0.3 times the total Sunset-EC from Eq. S9.
- 173 iii)  $EC_{f,estimated}$  (*y*-axis) for all sites by subtracting  $EC_{nf,estimated}$  from total Sunset-EC and compared with  $HOC_{estimated}$  (*x*-axis) obtained 174 from approach 1. A slope of 0.6 was obtained which is further used to calculate  $HOC_{estimated}$  for downwind Kanpur site outside Delhi, 175 where only total *EC* was available.

171

#### 176

#### 181 *C) Multi-linear regression for fossil and non-fossil fractions of OA sources*

- 182 The uncertainty-weighted mass concentration time series (*i*) of species-specific recovery corrected factors obtained after AMS-PMF were used in 183 Eq. S10. BBOC was assumed to be completely non-fossil.
- 184

 $OC_{n f i} - BBOC_i = a * CPOC_i + c * COOC_i + d * UOOC_i$  **Eq. S10** 

## 193  $OC_{f,i} - HOC_{est,i} = a * CPOC_i + b * BBOC_i + c * COOC_i + d * UOOC_i$  **Eq. S11**

 Due to on-site contamination, only six (Mn, Cu, Cd, Sn, Sb, and Pb) of the measured elements had sample concentrations higher than the threshold (field blank average + 3 standard deviations). The spatial and temporal variation, and relative contribution of these selected trace elements with 199 two water soluble marker ions (Na<sup>+</sup> and K<sup>+</sup>) is shown in Supplementary Fig. 11. The Pearson's *r* between water-soluble K<sup>+</sup> ions (IC) and total K

The results from multilinear regression on both the fossil and non-fossil fractions were similar, as shown in Supplementary Table 6.

#### *D) Sources of trace elements*

from acid-digested samples (ICP-MS) is 0.7. The values for the acid-digested samples are nearly twice those of the water-soluble ones.

201 To determine the sources of the trace elements, an unconstrained PMF was performed for 3-14 factors based on elements including  $Na<sup>+</sup>$ ,  $K<sup>+</sup>$ , Li, Mg, Ca, Ti, V, Cr, Mn, Fe, Co, Ni, Cu, Se, Sr, Zr, Mo, Pd, Cd, Sn, Sb, Cs, Ba, La, Ce, Pt, and Pb from all five sites. We used all measured elements in the PMF matrix to estimate the contribution of elements in the identified contamination factors. The data matrix (*j*=294 samples) included the 204 mass concentrations of the mentioned elements and the error matrix was computed by using Eq. 3 and  $4^{22,23}$ . The field blanks were treated as real 205 samples and included with the aerosol samples in the data and corresponding error matrix. MDL is the element-specific minimum detection limit 206 of instrument and  $p$  is the analytical uncertainty (3%). Two random seed runs were performed for each factor solution making a total of 24 runs. Supplementary Fig. 12 shows the optimum base-case solution chosen after investigating the preliminary diagnostics. The optimum solution with 208 the lowest  $Q/Q_{exp}$  was observed for the 8-factor solution, after which the relative change ( $\Delta Q/Q_{exp}$ ) is small (<0.25). Three factors identified as 209 K<sup>+</sup>-Na<sup>+</sup> rich, Cu-Cd rich, and Pb-Sn-Sb rich and are discussed below, and the remaining five (factors 2a, 2b, 2c, 2d and 2e) were treated as contamination due to their similar contribution in both the field blanks and aerosol samples, as shown in Supplementary Fig. 12.

 $K^+$ -Na<sup>+</sup> rich: Na<sup>+</sup>- and K<sup>+</sup>-dominated factor could potentially be present as a salt when combined with another anion (Cl<sup>-</sup>) or inorganic ion (NO<sub>3</sub><sup>-</sup> 

212 or  $SO_4^2$ ). At coastal sites (marine aerosols), sea salt is the major source of Na<sup>+</sup> ion, however this is unlikely in the present study as our sampling

214 coal/oil and biomass combustion, and vehicular emissions, as the seasonal  $K^+$  to Na<sup>+</sup> ratios are in the range of 0.5-1.7<sup>24</sup>, much higher than that of sea salt (0.036).

