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Speciation of organic fractions does matter for aerosol source apportionment. Part 2: intensive short-term campaign in the Paris area (France)

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Abstract

The present study aimed at performing PM$_{10}$ source apportionment, using positive matrix factorization (PMF), based on filter samples collected every 4h at a sub-urban station in the Paris region (France) during a PM pollution event in March 2015 (PM$_{10}$ > 50 µg m$^{-3}$ for several consecutive days). The PMF model allowed to deconvolve 11 source factors. The use of specific primary and secondary organic molecular markers favoured the determination of common sources such as biomass burning and primary traffic emissions, as well as 2 specific biogenic SOA (marine + isoprene) and 3 anthropogenic SOA (nitro-PAHs + oxy-PAHs + phenolic compounds oxidation) factors. This study is probably the first one to report the use of methylnitrocatechol isomers as well as 1-nitropyrene to apportion secondary OA linked to biomass burning emissions and primary traffic emissions, respectively. Secondary organic carbon (SOC) fractions were found to account for 47% of the total OC. The use of organic molecular markers allowed the identification of 41% of the total SOC composed of anthropogenic SOA (namely, oxy-PAHs, nitro-PAHs and phenolic compounds oxidation, representing 15%, 9%, 11% of the total OC, respectively) and biogenic SOA (marine + isoprene) (6% in total). Results obtained also showed that 35% of the total SOC originated from anthropogenic sources and especially PAH SOA (oxy-PAHs + nitro-PAHs), accounting for 24% of the total SOC, highlighting its significant contribution in urban influenced environments. Anthropogenic SOA related to nitro-PAHs and phenolic compounds exhibited a clear diurnal pattern with high concentrations during the night indicating the prominent role of night-time chemistry but with different chemical processes involved.

Keywords: Aerosol, Source apportionment, PMF, SOA, Molecular markers.
1. Introduction

Atmospheric particulate matter (PM) plays an important role on climate and air quality (Boucher et al., 2013; Heal et al., 2012). To design effective PM concentration reduction strategies, their sources and contributions from each source need to be known thoroughly. Thus, several source apportionment methods have been developed for this purpose. Receptor-oriented modelling is one of the approaches that has been extensively used for PM source apportionment studies in the past decades (Belis et al., 2015; Hopke, 2016). The most commonly used receptor models include chemical mass balance (CMB) (Chow and Watson, 2002; Watson et al., 2002), positive matrix factorization (PMF) (Paatero and Tapper, 1994), and UNMIX (Henry, 1997). PMF is a powerful multivariate method that can resolve the dominant positive factors without prior knowledge of sources (Hopke et al., 2006; Kim et al., 2003; Shrivastava et al., 2007; Zhang et al., 2009).

Typically, PMF uses trace elements, and organic and elemental carbon (OC/EC), as well as secondary ions and metals as the input data matrix to explore the “co-variances” between species (Kim and Hopke, 2004; Kim et al., 2003). The use of tracers with high source specificity in the PMF model can enhance the interpretation of the factors. For example, levoglucosan is often used to trace the biomass burning source (Simoneit, 2002). The use of organic molecular markers in the PMF has resulted in considerable progress in the understanding of the organic aerosol (OA) fraction (Jaeckels et al., 2007; Laing et al., 2015; Schembari et al., 2014; Shrivastava et al., 2007; Srimuruganandam and Shiva Nagendra, 2012; Waked et al., 2014; Wang et al., 2012; Zhang et al., 2009). However, these studies have often been based on 12 or 24 h sampling periods, making difficult to capture the information on “fast” chemical processes related to OA emissions and formation. The use of a higher time-resolution datasets (e.g. filter samplings every 6 h or less) may
facilitate the understanding of the processes involved, for both primary and secondary sources, and the analysis of their diurnal cycles.

This paper is the second paper of a two-part series demonstrating that the speciation of the OA fraction is important for PM source apportionment. Note that, the use of organic molecular markers in source apportionment studies is based on the assumption that these compounds are chemically stable in the atmosphere (i.e. tracer compounds) (Schauer et al., 1996). Some molecules can undergo a decay in the atmosphere by photochemical processes involving sunlight and atmospheric oxidants and their use may cause a bias in the source apportionment. Nevertheless, the first paper (Srivastava et al., 2018) highlighted the advantage of using primary and secondary organic molecular markers to resolve sources rarely apportioned in the literature such as two types of primary biogenic organic aerosols (fungal spores and plant debris), as well as specific biogenic and anthropogenic secondary OA (SOA). This second paper focuses on the identification of sources during a major PM pollution event using high resolution filter data in order to apportion specific primary (POA) and secondary OA fractions using various and distinctive markers and to understand the atmospheric chemical processes involved.

