#### 1 Supplementary information for manuscript

#### Identification and quantification of organic aerosol from 2 cooking and other sources in Barcelona using aerosol mass 3 spectrometer data 4

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# **1** Collection efficiency (CE)



31 concentration of species measured by AMS (right axis).



37 Figure S2: Scatterplot of combined time series of total AMS species (HR) and BC (y-axis) and Grimm PM<sub>1</sub>.

38 The data were fitted with a least orthogonal distance fit (red line).

**3 PMF** 

## **3.1 UMR solution**



42 Figure S3: Mass spectra of the UMR 5-factor-PMF solution.



44 Figure S 4: Time series of the UMR 5-factor-PMF solution and ancillary data.

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### 46 **3.2** Comparison of UMR and HR PMF solution

The  $R^2$  of the correlation of the mass spectra of the UMR and HR PMF solution range from 0.80 (COA) to 0.99 (LV-OOA), confirming their similarities. Bigger differences can be seen in the time series of the corresponding factors. The COA time series show discrepancies in the total mass especially in the beginning of the campaign (until 02 March 2009), visualized in the data points with a much lower slope in Fig. S5 h. For the BBOA, the UMR time series features peaks not inherent to the HR time series. Concerning the mass attribution to each factor, HR generally assigns more mass to the primary OA factors and less to the OOA factors. Here the higher resolution and, relied to that, the signal on an individual ion basis of the HR data matrix adds additional information to the HR data matrix and thus allows for a better quantification of primary and secondary OA.



58 Figure S5: Scatter plots of UMR and HR PMF spectra (a-e), time series (f-j) and a comparison of the mass

59 attributed to each factor relative to OA (k-o).

#### 61 **3.3 UMR solution criteria**

62 The PMF solution for a chosen number of factors p is a weighted iterative least squares fit

- 63 minimizing Q as in Eq. (1), with m and n denoting the rows and columns of the input matrices,
- 64 respectively. The known standard deviations  $\sigma_{ij}$  of the measured input values  $x_{ij}$  are used to
- 65 determine the weights of the residuals  $e_{ij}$ .

66 
$$Q = \sum_{i=1}^{m} \sum_{j=1}^{n} \left( e_{ij} / \sigma_{ij} \right)^{2}$$
(1)

67 If the model is appropriate and the data uncertainties estimations are accurate,  $(e_{ij} / \sigma_{ij})^2$  is ~1 and

68 the expected Q (*Qexpected*) =  $mn-p(m+n) \approx mn$ , the degrees of freedom of the fitted data. The *Q*-

69 value is one mathematical criterion for the quality of the PMF solution: Q/Qexpected >> 1

70 indicates an underestimation, *Q/Qexpected* <<1 an overestimation of errors in the input data

71 (Paatero et al., 2002). The mathematically correct value of *p* in PMF would be where the line

- changes the slope in the plot of a series of p values versus their respective minimized Q (Fig. S6
- a). However, a PMF solution has to be feasible in an ambient context and thus does not
- necessarily correspond to the mathematically correct value of *p*.
- Rotational ambiguity is a significant problem in the use of factor analysis (Paatero et al., 2002). PMF solutions are not unique since linear transformation (still conserving the non-negativity constraint) are possible ( $\mathbf{GF} = \mathbf{GTT}^{-1}\mathbf{F}$ ). The rotational freedom of the chosen solution can be explored through a non-zero valued user-specified rotational parameter *fpeak*. *Fpeak* > 0 tries to impose rotations on the emerging solutions using positive coefficients *r* in **T**, *fpeak* < 0 vice versa. *Fpeak* = 0 produces the most central solution (Fig. S6 b).
- 81 The number of factors p was chosen to be 6 for the UMR dataset (Fig. S7). In the solution with 82 p = 5 (Fig. S8 a), the spectra of BBOA, HOA, and COA are less clearly separated (e. g. high 83 signal at m/z 57 in the top factor resembling BBOA, but very little signal at m/z 57 in the red factor resembling COA). Figure S8 b shows the time series of the 5-factor solution - they are 84 85 less clearly distinct than those of the 6-factor solution. The 7-factor solution (Fig. S9, a) features a factor consisting mostly of signal at m/z 43 and a factor (orange) with single, isolated peaks 86 87 inconsistent with regular ion series. The time series show a more similar evolution (Fig. S9 b), 88 indicating a split of factors.

For the PMF solution presented in the manuscript, the 6-factor solution was chosen and the two factors assigned to SV-OOA (black and purple) regrouped to one SV-OOA, using the sum for

91 the time series and the loadings-weighted average of the spectra.

