Quantification of primary and secondary organic aerosol sources by combined factor analysis of extractive electrospray ionisation and aerosol mass spectrometer measurements (EESI-TOF and AMS)

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Figure S1. The time series and factor profile of AMS-only PMF solution is shown in a) and b), and the time series and factor profile of EESI-TOF-only PMF solution is shown in c) and d), respectively. Green lines are the factor time series of AMS-only and EESI-TOF-only PMF solution in this study, whereas the blue lines are the factor time series of solution in Stefenelli et al. (2019). Note, in addition to the original AMS PMF solution from Stefenelli et al. (2019), the AMS-only PMF solution from this study yields an inorganic nitrate factor (InorgNit), because NO⁺ and NO₂⁺ are included in the input matrix.





Figure S2. The time series and factor profile of AMS-only PMF solution is shown in a) and b), and the time series and factor profile of EESI-TOF-only PMF solution is shown in c) and d), respectively. Green lines are the factor time series of AMS-only and EESI-TOF-only PMF solution in this study, whereas the blue lines are the factor time series of solution in Qi et al. (2019). Note, in addition to the original AMS PMF solution from Qi et al. (2019), the AMS-only PMF solution in this study yields a inorganic nitrate factor (InorgNit), because NO⁺ and NO₂⁺ are included in the input matrix.





Figure S3. Scaled residual distribution of AMS (red line) and EESI (blue line) in an 8-factor solution from joint dataset in Zurich summer and corresponding overlap fractions when C_{EESI} is equal to 0.1 in a), 2 in b), and 10 in c), respectively. Balanced solution is shown in b), whereas in a) and c), AMS and EESI is overweighted, respectively.



Figure S4. Factor time series in a) and mass spectra in b) for 6-factor solution with C_{EESI} of 0.8 for Zurich summer dataset.



Figure S5. Factor time series in a) and mass spectra in b) for 6-factor solution with C_{EESI} of 1 for Zurich summer dataset.



Figure S6. Factor time series in a) and mass spectra in b) for 7-factor solution with C_{EESI} of 0.8 for Zurich summer dataset.



Figure S7. Factor time series in a) and mass spectra in b) for 7-factor solution with C_{EESI} of 1 for Zurich summer dataset.



Figure S8. Factor time series in a) and mass spectra in b) for 8-factor solution with C_{EESI} of 0.8 for Zurich summer dataset.



Figure S9. Factor time series in a) and mass spectra in b) for 8-factor solution with C_{EESI} of 1 for Zurich summer dataset.



Figure S10. Factor time series in a) and mass spectra in b) for 8-factor solution with C_{EESI} of 2 for Zurich summer dataset.



Figure S11. Factor time series in a) and mass spectra in b) for 9-factor solution with C_{EESI} of 0.8 for Zurich summer dataset.



Figure S12. Factor time series in a) and mass spectra in b) for 9-factor solution with C_{EESI} of 1 for Zurich summer dataset.



Figure S13. Factor time series in a) and mass spectra in b) for 9-factor solution with C_{EESI} of 2 for Zurich summer dataset.



Figure S14. Factor time series in a) and mass spectra in b) for 10-factor solution with C_{EESI} of 0.01 for Zurich winter dataset.



Figure S15. Factor time series in a) and mass spectra in b) for 11-factor solution with C_{EESI} of 0.001 for Zurich winter dataset.



Figure S16. Factor time series in a) and mass spectra in b) for 11-factor solution with C_{EESI} of 0.005 for Zurich winter dataset.



Figure S17. Factor time series in a) and mass spectra in b) for 11-factor solution with C_{EESI} of 0.01 for Zurich winter dataset.



Figure S18. Factor time series in a) and mass spectra in b) for 12-factor solution with C_{EESI} of 0.05 for Zurich winter dataset.



Figure S19. Factor time series in a) and mass spectra in b) for 13-factor solution with C_{EESI} of 0.05 for Zurich winter dataset.





Figure S20. Scaled residual distribution of AMS (red line) and EESI (blue line) in a 12-factor solution from joint dataset in Zurich winter and corresponding overlap fractions when C_{EESI} is equal to 0.005 in a), 0.005 in b), and 1 in c), respectively. Balanced solution is shown in b), whereas in a) and c), AMS and EESI is overweighted, respectively.



Figure S21. Fraction of unmixed solution selected for further analysis as a function of confidence level (p) for summer dataset in a) and winter dataset in b).