2. Experimental

2.1. Monitoring site and sampling period

Measurements were conducted at the ACTRIS SIRTA atmospheric supersite (Site Instrumental de Recherche par Télédétection Atmosphérique, 2.15° E; 48.71° N; 150 ma.s.l; http://sirta.ipsl.fr; Haeffelin et al. (2005)). This site is located approximately 25 km southwest from Paris city centre (Figure S1), surrounded by forests, agricultural fields and small villages, and is representative of suburban background conditions of the Paris region (Crippa et al., 2013a; Petit et al., 2017a; Petit
et al., 2014; Sciare et al., 2011). An intensive campaign was performed from 6-21, March 2015. The late winter-early spring period was chosen on purpose as intense PM pollution events are usually observed in Northern France (and Europe) during this period of the year due to the combination of significant residential emissions, manure spreading, stagnant atmospheric conditions favouring the accumulation of pollutants and possible photochemical processes enhancing the formation of secondary aerosols (Bressi et al., 2013; Crippa et al., 2013a; Dupont et al., 2016; Favez et al., 2012; Fröhlich et al., 2015; Petit et al., 2017a; Petit et al., 2014; Sciare et al., 2011; Waked et al., 2014).

2.2. Sample collection and co-located measurements

PM$_{10}$ samples (Tissu-quartz fibre filter, Pallflex, Ø=150 mm) were collected every 4 h from 6-21, March 2015 using a high-volume sampler (DA-80, Digitel; 30 m$^3$h$^{-1}$). Prior to sampling, quartz fibre filters were pre-heated at 500 °C for 12 h. After collection, samples were wrapped in aluminium foils, sealed in polyethylene bags, and stored at −20 °C until analysis. Shipping of the samples to the different laboratories for analyses have been done by express post using cool boxes (<5°C). A total of 92 samples and 5 field blanks were collected and analysed for an extended chemical characterization following the protocols described in section 2.3.

PM$_{10}$, Black Carbon (BC), NO$_x$ and O$_3$ concentrations were measured using co-located online analysers: TEOM-FDMS (1405F model, Thermo), multi-wavelength aethalometer (AE33 model, Magee Scientific), T200UP and T400 monitors (Teledyne API), respectively. Moreover, assuming that biomass burning and fossil fuel combustion were the two predominant combustion sources, BC from wood burning (BC$_{wb}$) and fossil fuel (BC$_{ff}$) emissions were estimated using the so-called “aethalometer model” (Drinovec et al., 2015; Sandradewi et al., 2008). The AE33 instrument uses
a dual-spot technology which provides the automatic compensation (k) of the aerosol loading effect (Drinovec et al., 2015) over the 7 wavelengths of measurements of BC (from near UV to near IR). An inaccurate automatic compensation has been observed for several days due to high scattering during the pollution episode linked to high ammonium nitrate concentrations. The associated data were thus manually corrected with fixed k values as explained by Petit et al. (2017a) to improve the separation between BC$_{wb}$ and BC$_{ff}$. Finally, meteorological parameters such as temperature, relative humidity (RH), wind direction, and wind speed were obtained from nearby weather station (about 5 km).

2.3. Analytical procedure

A total number of 71 different chemical species have been quantified on filter samples. Major ions (Cl$^-$, NO$_3^-$, SO$_4^{2-}$, NH$_4^+$, Ca$^{2+}$, Na$^+$, Mg$^{2+}$, K$^+$), methanesulfonic acid (MSA) and oxalate (C$_{2}$O$_4^{2-}$) were analysed using ion chromatography (Guinot et al., 2007). EC/OC was measured using a Sunset lab analyser and the EUSAAR-2 thermal protocol (Cavalli et al., 2010). Seven metal elements (namely Ca, Ti, Mn, Fe, Ni, Cu, and Pb) were quantified by PIXE (particle-induced X-ray emission) (Lucarelli et al., 2017; Lucarelli et al., 2011). Sugars, including known biomass burning markers (levoglucosan, mannosan and galactosan) and 3 polyols (arabitol, sorbitol and mannitol), were quantified using LC-PAD (Verlhac et al., 2013; Yttri et al., 2015). Nine PAHs, 14 oxy-PAHs, and 8 nitro-PAHs were quantified by UPLC/UV-Fluorescence and GC/NICI-MS (Albinet et al., 2006; Albinet et al., 2014; Albinet et al., 2013; Tomaz et al., 2016) using QuEChERS-like (Quick Easy Cheap Effective Rugged and Safe) extraction procedure (Albinet et al., 2014; Albinet et al., 2013). Slight modifications have been made in the purification step and the GC column used for oxy- and nitro-PAHs analysis allowing notably the clear separation of 2-
and 3-nitrofluoranthenes. Thirteen SOA markers (notably including α-methylglyceric acid, pinic acid, and methylnitrocatechols) (Nozière et al., 2015) were analysed by GC/EI-MS using authentic standards (Srivastava et al., 2018). Details of the analytical procedures and sample preparation for the analysis of PAHs and their derivatives, and SOA markers are provided in the Supplementary material (SM) (TablesS1toS4).

