Seasonal variability and source apportionment of volatile organic compounds (VOCs) in the Paris megacity (France)

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Abstract. Within the framework of air quality studies at the megacity scale, highly time-resolved volatile organic compounds (C2 - C8) measurements were performed in downtown Paris (urban background sites) from January to November 2010. This unique dataset included non-methane hydrocarbons (NMHCs) and aromatic/oxygenated species (OVOCs) measured by a GC-FID (Gas Chromatograph with a Flame Ionization Detector) and a PTR-MS (Proton Transfer Reaction - Mass Spectrometer), respectively. This study presents the seasonal variability of atmospheric VOCs being monitored in the French megacity and their various associated emission sources. Clear seasonal and diurnal patterns differed from one VOC to another as the result of their different origins and the influence of environmental parameters (solar radiation, temperature). Source Apportionment (SA) was comprehensively conducted using a multivariate mathematical receptor modeling. The United States Environmental Protection Agency’s Positive Matrix Factorization tool (US EPA, PMF) was used to apportion and quantify ambient VOC concentrations into six different sources. The modeled source profiles were identified from near-field observations (measurements from three distinct emission sources: inside a highway tunnel, at a fireplace and from a domestic gas flue, with hence a specific focus on road-traffic, wood burning activities and natural gas emissions) and hydrocarbon profiles reported in the literature. The reconstructed VOC sources were cross-validated using independent tracers such as inorganic gases (NO, NO2, CO), black carbon (BC) and meteorological data (temperature). The largest contributors to the predicted VOC concentrations were traffic-related activities (including motor vehicle exhaust, 15 % of the total mass on the annual average, and evaporative sources, 10 %), with the remaining emissions from natural gas and background (23 %), solvents use (20 %), wood burning (18 %) and a biogenic source (15 %). An important finding of this work is the significant contribution from wood burning, especially in winter, where it could represent up to ~ 50 % of the total mass of VOCs. Biogenic emissions also surprisingly contributed up to ~ 30 % in summer (due to the dominating weight of OVOCs in this source). Finally, the mixed natural gas and background source exhibited a high contribution in spring (35 %, when continental air influences were observed) and in autumn (23 %, for
home heating consumption).

Keywords : Urban atmospheric pollution, Volatile organic compounds (VOCs), Ozone precursors, Vehicular emissions, Positive Matrix Factorization (PMF), Source apportionment.

5 1 Introduction

More than half of the world’s population is now living in urban areas and about 70% will be city-dwellers by 2050 (United Nations, 2014). Many of these urban centers are ever-expanding, leading to the gradual growth of megacities. Strong demographic and economic pressures are exerting increasing stress on the natural environment, with impacts at local, regional and global scales. Megacities are hotspots of atmospheric gaseous and particulate pollutants, which are subjects of concern for sanitary, scientific, economic, societal and political reasons. The adverse health effects of outdoor air pollutants are recognized today. Indeed, ambient air pollution has been classified as carcinogenic to humans by the International Agency for Research on Cancer since October 2013 (IARC, 2013). In recent decades, air pollution has become one of the most widespread problems in many megacities and should be more investigated.

The understanding of the pollutants in urban areas remains complex given the diversity of their emission sources (unequally distributed in space and time) as well as their formation and transformation processes. Volatile organic compounds (VOCs) are of a great scientific interest because they play an important role in atmospheric chemistry. In the troposphere, primary VOCs take part in chemical and/or photochemical reactions, thus contributing to the formation of ground-level ozone ($O_3$) (Logan et al., 1981; Liu et al., 1987; Chameides et al., 1992; Carter et al., 1994) and secondary organic aerosols (SOA) (Tsigeridis and Kanakidou, 2003 and references therein; Ng et al., 2007). While some megacities face very poor air quality (such as Beijing, Gurjar (2014)) with pollutant concentrations way above recommended thresholds, European megacities experience stagnant pollution levels at the annual scale. However, pollution episodes related to high $O_3$ and PM concentrations still regularly occur, leading to detrimental health consequences.

Epidemiological studies revealed that outdoor air pollution, mostly from PM$_{2.5}$ and $O_3$, could lead to 17800 premature deaths in France (with 3100 for the Paris megacity) in 2010 and projections for the future are even worse (3800 in 2025 and 4600 in 2050 for Paris) (Lelievre et al., 2015). Paris and its surroundings (also called the Île-de-France region) constitute the second largest European megacity with about 12 million inhabitants, representing 20% of the French national population distributed over only 2% of its territory (Eurostat, 2014). Although this region is surrounded by a rural belt, it is considered as a large central urban area where a strong pollution signal can be detected. Deguillaume et al. (2008) have shown that the urban area of Paris was frequently associated with a “VOC-sensitive” chemical regime (also called “NO$_x$-saturated” regime according to Sillman, 1999), for which VOC anthropogenic emission reductions are more effective to decrease ozone levels than NO$_x$ anthropogenic emission reductions. Obtaining accurate knowledge on VOC emissions and sources is consequently essential for $O_3$ and SOA abatement measures.
Qualitative and quantitative assessments of VOC variability and sources have already been conducted within the Paris area during May-June 2007 (Gros et al., 2011; Gaimoz et al., 2011). This study concluded that road-traffic activities (traffic exhaust and fuel evaporation) influenced the total VOC fingerprint, with a contribution of ~39%. This finding was considered as being in disagreement with the local emission inventory provided by the air quality monitoring network AIRPARIF (http://www.airparif.asso.fr), for which the main contribution was related to solvents usage (from industries and from residential sectors). However, this work was performed over a short period of time (only few weeks). Although it provided valuable information about ambient VOC emissions and sources during a specific period (spring), it did not show their seasonal variations over longer time scales. More resolved observations are therefore required to check the representativity of these first conclusions.

In this context, the EU-F7 MEGAPOLI (Megacities: Emissions, urban, regional and Global Atmospheric POLLution and climate effects, and Integrated tools for assessment and mitigation) (Butler, 2008) and the French PRIMEQUAL-FRANCIPOL research programs involving several (inter)-national partners in the atmospheric chemistry community have been implemented. These MEGAPOLI - FRANCIPOL projects partly consisted in documenting a large number of gaseous and particulate compounds and determining their concentration levels, variabilities, emission sources and geographical origins (local or imported) within the Paris urban area. These experiments go therefore beyond the scope of this paper and a full description of scientific studies conducted under the programme can be found in the special issue “MEGAPOLI - Paris 2009/2010 campaign” available in the Atmospheric Chemistry and Physics (ACP) journal (e.g. Crippa et al., 2013; Skyllakou et al., 2014; Ait-Helal et al., 2014; Beekmann et al., 2015 and references therein).

Here, this work presents near real-time measurements of VOCs performed at urban background sites in downtown Paris from 15 January to 22 November 2010. Its objectives are to (1) assess ambient levels of a VOC selection, (2) describe their temporal (seasonal and diurnal) variabilities, (3) identify their main emission sources from statistical modeling and (4) quantify and discuss their source contributions on yearly and seasonal bases.

In order to identify and apportion ambient VOC levels to their emission sources, the advanced multivariate receptor modeling technique Positive Matrix Factorization (PMF) was applied. As no prior knowledge of the number or the chemical nature of source profiles is explicitly required (Paatero and Tapper, 1994), the identification of PMF source profile outputs must be made a posteriori. It usually relies on speciation profiles available in the literature. Within this study, near-field additional measurements (at source points: inside a highway tunnel, at a fireplace and from a domestic gas flue) were performed to help strengthen this identification of VOC profiles derived from PMF simulations. This experimental approach is dedicated to provide a specific fingerprint of VOC sources related to road-traffic, residential wood burning activities and domestic natural gas consumption, respectively. The originality of this work stands in using these near-field speciation profiles to refine the identification of apportioned sources.

First, Section 2 will describe (i) sampling sites, (ii) analytical techniques conducted and (iii) two combined approaches for identifying and characterizing the main VOC emission sources. Then, Section 3 will investigate VOC levels and their seasonal and diurnal patterns from long-term ambient air measurements. An accurate identification of PMF factors to real physical sources will be proposed in the Sub-section 3.4. Finally, yearly and seasonal contributions of each modeled source will be
discussed in the Sub-section 3.5 and compared to previous studies performed in Paris and widely in Europe in Sub-sections 3.6 and 3.7.

2 Material & Methods

2.1 Sampling sites description

As part of the European EU-F7 MEGAPOLI (Megacities: Emissions, urban, regional and Global Atmospheric POLLution and climate effects, and Integrated tools for assessment and mitigation (http://www.megapoli.info, 2007-2011)) program, a winter campaign involved measurements of a large amount of atmospheric compounds – with techniques including GC-FID and PTR-MS for VOCs – from 15 January to 16 February 2010 at the “Laboratoire d’Hygiène de la Ville de Paris” (LHVP) (Baklanov et al., 2010; Beekmann et al., 2015). Located in the southern part of Paris centre (13th district – 48°82’ N, 02°35’ E – 15 m above ground level, a.g.l.), LHVP dominates a large public garden (called “Parc de Choisy”) at approximately 400 m from Place d’Italie (grouping a shopping centre and main boulevards).