216 *Cu-Cd rich:* This factor potentially originates from either electrical/electronic equipment waste incineration or burning plastic/open waste<sup>25</sup>.

 *Pb-Sn-Sb rich*: This factor has multiple potential sources. It can originate from open waste/plastic burning, lead smelting industries and brake wear 218 (non-tailpipe traffic emissions)<sup>25</sup>.

219 In the 7-factor solution, the Pb-Sn-Sb factor is mixed with Cd, and the  $K^+$ -Na<sup>+</sup> rich factor is mixed with Cu, the combination of which in the 8- factor solution results in separate factors that have already been identified as separate sources from electronic and electrical waste incineration and open waste/plastic burning, respectively, at the same site (i.e., IITD) during real-time monitoring and source apportionment of trace elements 222 (offline vs. online PMF  $(n=35)$ ;  $R^2$ : 0.76 (Cu-Cd rich) and 0.77 (Pb-Sn-Sb rich)<sup>25</sup>. Further, in the 9-factor solution, a split of the contamination factor 2d was identified. Supplementary Fig. 12 demonstrates the trace element factor profiles from the final 8-factor solution, as well as their concentratons and relative contributions.

sites are quite far (1500 km) away from coastal areas. It could be related to multiple anthropogenic sources such as open trash/refuse burning,

241 Supplementary Fig. 1: Biomass burning corridors (27). Levoglucosan to mannosan ratio vs. levoglucosan to K<sup>+</sup> ratio to investigate the molecular diversity in biomass burning and spatiotemporally distinguish the change in the type of biofuels combusted.

#### *E) Stepwise linear regression modeling of PM2.5-OP sources*

 The model adds and removes predictor variables by forward and backward stepwise regression based on the "Akaike information criterion (AIC)" and prevents over-fitting. We have not included any interaction term among predictors and response variables although earlier studies have shown 228 the possibility of both synergistic and antagonistic effects due to organics-metal and metal-metal interaction<sup>26</sup>. In this least-square estimation method, AIC is defined as the sum of twice the number of predictor variables, *k*, and *N* times the log of the variance of the noise, as given in Eq. S12.

$$
^{231}
$$

$$
AIC = 2k + N * \log\left(\frac{RSS}{N}\right) \qquad \qquad \mathbf{Eq. S12}
$$

 Where RSS is the residual sum of squares and *N* corresponds to the sample size. Model inputs included averaged PMF-derived time series (from BS runs) of OA (CPOA, BBOA, COOA, UOOA and HOAestimated), trace elements (3) sources as well as their contamination factors and measured 234 volume-normalized OP concentrations (DCFH<sub>v</sub>, DTT<sub>v</sub>, and AA<sub>v</sub>). The linear least-square solver "lsqlin" was used to put non-negative constraints 235 on the predictor variables in order to obtain the final values of the OP strength  $OP_m$ , nmol  $\mu g^{-1}$ ) and the contribution of the OA and trace elements sources (shown as cumulative density function (CDF) plots in Supplementary Fig. 4a). We observed that unconstrained fitting does not affect the coefficient of predictor variables corresponding to the ambient factors.

#### **Supplementary Figures**



#### 

**Supplementary Fig. 2: Seasonal variation of (a) mass- (OPm, per µg of PM2.5) and (b) volume-normalized (OPv, per m<sup>3</sup> of air) OP values**  246 **for all 3 assays (DCFH, DTT and AA) for cold and warm period.** Box whisker plots [line/box: median and 25<sup>th</sup>-75<sup>th</sup> percentile; upper and lower



247 end of whisker:  $5<sup>th</sup>$ -95<sup>th</sup> percentile] are prepared from spatial- and temporal-values.







