# Airborne measurements of the spatial distribution of aerosol chemical composition across Europe and evolution of the organic fraction: Supplementary material

- W. T. Morgan<sup>1</sup>, J. D. Allan<sup>1,2</sup>, K. N. Bower<sup>1</sup>, E.J. Highwood<sup>3</sup>,
   D. Liu<sup>1</sup>, G. R. McMeeking<sup>1</sup>, M. J. Northway<sup>3</sup>, P. I. Williams<sup>1,2</sup>,
   R. Krejci<sup>4</sup> and H. Coe<sup>1</sup>
- 6 1. Centre for Atmospheric Science, University of Manchester, Manchester, UK
- 7 2. National Centre for Atmospheric Science, University of Manchester, UK
- 8 3. Department of Meteorology, University of Reading, UK
- Department of Applied Environmental Science, Atmospheric Science Unit, Stock holm University, Sweden

# 11 **1** Scope

The supplementary material outlined in this document is provided in order to present 12 the meteorological context of the flight operations and support the analysis techniques 13 and data quantification steps outlined in the main paper. The meteorological fields 14 corresponding to each flying period are presented and further information regarding 15 the photochemical context of the operations is presented. Further details regarding 16 the volume closure between the Aerosol Mass Spectrometer (AMS) and the Passive 17 Cavity Aerosol Spectrometer Probe (PCASP) are discussed. Comparison of the esti-18 mated HOA with primary combustion tracers is included. The relationship between 19 the fractional contribution of Low-Volatility Oxygenated Organic Aerosol (LV-OOA) 20 to the organic mass and the normalised organic signal at m/z 44 is also shown. Further 21 information is provided regarding the Positive Matrix Factorisation (PMF) analysis ex-22 amples from the main text, as well as a summary of some PMF diagnostics for the 23 whole dataset. The PMF analysis was performed using the tools presented by Ulbrich et al. (2009). 25

### <sup>26</sup> 2 Meteorological summary

Figs. S1 and S2 display the typical meteorological conditions prevalent during each
period considered by the analysis. The periods are relatively consistent in terms of
their transport patterns, with the air masses transporting pollution from continental
Europe downwind to either the UK region or into the Eastern Atlantic Ocean. Thus the
flights are predominantly focused upon either sampling such pollution over continental
Europe itself or at a range of scales downwind.

The evolution of the aerosol chemical composition during the LONGREX-2 period 33 was examined based upon the relatively consistent transport patterns prevalent during 34 the period. Fig. S3a displays the back trajectories for each flight during this period 35 based upon Straight and Level Runs (SLRs) during each flight. The trajectories display 36 highly consistent behaviour during the period, which is unsurprising given the relative 37 stability of the high pressure system located over Northern Europe during this period. 38 Fig. S3b highlights the back trajectory from the 14 May 2008, which was initialised 39 from a SLR during B374 in the Eastern Atlantic Ocean. B374 represented the end-40 point in our operations during this period both in terms of the geographical location 41 of the missions and also the distance from continental European sources i.e. the most 42 aged polluted air mass. The back trajectory indicates that the spatial coverage of the 43 flight operations closely matches the air mass transport during the period leading up to 44 the 14 May 2008. Specifically, flights B370-B374 took place across Northern Europe 45 during this period covering close to 5 days of air mass transport. 46

## **3 Photochemical context**

The relationship between  $O_3$  and CO with the  $O_3:NO_x$  ratio discussed in the main paper is presented in Fig. S4. The results indicate that  $O_3$  increases and CO decreases steadily in the 1-100  $O_3:NO_x$  range, which is a reflection of photochemistry and dilution respectively. Beyond an  $O_3:NO_x$  ratio of 100, the concentrations decrease with CO returning to background levels and  $O_3$  remaining relatively constant in the 40-60 ppb range.