A second measurement campaign involving less instrumentation (only PTR-MS for VOCs) was conducted at the LHVP site from 24 March to 22 November 2010 (as part of the French PRIMEQUAL-FRANCIPOL program: Impact of long-range transport on particles and their gaseous precursors in Paris and its region (http://www.primequal.fr, 2010-2013)). At the same time, hydrocarbon measurements by GC-FID were carried out by the regional air quality monitoring network AIRPARIF at the “Les Halles” subway station (48°51’ N, 02°20’ E – 2.7 m a.g.l.) located 2 km away from LHVP. The location of these two sampling sites is presented in Figure 1.

2.2 Representativeness of sampling sites

Due to the low intensity of the surrounding activities, the LHVP and Les Halles sampling sites were considered as urban background stations by AIRPARIF and by previous scientific studies (Favez et al., 2007; Sciare et al., 2010; Gros et al., 2011). In accordance with the 2008/50/EC European Directive, this station typology is based on two main criteria: (1) the population density is at least 4 000 inhabitants per square kilometre within a 1-km radius of the station and (2) no major traffic road is located within 300 metres.

This characterization of site typologies can be confirmed by studying the nitrogen monoxide (NO): nitrogen dioxide (NO₂) ratio. NO is known to be a vehicle pollution indicator whereas NO₂ has an important secondary fraction. To consider a station as an urban background site, the ratio R between annual average NO and NO₂ concentrations (NO/NO₂) should be less than 1.5 ppb:ppb, as indicated in a report (Mathé, 2010) written at the national level for regulatory purposes.

We observed a very close NO: NO₂ ratio (expressed as ppb: ppb) between both sites in 2010: 0.40 (LHVP) versus 0.38 (Les Halles). The same conclusion can be made for the years 2008 and 2009 (0.45 ± 0.01 for LHVP and 0.48 ± 0.01 ppb: ppb for Les Halles). With NO: NO₂ ratios very similar and less than 1.5, this confirms that these two locations have the same site
typology and can be considered as urban background stations.

To prove that it was valid to merge these two specific datasets from different locations, we opted for a comparison of PMF results derived from MEGAPOLI-FRANCIPOL (as performed in the present paper) and FRANCIPOL data files, respectively. PMF modeling simulations were performed using only the FRANCIPOL dataset. A good agreement was found for the majority of the emission sources. The graphs showing the comparison of PMF results between MEGAPOLI-FRANCIPOL and FRANCIPOL are reported in the Supplementary Section S1.

2.3 Experimental set-up

2.3.1 VOC measurements using a Proton Transfer Reaction-Mass Spectrometer (PTR-MS)

Within the MEGAPOLI and FRANCIPOL projects, online high-sensitivity Proton Transfer Reaction-Mass Spectrometers PTR-MS (Ionicon Analytik GmbH, Innsbruck, Austria) were used for real-time (O)VOC measurements. As this instrument has widely been described in recent reviews (Blake et al., 2009 and references therein), only a description of analytical conditions relating to ambient air observations is given here.

During these two intensive field experiments, a PTR-MS was installed in a small room located on the roof of the LHVP site (15 m a.g.l.). For the MEGAPOLI winter campaign, PTR-MS measurements performed by the Laboratoire de Chimie Physique (LCP, Marseille, France) have already been described in Dolgorouky et al. (2012). As those performed by the Laboratoire des Sciences du Climat et de l’Environnement (LSCE) during FRANCIPOL have not yet been described elsewhere, more technical details are presented below.

Air samples were drawn up through a Teflon line (0.125-cm inner diameter) fitting into a “DEKABON” tube in order to protect it from light. A Teflon particles filter (0.45-µm pore diameter) was settled at the inlet to avoid aerosols and other fragments from entering the system. The PTR-MS was operating at standard conditions: a drift tube held at 2.2 mbar pressure, 60°C temperature with a drift field of 600 V voltage to maintain an $E/N$ ratio of ~ 130 Townsend (Td) [$E$: electrical field strength [V cm$^{-1}$]; $N$: buffer gas number density [molecule cm$^{-3}$]; 1 Td = $10^{-17}$ V cm$^2$]. First water cluster ions H$_3$O$^+$H$_2$O (at m/z 37.0) and H$_5$O$^+$H$_2$O H$_2$O (m/z 55.0) were also measured as well as NO$^+$ and O$_2^+$ masses to indicate any leak into the system and assess the PTR-MS performances.

(O)VOC measurements performed in a full-scan mode were enabled to browse a large range of masses (m/z 30.0 – m/z 150.0). Eight protonated target masses were considered here: methanol (m/z = 33.0), acetonitrile (m/z = 42.0), acetaldehyde (m/z = 45.0), acetone (m/z = 59.0), methylvinylketone (MVK) + methacrolein (MACR) + isoprene hydroxy hydroperoxides (ISOPOOHs) (m/z = 71.0), benzene (m/z = 79.0), toluene (m/z = 93.0) and xylenes (m+p-, o-) + C$_8$ aromatics (m/z = 107.0).

With a dwell time of five seconds per mass, a mass spectrum was obtained every two to ten minutes for MEGAPOLI and FRANCIPOL campaigns, respectively. Around 80% of PTR-MS data were validated. Missing data were partly due to background measurement periods and calibrations. The PTR-MS background for each mass was monitored by sampling zero air through a
catalytic converter heated to 250°C to remove chemical species. Daily background values were averaged and subtracted from ambient air measurements.

In order to regularly ensure the analytical stability of the instrument, injections from a standard containing benzene (5.7 ppbv [parts per billion by volume] ± 10 %) and toluene (4.1 ppbv ± 10 %) were performed approximately once a month from March to November 2010. These measurements have shown that the analyzer stability remained stable during the year with variations within ± 10 %. In addition, two full calibrations were performed before and during the intensive field campaign with a Gas Calibration Unit (GCU, Ionicon Analytik GmbH, Innsbruck, Austria). The standard gas mixture provided by Ionicon contained 17 VOCs at 1 ppmv [parts per million by volume]. These calibration procedures consisted of injecting defined concentrations (in the range from 0 to 10 ppbv) of different chemicals (previously diluted with synthetic air) with a relative humidity at 50 %.

Gas calibrations allowed to determine the repeatability of measurements, expressed here as a mean coefficient of variation. This coefficient was less than 5 % for most of the masses. Slightly higher coefficients were observed for m/z 69.0 (isoprene) and m/z 71.0 (crotonaldehyde) with 5.6 % and 5.2 %, respectively. Observed differences between both full calibration procedures were from 1.1 % for methanol to 9.8 % for toluene, thus illustrating good analyzer stability over time.

Detection limits (LoD) were calculated as three times the standard deviation of the normalized background counts when measuring from the catalytically converted zero air. For the MEGAPOLI winter campaign, LoD ranged from 0.020 to 0.317 µg m⁻³ (0.007 – 0.238 ppb) whereas they were estimated between 0.020 to 0.330 µg m⁻³ (0.007 – 0.248 ppb) during the FRANCIPOL intensive campaign. The analytical uncertainty on all data was estimated by taking into account errors on standard gas, calibrations, blanks, reproducibility/repeatability, linearity and relative humidity parameters. The measurement uncertainty was estimated at ± 20 % in agreement with previous studies (Gros et al., 2011; Dolgorouky et al., 2012).

2.3.2 NMHC on-line measurements by Gas Chromatography (GC)

Two different automated Gas Chromatographs equipped with a Flame Ionization Detector (GC-FID) were used in order to continuously measure light (C₂ - C₆) VOCs in ambient air. The AirmoVOC C₂ - C₆ analyzer (Chromatotec, Saint-Antoine, France), provided by LSCE, was installed near the PTR-MS at the LHVP site from January to February 2010 (e.g. MEGAPOLI period). An in-depth description of the analyzer, sampling set up and technical information (sampling flows, preconcentration, desorption-heating times, types of traps and columns . . . ) can be found in Gros et al. (2011). For each half-hour analysis, more than 20 VOCs were monitored. A certified standard gas mixture (NPL, National Physical Laboratory, Teddington, Middlesex, UK) containing in average 4.00 ppbv of major C₂ - C₉ NMHCs was used for calibration procedures. The injections of this standard allowed checking compound retention times, testing the repeatability of atmospheric measurements and calculating average response factors to calibrate all the measured ambient hydrocarbons. Detection limits were in the range of 0.013 (n-hexane) – 0.060 µg m⁻³ (iso-/n-pentanes) (0.004 – 0.020 ppb). The analytical uncertainties on LSCE measurements were estimated from laboratory tests (i.e. memory effects, repeatability, accuracy of the gas standard) to be ± 15 % (Gros et al., 2011).

