



Supplement of

Evidence for ambient dark aqueous SOA formation in the Po Valley, Italy

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47 AMS Organic Aerosol Source Apportionment

48 Source apportionment analysis on the high resolution organic aerosol (OA) mass spectra provided by the AMS was made using the Multilinear Engine algorithm (ME-2) developed by 49 Paatero [1999] and the interface Solution Finder (SoFi 4.9) [Canonaco et al., 2013]. Prior to 50 analysis, the organic matrix was prepared according to the recommendations of Ulbrich et al. 51 [2009]. First, isotope ions were removed and a minimum counting error was applied. Fragments 52 with a signal-to-noise ratio (SNR) below 0.2 were down-weighted by a factor of 10 and 53 54 fragments with a SNR between 0.2 and 2 were down-weighted by a factor of 2. Finally, the fragments related to ion CO_2^+ were also down-weighted since they are calculated as a constant 55 fraction of the ion CO_2^+ [Allan et al., 2004]. Elemental analysis on the mass spectra of the 56 identified factors was performed using the Analytic Procedure for Elemental Separation (APES 57 vers. 1.06) based on Aiken et al. [2007, 2008] and including the improved estimation from 58 Canagaratna et al. [2015]. 59 60

For the first attempt, a non-constrained approach was investigated using a factor number ranging from 1 to 6 and applying 10 seeds (Figure S1). The best solution was obtained for the 4-61 factors solution (Figure S2) including 3 different oxygenated OA (OOA-a, OOA-b and OOA-c) 62 63 and a mixed-sources factor (mix-OA). The mix-OA factor contributes to 16% of the total OA and it has a mass spectrum with ions typically associated with hydrocarbon-like OA and shows 64 the lowest O/C (oxygen/carbon) ratio (0.28) compared to the other factors. Although, it indicates 65 66 that this factor can be related to primary OA, its elemental ratios are higher than reported Hydrocarbon-like OA (HOA) factors [Canagaratna et al., 2015]. This factor also has a large 67 contribution of oxygenated fragments at m/z 43 (CHO⁺) and 44 (CO₂⁺) compared to previously 68 reported HOA factors. Regarding its time variation, this factor correlates relatively well with 69 gas-phase primary emissions tracers (e.g., benzene (r=0.35), toluene (r=0.48)) and particulate 70 black carbon (BC, r=0.49) as well as with semi-volatile inorganic compounds (e.g., nitrate 71 72 (r=0.57)). Therefore, considering the mass spectrum and time series particularities, this factor 73 was identified to represent a mixture of Hydrocarbon-like OA (HOA) and semi-volatile OA (SV-OOA). The three OOA factors have quite similar mass spectra, but they present clear distinct 74 75 time trends. Therefore, they are considered as separate factors and identified as follows: 76

- OOA-a (10% of total OA) appears to be specific to a certain time period of the campaign

characterized by high temperature, a high pressure system, and stagnant air masses. Therefore,

this OOA-a factor can be related to an accumulation of aged particles on the regional

80 background. The OOA-a mass spectrum is dominated by oxygenated ions and shows the highest

- 81 O/C ratio (1.02) in agreement with aged OA.
- 82

- OOA-b (30% of total OA) is the least oxygenated OOA factors (O/C = 0.55). It also correlates well with sulfate (r=0.58), but also with methanesulfonic acid (MSA, r=0.60). Therefore, this suggested that OOA-b might be related to marine OA rather than continental OA. This is in agreement with previous measurements made at the same location by *Saarikoski et al.* [2012],

87 who reported a factor with a source originating from the Mediterranean Sea.

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- OOA-c (44% of total OA) correlates with particulate sulfate (r=0.55) but not with MSA,

90 opposite to OOA-b, and therefore can be linked to more "continental" SOA formation.

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Increasing the number of factors did not provide a significant change on the mix-OA 92 93 factor as illustrated in Figure S3 but rather a change in the split of the different OOA factors. Therefore, as a second attempt, the source apportionment was performed in a semi-constrained 94 95 mode in order to dissociate primary OA from semi-volatile OA more clearly. The principal primary OA source expected is the HOA factor. In contrary to Saarikoski et al. [2012] who 96 reported the contribution of a Biomass Burning OA (BBOA) factor in the spring season, here no 97 BBOA is expected since the contribution of the fragment m/z 60 (a tracer for BBOA) to total OA 98 99 was systematically below the background level of 0.3% defined by *Cubison et al.* [2011]. Consequently, a reference HOA mass spectrum corresponding to an average of 2 HOA factors 100 previously identified in this area (M. Rinaldi, personal communication) was used as a priori 101 102 information to partially constrain the model.