# **4** HOA estimation and LV-OOA interpretation

Included in Fig. S5a are correlations of the estimated HOA concentration with Black 55 Carbon (BC), NO<sub>x</sub> and CO. These indicate that for 8 of the flights, the correlations 56 of the estimated HOA with these primary emission tracers are greater than 0.5. Cor-57 relations lower than 0.5 are generally encountered on flights where these tracers and 58 the estimated HOA are very low, thus the correlations break down at values when the relationships exhibit enhanced noise due to low signal. This is demonstrated in 60 61 Fig. S5b and c, where at low concentrations the relationships are relatively flat but at enhanced concentrations, the correlation is greater. Given the simple nature of the 62 HOA estimate, these correlations and relationships do provide some confidence that 63 the estimated HOA provides a qualitative indicator of the contribution of HOA to the 64

of OM burden. Also shown in Fig. S5a are the correlations between the Low-Volatility

66 Oxidised Organic Aerosol (LV-OOA) organic mass fraction and the m/z 44:OM ratio

67 described in the main paper.

# **5** AMS versus PCASP comparison

Validation of the collection efficiency treatment applied to the dataset following the 69 principles developed by Matthew et al. (2008) is accomplished by comparing the AMS 70 data with the volume estimated concentrations from the PCASP instrument. The AMS 71 total mass concentrations were converted to total volume concentrations using the den-72 sities reported by Cross et al. (2007), which correspond to 1.27 g cm<sup>-3</sup> for organics 73 and  $1.77 \text{ g cm}^{-3}$  for inorganics. A comparison of the estimated volume from the AMS 74 and PCASP is shown for SLRs below 3000 m in Fig. S6. Over all of the considered 75 flights, the estimated AMS volume concentrations were 26% higher than the estimated 76 PCASP volumes. This average agreement is predominantly determined by the LON-77 GREX flights, which were quite consistent in terms of the agreement from flight-to-78 flight. The ADIENT flying periods sit on either side of the overall regression slope, 79 with the ADIENT-2 flying indicating that the PCASP volume was 48% of the AMS 80 volume. These discrepancies are considered tolerable given the large uncertainties pre-81 viously reported in the literature for PCASP volume estimates (e.g. Moore et al., 2004; 82 Hallar et al., 2006) and the uncertainties in the AMS volume estimates. 83

For B357, the PCASP volume estimate was more than two times greater than the 84 AMS volume estimate, which is outside of the bounds of uncertainty for the two instru-85 ments. The reason for this discrepancy is unknown but could reflect an artifact in either 86 instrument or the presence of material that is not detected by the AMS. The discrep-87 ancy between the two instruments is also reflected in the calculated volume-scattering 88 relationship when comparing the measurements with a nephelometer system. The main 89 difference between B357 and the other flights in the dataset is the sampling altitude of 90 the aircraft, where in B357, the aircraft operated at a constant altitude of 200 m for 91 the majority of the flight. The other flights in the dataset operated at altitudes higher 92 in the boundary layer. Potentially, the nephelometer and PCASP measurements could 93 be perturbed by the constant low-level flying in a humid environment as the aerosol 94 sampled may not be sufficiently dried in the inlet lines and by the heater respectively. 95 The AMS volume estimates do not include water, so this could potentially cause the 96 discrepancy. Additionally, the PCASP and nephelometer may be measuring refractory 97 material or particles above the cut off of the AMS aerodynamic lens. This would also lead to the AMS underestimating the volume relative to the PCASP. 99

# **6** Additional organic factor analysis diagnostics

A summary of the  $Q/Q_{expected}$  parameter is shown in Fig. S7, which indicates that the parameter is generally less than 2. Four flights had  $Q/Q_{expected}$  values greater than 2, with the largest value being 5.01 (B374). Such values are greater than is generally considered optimal if attempting to produce a perfect characterisation of the dataset but given the difficulty in deriving robust results when more than 2 factors are chosen, it is
 not possible to reduce Q further. Consequently, the additional Q contribution prevalent
 in the dataset is considered as 'model error'.

#### **108** 6.1 Example PMF solutions

The time series, mass spectra and residuals for the two factor solutions for the three example flights used to describe the factor analysis in the main paper are presented here. B369 is also included as an additional example, reflecting background conditions for comparison with the more polluted examples. The results for B357 are shown in Fig. S8, for B362 in Fig. S9, for B369 in Fig. S10 and B406 in Fig, S11. All the examples show excellent agreement between the measured and reconstructed organic mass concentrations.