From March to November 2010 (e.g. FRANCIPOL period), a GC-FID coupled to a thermo-desorption unit was in operation at the “Les Halles” subway station monitored by the regional air quality network AIRPARIF. Air samples were drawn up at
2.7 m a.g.l. 29 hydrocarbons from C\textsubscript{2} - C\textsubscript{9} were measured during this experiment. A full calibration was performed once a month with a standard gas mixture containing only propane during 6 hours. As the FID response is proportional to the Effective Carbon Number (ECN) in the molecule, calibration coefficients were calculated for each compound and regularly checked so that they drifted no more than ± 5% (tolerance threshold). In addition, a zeroing was carried out every 6 months using a zero air bottle in order to detect any instability or problem with the GC-FID system. LoD were assessed at 0.024 µg m\textsuperscript{-3} for all the selected compounds, except for n-hexane (0.013 µg m\textsuperscript{-3}) (0.022 – 0.004 ppb). A comprehensive calculation of uncertainties could not be performed due a lack of sufficient information from AIRPARIF. Nevertheless, from previous experimental tests, NMHC measurements were provided with an accuracy of ± 15%, which happened to be in agreement with the uncertainty value for the LSCE GC-FID.

### 2.3.3 Additional data available

Some ancillary pollutants and parameters were also measured and used as independent tracers with the aim of strengthening the identification of VOC emission sources derived from the receptor modeling.

Black carbon (BC) was measured using a 7-wavelength (370, 470, 520, 590, 660, 880 and 950 nm) AE31 Aethalometer (Magee Scientific Corporation, Berkeley, CA, USA) with a time resolution of five minutes. BC data were acquired by this instrument from 15 January to 10 September 2010. Raw data were corrected using the algorithm described in Weingartner et al. (2003) and Sciare et al. (2011). BC concentrations issued from fossil fuel and wood burning (BC\textsubscript{ff} and BC\textsubscript{wb}, respectively) were assessed in accordance with their own absorption coefficients using the “Aethalometer” model described by Sandradewi et al. (2008).

Carbon monoxide (CO) measurements were performed using an analyzer based on InfraRed absorption (42i-TL instrument, Thermo Fisher Scientific, Franklin, MA, USA) with a time resolution of five minutes. Nitrogen monoxide and dioxide (NO, NO\textsubscript{2}) were measured by chemiluminescence using an AC31M analyzer (Environment SA, Poissy, France) and ozone (O\textsubscript{3}) was monitored with an automatic UltraViolet absorption analyzer (41M, Environment SA, Poissy, France). NO, NO\textsubscript{2} and O\textsubscript{3} measurements were provided with a 1-min time resolution by the local air quality network AIRPARIF. In addition, Gas Chromatography-Mass Spectrometry (GC-MS) measurements were performed to measure C\textsubscript{3} - C\textsubscript{7} OVOCs, including aldehydes, ketones, alcohols, ethers and esters during the MEGAPOLI winter campaign. This instrument has been described in details by Roukos et al. (2009). As the measurement frequency was different for each analyzer, a common average time was defined to get all data sets on a similar time step of one hour.

Standard meteorological parameters (such as temperature, relative humidity as well as wind speed and direction) were provided by the French national meteorological service “Météo-France” from continuous measurements recorded at the Paris-Montsouris monitoring station (14th district – 48°49' N, 02°20' E), located about 2 km away from the LHVP site.

In order to determine the air masses origin, 5-day back trajectories were calculated every 3 hours from the PC based version of the Hybrid Single Particle Lagrangian Integrated Trajectory (Hysplit) model (Stein et al., 2015) with Global Data Assimilation System (GDAS) meteorological field data. Back trajectories were set to end at Paris coordinates (49°02’ N, 02°53’ E) at 500 m a.g.l.
2.4 Two combined approaches for characterizing VOC emission sources

2.4.1 Bilinear receptor modeling: Positive Matrix Factorization (PMF) tool

Developed about 20 years ago, Positive Matrix Factorization (PMF) is an advanced multivariate factor analysis tool widely used to identify and quantify the main sources of atmospheric pollutants. Concerning VOCs, PMF studies have been conducted in urban (e.g. Brown et al., 2007 – LA, USA; Lanz et al., 2008 – Zürich, Switzerland; Morino et al., 2011 – Tokyo, Japan; Yurdakul et al., 2013 – Ankara, Turkey) and rural areas (e.g. Sauvage et al, 2009 – France; Lanz et al., 2009 – Jungfraujoch, Switzerland; Leuchner et al., 2015 – Hohenpeissenberg, Germany). For this current study, the PMF 5.0 software developed by the EPA (Environmental Protection Agency) was used in the robust mode from ambient air VOC measurements from January to November 2010. A more detailed description of this PMF analysis is given in Appendix A.

PMF mathematical theory was extensively described in Paatero and Tapper (1994). Briefly, this statistical method consists in decomposing an initial chemically speciated dataset into factor profiles and contributions. Equation (1) summarizes this principle in its matrix form:

\[ X = GF + E \]

Where \( X \) is the input chemical dataset matrix; \( G \) is the source contribution matrix; \( F \) is the source profiles matrix and \( E \) the so-called residual matrix.

The initial chemical database used for this statistical study contains a selection of 19 hydrocarbon species and masses divided into ten compound families: alkanes (ethane, propane, iso-butane, n-butane, iso-pentane, n-pentane and n-hexane), alkenes (ethylene and propene), alkyne (acetylene), diene (isoprene) aromatics (benzene, toluene, xylenes + C\(_8\) species), alcohol (methanol), nitrile (acetonitrile), aldehyde (acetaldehyde), ketone (acetone) and enones (methylvinylketone, methacrolein and isoprene hydroxy hydroperoxides), which have been measured from 15 January to 22 November 2010 (\( n = 6445 \) with a 1h-time resolution). This combination of hydrocarbon species and masses is similar to that from Gaimoz et al. (2011) except for iso-butene. Each missing data point was substituted with the median concentration of the corresponding species over all the measurements and associated with an uncertainty of four times the species-specific median, as suggested in Norris et al. (2014). The proportion of missing values ranges from 19% (especially for compounds measured by PTR-MS) to 41% (only for isoprene). This high percentage for isoprene can be mainly explained by analytical problems on GC-FID in July. Despite this limitation, it was decided to take into account this compound as isoprene is a key tracer related to biogenic emissions.

The uncertainty matrix was built upon the procedure described by Norris et al. (2014), adapted from Polissar et al. (1998). This matrix requires the Method Detection Limit (MDL, here in µg m\(^{-3}\)) and the analytical uncertainty (\( u \), here in %) for each selected species. MDLs were calculated as 3\( \sigma \) baseline noise and in some cases were homogenized to keep consistency in uncertainty calculations. Species MDLs were ranged from 0.013 to 0.060 µg m\(^{-3}\) (0.004 – 0.020 ppb) for NMHCs measured by
GC-FID and from 0.020 to 0.330 µg m⁻³ (0.007 – 0.248 ppb) for VOCs measured by PTR-MS. Their analytical uncertainties were respectively estimated at 15% and 20% and kept constant over the experiments.

Single species additional uncertainties were also calculated using an equation-based on the Signal-to-Noise (S/N) ratio. As a first approach, Paatero and Hopke (2003) suggested categorizing a species as Bad if the Signal to Noise Ratio (SNR) was less than 0.2, Weak if it was between 0.2 and 2, and Strong if it was greater than 2. Bad variables are excluded from the dataset, Weak variables get their uncertainties tripled while uncertainties of Strong variables stay unchanged. Here, all species exhibited a SNR greater than 3, except for isoprene which had a ratio of 1.7 due to its 41% missing values recorded. To address this lack of isoprene data, several empirical tests (e.g. simulating an averaged seasonal/diurnal cycle of isoprene or increasing the analytical uncertainty of raw data from 15% up to 30%) were conducted within PMF simulations with the aim of better modeling the variability of this compound. As a consequence of these tests, no significant improvement on the quality of modeling isoprene was observed. Finally, isoprene is still categorizing as strong here. In addition to isoprene, acetonitrile exhibited a SNR less than 3 (S/N ratio of 2.7). It was the only VOC to be defined as weak because it may be eventually contaminated with local emissions from laboratory exhausts (although visible spikes of acetonitrile were excluded from the initial dataset). Keeping in mind these limitations for isoprene and acetonitrile, it was decided to include these two compounds into the PMF model because they are considered as relevant tracers for biogenic and wood burning activities, respectively. ΣVOC was defined as Total Variable and automatically categorized as weak to lower its influence in the final PMF results. No optional Extra Modeling Uncertainty was applied here.

All the Q values (Qtrue, Qrobust, Qexpected), scaled residuals, predicted versus observed concentrations interpretation and the physical meaning of factor profiles were investigated to determine the optimum number of factors. Although some mathematical indicators pointed towards a 5-factor solution, a source mixing solvents use activities, natural gas and background emissions was detected. In order to split individually each emission source, the 6-factor solution was then investigated and chosen in terms of interpretability and fitting scores. More technical details are reported in Appendix A, Sub-section A2.