251

252 **Supplementary Fig. 3: OP<sup>m</sup> for OA and elemental sources, and comparison of measured vs. modelled OPv. (a)** OP<sup>m</sup> (OP per unit aerosol 253 mass; nmol min<sup>-1</sup>  $\mu$ g<sup>-1</sup>) for different OA and elemental sources (displayed as cumulative density functions, CDF) obtained from multi-linear 254 regression (1000 bootstrap runs) between measured  $OP_v$  (as predictor variable), and OA and elemental source concentrations (as response variable) 255 and their contamination sources. OA sources consist of COOA, UOOA, BBOA, CPOA, and HOAest, and elemental sources of Cu-Cd rich, Pb-Sn-256 Sb rich, and K<sup>+</sup>-Na<sup>+</sup> rich profiles. Factors 1a and 1b are the contamination factors in OA-PMF and Factors 2a-2e in elemental PMF. **(b)** Comparison 257 of source-specific stacked OP<sub>v</sub> (as the product of real source mass contribution after subtracting contamination sources and OP<sub>m</sub>) and total measured 258 OP<sup>v</sup> for AA, DCFH and DTT.



 **Supplementary Fig. 4: Comparison of source contribution to PM2.5 and its OPv.** Site-specific monthly averaged concentration and fractional 261 contribution of OA and elemental sources and total inorganics to total PM<sub>2.5</sub> and OP<sub>v</sub> (AA, DTT and DCFH). The fractional contribution is 262 estimated by considering the median  $OP_m$  of each source for each assay obtained from 1000 bootstrap runs while performing multi-linear regression, and OA sources as well as PM2.5 spatial and temporal concentrations.



**Supplementary Fig. 5: Typical sample measurement cycle of HR-AMS and EESI-ToF-MS.** Each sample was measured for 12 minutes with

 18 minutes of preceding water blank spending a total of 30 minutes on each sample by AMS and EESI. During EESI measurements, intermittent HEPA filter switching was performed to determine interferences due to sticky residuals from the previous sample.

















 **Supplementary Fig. 6: HR-AMS PMF diagnostics. a1)** Diagnostic plot showing the relative change in *Q*/*Q*exp and relative standard deviation for unconstrained solution vs. number of factors, respectively, and **a2)** Scaled residual distribution for the chosen 6-factor solution.s B) Absolute difference in *Q*/*Qexp* between different number of factor solutions. **c)** Time series of scaled residuals for different number of factors (3-7). Factor profiles for unconstrained (*m/z* > 44) 6- **(d)** and 7-factor **(e)** solution. **f)** Factor profiles for 6-factor solution constrained using the CPOA profile (*a*-value = 0.5). **g)** Averaged (10 seed runs) optimum 6-factor solution (*m/z* 12-180) constrained with CPOA profile and passing upper and lower limits of the factor timeseries obtained from 100 BS runs of constrained PMF of HR ions with *m/z* > 44 **(e)**. **h)** Box and whisker plot of the Coefficient of variation (CoV) obtained for 100 BS runs of the optimum AMS-PMF solution obtained in **(g)**. **i)** Box whisker plot of water-soluble fraction of identified factors for all five sites. Unlike other factors, the OC concentrations of the contamination factors (Factor 1a and Factor 1b) have a similar range for field samples and field blanks. To identify unique ions related to single factors or a group of factors, 2D-hierarchical clustering was performed on AMS matrix of PMF (optimum solution shown in **g**) profiles and standardization (z-score) was done along the ions. A two-step clustering process (MATLAB 2016a); first along the columns (factors) and then the rows (ions) was done, where (i) the Euclidean distance was calculated for similarity/dissimilarity between pairs of ions/factors and (ii) then the Ward's method was used to form the linkages of ion/factor pairs. This resulted into **j)** dendograms with clusters of AMS ions on the y-axis (categorized into traditional AMS ion families with color scheme similar to **g)** and AMS-factors on the x-axis. It is interesting to note that CPOA is not only dominated by the CH and CHN ion groups (green cluster) but includes the majority of the fitted ions (CHOgt1 family in pink cluster and CHOgt1 and CHNO families in dark cyan cluster). The contamination factors (Factor 1a and 1b) are clustered together and the factors (UOOA and COOA) identified in the AMS-PMF on the basis of seasonal variation are clustered separately (see Materials and Methods). **k)** AMS measured-raw water-soluble mass concentration of each factor across all sites. Box whisker plot represent values obtained from 100 bootstrap runs.