2.4. Source apportionment methodology

2.4.1. Receptor modelling: PMF

The U.S. Environmental Protection Agency (US-EPA) PMF 5.0 software has been used to perform PM$_{10}$ source apportionment. PMF is based on a weighted least squares fit, where the weights are derived from the analytical uncertainty and provides the optimal solution by minimizing the residuals. Detailed information on this receptor modelling can be found elsewhere (Paatero, 1997; Paatero and Tapper, 1994) and in the SM, and also in the companion paper together with uncertainty calculation details (Srivastava et al., 2018). PM$_{10}$ concentrations were included as the total variable with low weight (weak variable) in the model to determine the source contributions.

2.4.2. Criteria for the selection of species

The choice of the species used as input data for the PMF analysis is a crucial step, which can significantly influence the model results (Lim et al., 2010). Usually, the selection of the species is based on the signal-to-noise ratio (S/N) (Paatero and Hopke, 2003). Here, only species with a S/N ratio above 0.2 were considered (Table S5). In addition, the following set of criteria were used for the final selection of the input species: major PM chemical species, compounds with at least 40%
of total data points above the detection limit, and those being considered as specific markers of a given source (e.g., α-methylglyceric (α-MGA) and 2-methylerythritol acid (2-MT) (SOA markers of isoprene oxidation), 2,3-dihydroxy-4-oxopentanoic acid (DHOPA) (SOA marker of toluene oxidation), methylnitrocatechol isomers (SOA from phenolic compound oxidation mainly emitted by biomass burning)) (Carlton et al., 2009; Iinuma et al., 2010; Kleindienst et al., 2004) (Table S6).

Furthermore, when several specific markers of a given source were available, criteria previously described by Srivastava et al. (2018), such as the selection of only one or two representative species per expected source and the selection of markers mainly present in the particulate phase, have been applied.

Finally, a total number of 34 chemical species (listed in Table S6) were used in the present PM$_{10}$ source apportionment study.

### 2.4.3. Optimization of the final solution

Optimization of the final solution was based on the application of constraints to obtained clear factor chemical profiles. The general framework for applying constraints to PMF solutions has already been discussed elsewhere (Amato and Hopke, 2012; Amato et al., 2009). A priori information is introduced into the model as auxiliary terms in the object function (Q) (see the SM for details) by the implementation of so-called “pulling” (Paatero and Hopke, 2009). “Soft pulling” (species pulled up maximally and/or pulled down maximally)” and “hard pulling” (with defined limits) type of constraints were applied, where species in the factors were selectively pulled down or up. Details related to the constraints applied to each factor profile are given in Table S7.

The change in the Q values, were considered here as a diagnostic parameter to provide insight into the rotation of factors. All model runs were carefully monitored by examining the Q values
obtained in the robust mode. The observed change in the Q-robust was approximately 5% (Norris et al., 2014).

Three criteria, including correlation coefficient (r) between the measured and modelled species, bootstrap, and t-test (two-tailed paired t-test) performed on the base and constraint runs, were used to select the optimal solution, as explained previously (Srivastava et al., 2018).

A threshold of 80% for the bootstrap was considered to indicate that the chosen solution may be appropriate. The species showing poor correlations (r<0.5) between observed and modelled concentrations were carefully examined to determine whether the species should be down-weighted or excluded. Student's t-test was used to evaluate the effectiveness of the applied constraints and to verify if the differences were statistically insignificant for all source profiles (two-tailed paired t-test significance test at p<0.05 probability).

The comparison between the reconstructed and measured input species showed very good agreement except for 2-methylerythritol (r<0.50, n=92, p<0.05) (Table S8). Bootstrapping on the final solution showed stable results with ≥ 85 out of 100 bootstrap mapped factors (Table S9). No significant difference (p values in the range 0.07-0.40) was observed in the source chemical profiles between the base and the constrained runs (Table S10, Figures S2 and S3).

2.5. Back trajectories and geographical origins

A study of the geographical origin of selected identified sources has been performed by concentration-weighted trajectory (CWT). Such approach combines concentration data measured at the receptor site (in this case for each PM source) with back trajectories and helps to localize the air parcel responsible for high observed concentrations. For this purpose, back trajectories were calculated every 3 h using the standalone version of HYSPLIT v4.1 (Draxler, 1999; Stein et al.,
and CWT calculations, and cluster analysis, were performed using the ZeFir Igor package (Petit et al., 2017b). Details about all these calculations have been reported previously (Petit et al., 2017a).