PMF output uncertainties were estimated using two Error Estimation methods starting with DISP (dQ-controlled displacement of factor elements) and finally processing to BS (Classical bootstrap). The DISP analysis results were considered validated: no error could be detected and no drop of Q was observed. As no swap occurred, the 6-factor PMF solution was considered sufficiently robust to be used. Bootstrapping was then carried out, executing 100 iterations, using a random seed, a block size of 874 samples and a minimum Pearson correlation coefficient (R-Value) of 0.6. All the modeled factors were well reproduced through this bootstrap technique over at least 88 ± 2% of runs, hence highlighting their robustness. A low rotational ambiguity of the reconstructed factors was found by testing different degrees of rotations of the solution using the Fpeak parameter (Fpeak = 0.1).
### 2.4.2 Determination of source profiles from near-field observations

As emphasized in Paatero and Tapper (1994), a Positive Matrix Factorization analysis does not require a priori knowledge on the chemical nature of factor profiles. To help strengthen the identification of VOC emission sources derived from this statistical tool, near-field additional measurements (at limited source points) were worthwhile. These in-situ observations were performed close to specific local emission sources in real conditions as far as possible. They aim at providing chemical fingerprints (considered as reference speciation profiles) of three significant VOC sources representative within the Paris area: road traffic (i), residential wood burning (ii) and domestic natural gas consumption (iii). The speciated profiles of these different anthropogenic sources and their representativity are given here. All the technical details of these experiments are reported in Section S2 in the Supplement.

#### Highway tunnel experiment

Road-traffic was considered to be one of the most significant sources of primary hydrocarbons in many megacities (Seoul – Na, 2006; Los Angeles – Brown et al., 2007; Zürich – Lanz et al, 2008), including Paris (Gros et al., 2007, 2011; Gaimoz et al., 2011). The accurate characterization of the vehicle fleet footprint is therefore important. Consequently, near-field VOC measurements (within the “PRIMEQUAL – PREQUALIF” program) were conducted inside a highway tunnel located about 20 km southeast of inner Paris centre in autumn 2012.

This first experiment has the advantage of supplying a realistic assessment of the average chemical composition of vehicular emissions, as these in-situ measurements were performed under on-road real driving conditions. Most of VOCs emitted from road-traffic are representative of local primary emissions (due to their relatively short lifetimes). Photochemical reactions leading to changes in the initial composition of the air and to the formation of secondary products can be considered of minor importance.

VOC levels during traffic jam periods (07-09 h – 17:30-19 h, Local Time) were considered as the most representative values of vehicular emissions. In order to omit any local background, nighttime values (as suggested in Ammoura et al., 2014) were subtracted from the peak VOC concentrations. The mass contribution of 19 selected compounds was calculated and reported in Figure 2. Each compound is expressed in terms of weight out of the weight (w/w) of the Total VOC (TVOC) mass. The two predominant species measured inside the highway tunnel were toluene (19.4 %, 27.1 µg m\(^{-3}\) (7.1 ppb) on average) and \textit{iso}-pentane (18.6 %; 26.0 µg m\(^{-3}\) (8.7 ppb)). The next most abundant VOCs were aromatics (benzene, \(C_8\)) and oxygenated compounds (acetaldehyde and methanol – 5.0 - 10 %; 5-9 µg m\(^{-3}\) (1.1 - 4.6 ppb)). In addition, significant contributions of light alkenes (ethylene, propene, acetylene; 3.3 % – 4.5 %; 4.6 – 6.2 µg m\(^{-3}\) (4.2 – 5.4 ppb) and alkanes (such as butanes, \(n\)-pentane, \(n\)-hexane, methyl alkanes) were also noticed. These observations were found to be consistent with the literature (Na, 2006; Araizaga et al., 2013 concerning NMHCs and Legreid et al., 2007 for OVOCs measurements) and more importantly with the study from Touaty and Bonsang (2000), for which \textit{iso}-pentane, ethene, acetylene, propene and \(n\)-butane were considered as the major aliphatic compounds observed in the same highway tunnel in August 1996 (Aromatics and OVOCs were not
Fireplace experiment

Residential wood burning activities have been shown to be a significant source of (O)VOCs to local indoor and outdoor air pollution during winter months in urban areas (Evtyugina et al., 2014). Currently, only a few studies about the characterization of VOCs from wood burning have been conducted in Europe (Gustafson et al., 2007; Gaeggeler et al., 2008). After being subject to lively debates within the Île-de-France region, an in-depth investigation of this source would therefore appear necessary to better understand its emission specificities and its potential impacts on atmospheric chemistry.

In order to complete the information on wood burning activities, VOC measurements (within the “CORTEA-CHAMPROBOIS” program) were performed at a fireplace facility located ~ 70 km northeast of the inner Paris centre in March 2013. On-line (PTR-MS) and off-line (sampling flasks analyzed later on at the laboratory with a GC-FID) measurements were performed. These in-situ observations represent a more qualitative (predominant species identification) than quantitative approach as the resulting speciation profile is based on a limited number of data. As illustrated in Figure 3, 19 VOC species could be detected. C₂ hydrocarbons (ethylene, acetylene), C₆ - C₈ aromatics (benzene, xylenes) and oxygenated species (methanol, acetaldehyde and acetone) can be considered as predominant compounds from domestic wood burning. This finding is still consistent with intensive field studies of wood burning performed in Europe (Barrefors and Petersson, 1995; Gaeggeler et al., 2008; Evtyugina et al., 2014; Nalin et al., 2016).

Natural gas experiment

Natural gas is predominately composed of methane (CH₄) accounting for at least 80 % of the total chemical composition. It is also a mixture of other pollutants including lightweight VOCs and lower paraffins (approximately 10 % by volume).

In a first approach to determine the speciation profile from natural gas used in Paris, near-field samplings were performed from a domestic gas flue using three stainless-steel flasks, which have been analyzed by GC-FID at the laboratory. Main results (Figure 4) show a large dominance of alkanes, such as ethane (~80 %), propane (~11 %) and heavier hydrocarbons (like butanes, pentanes) ranging from 4.5 to 0.4 %. Ethane and propane therefore appear as a significant profile signature of natural gas leakages (Passant, 2002).

3 Results and discussions

3.1 Meteorological conditions and air mass origins

Meteorological parameters (e.g. temperature, relative humidity, rainfall, sun exposure, boundary-layer height, wind speed and direction) are known to be key factors governing seasonal and diurnal variations of air pollutant levels.

Air temperatures observed during the campaign were comparable to standard values determined by the French national meteorological service “Météo-France” (available at: http://meteofrance.com), with however an uncommon cold wintertime (Bressi et al., 2013 - Fig.S1a). Temperatures recorded in January and February 2010 were respectively between -2°C and -3.5°C.
below normal values (see. Section S3 in the Supplement). Extreme unusual cold-air outbreaks and a few snow flurries affected the Paris region, thus explaining higher temperature anomalies during that period. Levels of hours of sunshine and rainfall were globally consistent with standard values, with however some discrepancies in winter/autumn and spring, respectively (Fig. S3). In addition, atmospheric boundary layers showed seasonal changes with mean heights up to ~ 800 m in winter and up to 1 600 m in spring and summer occurring during the afternoon (see Section S4 in the Supplement). These average seasonal heights are expected to play a key role in pollutant dispersion and consequently impact ambient VOC concentrations.

For the year 2010, Paris was mainly influenced by air masses coming from the west (62 %) and usually associated with clean marine air influences from the Atlantic Ocean (see Figure 5). They are typically representative of local and regional pollution conditions, as already observed in Gros et al., 2011, Gaimoz et al., 2011; Dolgorouky et al., 2012 and Petit et al., 2015. To a lesser extent, Paris can be affected by northeast air masses (26 %) originating from Eastern France, the Benelux area, Northern Germany and indicative of continental imports of long-lived pollutants (Gaimoz et al., 2011). Air masses coming from the west are generally observed in summer and autumn (32 – 41 %) whereas northeast air masses are found to be significant in winter (34 %) and most frequently in spring (ca. 40 %), due to the stagnation of an anticyclone surrounding the British Isles (Monthly weather report for Paris and its surroundings during April 2010, Météo-France) during that period.