For this approach, the number of factors was varied from 5 to 7 since at least 5 factors are 103 expected based on previous AMS measurements in the Po Valley (HOA and 4 types of OOA). In 104 order to test the sensitivity of the results, the difference in the degree of variation for the various 105 fragments for the output HOA factor to the input reference mass spectra (the so-called a-value) 106 was investigated for a-values ranging from 0.05 (i.e., extremely constrained run where fragments 107 of the resulting HOA factor can only vary from 5% compared to the reference HOA) to 0.5 (50% 108 variation). The contribution of the HOA to the total OA was extremely stable over the 109 investigated a-value range indicating that identification of the HOA factor is quite robust (Figure 110 111 S4). The 5-factors solution (with an a-value of 0.1) was considered as the final solution (Figure S5). This solution corresponds to better discrimination between HOA and the semi-volatile OA 112 (referred in the following as OOA-1), while the 3 others OOA factors correspond to the 113 previously identified ones in the non-constrained model and here are referred to as OOA-2 (12% 114 of OA), OOA-3 (28% of OA), and OOA-4 (45% of OA) in order to avoid confusion when 115 referring to the first (unconstrained) analysis (Figures S6 and S7). Increasing the number of 116 117 factors to 6 or 7 solely leads to a further splitting of the OOA factors without a clear identification. 118

The HOA factor (4% of OA) now better follows the time trend of benzene (r=0.58), 119 while no real improvement of the correlation with BC (r=0.50) and toluene (r=0.49) can be 120 reported. However, the semi-volatile OOA-1 is now better correlated with nitrate (r=0.74) than 121 HOA (r=0.36) confirming the presence of these two factors in the previously identified mix-OA. 122 Although the OOA-1 factor (12% of total OA) is related to semi-volatile OA, its mass 123 spectrum appears to be more oxygenated (higher contribution of the CO_2^+ fragment compared to 124 the CHO⁺) than classical SV-OOA (CHO⁺ > CO₂⁺). However, this is quite similar to the 125 previously reported semi-volatile OOA measured at SPC by Saarikoski et al. [2012]. 126 Although contributions of the 3 others OOA factors (OOA-2, OOA-3, and OOA-4) to the 127

total OA are quite similar to the contribution of their corresponding factors in the non-128 constrained mode (12%, 28% and 45%, respectively), some small differences can be reported 129 130 either in terms of their mass spectra (and consequently their elemental ratios) or their time trends. These differences can be explained by a small contribution of the non-constrained OOA factors 131 (i.e., OOA-a, OOA-b, and OOA-c) to OOA-1. The most stable factor is OOA-2 which correlates 132 quite well with the previously identified OOA-a, even if the final factor has a lower oxidation 133 state. Although OOA-2 contributed to 12% over the entire time period, during its prevalent 134 period it accounted for up to more than half of the OA. 135

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230 Figure Captions

- Figure S1. Evolution of the Q/Q_{exp} ratio (top) and factor contribution (bottom) over the investigated factor range for the non-constrained model.
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Figure S2. Time series (top) and mass spectra colored by fragments family (bottom) for the non-constrained 4-factors solution.

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Figure S3. Evolution of the mix-OA factor time series (top) and mass spectra (bottom) for

different factor solutions (from 3 to 6). Numbers in parentheses on the bottom plots (following
the number of the factor solution) correspond to the slope of the regression line compare to the
selected 4-factors solution.

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Figure S4. Evolution of the Q/Q_{exp} ratio (top) and factor contribution (bottom) over the investigated factor range for the partially-constrained model.

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Figure S5. Overview of the partially-constrained factor solution including (a) time series of the factors and corresponding tracers, (b) mass fraction of the different factors to the total OA, and (c) mass spectra of the factors colored by fragment family.

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Figure S6. Comparison between the time trends of the factors identified for the non-constrained
(y-axis) and the ones identified for the partially-constrained (x-axis) analysis.

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Figure S7. Comparison between the mass spectra of the factors identified for the non-

constrained (y-axis) and the ones identified for the partially-constrained (x-axis) analysis.

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Figure S1



seed







Figure S4



seed