3. Results and discussion

3.1. Overview of the PM\textsubscript{10} chemical composition

The daily PM\textsubscript{10} mass concentrations ranged from 12-130 μg m\textsuperscript{-3}, with an average of 49 μg m\textsuperscript{-3} during the campaign. The PM chemical composition showed a large predominance of secondary inorganic species, especially ammonium nitrate, highlighting the significance of secondary processes throughout the studied period (Petit et al., 2017a). OM concentrations ranged from 2 to 25 μg m\textsuperscript{-3}, with an average value of about 12 μg m\textsuperscript{-3} (Figure 1). Note that the slight differences observed between the measured and the reconstructed PM\textsubscript{10} mass concentrations may be due to the PM water content and/or some sampling artefacts together with the measurement uncertainties (Schwab et al., 2006). The measurement period can be divided into 3 sub-periods according to the air mass origins. The SO\textsubscript{4}/EC ratio was considered as a proxy to distinguish local from regional influences (Petit et al., 2015). This ratio showed a minimum value of about 2 during the period from 03/06 to 03/11, emphasizing the role of local emissions such as residential wood burning at the beginning of the studied period. This was notably supported by the high values of the levoglucosan/PM\textsubscript{10} ratio observed during this period together with low wind speeds (recirculation) (Figure S4). Nitrate concentrations started to slightly increase from 03/11 and then the increase was more pronounced from 03/13 together with the rise of the SO\textsubscript{4}/EC ratio, and associated with higher wind speeds coming from the NE direction, suggesting an influence of medium range transport. A substantial change was noticed in the PM composition during the most intense part of
the campaign (03/18-03/21). The air masses originated mostly from the NNE direction with relatively high wind speeds, and high SO$_4$/EC ratio (of about 8), indicating the advection of aged aerosols over the Paris region and highly impacted by long-range transport. In addition, the high concentration of oxalate (up to 0.5 $\mu$g m$^{-3}$) observed during this 3$^{rd}$ period (Figure S5) suggests the significant role of photochemical processes enhancing the formation of more oxidized products.

3.2. Description of PMF factors

Based on the methodology described before (section 2.4), a 11-factor solution provided the most reasonable PMF result. The use of specific molecular organic markers in the PMF model allowed to deconvolve common aerosol sources such as primary traffic emissions, biomass burning, dust, mixed secondary aerosols, nitrate-rich factor, and sea salt, as well as 2 specific biogenic- and 3 anthropogenic-SOA sources. The sum of all these apportioned PMF factors (reconstructed PM$_{10}$) showed a very good agreement with the measured PM$_{10}$ concentrations ($r=0.97$, $n=92$, $p<0.05$) (Figure 2 and Table S8). Identified aerosol sources, their chemical profiles and temporal evolutions are shown on Figures 3 and 4 and discussed individually hereafter.

3.2.1. Mixed secondary aerosols

This factor was obtained as the predominant one, with an average concentration of 18.2 $\mu$g m$^{-3}$ and accounting for approximately 37% of the PM$_{10}$ mass (Figure 2). It was characterized by high contributions of NO$_3^-$, SO$_4^{2-}$ and NH$_4^+$ (43%, 67% and 56% of the total mass, respectively) and showed a well-marked temporal variation, with very high concentrations during the second half of the campaign after 03/14 (Figure 3). In addition, this factor showed significant contributions of oxalate (36% of the species within this factor), $\alpha$-MGA (25%) and DHOPA (20%), known to be
typical secondary organic species or to be markers of SOA formation from the oxidation of isoprene and toluene (Carlton et al., 2009; Kawamura and Ikushima, 1993; Kleindienst et al., 2012).

Such a high contribution of this secondary factor is expected to be related to the aging of air masses during long range transport. This was also confirmed by the high SO$_4$/EC ratio (=50) observed in the factor chemical profile. As detailed in Petit et al. (2017a), a dense cloud cover was observed over Northern France during this period, with a limited amount of sunshine, and relatively high wind speeds (Figure 1) favouring the convective potential of lower atmospheric layer formation and advection of aged aerosol over the Paris region. This is furthermore supported by the CWT analysis showing low contribution of local/regional sources (Figure S6), which is consistent with the advected pattern of sulfate in Western Europe. This is also seconded by the findings published previously for the same period using on-line instrumentation (aerosol chemical speciation monitor, ACSM) (Petit et al., 2017a).

3.2.2. Nitrate-rich factor

This factor was characterized by the large proportions of NO$_3^-$ and NH$_4^+$ (51 and 40% of the total mass of each compound in this factor) (Figure 3). This source represents nitrate-rich aerosols due to the dominant fraction of ammonium nitrate and accounted for 23% of the total PM$_{10}$ mass (11.2 µg m$^{-3}$) (Figure 2). This factor also included small contributions of species such as oxalate (5%) and 2-NF (15%), known to be secondarily formed from the oxidation of PAH (fluoranthene) (Arey et al., 1986; Atkinson et al., 1987), and also benzo[a]fluorenone (B[a]Fone) (9%), species probably of secondary origin during the studied period (see section 3.2.9).

High contributions of this factor at low wind speeds were observed (Figures 1 and 3), indicating the significant role of rather local formation of ammonium nitrate during this period. This was
further supported by the CWT analysis (Figure S6), and also confirmed by the previous observations at the same site for the same PM pollution event as specified above (Petit et al., 2017a).