Figure S22. Acceptance probability (i.e., all criteria satisfied simultaneously) calculated from all runs in multi-2D scans as a function of *a*-value of a) $COA_{S,C}$, b) InorgNit_{S,C} and c) HOA_{S,C}. To maintain consistency with *a*=0.1 to *a*=1.0, the *a*=0 point considers only runs in which the factor in question is being scanned against a single other factor, discarding runs for which the factor in question is fixed at *a*=0 while two other factors are scanned. Within each sub-figure, the response to different criteria thresholds are shown. Final selected values for criteria thresholds are displayed as a thicker line, while vertical dashed lines denote the final selected upper limit for *a*-value randomisation in the subsequent bootstrap analysis. Note that due to the requirement that all criteria be satisfied simultaneously, the thicker lines (and only the thicker lines) are identical across all panels in a sub-figure. Acceptance requires that a run fulfil all criteria simultaneously.





Figure S23. Criteria to select *a*-value range for $CSOA_{W,C}$ factor in Zurich winter dataset. Distribution of mass weighted fraction of nicotine in $CSOA_{W,C}$ for 726 runs, with 396 runs with *a*-value of 0 and 33 runs for each a-value from 0.1 to 1. The mass weighted fraction of nicotine apportioned to $CSOA_{W,C}$ in most runs are higher in 0.96.









Figure S24. Acceptance probability (i.e., all criteria satisfied simultaneously) calculated from all runs in multi-2D scans as a function of *a*-value of a) $HOA_{W,C}$, b) $COA_{W,C}$, c) InorgNit_{W,C} and d) $CSOA_{w,C}$. To maintain consistency with *a*=0.1 to *a*=1.0, the *a*=0 point considers only runs in which the factor in question is being scanned against a single other factor, discarding runs for which the factor in question is fixed at *a*=0 while two other factors are scanned. Within each sub-figure, the response to different criteria thresholds are shown. Final selected values for criteria thresholds are displayed as a thicker line, while vertical dashed lines denote the final selected upper limit for *a*-value randomisation in the subsequent bootstrap analysis. Note that due to the requirement that all criteria be satisfied

simultaneously, the thicker lines (and only the thicker lines) are identical across all panels in a subfigure. Acceptance requires that a run fulfil all criteria simultaneously.



Figure S25. Diurnal cycle of median accepted 764 runs in Zurich Summer dataset, represented in red lines, shaded area indicates the interquartile range, dashed lines are the maximum and minimum value of diurnal cycle calculation.



Figure S26. Average factor profiles of 764 accepted bootstrap runs in Zurich Summer dataset, coloured by different ion families, with error bars of mean±standard deviation. Note, both AMS and EESI-TOF factor profiles are normalised, according to Eq. 9.



Figure S27. Average factor profiles of 308 accepted bootstrap runs in Zurich winter dataset, coloured by different ion families, with error bars of mean±standard deviation. Note, both AMS and EESI-TOF factor profiles are normalised, according to Eq. 9.



Figure S28. Factor profiles of LO-OOA_{S,A} and MO-OOA_{S,A} from AMS-only PMF analysis and four SOAs (DaySOA1_{S,A}, DaySOA2_{S,A}, NightSOA1_{S,A} and NightSOA2_{S,A}) resolved from combined PMF analysis for summer.





Figure S29. Comparison of four summer SOA factors (DaySOA1_{S,E}, DaySOA2_{S,E}, NightSOA1_{S,E} and NightSOA2_{S,E}) resolved from EESI-TOF-only PMF analysis to the corresponding factors (DaySOA1_{S,C}, DaySOA2_{S,C}, NightSOA1_{S,C} and NightSOA2_{S,C}) resolved from the combined PMF analysis.



Figure S30. Comparison of two LABB_{W,C} factors resolved from combined dataset in Zurich winter. AMS part mass spectra comparison without NO^+ and NO_2^+ is shown in a), and EESI-TOF part mass spectra comparison is shown in b). Both figures are coloured by different ion groups.



Figure S31. Comparison of two LABB_{W,C} factors resolved from the combined dataset in Zurich winter. Scatter plot of AMS ions without NO^+ and NO_2^+ is shown in a), and scatter plot of EESI ions is shown in b). In both figures, correlation of ion intensity of LABB1_{W,C} and LABB2_{W,C} is higher than 0.99 with slope around 1.