**Supplementary Fig. 7: Species-specific factor recoveries. (a) Box and whisker plot**  $(10^{th}, 25^{th}, 50^{th}, 75^{th}$  **and**  $90^{th}$  **percentile) of the Coefficient**  of variation (CoV) obtained for the optimum PMF solution for estimating species-specific factor recoveries. 100 BS runs were performed on water- soluble AMS ions and water-insoluble OC (WIOC) while constraining the solution with upper and lower concentration limits for each factor. The higher value for BBOA is observed in the runs where the sample concentration decreases by a factor of 100. **(b)** Mass closure of reconstructed recovery corrected AMS-PMF-derived OC mass and Sunset-measured OC mass.



 

 **Supplementary Fig. 8: Near-molecular level chemical fingerprints of OA sources. (a)** Averaged (10 seed runs) optimum 6-factor solution (*m/z* 120-440) constrained with BBOA profile (from 9-factor solution; *a* = 0.3) and AMS-derived time series of contamination factor-1b (*a* =0.39). The 317 signal intensity (SI) of  $C_6H_{10}O_5Na^+$  in all source profiles was reduced by a factor 10 to clearly show the other contributing marker ions. Further, atomic H:C vs O:C ratios of individual sources: **(b)** UOOA, and **(c)** COOA. The lines represent different groups defined on the basis of modified

319 aromaticity index (AI<sub>mod</sub>) and H:C ratio where Group 1 representing combustion-derived condensed polycyclic aromatics (AI<sub>mod</sub> > 0.66) lie below 320 the dotted line, Group 2 representing vascular plant-derived polyphenols  $(0.66 \geq AI_{mod} > 0.50)$  lie between the dotted and short dashed line, Group 321 3 with highly unsaturated and phenolic compounds ( $AI_{mod} \leq 0.50$  and H:C < 1.5) between the short dashed and medium dashed line, Group 4 of 322 aliphatic compounds  $(2.0 \geq H.C \geq 1.5)$  is between the medium dashed and solid line, and Group 5 representing saturated fatty and carbohydrates 323 (H:C  $>$  2.0) lie above the solid line. Here, all ions having a relative contribution  $(f_c)$  of the factor greater than 0.35 were selected for COOA whereas for UOOA, those having *f<sup>c</sup>* of UOOA greater than 0.4 and *f<sup>c</sup>* of other remaining factors lower than 0.3, were selected. The marker size represents the selected ions' fractional contribution to the factor profile.





 **Supplementary Fig. 9: Meteorological parameters measured at three sites in Delhi.** Temporal variation of **(a)** temperature (°C) and relative 330 humidity (%), (b) wind speed (m s<sup>-1</sup>) and wind direction (degrees) measured at IITD, IITMD (from the DPCC station at Mandir Marg, roughly 2.7 km air distance) and MRIUF from January to February, 2018 using automated weather station installed at the building rooftop.







 **Supplementary Fig. 10: Fossil and non-fossil fraction of OC and EC.** Relative contributions [left panels] and absolute concentrations [right panels] of the fossil and non-fossil fractions of OC and EC at three sites (CCSHAU, IITMD and IITD).



**Supplementary Fig. 11: Total elemental concentration and its components.** (A) Spatially resolved (all 5 sites) stacked time series of trace

elements (except contaminated ones) and water-soluble marker ions. The trace elements include Cu, Mn, Cd, Sn, Sb, and Pb measured by ICP-

342 MS. Na<sup>+</sup> and K<sup>+</sup> (here, divided by a factor of 10 for demonstration purpose) were the two water-soluble marker ions measured by ion chromatography.