- 92 Figure S10 presents the explained variance of the organics as a function of *fpeak* for the chosen
- 6-factor solution. *fpeak* was chosen to b -0.7 based on correlations of the corresponding factorswith reference spectra.
- 95 A boxplot of the scaled residuals (boxes are +/- 25% of points) per m/z is shown in Fig. S11,
- 96 time series of the residuals and *Q/Qexpected* are shown in Fig. S12. On 16 March 2009, a power
- 97 failure led to a breakdown of the instrument and a subsequent pumping down effect (Fig. S12).
- 98 Downweighting this period in the input for PMF did not alter the solution.
- 99 The solution space for the chosen p = 6 (central rotation) was explored by running PMF with 50 random initial values (SEED) at iteration start (Figs. S13 - 14). Roughly three solution groups 100 101 can be identified (numbers in Fig. S14). Groups 1 and 2 feature a factor spectrum predominantly 102 consisting of m/z 43 and two spectra that are basically identical. The spectrum with BBOA-like 103 features shows no contributions at m/z 44, which is inconsistent with previous studies. For group 104 3, all spectra not assigned to OOA show very high similarities. The solution with a central 105 rotation (fpeak = 0) was thus discarded regardless of SEED values. Similar information was also 106 published in the supplementary information of Mohr et al. (2011).



107



109 UMR solution.



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111 Figure S7: 6-factor UMR solution chosen, mass spectra (a) and time series (b). The black and the purple

- 112 factor (SV-OOA 1 and 2) were regrouped to SV-OOA.
- 113



115 Figure S8: 5-factor UMR solution, mass spectra (a) and time series (b).



117 Figure S9: 7-factor UMR solution, mass spectra (a) and time series (b).





120 Figure S10: Variance explained by PMF due to the 6-factor UMR solution as a function of *fpeak*. For the

- 121 solution presented, fpeak =-0.7.



126 Figure S11: Median black strokes) and lower/upper quartiles (boxes) of the scaled residuals per m/z.







132 Figure S13: *Q/Qexpected* as a function of different *SEED* values.



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Figure S14: Variance explained by PMF due to the 6-factor UMR solution as a function of SEED. The numbers 1, 2 and 3 denote the three solution groups identified (see text).

### 138 **3.4 HR solution criteria**

Figure S15 shows the chosen 6-factor solution for the HR dataset. The SV-OOA factors (green and purple) were regrouped to a single SV-OOA factor as described in the previous UMR section. The 5-factor solution (Fig. S16) was discarded due to the high similarity of two factors (green and blue, spectra and time series). The 7-factor solution (Fig. S17) features three LV-OOA spectra: PMF seems to assign an individual LV-OOA factor to three different sections in the time series.

145 As the variance in the solution space p = 6 as a function of *fpeak* is negligible (Figs. S18 – 19),

146 the most central solution (fpeak = 0) was chosen and different SEED values were explored (Figs.

147 S20 – 21). SEED = 64 was chosen as the best solution due to correlations with reference spectra.

148 A boxplot of the scaled residuals (boxes are +/- 25% of points) per m/z is shown in Fig. S22,

149 time series of the residuals and  $Q/Q_{exp}$  are shown in Fig. S23. The same irregularities as for the

- 150 UMR data can be observed as well in the HR data.
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- 152



154 Figure S15: Chosen 6-factor solution of the HR dataset, mass spectra (a) and time series (b).



157 Figure S16: 5-factor solution for the HR dataset, mass spectra (a) and time series (b).







163 Figure S18: *Q/Qexpected* versus the number of factors *p* (a) or *fpeak* (b), HR PMF. The orange circle denotes

- 164 the chosen solution.



Figure S19: Variance explained by PMF due to the 6-factor HR solution as a function of *fpeak*. For the
solution presented, *fpeak* =0.



171 Figure S20: *Q/Qexpected* versus *SEED* for the HR solution.





181 Figure S23: Time series of scaled residuals (top panel) and *Q/Qexpected* (lower panel) for the HR solution.
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184 Figure S24: Scatter plot of the time series of b<sub>abs</sub>(880 nm)<sub>traffic</sub> and HOA. The red line is the least orthogonal

- $185 \qquad \text{distance fit where the circle data points were removed.}$
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188 Figure S 25: m/z 55/Org ( $f_{55}$ ) plotted against m/z 57/Org ( $f_{57}$ ).

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Figure S26. Signal at m/z 55 in the HR spectra of meat cooking sources (a) and vehicle engine sources (b). In the engine exhaust spectra, the signal is almost entirely due to the reduced hydrocarbon ion C<sub>4</sub>H<sub>7</sub><sup>+</sup>, whereas in

- 193 the cooking spectra there is also substantial contribution from the oxygen-containing ion  $C_3H_3O^+$ . Reprinted
- 194 from Mohr et al. (2009).
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