### 3.2.3. Primary traffic emissions

Given the constraints applied, this factor showed a significant amount of EC (35% of species in this factor) along with relatively high content of metal elements such as Fe (82%), Cu (53%), and Mn (28%) (Figure 3). These species are typically associated with road transport in (peri-) urban environments. Fe, Cu, and Mn may originate from brake and tire abrasion and/or road dust resuspension (Garg et al., 2000; Hildemann et al., 1991; Pant and Harrison, 2012; Srivastava et al., 2016; Sternbeck et al., 2002), notifying the influence of both, exhaust and non-exhaust traffic emissions in this factor. This source attribution is furthermore supported by significant correlations with BC$_{ff}$ ($r=0.5$, $n=92$, $p<0.05$) and NO$_x$ ($r=0.73$, $n=70$, $p<0.05$) (Figure S7).

This factor accounted for 5% of the PM$_{10}$ mass during the studied period, corresponding to an average concentration of 2.2 µg m$^{-3}$ (Figure 2). This value is in good agreement with the previous observations made in the Paris region (average concentrations of about 2.3 µg m$^{-3}$ in spring season according to Bressi et al. (2014).

Interestingly, the highest fraction of 1-nitropyrene (1-NP) (53%), likely to be a good marker of diesel emissions (Keyte et al., 2016; Schulte et al., 2015; Zielinska et al., 2004a; Zielinska et al., 2004b), was associated with this factor. Irrespective of the constraint applied on 1-NP, it always showed maximum attribution with this factor along with species mentioned above. In addition, 1-NP showed a good correlation with NO$_x$ ($r=0.71$, $n=70$, $p<0.05$) (Figure S7). The traffic vehicle fleet in France was composed of about 61.5% of diesel engines in 2015 and up to 80%, taking into
account all vehicle categories (including heavy trucks) (CCFA, 2016). These results showed that 1-NP can be used as a good primary organic molecular marker to trace traffic (diesel) exhaust emissions. To the best of our knowledge, this is the first report of the use of 1-NP to resolve the sources of primary traffic emissions in PMF model.

3.2.4. Dust

This factor was found as one of the major PM$_{10}$ contributor, with an average concentration of 8.7 µg m$^{-3}$ and accounting for 18% of the PM$_{10}$ mass (Figure 2). It was notably composed of metals and cations, such as Ti, Ca$^{2+}$, Mg$^{2+}$, Cu, Mn and Fe (68%, 75%, 28%, 31%, 39% and 18% of species in this factor respectively), together with a significant amount of OC (20%) (Figure 3). Therefore, this source was identified as dust, including the influence of construction work, soil abrasion and resuspension (Amato et al., 2016a; Amato et al., 2016b; Andersen et al., 2007; Mossetti et al., 2005; Querol et al., 2004; Yin et al., 2005). High concentrations observed during the entire campaign corresponded to the air masses originated from the NE to N directions (Figure 1). Back trajectories analysis showed negligible impact from Sahara during this period (Figure S8).

3.2.5. Sea salt

This factor was characterized by high contributions of Na$^+$ (76%), Mg$^{2+}$ (54%) and Cl$^-$ (58%) (Figure 3). Cl$^-$/Na$^+$ and Mg$^{2+}$/Na$^+$ ratios of 1.0 and 0.1, respectively, were on the same order of magnitude as the standard sea water composition (1.2 and 0.1, respectively) (Tang et al., 1997), and also similar to the values reported for the Paris region in previous study (Cl$^-$/Na$^+=0.96$; Mg$^{2+}$/Na$^+=0.13$) (Bressi et al., 2014). The lower proportion of chloride can be explained by the occurrence of acid–base reactions between sea-salt particles and sulfuric and/or nitric acids, leading
to the volatilization of HCl (Seinfeld and Pandis, 2012) during the transport of air masses from marine regions to the Paris area.

This source accounted for 2% of the total PM$_{10}$ mass (1.8 µg m$^{-3}$) (Figure 2) and showed its highest concentrations during the period from 03/10-03/12 when the origin of air masses changed from recirculation to the NE (Figures 1 and 3). The results from the CWT analysis highlighted the geographical origin of this source from the Atlantic Ocean and to a lesser extent from the North Sea (Figure S6).

### 3.2.6. Biomass burning

This factor was characterized by typical molecular markers from cellulose combustion with significant amounts of levoglucosan (levo, 44%) and mannosan (manno, 38%) (Simoneit, 2002; Simoneit, 1999) (Figure 3). It also included significant contributions of PAHs (B[a]F, B[ghi]P, and In[1,2,3-cd]P) (21-29%) and K$^+$ (18%). Biomass burning concentrations were especially high at the beginning of the PM pollution event (Figure 3). This source accounted for 8% of the PM$_{10}$ mass, with an average concentration of 3.8 µg m$^{-3}$ (Figure 2). This factor also showed a good correlation with BC$_{wb}$ ($r=0.63$, $n=92$, $p<0.05$) (Figure S9) even if the BC$_{wb}$/BC$_{ff}$ separation was probably not optimal during the last period of the campaign due to high loading of inorganic aerosols (Petit et al., 2017a).