349<br>350

350 **Supplementary Fig. 12: Sources of trace elements.** Unconstrained PMF of 27 elements, including field blanks and aerosol samples. **(a)** 351 Diagnostic plot showing the relative change in  $Q/Q_{exp}$ . (b) Factor profiles for 7-factors solution with Pb-Sn-Sb rich mixed with Cd and K<sup>+</sup>-Na<sup>+</sup> 352 mixed with Cu. **(c)** Factor profiles for optimum 8-factor solution. **(d)** Factor profiles for 9-factors solution with Pb-Sn-Sb rich mixed with Cd and 353 splitting of Factor 2d. **(e)** Box whisker plot of elemental sources concentrations in field and blank samples of all five sites. **(f)** Relative contribution 354 of elemental sources.

355

356

#### 357 **Supplementary Tables**

- 358 **Supplementary Table 1. Comparison of wintertime total OP<sup>v</sup> (AA, DTT), PM2.5 mass and PM2.5 oxidative strength across different**
- 359 **regions of the world.**





#### 361 **Supplementary Table 2. OA sources vs. targeted organic compounds correlation matrix.** Fossil (*f*) and non-fossil *(nf)* fraction of OA

362 sources are correlated to organic markers measured by GC-MS and LC-MS.





363

#### 364 **Supplementary Table 3. Average DCFHm, DTTm and AA<sup>m</sup> values determined by multilinear regression (MLR) on volume-normalized**

365 **measured assays (as response variable) and OA and elemental sources (as predictor variable).** MLR was performed by providing 0 as lower

366 bound (averaged over 100 bootstrap runs).



367

368 **Supplementary Table 4. Chemical composition and other auxiliary measurements conducted on the collected filters.** A total of 300 filter

369 samples (including field blanks) referred here as "All" were analyzed for EC-OC, WSOC, water soluble inorganic carbon, ions, and trace elements.

370 Bulk and near-molecular level information on organic aerosols (OAs) was obtained using the offline technique developed at PSI<sup>28,29</sup>. A subset of

371 140 filters selected from four sites (CCSHAU, IITMD, IITD and MRIUF) were analyzed for targeted organic compounds. Another subset of 193

filters from CCSHAU, IITMD and IITD were analyzed for  ${}^{14}C$  of TC and 44 filters out of 193 for  ${}^{14}C$  of EC.







## 374 **Supplementary Table 5. Variation of factor-specific recoveries (**ܴ, **in percentage) and OM:OC ratio.**



375

## 376 **Supplementary Table 6. Fitting coefficients of relative (A) fossil and (B) non-fossil contributions of different factors determined by**

377 **multilinear regression.**



378 379

380 **Supplementary Table 7. AMS vs. EESI PMF correlation matrix.** AMS and EESI PMF factor time series from the final solution were 381 normalized to labelled  $\text{Na}_2{}^{34}\text{SO}_4$  and  ${}^{15}\text{NO}_3$  or  ${}^{34}\text{SO}_4{}^{2}$ , respectively.

	CPOA <sub>AMS</sub>	BBOAAMS	<b>COOA</b> <sub>AMS</sub>	<b>UOOA</b> <sub>AMS</sub>	Factor 1a <sub>AMS</sub>	Factor 1b <sub>AMS</sub>
<b>CPOAEESI</b>	0.87	0.65	0.43	$-0.04$	0.29	$-0.16$
<b>BBOAEESI</b>	0.43	0.73	0.16	$-0.06$	$-0.16$	0.32
<b>COOAEESI</b>	0.25	0.18	0.81	0.002	0.23	$-0.10$
<b>UOOAEESI</b>	$-0.10$	0.003	$-0.24$	0.42	$-0.24$	0.25
Factor 1a <sub>EESI</sub>	$-0.03$	0.04	$-0.23$	0.08	$-0.45$	0.62
Factor 1b <sub>EESI</sub>	$-0.10$	$-0.07$	$-0.17$	$-0.20$	0.42	0.30