#### **116 6.2 Bootstrapping analysis**

Numerical stability of the 2-factor solutions was quantitatively evaluated using a boot-117 strapping analysis (Ulbrich et al., 2009, and references therein) where random resam-118 pling of the data matrix is performed in the time dimension. This analysis was per-119 formed using 20 iterations, with the results being grouped according to the uncentred 120 correlation coefficient between mass spectral profiles. The results from the bootstrap-121 ping analysis are shown for B357 in Fig. S12, for B362 in Fig. S13, for B369 in Fig. 122 S14 and for B406 in Fig. S15. The bootstrapping results for all the flights are sum-123 marised in Table S1. The variance of the solutions in both the time series and factor 124 profile dimensions is evaluated using suitable metrics. The time series diagnostic is the 125 mean of the standard deviation for each factor, reported as a percentage of the over-126 all mean mass concentration. For the mass spectra, the greatest standard deviation for 127 each factor profile from the bootstrapping analysis is reported. A mean is not deemed 128 appropriate to evaluate the stability of the mass spectra as the chemical assignment of 129 factors is performed based on a limited number of peaks (i.e. less than 10). 130

The OOA-1 (LV-OOA) factor profiles are highly robust with little deviation be-131 tween the average mass spectrum from the bootstrapping analysis and the base solu-132 tion. Furthermore, the standard deviations are typically low. This is a consistent result 133 throughout the dataset, which is shown by the low scores for the diagnostics in Table 134 S1. The OOA-2 factor profiles are more variable for B362, B369 and B406, with less 135 stability in the signal intensity at m/z 44. This reflects the continuum nature of the OM 136 discussed in the main text, whereby there is significant variability on a given flight in 137 the level of oxidation. Thus by randomly resampling the dataset in the time dimension 138 using the bootstrapping procedure, conditions with either enhanced or diminished m/z 139 44 in the OOA-2 may be encountered and this is reflected by enhanced standard devi-140 ations in both the time series and mass spectra. The normalised standard deviation for 141 the time series of OOA-2 for B369 is much greater than the other flights in the dataset. 142 This is predominantly a result of the low concentrations encountered during the flight, 143 coupled with large standard deviations which are associated with large changes in the 144 composition of the OOA-2 component in this instance. For much of the flight, signal 145 at m/z 57 is close to zero indicating that HOA makes a minimal contribution to the 146

OOA-2 component. However, during the peak OM event at 15:47 UTC, signal at m/z 147 57 is enhanced and thus the contribution of HOA is likely elevated. This event co-148 incides with the maximum in the NO<sub>x</sub> concentration and is associated with potential 149 sampling of low-level urban outflow from Stockholm into the Baltic Sea. During this 150 period, the standard deviations for the OOA-2 factor increase. Thus such changes in 151 the OOA-2 composition from SV-OOA dominated to HOA dominated are reflected by 152 large increases in the standard deviation values from the bootstrapping analysis. The 153 much lower normalised standard deviation values associated with the time series of 154 the OOA-2 components for the rest of the dataset suggest this feature to be atypical. 155 This is a reflection of the regional nature of the measurements, with few instances of 156 prolonged exposure to intense urban signatures. 157

The results presented here demonstrate the robustness of the chosen 2-factor solutions in terms of both the mass spectra and time series of the components. The OOA-1 (LV-OOA) components are highly numerically stable, while the decreased numerical stability of the OOA-2 (SV-OOA and HOA) components is entirely consistent with the continuum of oxidation/aging discussed in the main text.

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Flight	$SD_{ts}/TS$ (%)		Max $(SD_{ms})$ (%)	
	OOA-1	OOA-2	00A-1	OOA-2
B357	2.0	2.5	0.77	0.85
B362	5.6	9.7	1.00	2.44
B365	2.8	3.0	0.62	1.31
B366	13.4	11.8	1.66	1.46
B369	19.1	45.9	2.23	1.66
B370	2.9	3.6	0.52	0.46
B371	4.8	8.0	0.41	1.82
B373	2.3	3.9	1.06	0.86
B374	1.1	1.7	0.44	0.68
B379	6.6	9.3	0.61	1.63
B380	1.4	2.4	0.36	0.37
B401	11.1	13.3	0.24	0.32
B402	5.5	14.9	0.76	1.23
B406	1.4	1.8	1.06	0.79

Table S1: Summary of the diagnostics relating to the bootstrapping analysis from the dataset for each flight.