3.2 VOC concentration levels in ambient air

The main results of descriptive statistics for all the measured VOCs (from both GC-FID and PTR-MS instruments) on the whole sample set were summarized in Table 1. The average composition of VOCs was mainly characterized by oxygenated species (0.7 – 5.9 µg m\(^{-3}\) (0.4 – 4.4 ppb); 36.5 % of the TVOC mass), alkanes (0.5 – 4.6 µg m\(^{-3}\) (0.1 – 3.7 ppb); 39.1 %) followed by aromatics (1.1 – 3.3 µg m\(^{-3}\); 16.9 % (0.3 – 0.9 ppb) ) and to a lesser extent by alkenes, alkynes and dienes (0.3 – 1.6 µg m\(^{-3}\) (0.1 – 1.3 ppb); 7.5 %). Both alkanes and OVOCs significantly contribute up to 75 % of the TVOC concentrations. With ethane (10.9 %, 4.6 µg m\(^{-3}\) (3.7 ppb) on average) being the main alkane, methanol (14.0 %, 5.9 µg m\(^{-3}\) (4.4 ppb)) and acetone (11.6 %, 4.9 µg m\(^{-3}\) (2.0 ppb)) are considered to be the two major oxygenated compounds measured in this study. This conclusion is in agreement with previous VOC measurements performed in downtown Paris in 2007 (Gros et al., 2011).

The comparison between these average ambient levels and VOC measurements reported in the literature for different urban areas is restricted here to PTR-MS data as they constitute the most original dataset of this study. Most atmospheric studies were indeed conducted in urban metropolitan areas by investigating only NMHC measurements.

Table 2 summarizes PTR-MS data collected during the intensive experiment together with average VOC levels reported in ppb from other cities around the world. For the Paris megacity, a significant decrease in VOC concentrations was observed between spring 2007 and spring 2010 (from - 53.8 % for xylenes and C\(_8\) aromatics to - 25 % for benzene and MVK+MACR+ISO-POOHs), excepted for methanol and acetaldehyde (+ 11.8 %, + 35.7 %, respectively). These differences in aromatics concentration levels are consistent with decreasing downward trends of NMHC recorded during springtime in Paris (Waked et al., 2016). We note that the study of Waked et al. (2016) on VOC trends in Paris only concerns NMHC and not oxygenated species. As these OVOCs are significantly impacted by biogenic and secondary sources, it is not surprising to observe a different variation between 2007 and 2010.
Among selected species, benzene (as a carcinogenic agent) is one of the few regulated VOCs. According to the Directive 2000/69/EC, the annual mean benzene concentration in ambient air should not exceed 5 µg m$^{-3}$ (1.5 ppb). Background levels of benzene were relatively stable in recent years, with an annual average concentration of 1.1 µg m$^{-3}$ (0.3 ppb) (Airparif, 2015).

Average VOC concentrations were also calculated in line with sampling periods of the other European and global studies over different years (see Table 2). In this study, measured VOC levels were in the range of those found within some European cities (Barcelona, London – from 0.1 to 2.1 ppb concentration differences). However, average VOC levels observed in Paris were significantly lower than those measured in Houston (USA – from 0.1 to 6.9 ppb concentration differences) and more particularly in Beijing (China – from 2.5 to 8.9 ppb), in Mexico City (Mexico – from 0.1 to 27.4 ppb) and in Mohali (India – from 0.9 to 32.7 ppb).

3.3 Seasonal and diurnal variations

The variability in VOC concentration levels is controlled by a combination of factors including source strengths (e.g. emissions), dispersion and dilution processes as well as photochemical reaction rates with OH radicals and other oxidants (Filella and Peñuelas, 2006). Variations of selected trace gases (nitrogen/carbon monoxide, NO/CO – Figure 6) and VOCs illustrating contrasting emission sources and atmospheric lifetimes were analyzed at different time scales. As (O)VOCs measured by PTR-MS constitute the most original data of this study (and representing ~ 37% of the TVOC mass), a discussion on their variations (Figure 7) and their respective sources is given here. For information, an overview of seasonal and diurnal profiles of lighter hydrocarbons ($C_2$ - $C_6$) measured by GC-FID is reported in Section S6 in the Supplement.

Known as combustion tracers (traffic/wood burning), nitrogen monoxide (NO) and carbon monoxide (CO) exhibit higher median concentrations during winter and in late autumn, while lower concentrations appear in summer (Fig. 6a, 6b). These low levels can be explained by greater photochemical reaction rates (linked to higher solar radiation) combined with a stronger vertical atmospheric mixing compared to the other seasons. Another explanation of this variability is the increase in NO and CO emissions due to home heating fuels consumed in winter. NO concentrations are significantly enhanced between 06 and 12 h LT (with its maximum around 09 h). Contrary to NO, the diurnal pattern of CO is characterized by a “double wave” profile with an initial peak at 07-10 h (maximum at 09 h) and a second one at the end of the afternoon between 16 and 21 h. These increases typically correspond to morning and evening rush-hour traffic periods, as previously observed in Ammoura et al. (2014). The evening peak is smaller in magnitude than the morning one partly due to a higher Planetary Boundary-Layer (PBL) height in the afternoon (leading to dispersion and dilution processes) and to more disperse traffic periods. After 21 h, CO levels stay quite high (240 – 260 ppb) due to several factors: on-going emissions (traffic and wood-burning activities), lower photochemical reactions and atmospheric dynamics (the shallower boundary layer leads to more accumulation of CO (and other co-emitted species)). The evening event is not observed for NO as during this time ozone ($O_3$) presents its highest concentrations, leading to the titration of NO.

Good correlations between CO and some alkanes (iso-/n-pentane, n-hexane), alkenes (ethylene, propene), acetylene and aromatics (benzene, toluene and xylenes + $C_8$) were found when considering a Pearson’s correlation coefficient $r$ greater
than 0.6. All these compounds follow a similar seasonal and diurnal pattern, indicating that they share some or almost all common sources related to anthropogenic combustion processes (e.g. road traffic and/or wood burning). These observations are in agreement with the conclusions from Gros et al. (2011) and Gaimoz et al. (2011).

With atmospheric lifetimes from a few hours to several days, oxygenated species (OVOCs) are emitted from primary sources, mainly of biogenic origins, and significant secondary sources related to the oxidation of hydrocarbons. High concentration levels of OVOCs (for instance, methanol) and CO were observed during winter months (the season with coldest temperatures and where wood burning-related activities can play an important role). The low height of the PBL is also a relevant factor to consider as it can lead to the accumulation and the stagnation of VOC species into the troposphere during that season. In addition, significant OVOC levels were observed from April to September. In springtime, elevated baseline levels were measured when the Paris region was mostly influenced by northeast influences (see Fig.5), suggesting that they partly depended on continental imported and already processed air-masses. Biogenic emissions indeed contributed to high OVOC concentrations during this season and in summer.

Methanol is usually released into the atmosphere by vegetation and man-made activities contributing to a relatively high background levels during most of the year. This compound displays a specific diurnal pattern depending on the season and atmospheric dynamics (see Section S4 in the Supplement). In winter, methanol shows a “double wave” profile with two peaks at 10-11 h and 19-20 h (see Fig. 7c, top right), suggesting the influence of anthropogenic activities (e.g. road-traffic, wood burning sources). A slight delay (1-2 hours) is observed for methanol in comparison to other primary species (for instance, aromatics), that highlights the secondary origin of this species. In summer, methanol is characterized by high concentrations during night hours (00 - 06 h), followed by a significant decrease until the early afternoon and another increase from 18 h to midnight (Fig. 7c, bottom right). This night-time maximum of methanol has already been observed in urban environments, however, with no clear explanation (Solomon et al., 2005). This diurnal cycle can possibly be interpreted as the accumulation of species concentrations during the night from a local source under a shallow inversion layer, which is decreasing when the PBL is increasing (as dilution and dispersion processes occurring). However, the corresponding nighttime source has not been yet identified.

With a relatively short lifetime (~ 9 hours), acetaldehyde shows a diurnal cycle fairly comparable to acetone (Fig. 7d, 7e). Lower concentrations were observed during the night and from 18 h. Average levels increase from sunrise to a maximum at noon and slightly decrease in the afternoon. For these two OVOC species, the reduction of concentrations does not occur in the same way. From 12 – 18 h, average acetaldehyde concentrations are linearly decreasing (~1.0 µg m$^{-3}$ or 0.5 ppb) while mean acetone levels show a slower decline rate (~0.5 µg m$^{-3}$ or 0.25 ppb) with a tiny raise at 17h. This finding depends on their emission sources and strengths (e.g. biogenic, solvents use), but also on their respective photochemical reaction rates (1.5 x 10$^{-11}$ cm$^3$ molecule$^{-1}$ s$^{-1}$ for acetaldehyde and 1.8 x 10$^{-13}$ cm$^3$ molecule$^{-1}$ s$^{-1}$ for acetone) [Atkinson et al., 2006]. As acetone has a relatively long atmospheric lifetime (~ 68 hours), concentration levels are often more homogeneous.

Finally, methylvinylketone, methacrolein and isoprene hydroxy hydroperoxides (MVK+MACR+ISOPOOHs), three secondary products of isoprene photo-oxidation (as good indicators of biogenic activities), exhibit high levels in the late afternoon
due to the oxidation of daytime isoprene. The formation of these compounds mostly occurs in summer, but also in winter (Fig. 7f). This fact could eventually be related to anthropogenic activities such as wood burning (see Sub-section 2.4.2, Fig. 3).