	All sites		<b>CCSHAU</b>		<b>IITMD</b>		<b>IITD</b>			<b>MRIUF</b>			<b>IITK</b>					
	<b>DCFH</b> <sub>v</sub>	AA <sub>v</sub>	DTT <sub>v</sub>	<b>DCFH</b> <sub>v</sub>	AA <sub>v</sub>	$DTT_v$	DCFH <sub>v</sub>	AA <sub>v</sub>	DTT <sub>v</sub>	DCFH <sub>v</sub>	AA <sub>v</sub>	$DTT_v$	<b>DCFH</b> <sub>v</sub>	AA <sub>v</sub>	DTT <sub>v</sub>	DCFH <sub>v</sub>	$AA_v$	DTT <sub>v</sub>
$AA_v$	$0.60**$			$0.72**$			$0.49**$			$0.42**$			$0.82**$			$0.82**$		
$DTT_v$	$0.43**$	$0.56**$		$0.64**$	$0.53**$		0.18	$0.57**$		$0.57**$	$0.64**$		$0.54**$	$0.64**$		$0.53**$	$0.71**$	
$PM_{2.5}$	$0.46**$	$0.54**$	$0.53**$	0.12	$0.49**$	$0.38**$	$0.43**$	$0.52**$	$0.35**$	$0.53**$	$0.50**$	$0.65**$	$0.57**$	$0.55**$	$0.73**$	$0.66**$	$0.69**$	$0.68**$
AA <sub>m</sub>	0.23	$0.59**$	0.02	0.18	$0.56**$	$-0.15$	$0.44*$	$0.64**$	0.17	$-0.09$	$0.62**$	0.04	$-0.12$	$0.65***$	0.21	0.40	0.37	$-0.11$
DTT <sub>m</sub>	$-0.07$	0.06	0.14	0.10	0.23	0.01	0.08	0.16	$0.49*$	$-0.18$	0.06	0.15	$-0.16$	0.11	0.11	$-0.21$	$-0.26$	$-0.23$
DCFH <sub>m</sub>	$0.64**$	0.20	$-0.10$	$0.56**$	0.20	$-0.15$	$0.70**$	0.07	$-0.24$	0.21	$-0.16$	$-0.28$	$0.66*$	$-0.07$	$-0.04$	$0.67**$	0.30	$-0.04$
<b>OC</b>	$0.55***$	$0.68**$	$0.55**$	0.31	$0.65**$	0.47	0.46	0.57	0.35	$0.60**$	$0.65**$	$0.72**$	0.47	$0.65***$	$0.77**$	$0.70**$	$0.80**$	$0.65**$
EC	$0.31**$	$0.53**$	$0.35**$	$0.32*$	$0.44**$	$0.33*$	$0.34*$	$0.57**$	$0.55***$	$0.35*$	$0.80**$	$0.58**$	0.12	$0.65**$	$0.68**$	$0.53**$	$0.50**$	$0.41*$
NO <sub>3</sub>	$0.31**$	$0.24**$	$0.30**$	0.05	0.17	0.08	$0.50**$	$0.44**$	0.20	$0.43*$	0.13	$0.35*$	$0.65**$	0.30	$0.51**$	$0.55**$	0.45	$0.57**$
SO <sub>4</sub> <sup>2</sup>	$0.31**$	$0.27**$	$0.38**$	0.05	0.20	$0.27**$	$0.50**$	$0.41**$	$0.30*$	$0.37**$	0.12	0.26	$0.50**$	0.36	$0.58**$	0.20	0.13	0.52
$Cl^{\cdot}$	0.03	0.05	0.13	$0.50**$	$0.57**$	$0.43**$	$-0.12$	0.05	0.09	$-0.12$	$-0.10$	0.03	0.28	0.26	0.40	$0.58**$	$0.67**$	$0.63**$
<b>CPOA</b>	$.04**$	$0.59**$	$0.39**$	0.08	$0.68**$	$0.35*$	$0.28*$	$0.40**$	0.25	0.18	0.21	0.26	$0.45**$	$0.60**$	$0.65***$	$0.66**$	$0.86**$	$0.56**$
<b>BBOA</b>	$0.47**$	$0.63**$	$0.38**$	0.07	$0.57**$	0.23	0.25	$0.34**$	$0.26*$	$0.39**$	$0.65**$	$0.54**$	$0.34**$	$0.67**$	$0.54*$	$0.72**$	$0.83**$	$0.56**$
<b>COOA</b>	$0.32**$	$0.32**$	$0.47**$	0.05	$0.40**$	$0.53**$	$0.29*$	$0.36**$	0.15	$0.40**$	0.15	$0.29*$	$0.32**$	$0.30*$	$0.53**$	$0.52**$	$0.50**$	$0.62**$
<b>UOOA</b>	$0.23**$	0.07	0.02	$0.36**$	$-0.04$	$-0.10$	$0.46**$	$0.25*$	$0.33*$	$-0.01$	0.001	0.03	0.07	0.14	0.17	$-0.30*$	$-0.50*$	$-0.30$