Figure S1: Flight tracks of the BAe-146 considered by this analysis for the ADIENT periods. Also shown are ECMWF 850 hPa geopotential height fields for each period considered by the analysis, where the field is either pertinent to a particular flight or is representative of the overall meteorological regime of the period. All geopotential height fields are for 12UTC. (a) summarises the flights for the UK-based ADIENT flying in April 2008 and the geopotential height field is from 16 April 2008 (B357). (b) summarises the flights for the UK-based ADIENT flying in September 2008 and the geopotential height field is from 25 September 2008 (B406).



Figure S2: Same plots as Fig. S1 but now for the LONGREX flying period. (a) summarises the flights for the LONGREX-1 period with a geopotential height field for the 06 May 2008. (b) summarises the flights for the LONGREX-2 period with a geopotential height field for the 14 May 2008. (c) summarises the flights for the LONGREX-3 period with a geopotential height field for the 22 May 2008.



Figure S3: (a) Back trajectories at 850 hPa initialised from 1200 UTC on each flight day from the LONGREX-2 period. The initialisation points correspond to several SLRs during each flight. (b) Air mass back trajectory initialised from 1200 UTC on 14 May 2008 at 850 hPa. The numbered points relate to the number of days passed since the air mass was in that location.



Figure S4: (a) Relationship between  $O_3$  and  $O_3:NO_x$  for all flights. (b) Relationship between CO and  $O_3:NO_x$  for all flights expect for ADIENT-2. The red dashed lines refer to the distance from source boundaries discussed in the main paper.



Figure S5: (a) Summary of correlations for estimated HOA with Black Carbon (BC),  $NO_x$  and CO. Also shown are the correlations for LV-OOA organic mass fractions with the normalised organic signal at m/z 44. (b) and (c) display the median estimated HOA versus  $NO_x$  and BC respectively for SLRs across the whole dataset. The error bars reflect the  $25^{th}$  and  $75^{th}$  percentiles on each SLR. The black dotted lines in (b) correspond to the literate POA:NO<sub>x</sub> ratios reported by Kirchstetter et al. (1999).



Figure S6: Comparison of inferred volume from the AMS with estimated sub-micron volume derived from the PCASP. The markers refer to SLRs below 3000 m i.e. within the boundary layer. The text markers refer to the flight which the point is from. Linear regression lines are shown for both individual flying periods and the study as a whole.



Figure S7: Summary of the  $Q/Q_{expected}$  parameter for each flight in the dataset. All values are for the two-factor solutions with an fPeak of zero.



Figure S8: PMF solution for the two factor case from flight B357 including factor component time series (a) and mass spectra (b). The absolute residual is also shown in (c), whilst a comparison between the factor analysis reconstructed mass and measured organic signal is displayed in (d).



Figure S9: PMF solution for the two factor case from flight B362 including factor component time series (a) and mass spectra (b). The absolute residual is also shown in (c), whilst a comparison between the factor analysis reconstructed mass and measured organic signal is displayed in (d).



Figure S10: PMF solution for the two factor case from flight B369 including factor component time series (a) and mass spectra (b). The absolute residual is also shown in (c), whilst a comparison between the factor analysis reconstructed mass and measured organic signal is displayed in (d).



Figure S11: PMF solution for the two factor case from flight B406 including factor component time series (a) and mass spectra (b). The absolute residual is also shown in (c), whilst a comparison between the factor analysis reconstructed mass and measured organic signal is displayed in (d).



Figure S12: Results of the bootstrapping analysis for the two factor solution mass spectra for flight B357. (a) displays the mass spectrum for OOA-1, while (b) displays the mass spectrum for OOA-2.



Figure S13: Results of the bootstrapping analysis for the two factor solution mass spectra for flight B362. (a) displays the mass spectrum for OOA-1, while (b) displays the mass spectrum for OOA-2.



Figure S14: Results of the bootstrapping analysis for the two factor solution mass spectra for flight B369. (a) displays the mass spectrum for OOA-1, while (b) displays the mass spectrum for OOA-2.



Figure S15: Results of the bootstrapping analysis for the two factor solution mass spectra for flight B406. (a) displays the mass spectrum for OOA-1, while (b) displays the mass spectrum for OOA-2.