### 3.4 Source apportionment

#### 3.4.1 Motor Vehicle Exhaust factor

The speciation profile of Factor 1 (see Fig. 8a) exhibits high contributions of alkanes, such as pentanes (iso-; n-) and n-hexane with on average ~ 50% of their variabilities explained by this factor. Aromatic compounds (toluene, xylene + C_8, benzene; ~ 35%) and light alkenes (ethylene, propene), which are considered as typical combustion products, are also the predominant species in this factor.

To evaluate the relevance of this factor, a comparison between speciated profiles from tunnel measurements (Fig. 1) and PMF simulations was done and is reported in Figure 9. Traffic profiles are in general coherent and consistent amongst themselves, thus allowing to label this factor as a “Motor Vehicle Exhaust” source. Indeed, a good agreement is observed between these two profiles for the major species such as iso-/n-pentane, toluene, ethylene, propene. Instead, significant differences in mass contributions for ethane (almost a factor of 10), acetylene (considered as a key combustion compound emitted from traffic not identified in the PMF profile), isoprene (represented by evaporative sources) and oxygenated species were found. These differences can potentially be explained by several reasons. Firstly, the proportion of VOC emitted from traffic may be different depending upon the types of vehicles/engines/fuels (Montero et al., 2010). VOC emissions can also be dependent on the use of vehicles (age, maintenance), driving situations and thermal conditions (hot soak). Secondly, the vehicle fleet composition is different in the centre of Paris and in a highway tunnel. Although the proportion of passengers cars and Light Duty Vehicles (LDV) accounts for 60–90% of the total composition of the fleet in circulation in both cases, the share of two-wheelers and heavy goods vehicles can be different. Indeed, heavy vehicles are subject to traffic limitations prohibiting their entry in Paris whereas as they are allowed in the highway tunnel (5%). The proportion of two-wheelers is significant in Paris (10–20%) (Airparif, 2013) while they represent less than 2% of the total vehicles in the tunnel. Finally, PMF artefacts cannot be excluded. We suppose that the contribution of ethane in the “Motor Vehicle Exhaust” factor is overestimated (same as the “Wood Burning” factor) at the expense of the mixed “Natural Gas and Background” source (for which ethane contributions seem to be underestimated).

This factor 1 displays fair correlations with nitrogen monoxide/dioxide (NO/NO_2), carbon monoxide (CO) and black carbon (from its fossil fuel fraction), which are known to be relevant vehicle exhaust markers (0.53 < r < 0.64). The average contribution of this factor is rather stable throughout the year (5.8 µg m^{-3} - see Figure 10, panel 1 and Section S7 in the Supplement). A smaller contribution is found during winter (3.2 µg m^{-3}) whereas the highest emissions from motor vehicle exhaust occur in autumn (8.6 µg m^{-3} on average) with a contribution of up to 10.1 µg m^{-3} in September. This seasonal cycle has already been observed and described in Bressi et al. (2014) for the road-traffic source of fine aerosols in Paris. The diurnal variation of this source is characterized by a “double wave” profile with an initial increase at 07-10 h (LT, Local Time) and a second increase at the end of the afternoon between 16 and 19 h (LT). These increases correspond to morning and evening rush-hour traffic
periods. After 21h, the absolute contributions of this factor stay quite high (7.0 – 8.0 µg m\(^{-3}\)) due to several factors: on-going emissions (until midnight), lower photochemical reactions and atmospheric dynamics (the shallower boundary layer leads to more accumulation of pollutants at night). Lower contributions are generally displayed on late mornings/early afternoons and at night. This reduction in factor contributions could be mainly explained by dilution and OH oxidation processes of more reactive species, which are not being balanced by additional vehicular emissions. This pronounced cycle has already been reported in previous studies (Gaimoz et al., 2011 and references therein). The temporal source strength variation is usually much more pronounced during weekdays than the weekend.

3.4.2 Evaporative sources factor

The profile of Factor 2 (see Fig. 8b) exhibits a high contribution from propane and iso-/n-butanes, with more than 47% of their variabilities explained by this factor. It was already identified by Gaimoz et al. (2011) and is used here as reference profile from “gasoline evaporation” emissions (including storage, extraction and distribution of gasoline or Liquid Petroleum Gas (LPG)). The generic term "Evaporative sources" is here used to take into account these types of evaporative emissions. Factor 2 also includes a significant proportion of isoprene (20%). This finding is still consistent with the conclusions of Borbon et al. (2001), which have shown that traffic activities emit a small amount of isoprene. In the same way, oxygenated compounds (acetaldehyde (4%), acetone (6.6%)) were found in fugitive evaporative emissions in agreement with what was observed during the highway tunnel experiment (see Fig. 2).

Among independent tracers used, only NO displays a fair agreement with this factor (\(r = 0.35\)). A correlation between F2 and F1 can also be noted (\(r = 0.36\)), thus indicating that these two factors are related to a common source (e.g. road-traffic). This source is in the range of \(~ 3.9\) µg m\(^{-3}\) over the whole studied period (see Fig. 10, panel 2). The annual trend of F2 seems to be consistent with the motor vehicle exhaust factor (F1), even though its monthly change remains ambiguous. Indeed, lower evaporative contributions are recorded both in winter and in early summer with minimum average contributions in June and July (1.7 µg m\(^{-3}\)). This finding was already identified by Frachon H. (2009; pers. communication). However, this value in June is somewhat puzzling as road-traffic emissions are usually significant (4.9 µg m\(^{-3}\)). In July, propane and butanes (iso-/n-) values were missing due to analytical problems on the operating GC-FID. Consequently, these compounds were simulated by the PMF model (e.g. missing values were virtually substituted by median values) which may underestimate the contribution of this factor during this specific period of time. However, high contributions of this source occur in August (6.4 µg m\(^{-3}\)). Although exhaust emissions are not particularly important, this observation could be eventually explained by gasoline storage and distribution sources, which may have increased with higher temperatures during that month. Maxima temperatures have generally been in the range of 16 to 32°C. The source contribution is in average higher in autumn (6.1 µg m\(^{-3}\)) with a contribution of up to 6.3 µg m\(^{-3}\) in October.

The diurnal variation of this factor contribution is characterized by a nighttime minimum, an increase from 07 to 10 h (consistent with the motor vehicle exhaust factor, F1) and a much slower decrease in emissions during the afternoon than those observed for the vehicle combustion profile. This second factor therefore represents the emissions of less reactive species (OVOCs, propane, butanes), for which concentrations cannot be expected to be consumed photochemically in short transport
times. The temporal source strength variation is less pronounced on weekends than weekdays, which is typical of mobile source activity patterns.

According to the Copert IV (European Environment Agency, EEA) program for the calculation of air pollutant emissions from road transport, gasoline evaporation emissions can be explained by the evaporation of VOCs due to temperature, vehicle refueling, running losses, diurnal and “hot soak” reactions (when a hot engine is switched off). It was speculated that hot engines would emit more in the morning than in the evening, considering typical conditions of active inhabitants going to and from their workplace. Fugitive gasoline emissions from the loading of tank trucks, transportation and unloading from tank trucks at service stations and distributions depots can also be likely sources of this factor. In summary, this source depends on several parameters (related to road-traffic conditions, the vehicle fleet composition, economic activities and meteorological observations), which can make the interpretation of its seasonal variability difficult.

### 3.4.3 Wood Burning factor

In Paris, domestic wood burning represents a non-negligible part (about 5%) of the energy consumption by fuel used for home heating (Airparif, 2011). The chemical profile of this source (Factor 3), shown in Fig. 8c, is mainly dominated by acetylene with approximately 80% of its variability explained by this factor. It also includes ethylene (57.4%), benzene (22.7%) and oxygenated compounds, such as acetonitrile, acetaldehyde and methanol (with 18.3%, 12.6% and 8.2%, respectively). Acetonitrile is a hydrocarbon commonly used as a marker of biomass burning (Holzinger et al., 1999). All these chemical species typically reflect an anthropogenic source related to wood combustion processes (Lanz et al., 2008; Leuchner et al., 2015) in agreement with the fireplace emission profile (see Sub-section 2.4.2, Fig. 3). No full comparison between both speciation profiles was possible as the fireplace profile was based on a limited number of data. With this mind, only a qualitative approach allowed to identify predominant species emitted from this source and confirm the term "wood burning" assigned to this factor.

Biomass burning emissions are well correlated with black carbon originating from residential wood burning (BC\textsubscript{wb}) and carbon monoxide, a long-lived compound especially emitted from combustion reactions ($0.6 < r < 0.7$). In addition, they well co-vary with napthalene ($m/z$ 129.0 measured by PTR-MS) - a known polycyclic aromatic hydrocarbon emitted from combustion processes (industry, tailpipe emissions) including wood burning (Purvis and McCrillis, 2000). As expected, wood burning contributions display a distinct cycle with a winter maximum (20.5 µg m\textsuperscript{-3} on average) and a summer minimum (3.3 µg m\textsuperscript{-3}). Average contributions of this factor are rather stable in both spring and fall (6.9 and 5.9 µg m\textsuperscript{-3}, respectively).