This factor also included small contributions of known secondary organic species such as oxalate (17%) and DHOPA (13%), and also of benzo[a]fluorenone (B[a]Fone) (11%), benzo[b]fluorenone (B[b]Fone) (12%) and 9-nitroanthracene (9-NA) (19%), species mainly of secondary origin during this specific event (see sections 3.2.9 and 3.2.10). Thus, this factor may contain a part of oxidized primary organic aerosol (OPOA), arising from the oxidation of organic
compounds between the emission point and their introduction in ambient air (Nalin et al., 2016), but not resolved by the model.

3.2.7. Biogenic SOA-1 (marine)

This factor contained 100% of methanesulfonic acid (MSA), a known secondary oxidation product of dimethylsulfide (DMS), which is emitted by phytoplankton and several types of anaerobe bacteria and released from the ocean into the atmosphere (Carlton et al., 2016; Charlson et al., 1987; Chasteen and Bentley, 2004; Crippa et al., 2013b).

The presence of DMS in seawater has been seen to covary with biological productivity and incident solar radiation at the Earth's surface (Andreae and Raemdonck, 1983; Bates et al., 1987). The CWT analysis showed a geographical origin of this factor from the North Sea (Figure S6). The worldwide distribution of chlorophyll, used as a phytoplankton indicator, estimated using satellite-based measurements, showed a hot spot of phytoplankton bloom near the North Sea in March 2015 and further confirmed the identification and the geographical origin of this SOA source from marine emissions (Figure S10).

The presence of SO$_4^{2-}$ in this factor is also considered to be partly related to the oxidation of DMS, as the MSA/SO$_4^{2-}$ ratio for this factor of 0.14 was similar to values previously reported in the literature for such source (e.g., about 0.08 in Bove et al. (2016)).

This factor accounted for 2% of the PM$_{10}$ mass (0.8 µg m$^{-3}$) (Figure 2) and followed a clear temporal variation with higher contributions during the period from 03/17 to 03/21 (Figures 1 and 4). This was in agreement with long range transport and aging processes as highlighted by the high concentrations of NO$_3^-$ and SO$_4^{2-}$ during the last days of the PM pollution event (Petit et al., 2017a).
3.2.8. **Biogenic SOA-2 (isoprene)**

This factor was characterized by significant contributions of oxidation products of isoprene ($\alpha$-MGA and 2-MT; 65 and 66%, respectively) (Carlton et al., 2009) (Figure 4). As expected for the late winter and early spring seasons, biogenic SOA factor showed a very low contribution to the PM$_{10}$ mass (<1%) corresponding to a concentration of 0.4 $\mu$g m$^{-3}$ (Figure 2).

3.2.9. **Anthropogenic SOA-1 (oxy-PAHs)**

The sources of anthropogenic SOA factor were resolved using oxy-PAHs (dibenzo[b,d]pyran-6-one (6H-DPone) (84%), benzo[a]fluorenone (53%) and benzo[b]fluorenone (60%) (Figure 4). Dibenzo[b,d]pyran-6-one, a product of phenanthrene photooxidation has been described as a good marker of PAH SOA formation (Lee and Lane, 2010; Perraudin et al., 2007; Tomaz et al., 2017) while benzo[a]fluorenone and benzo[b]fluorenone may originate from both primary and secondary processes (Albinet et al., 2007; Tomaz et al., 2017). Here, both compounds showed significant correlations ($r=0.55$-$0.57$, $n=92$, $p<0.05$) with dibenzo[b,d]pyran-6-one, indicating the probable secondary origin of these compounds during the PM pollution event (Figure S11). This factor also showed a relatively high amount (16% of species in this factor) of DHOPA, which is known as a SOA marker from toluene oxidation (Kleindienst et al., 2012). Finally, significant contributions of primary species such as PAHs (B[a]F, B[a]P, B[ghi]P, and In[1,2,3-cd]P) (23-41%), levoglucosan (19%), mannosan (22%) and 1-NP (17%) were also noticed. Thus, this factor seemed characteristic of PAH SOA from anthropogenic sources including combustion processes such as biomass burning and traffic. A similar factor was already resolved using oxy-PAHs in the companion paper (Srivastava et al., 2018).
Overall, anthropogenic SOA source (oxy-PAHs) accounted for approximately 2% of the total PM$_{10}$ mass with a concentration of 1.1 µg m$^{-3}$ on an average (Figure 2). High concentrations (up to 8 µg m$^{-3}$) were especially observed during the 03/13-03/18 period, when air masses originated from the NE direction (Figures 1 and 4).

3.2.10. **Anthropogenic SOA-2 (nitro-PAHs)**

This source was characterized by the high fraction of 2-nitrofluoranthene (2-NFlt) (78%) (Figure 4). 2-NFlt is only secondarily formed from the gas phase reaction between fluoranthene and NO$_2$ initiated by OH (day-time)/NO$_3$(night-time) radicals (Arey et al., 1986; Atkinson et al., 1987). Therefore, this factor was identified as another PAH SOA source but with different chemical processes involved (see section 3.3.2). In addition, this factor also included a significant contribution of 9-NA (41%) which may originated from both, primary and secondary sources (Albinet et al., 2007; Ringuet et al., 2012b), as both processes were dominant during the sampling campaign.