383 **Supplementary Table 8. Correlation matrix of total PM2.5 and its constituents and their sources with OPv.**



384 \*p value < 0.05, \*\* p value < 0.01. Pearson's  $r > 0.55$  are bold.

#### **Supplementary References**

- 1. Bhowmik, H. S. et al.Temporal and spatial variability of carbonaceous species (EC; OC; WSOC and SOA) in PM2.5 aerosol over five sites of Indo-Gangetic Plain, *Atmos. Poll. Res.* **12**, 375-390 (2021).
- 2. Tripathi, N. et al. Characteristics of VOC composition at urban and suburban sites of New Delhi, India in winter, J. *Geophys. Res.-Atmos.* **127** (2022).
- 3. DeCarlo P.F. et al. Field deployable, high-resolution, time-of-flight aerosol mass spectrometer, *Anal. Chem.,* **78**, 8281–8289 (2006).
- 4. Lopez-Hilfiker, F. D. et al. An extractive electrospray ionization time-of-flight mass spectrometer (EESI-TOF) for online measurement of atmospheric aerosol particles, *Atmos. Meas. Tech.* **12**, 4867–4886 (2019).
- 5. Lee, C. P. et al. Online Aerosol Chemical Characterization by Extractive Electrospray Ionization–Ultrahigh-Resolution Mass Spectrometry (EESI-Orbitrap), *Environ. Sci. Technol.* **54**, 3871-3880 (2020).
- 6. Sueper, D. et al. A community software for quality control and analysis of data from the aerodyne time-of-flight aerosol mass spectrometers (ToF-AMS), *Annual Conference of the American Association for Aerosol Research,Reno, Nev (2007)*.
- 7. Salazar, G., Zhang, Y. L., Agrios, K. & Szidat, S. Development of a method for fast and automatic radiocarbon measurement of aerosol samples by online coupling of an elemental analyzer with a MICADAS AMS, *Nucl. Instrum. Methods Phys. Res.* **361 (B),** 163-167 (2015).
- 400 8. Zhang, Y. L. et al. On the isolation of OC and EC and the optimal strategy of radiocarbon-based source apportionment of carbonaceous aerosols, *Atmos. Chem. Phys.* **12**, 10841–10856 (2012).
- 402 9. Szidat, S. et al. <sup>14</sup>C analysis and sample preparation at the new Bern Laboratory for the Analysis of Radiocarbon with AMS (LARA), *Radiocarbon* **56**, 561–566 (2014).
- 404 10. Agrios, K. et al. Online coupling of pure  $O_2$  thermo-optical methods  $-$  <sup>14</sup>C AMS for source apportionment of carbonaceous aerosols study, *Nucl. Instrum. Meth. Phys. Res. B.* **361**, 288–293 (2015).
- 11. Stuiver, M. & Polach, H. A. Reporting of C-14 data discussion, *Radiocarbon* **19**, 355–363 (1977).
- 12. Zhang, Y. L. et al. Fossil and nonfossil sources of organic and elemental carbon aerosols in the outflow from Northeast China, *Environ. Sci. Technol.* **50**, 6284–6292 (2016).
- 13. Zotter, P. et al. Radiocarbon analysis of elemental and organic carbon in Switzerland during winter-smog episodes from 2008 to 2012 Part 1: Source apportionment and spatial variability, *Atmos. Chem. Phys.* **14**, 13551–13570 (2014).
- 14. Cavalli, F., Viana, M., Yttri, K. E., Genberg, J. & Putaud, J. -P. Toward a standardised thermal-optical protocol for measuring atmospheric organic and elemental carbon: the EUSAAR protocol, *Atmos. Meas. Tech.* **3**, 79–89 (2010).