Wood burning emissions linked to home/building heating are obviously highly dependent on meteorological conditions and particularly on cold temperatures. A clear negative relationship between the wood burning factor and temperature is found ($r=-0.56$). The diurnal variation of this source exhibits a “double wave” profile. Average contributions increase from sunrise to a maximum in midmorning and decrease until 16-17 h. At the end of the day, a second increase is observed with another maximum contribution at 19 - 21 h. This diel cycle can be explained by domestic behaviors. An important finding is that the diurnal pattern of this source is fairly comparable to that of the motor vehicle exhaust factor. However, the wood burning factor does not display any distinct weekly variation. High contributions are observed all week (without any distinction between weekdays and weekends) compared to motor exhausts, for which vehicular emissions are less pronounced on weekends than
weekdays. In addition, it exhibits poor correlations with NO, NO$_2$ and BCff ($r = 0.30$, $0.29$ and $0.19$, respectively), thus indicating that the wood burning factor is completely independent of the motor vehicle exhaust source.

### 3.4.4 Biogenic factor

The profile of Factor 4 (see Fig. 8d) exhibits a high contribution from isoprene, a known chemical marker of biogenic emissions, with more than 79% of its variability explained by this factor. In addition, this factor profile includes isoprene’s oxidation products (methylvinylketone (MVK), methacrolein (MACR) and isoprene hydroxy hydroperoxides (ISOPOOHs)) with more than 48%, methanol and acetone - a selection of compounds having a large contribution from biogenic emissions (Kesselmeier and Staudt, 1998; Guenther, 2002). It also accounts a significant contribution of some light alkenes (e.g. ethylene and propene), which can be evenly emitted by plants (Goldstein et al., 1996). Consequently, this factor F4 is termed “biogenic factor”.

Amounts of light alkanes (butanes, iso-pentane, n-hexane) and acetonitrile were also found in this profile and could be attributed to a mixing with other temperature-related sources or artefacts from the PMF model (Leuchner et al., 2015).

Biogenic emissions are directly related to solar radiation (Steiner and Goldstein, 2007) and ambient temperature ($r > 0.7$). For that reason, the highest biogenic factor contributions occur in summer ($10.5 \mu g m^{-3}$ on average) with a contribution of up to $14.3 \mu g m^{-3}$ in July. Daily mean contributions gradually increase from 09 h. A slight delay (03 h) is observed in comparison with diurnal temperature/solar radiations variations (for which values increase from sunrise at 06 h). We assume that chemistry affects this source factor as it takes part in the formation of secondary species (MAK+MVK+ISOPOOH, for instance) from the oxidation of primarily emitted compounds (isoprene, OVOC). Diurnal contributions reach their maximum at the end of the day (19h). Highest nighttime contributions of this source can be explained by the presence of oxygenated species (long-lived compounds already present in the atmosphere and/or secondarily formed from the oxidation of isoprene) in the profile combined with lower photochemical reactions and atmospheric dynamics (a low PBL height) at night.

### 3.4.5 Solvents use factor

The profile of Factor 5, shown in Fig. 8e, is associated with a large contribution of selected OVOCs (acetaldehyde, methanol and acetone) with on average ~ 33% of their variabilities explained by this factor. Significant contributions from aromatic compounds (toluene, xylene $+$ C$_8$ and benzene) and some alkanes (pentanes, butanes, propane and n-hexane) are also observed. Toluene, in addition to road-traffic, is a good marker for solvents originating from an industrial source (Buzcu and Fraser, 2006). Benzene, due to its toxic and carcinogenic nature, was regulated in recent years and is strongly limited in solvent formulations. Current standards establish limits in benzene concentrations at 0.1% in cleaning products. However, PMF results point out the presence of benzene in this factor, suggesting that this compound might potentially still be in use by some manufacturers. Finally, the presence of these aforementioned species illustrates that this profile could be linked to industrial emissions, although a mixing of different sources cannot be excluded.

This factor co-varies well with ethanol, butan-2-one (also called methylethylketone – MEK), isopropyl alcohol or even ethyl acetate ($0.68 > r > 0.52$, respectively) – four organic compounds that were measured by GC-MS during the MEGAPOLI campaign (January-February 2010). These species are often used as solvents, diluents or cleaning fluids in industrial processes.
Zheng et al., 2013). Some manufactories can consume fossil fuels for their activities, which may explain the fairly good correlation between this factor and black carbon originating from fossil fuels (BC_ff, r = 0.50). Indeed, these fossil fuels could be used by industries as diverse as paints, paintings inks and lacquers (Tsai et al, 2001; Cornelissen and Gustafsson, 2004).

The highest contribution of this factor is observed during winter (14.2 µg m\(^{-3}\)) with a contribution of up to 20.9 µg m\(^{-3}\) in January. In winter, factor contributions increase at 06 h and reach their maximum between 11 h and 19 h (15 – 20 µg m\(^{-3}\)) before a long and gradual decline in the evening (see Fig. 10, panel 5 – top right). Higher contributions in winter can be explained by lower photochemical reactions (combined with weaker OH concentrations/UV radiations) and atmospheric dynamics. Indeed, a shallower PBL (and consequently, less intense vertical dynamics) leads to more accumulation of pollutants and thus to higher source contributions. The daily wintertime variability of this source is in agreement with the diel cycle of independent tracers (ethanol, butan-2-one).

Reconstructed contributions associated with this factor are also significant in summer (12.6 µg m\(^{-3}\) in July), which could be mainly explained by the evaporation of solvent inks, paints and other applications during that month due to higher temperatures. In spring/summer/autumn, factor contributions also increase at sunrise, but reach their maximum between 08 and 10 h (typical of anthropogenic activities). They progressively decrease during the afternoon (see Fig.10, panel 5 – bottom right). This gradual decline (not earlier observed in winter) is influenced by greater photochemical reactions and more intense vertical dynamics during these three seasons, leading to dispersion and dilution processes (and consequently, lower source contributions during the afternoon).

The temporal source strength variation is much more pronounced during weekdays than the weekend, except on Saturday morning. These diel and weekly patterns seem to be consistent with industrial source activities.

3.4.6 Natural Gas and Background factor

The profile of Factor 6, shown in Fig. 8f, is mainly dominated by ethane with around 45% of its variability explained by this factor. It also contains propane (14.7%) and light alkanes (butanes), which are key long-lived compounds known to be associated with natural gas leakages. Such species have already been identified in the natural gas experiment (see Sub-section 2.4.2, Fig. 4), thus allowing to confirm the identification of this profile. The diel pattern of this factor is mainly based on the diurnal variation of ethane, which is characterized by a nighttime maximum and a mid-afternoon minimum. Mainly due to its low reactivity, the behavior of ethane can be interpreted as homogeneous species levels during the night under a shallow inversion layer, then followed by concentration reductions caused by the increase of the PBL and vertical mixing - leading to dispersion and dilution processes. Average contributions of this factor were significantly higher when the PBL was low (~11.0 – 14.0 µg m\(^{-3}\)) and lower when the PBL was high (~ 6.0 µg m\(^{-3}\)).

This F6 profile is also characterized by the presence of oxidized pollutants (OVOCs including acetone and methanol) and aromatic compounds (like benzene), which have relatively long atmospheric residence times of respectively 53, 12 and 9 days (assuming OH = 2.0 x 10\(^{6}\) molecules cm\(^{-3}\)) [Atkinson et al., 2000]. Because of their low reactivity, all the species of this factor tend to accumulate in the atmosphere and show significant background levels, especially in the northern hemisphere.
The resulting emissions can be considered as a partly aged background air, implying a possible regional background and/or a long-range (intercontinental) transport.

The average contribution of this mixed source (combining both natural gas and background emissions) is in the range of 9.2 µg m$^{-3}$ during the whole studied period. Lowest source contributions were observed in winter, which does not fit with that reported in the literature. As mentioned in the “Motor vehicle exhaust” and “Wood burning” sub-sections (3.4.1 and 3.4.3, respectively), PMF artefacts cannot be ruled out. Indeed, a problem with the distribution of ethane (considered as the key species of the mixed source) within PMF factors was raised. We assumed that higher ethane contributions were partly assigned to the “motor vehicle exhaust” and “wood burning” factors. Consequently, we assumed that the “natural gas and background” factor contributions were underestimated (especially in winter) for the benefits of the “wood burning” factor (another source significantly contributing during this season).

The highest contributions occur in spring (13.3 µg m$^{-3}$) when the Paris region is mostly influenced by prevailing air masses originating from the north and the northeast parts of Europe passing over Germany and the Benelux area (see Fig. 5). These continental imports constitute background events, which significantly impact baseline levels of ethane and oxygenated species. Slightly lower reconstructed mass contributions of this factor F6 were also observed in autumn. This fact can be explained by the consumption of natural gas (for home heating) during this season as average temperatures are progressively going down. No significant continental influences occur during the fall period as main air masses were coming from the west, south and southeast sectors, thus illustrating the importance of local pollution emissions during this season.