Overall, this source had a very low contribution to the PM$_{10}$ mass (0.6%), corresponding to a concentration of 0.3 µg m$^{-3}$ (Figure 2). The discernible temporal pattern with higher concentrations during the night indicated the prominent role of night-time chemistry in the formation of nitro-PAHs during this period (Figure 4) as previously observed at the same site (Ringuet et al., 2012a).

3.2.11. **Anthropogenic SOA-3 (phenolic oxidation)**

This factor was resolved using 4-methyl-5-nitrocatechol (4-Me5Nc) and 3-methyl-5-nitrocatechol (3-Me5Nc) with very high contributions of both compounds (94-100%) (Figure 4). These species are typical by-products of SOA formation from the photooxidation of cresols mainly
originating from biomass burning emissions (Bruns et al., 2016; Iinuma et al., 2010). In addition, this factor also showed significant contributions of levoglucosan (19%), mannosan (15%), and DHOPA (13%), mainly illustrating the characteristic of biomass burning SOA. To the best of our knowledge, this study is probably the first one reporting the use of methylnitrocatechol isomers for the apportionment, using PMF, of anthropogenic SOA linked to biomass burning.

The source showed a quite low contribution to the PM$_{10}$ mass of 2% on an average corresponding to a concentration of 0.8 µg m$^{-3}$ (Figure 2). However, high concentrations and contributions were observed during the recirculation period (up to 1.4 µg m$^{-3}$ corresponding to 5% of PM) in agreement with the dominant period of biomass burning emissions.

### 3.3. Focus on organic aerosols (OA)

#### 3.3.1. OA sources and formation processes

Assuming a OC-to-OA conversion factor around 1.8 (Sciare et al., 2011), OA represented about 25% of the total PM$_{10}$ mass during the studied period. Major contributors were dust (representing 20% of total OC on average for the campaign), biomass burning (19%), and mixed secondary aerosols (17%), followed by primary traffic emissions (14%) and nitrate-rich factor (11%). Only sea salt did not show any contribution to OC mass (Figure S12). Primary OC (POC) was estimated as the sum of OC fractions associated with dust, biomass burning, and traffic, whereas OC present in the other factors (namely biogenic SOA-1 & 2, anthropogenic SOA-1, 2 & 3, mixed secondary aerosols and nitrate-rich factor) was considered as representing secondary OC (SOC). These calculations led to almost equivalent contributions of POC and SOC to the total OC (53% and 47% on average for the campaign, respectively).
As shown in Figure 5, the use of specific organic molecular markers allowed the clear identification of 41% of the total SOC composed of anthropogenic SOA (oxy-PAHs, nitro-PAHs, phenolic oxidation, corresponding to average OC contributions of 15%, 9%, 11%, respectively) and biogenic SOA (marine + isoprene, 6% in total). The low contributions of these last two sources may be related to the low biogenic activity during the sampling period (end of winter, beginning of spring). Anthropogenic SOAs were thus considered as the dominant SOC contributors identified and in particular PAH SOA, accounting for 24% of the total SOC. This results highlights the significant contribution of anthropogenic SOA in urban influenced environments, in good agreement with the literature (Shakya and Griffin, 2010; Srivastava et al., 2018; Zhang and Ying, 2012).

Secondary processes leading to the presence of OA within the nitrate-rich factor and mixed secondary aerosols remained unknown. However, previous studies already suggested that OC fractions associated to nitrate- and/or sulfate-rich factors can be considered as secondary (Ke et al., 2008; Lee et al., 2008; Pekney et al., 2006). This was confirmed here by the presence of oxalate, α-MGA, 2-NFt and DHOPA in both factors. Formation processes leading to OA directly associated with secondary inorganic aerosols may notably include condensation of semi-volatile organic compounds onto the surface of pre-existing particles (Amato et al., 2009; Favez et al., 2007; Surratt et al., 2007), and nucleation mechanisms from various gaseous precursors such as mono- and polyaromatic compounds, alkanes, alkenes…(Lim and Ziemann, 2005; Tkacik et al., 2012; Zhao et al., 2014), as well as the formation of organosulfates and/or organonitrates (Riva et al., 2015; Surratt et al., 2006). Further works are still needed to identify and quantify relevant and specific molecular markers that could help to evaluate the significance of these processes using source-receptor models such as PMF.
3.3.2. **Diurnal profiles**

Figure 6 shows the diurnal profiles of significant and relevant OA factors obtained within the present study. The biomass burning factor showed a pronounced peak at night-time, which is consistent with human activities. As expected too, traffic exhibited two pronounced peaks corresponding to morning and evening commuting periods.

It is known that oxy-PAH secondary formation involves photochemical reactions of parent PAHs with ozone or OH radicals (Vione et al., 2004; Walgraeve et al., 2010). However, no specific trend was observed for anthropogenic SOA-1 during the studied period.