Figure S32. Histogram of factor EESI-TOF relative sensitivity to two primary sources and four secondary sources normalised by EESI-TOF sensitivity to $COA_{S,C}$ in Zurich summer campaign. Relative sensitivity is calculated as apparent sensitivity (AS_k) to a factor over mean apparent sensitivity to COA.



Histogram of factor apparent sensitivity (AS) in winter

Figure S33. Histogram of factor relative sensitivity to two primary sources, two aged-biomass burning factors, two nitrogen-related sources, one event-specific factor, two secondary sources, and the sum of biomass burning related factors. Relative sensitivity is calculated as the apparent sensitivity (AS_k) to a factor over the mean apparent sensitivity to COA_{W,C}.



Figure S34. Histogram of ratio of the relative apparent sensitivity $AS/\overline{AS_{COA}}$ to the mean relative apparent sensitivity $\overline{AS/\overline{AS_{COA}}}$ for primary sources and secondary sources a) within 764 accepted runs in Zurich summer campaign and b) within 308 accepted runs in Zurich winter campaign. Note in b) the sum of biomass burning related factors (denoted Sum-BB_{W,C}) is calculated from MABB_{W,C}, LABB_{W,C}, NitOA2_{W,C}, NitOA2_{W,C} and EVENT_{W,C}. In this plot, all the distributions are re-centred to 1, and the width of the distribution represents the relative uncertainty of the corresponding factor.

Table S1. Summary of relevant quantities to estimate $frac_{ON,k}$ and corresponding $\sum_{j} (f_{k,j})_{AMS,k}$,

with different R_{ON} for each unconstrained factor from combined PMF analysis for summer and
winter. Note that the HOA, COA, and CSOA in both factors are constrained to have zero contribution
from NO^+ and NO_2^+ .

Summer								
NH ₄ NO ₃ reference:			$f_{\rm NO^+} = 0.487, \qquad f_{\rm NO^+_2} = 0.318, \qquad R_{\rm cal} = 0.688$					
			$R_{\rm ON} = 0.08$		$R_{\rm ON} = 0.14$		$R_{\rm ON} = 0.20$	
Factor	$f_{\rm NO^+}$	$f_{\rm NO_2^+}$	frac _{ON}	$\frac{\sum_{j} (f_{k,j})_{\text{AMS,OA}}}{\sum_{j} (f_{k,j})_{\text{AMS,all}}}$	frac _{ON}	$\frac{\sum_{j} (f_{k,j})_{\text{AMS,OA}}}{\sum_{j} (f_{k,j})_{\text{AMS,all}}}$	frac _{ON}	$\frac{\sum_{j} (f_{k,j})_{\text{AMS,OA}}}{\sum_{j} (f_{k,j})_{\text{AMS,all}}}$
HOA _{S,C}	0	0	0	1	0	1	0	1
COA _{S,C}	0	0	0	1	0	1	0	1
InorgNit _{S,C} ^{1*}	0.486	0.318	0.0358	0.225	0.0420	0.230	0.0496	0.236
CSOA _{S,C}	0	0	0	1	0	0.995	0	1
DaySOA1 _{S,C}	0.0137	0.00262	0.742	0.996	0.869	0.998	1.000	1.000
DaySOA2 _{S,C}	0.0111	0.00103	0.969	1.000	1.000	1.000	1.000	1.000
NightSOA1 _{S,C}	0.0445	0.00979	0.682	0.983	0.798	0.989	0.944	0.997
NightSOA2 _{S,C}	0.0679	0.00824	0.897	0.993	1	1	1	1
Winter								
	reference:		$f_{\rm NO^+} = 0.630, \qquad f_{\rm NO_2^+} = 0.248, \qquad R_{\rm cal} = 0.394$					
			$R_{\rm ON} = 0.08$		$R_{\rm ON} = 0.14$		$R_{\rm ON} = 0.20$	
Factor	$f_{\rm NO^+}$	$f_{\rm NO_2^+}$	frac _{ON}	$\frac{\sum_{j} (f_{k,j})_{\text{AMS,OA}}}{\sum_{j} (f_{k,j})_{\text{AMS,all}}}$	frac _{ON}	$\frac{\Sigma_j(f_{k,j})_{\text{AMS,OA}}}{\Sigma_j(f_{k,j})_{\text{AMS,all}}}$	frac _{ON}	$\frac{\Sigma_j(f_{k,j})_{\text{AMS,OA}}}{\Sigma_j(f_{k,j})_{\text{AMS,all}}}$
HOA _{W,C}	0	0	0	1	0	1	0	1
COA _{W,C}	0	0	0	1	0	1	0	1
InorgNit _{W,C} ^{1*}	0.636	0.263	0	0.101	0	0.101	0	0.101
CSOA _{W,C}	0	0	0	1	0	1	0	1
SOA1 _{W,C}	0.264	0.0347	0.799	0.940	1	1	1	1
SOA2 _{W,C}	0.570	0.238	$0^{2^{*}}$	0.192	$0^{2^{*}}$	0.192	$0^{2^{*}}$	0.192
MABB _{W,C}	0.133	0.00185	1	1	1	1	1	1
LABB _{W,C}	0.195	0.0176	0.958	0.991	1	1	1	1
NitOA1 _{W,C}	0.0781	0.00260	1	1	1	1	1	1
NitOA2 _{W,C}	0.143	0.293	$0^{2^{*}}$	0.564	$0^{2^{*}}$	0.564	$0^{2^{*}}$	0.564
EVENT _{W,C}	0.0736	0.0203	0.319	0.936	0.416	0.945	0.572	0.960