- 15. Alleman, L. Y., Lamaison, L., Perdrix, E., Robache, A. & Galloo, J-C. PM<sup>10</sup> metal concentrations and source identification using positive matrix factorization and wind sectoring in a French industrial zone, *Atm. Res.* **96**, 4, 612-625 (2010).
- 16. Borlaza, L. J. et al. Disparities in particulate matter (PM10) origins and oxidative potential at a city-scale (Grenoble, France) Part I: Source apportionment at three neighbouring sites. *Atmos. Chem. Phys.* **21**, 5415–5437 (2021)
- 17. Orasche, J., Schnelle-Kreis, J., Abbaszade, G. & Zimmermann, R. Technical Note: In-situ derivatization thermal desorption GC-TOFMS for direct analysis of particle-bound non-polar and polar organic species, *Atmos. Chem. Phys.* **11**, 8977–8993 (2011).
- 419 18. Calas, A. et al. Comparison between five acellular oxidative potential measurement assays performed with detailed chemistry on  $PM_{10}$  samples from the city of Chamonix (France), *Atmos. Chem. Phys.* **18**, 7863–7875 (2018).
- 19. Borlaza, L. J. et al. Disparities in particulate matter (PM10) origins and oxidative potential at a city scale (Grenoble, France) Part 2: Sources of PM<sup>10</sup> oxidative potential using multiple linear regression analysis and the predictive applicability of multilayer perceptron neural network analysis, *Atmos. Chem. Phys.* **21**, 9719–9739 (2021).
- 20. Campbell, S. J. et al. Atmospheric conditions and composition that influence PM2.5 oxidative potential in Beijing, China, *Atmos. Chem. Phys.* **21**, 5549–5573 (2021).
- 21. Murrant C. L. & Reid, M. B. Detection of reactive oxygen and reactive nitrogen species in skeletal muscle, *Microsc. Res. Tech*. **55**, 236-248, 2001.
- 22. Polissar, A. V., Hopke, P. K., Paatero, P., Malm, W. C. & Sisler, J. F. Atmospheric aerosol over Alaska: 2. Elemental composition and sources.
- *J. Geophys. Res. Atmos.* **103**, 19045–19057 (1998)
- 23. Reff, A., Eberly, S. I. & Bhave, P. V. Receptor modeling of ambient particulate matter data using positive matrix factorization: review of existing methods, *J. Air Waste Manage. Assoc.* **57**, 146–154 (2007).
- 24. Ooki, A., Uematsu, M., Miura, K. & Nakae, S. Sources of sodium in atmospheric fine particles, *J. Atmos. Environ.,* **36(27)**, 4367-4374 (2002).
- 25. Rai, P. et al. Real-time measurement and source apportionment of elements in Delhi's atmosphere. *Sci. of Tot. Environ.* **742**, 140332, 2020.
- 26. Gao, D., Godri Pollitt, K. J., Mulholland, J. A., Russell, A. G. & Weber, R. J. Characterization and comparison of PM2.5 oxidative potential assessed by two acellular assays, *Atmos. Chem. Phys.* **20**, 5197–5210, 2020.
- 27. Cheng, Y. et al. Biomass burning contribution to Beijing aerosol, *Atmos. Chem. Phys.* **13**, 7765–7781 (2013).
- 28. Dӓellenbach, K. R. et al. Characterization and source apportionment of organic aerosol using offline aerosol mass spectrometry, *Atmos. Meas. Tech.* **9**, 23–39 (2016).
- 29. Qi, L. et al. Organic aerosol source apportionment in Zurich using an extractive electrospray ionization time-of-flight mass spectrometer (EESI-TOF-MS) – Part 2: Biomass burning influences in winter, *Atmos. Chem. Phys.,* **19**, 8037–8062 (2019).