Both other anthropogenic SOA factors (nitro-PAHs and phenolic oxidation) exhibited a clear diurnal pattern with high concentrations at night with peaks at distinct times suggesting a significant role of night-time chemistry but with different chemical processes involved. For anthropogenic SOA-2 (nitro-PAHs), an increase of the concentrations from early night until early morning was observed. These diurnal profiles agreed with the ones usually reported for nitrate radical (Reisen and Arey, 2005), indicating the predominance of such processes in the secondary formation of nitro-PAHs. The peaks of concentrations for anthropogenic SOA-3 (phenolic oxidation) were observed in early night in coherence with the biomass burning emissions as the main source of phenolic SOA precursor compounds.

For both anthropogenic SOA-2 and-3, high RH conditions observed during the night (Figure S13) could also favour secondary formation processes (Jia and Xu, 2014). However, the aerosol liquid water content, together with the higher solubility of the phenolic precursor compounds as well as methylnitrocatechols, could explain the low concentrations of anthropogenic SOA-3 observed during the second part of the night until the morning. Nitro-PAHs are less water soluble
than methyl nitrocatechols and consequently, they may tend to partition into the solid phase instead of the aqueous one, explaining the differences observed in the diurnal profiles between both factors. Finally, the degradation of both category of compounds during the day-time under solar irradiation cannot be ruled out.

4. Conclusion

The use of specific primary and secondary molecular markers, with high time resolution data, in PMF model allowed to resolve 11 major PM$_{10}$ sources during a PM pollution event observed in March 2015 in the Paris region. Besides common factors (biomass burning, traffic, dust, sea salt, secondary inorganic aerosols), 2 specific biogenic SOA (marine + isoprene) and 3 anthropogenic SOA (nitro-PAHs + oxy-PAHs + phenolic oxidation) factors have been identified. Results obtained showed that 1-NP can be used as a good primary organic molecular marker to trace traffic emissions and notably diesel exhaust ones.

A special attention was put on SOA, roughly accounting for half of the total OC loading. Molecular markers used in the present study allowed elucidating the probable precursor’s origins of about 41% of the total SOC fraction. However, future studies are still needed to focus on the identification, and then the incorporation into PMF analysis, of molecular markers from other known SOA precursors (alkanes, alkenes, mono- and polyaromatic compounds) and of class of compounds which significantly contribute to the SOC fraction (i.e., organosulfates and organonitrates) to further discriminate the nature of OA associated with secondary inorganic sources.

The results obtained in this study highlighted that 35% of the total SOC originated from anthropogenic sources. In particular, PAH SOA (oxy-PAHs + nitro-PAHs) accounted for 24% of
the total SOC, highlighting the significance of such SOA precursors in urban influenced environments.

Finally, high time-resolution filter dataset available for the present study allowed a better understanding of the chemical processes involved, notably based on the investigation of diurnal variations. Anthropogenic SOA related to nitro-PAHs and phenolic compounds exhibited a clear temporal pattern with high contributions during the night but with different chemical processes involved. The effects of relative humidity and liquid aerosol content in the formation of these secondary compounds are still not fully understood and should be investigated.

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**Figures**

![Graphical representation of PM$_{10}$ chemical composition, SO$_4$/EC ratio and wind speed observed at Paris-SIRTA, France (March 2015). Background colours refer to air mass clusters. Cluster analysis performed using ZeFir based on back trajectories calculated every 3 h with HYSPLIT model.]

**Figure 1.** Temporal variation of PM$_{10}$ chemical composition, SO$_4$/EC ratio and wind speed observed at Paris-SIRTA, France (March 2015). Background colours refer to air mass clusters. Cluster analysis performed using ZeFir based on back trajectories calculated every 3 h with HYSPLIT model.
Figure 2. Average contributions (left) and temporal evolution (right) of the identified sources to PM$_{10}$ mass concentrations at Paris-SIRTA, France (March 2015).
Figure 3. Source profiles and temporal evolution of mixed secondary aerosols, nitrate-rich, primary traffic emissions, dust, sea salt and biomass burning factors identified at Paris-SIRTA, France (March 2015). Coloured bars and red dots represent the concentrations and the percentages of each species apportioned in the factor, respectively.
Figure 4. Source profiles and temporal evolution of biogenic SOA-1, biogenic SOA-2, anthropogenic SOA-1, anthropogenic SOA-2 and anthropogenic SOA-3 identified at Paris-SIRTA, France (March 2015). Coloured bars and red dots represent the concentrations and the percentages of each species apportioned in the factor, respectively.
Figure 5. Average contributions of the identified sources to the total OC (POC + SOC) (left) and SOC (right) mass concentrations at Paris-SIRTA, France (March 2015).
Figure 6. Diurnal profiles of PMF resolved anthropogenic SOA factors (1, 2, and 3), biomass burning and traffic identified at Paris-SIRTA, France (March 2015).