Note 1*: The $frac_{ON}$ for InorgNit is not 0 due to the uncertainties in the constraint. $\frac{\sum_j (f_{k,j})_{AMS,OA}}{\sum_j (f_{k,j})_{AMS,all}}$ in this factor is not 0 due to the uncertainties in the constraint and the CO₂⁺ resulted from NH₄NO₃ which is

also included in the reference profiles (Pieber et al., 2016).

Note 2*: according to Eq. 18, the $frac_{ON}$ is negative, therefore, we regarded all signals from NO⁺ and NO₂⁺ as inorganics, and set $frac_{ON}$ to be 0.

Factor (Standard deviation / mean factor concentration)median HOA_{S,C} 7.4% 4.5% COA_{S,C} 5.2% InorgNit_{S,C} CSOA_{S.C} 3.2% DaySOA1_{S,C} 5.4% DaySOA2_{S,C} 1.4% 0.6% NightSOA1_{S,C} NightSOA2_{S,C} 4.3% HOA_{W,C} 12.8% COA_{W,C} 14.5% InorgNit_{W,C} 26.5% 13.7% CSOA_{W,C} $SOA1_{W,C}$ 12.2% 18.5% $SOA2_{W,C}$ MABB_{W,C} 24.1% 18.1% LABB_{W,C} NitOA1_{W,C} 24.3% 35.1% NitOA2_{W,C} EVENT_{W,C} 3.5%

Table S2. Summary of the median value of the ratio of standard deviation to mean factor concentration over accepted runs at each time point for summer and winter factors.

Table S3. Summary of apparent sensitivities for factors retrieved from combined PMF analysis for summer and winter.

Factor	Factor apparent sensitivity (AS_k) (cps /(μ g m ⁻³))
COA _{S,C}	$5.09 (\pm 0.45) \times 10^2$
CSOA _{S,C}	$22.10 (\pm 1.27) \times 10^2$
DaySOA1 _{S,C}	$8.45~(\pm 0.32) imes 10^2$
DaySOA2 _{S,C}	$17.47 (\pm 1.20) \times 10^2$
NightSOA1 _{S,C}	$7.86 (\pm 0.43) \times 10^2$
NightSOA2 _{S,C}	$38.23 (\pm 4.33) \times 10^2$
Bulk OA _{S,C}	$12.54 (\pm 0.10) \times 10^2$
COA _{W,C}	$1.10 (\pm 0.13) \times 10^3$
CSOA _{W,C}	$4.56 (\pm 0.54) \times 10^3$
SOA1 _{W,C}	$1.29 (\pm 0.11) \times 10^3$
SOA2 _{W,C}	$2.25 (\pm 0.24) \times 10^3$
MABB _{W,C}	$2.62 (\pm 0.49) \times 10^3$
LABB _{W,C}	$6.51 (\pm 2.01) \times 10^3$
NitOA1 _{W,C}	$1.43 (\pm 0.57) \times 10^3$
NitOA2 _{W,C}	$1.12 (\pm 0.21) \times 10^3$
EVENT _{W,C}	$1.93 (\pm 0.18) \times 10^3$
Bulk OA _{W,C}	$2.27 (\pm 0.07) \times 10^3$

Reference

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