### 3.5 VOC source contributions

PMF simulations revealed the significant contribution of six VOC emission sources (e.g. five specific factor profiles and a mixed one, for which the natural gas source could not be isolated from background emissions). This Source Apportionment (SA) analysis concluded that the predominant sources at the receptor site were road-traffic-related activities (including motor vehicle exhaust, 15% of the Total VOC (TVOC) mass on the annual average, and evaporative sources, 10%), with the remaining sources from natural gas and background (23%), solvents use (20%), wood-burning (17%) and biogenic activities (15%). Each modeled factor exhibits distinct patterns due to the variations of the different source emissions and meteorological conditions. Monthly averaged contributions (expressed in %) of these factors to TVOC mass and all sources are reported in Fig. 11. Seasonal variations of the individual sources have already been commented in the previous sections. Therefore, only the most important features are reported here.

Road-traffic emissions were identified by PMF simulations to be the main source of VOCs in Paris. The sum of motor vehicle exhaust and evaporative sources accounted for a quarter of the TVOC mass. It showed higher contributions at the end of the year (21% and 15%, respectively), that is still consistent with the study from Bressi et al. (2014) and with long-term black carbon measurements (Petit et al., 2015) linked to enhanced traffic during autumn in Paris. Most importantly, it was observed that the wood burning source exhibited a significant contribution in winter months (almost 50% in January and February), which is still in agreement with wood-burning related-particle emissions (Favez et al., 2009). The biogenic source also displayed a significant contribution (~30%) in summer (mainly due to the weight of oxygenated species in the factor profile). The solvent
use source displayed high contributions during winter months (~33%, due to a lower PBL height and slower photochemical reactions during that period) and in July (due to the evaporation of solvents controlled by temperature). The source mixing natural gas and background showed a higher proportion in springtime (~34%) and lower proportions during autumn (~25%). This conclusion can be explained by pollution events that are both related to air masses imported from continental Europe (see Fig. 5) and/or specific meteorological conditions (low temperatures involving the use of home heating), respectively.

The reactivity of each modeled factor has also been investigated by considering the factor concentration of each species with their OH rate constant \( k_{\text{OH}} \) (Atkinson and Arey, 2003) and is reported in relative (and absolute) contributions in Figure. 11. Among all the emission sources identified by PMF, solvents use and motor vehicle exhaust factors appear as the main reactive sources [26% (33 s\(^{-1}\)) and 23% (40 s\(^{-1}\)), respectively]. This can be explained by high constant rates of aromatics and alkenes mainly associated to these two emission sources. The contribution of the biogenic source is surprisingly weak (17%). Although isoprene is an extremely reactive species, this factor exhibits a high weight of OVOCs for which constant rates can be low. Instead, the relative contribution of the mixed source “natural gas and background” is surprisingly high (16%) due to the presence of aromatics (toluene and xylenes) in the factor profile. The lower contribution of reactivity is represented by the evaporative sources factor [5% (13 s\(^{-1}\))] which contains more stable gases (propane, butanes).

### 3.6 Comparison with previous Source Apportionment (SA) studies performed in Paris

Based on 1-year daily PM\(_{2.5}\) measurements (September 2009 – September 2010), Bressi et al. (2014) also conducted a Source Apportionment (SA) analysis using the PMF method (EPA PMF 3.0) with the aim of identifying and characterizing major fine aerosols emission sources within the Paris area. Seven factors, namely Ammonium Sulfate (A.S.)-rich factor, Ammonium Nitrate (A.N.)-rich factor, heavy oil combustion, road-traffic, biomass burning, marine aerosols and metal industry were identified. Special attention is paid here to common modeled factor categories.

Primarily of local origin, the road-traffic source (resulting from exhaust and non-exhaust processes) constitutes approximately 14% of PM\(_{2.5}\) mass (~2.1 µg m\(^{-3}\), on average) over the whole sampling period. Its annual contribution was considered as significant but surprisingly low given the high traffic density in Paris and its surroundings. It exhibits stable averaged contributions throughout the year, with a smaller proportion in winter (6%, 1.3 µg m\(^{-3}\)) and higher in autumn (19%, 2.5 µg m\(^{-3}\)). This temporal source variation is still in agreement with the seasonal cycle of the road-traffic source (combining motor vehicle exhaust and evaporative running losses) issued from our VOC PMF analysis (see Sub-sections 3.4.1 and 3.4.2). The second common wood burning source is estimated for the first time over long periods and contributes to around 12% (1.8 µg m\(^{-3}\)) of the total PM\(_{2.5}\) mass. As expected, higher contributions were significantly observed during winter (22%, 4.7 µg m\(^{-3}\)) and in autumn (18%, 2.4 µg m\(^{-3}\)). This finding is still consistent with the seasonal pattern of the wood burning VOC source. Because of the daily time resolution of filter sampling, no diurnal variation of modeled sources was reported in Bressi et al. (2014), thus limiting any additional comparison with this study.

Based on 1-month VOC measurements (25 May – 14 June 2007) performed at the LHVP site, Gaimoz et al. (2011) also conducted a PMF analysis with the aim of identifying and apportioning major VOC sources in Paris. Seven factors, namely
vehicle exhaust, fuel evaporation, remote industrial sources, natural gas + background, local sources, biogenic and fuel evaporation and wood burning, were found. For an appropriate comparison between this study and our work, special attention is paid here to the modeled speciation profiles and source contributions.

Chemical profiles from Gaimoz et al. (2011) revealed consistent findings with this study. The fuel evaporation factor is mainly composed of butanes, propane and ethane whereas the vehicle exhaust factor includes iso-pentane, benzene, toluene, C₈ and C₉-aromatics and in lower proportions ethylene, propene and acetylene. These observations are consistent with modeled evaporative sources and motor vehicle exhaust profiles obtained in this work. A biogenic and fuel evaporation source is also identified and essentially made of isoprene, methanol, acetone and a high proportion of iso-pentane, suggesting that this factor is mixing up biotic emissions and road-traffic activities. Highly dependent on (continental) air-mass origins, a remote industrial factor (related to industrial activities and long-range transport of secondary VOCs) is found to exhibit high contributions of OVOCs (methanol, acetone), aromatic species (toluene, C₈-C₉ aromatics) and some light alkanes. Our PMF study emphasized a solvents use source, for which these aforementioned compounds were observed, in addition to benzene. The wood burning source includes only a high contribution of acetonitrile although ethylene, acetylene and benzene are significantly emitted - in accordance with findings from the fireplace experiment (see Sub-section 2.4.2). The mixed natural gas + background source is especially driven by ethane, methanol and acetone. No aromatic species appear in this factor profile. Finally, the local source (LPG – Liquefied Petroleum Gas) including propane and pentanes seems to be associated with fuel evaporation sources. These kinds of emissions have been included in the evaporative sources factor.

During May – June 2007, Gaimoz et al. (2011) concluded that the major VOC sources were related to road-traffic emissions (~39 % of the TVOC mass), with the remaining emissions from wood burning (2%), biogenic and fuel evaporation (5%), remote industrial sources (35%), natural gas and background (13%) and local sources (7%) during the whole studied period (Fig. 13, left pie chart). To accurately compare VOC sources proportions between 2007 and 2010 (for a similar combination of hydrocarbons and masses), the contribution of each main factor was recalculated for the specific time period May – June 2010 (Fig. 13, right pie chart).

Significant differences between biogenic and wood burning sources contributions could eventually be both explained by the weight of major OVOCs into speciation profiles (relative proportions of methanol, acetaldehyde, acetone in these factors are higher than those of the comparative study) and the differences in temperatures affecting the Paris region. The temperatures recorded in May – June 2007/2010 were 20°C and 16°C, respectively. This would explain a higher home heating consumption and consequently, a higher contribution of the wood burning factor in 2010 (9 % versus 2 % for the previous work). Regarding to solvents use source contributions, differences can also be explained by temperatures (as they constitute a relevant indicator in solvent emissions) and by the amount of solvents used in manufactories due to recent regulatory frameworks in place (20 % in 2010 versus 35 % in 2007). The difference of natural gas and background source contributions can be due to the importance of air masses coming from the north and northeast parts of Europe between 25 May and 14 June 2010. These air-mass origins were also observed in 2007 and could have affected remote industrial-related emissions and not the mixed source. Slight differences of the motor vehicle exhaust source between 2007 and 2010 (22 % versus 14 %) could be explained by densification strategies.
and technological innovations for reducing car use and emissions. Finally, observed differences for the evaporative sources factor (5% for 2010 and 17% in 2007) are related to emissions and high temperatures observed in 2007.

3.7 Comparison with some European SA studies

Yearly average contributions of the modeled VOC sources (see Section 3.5) were also compared with other Source Apportionment (SA) studies performed within urban areas in Europe and in the World. From the different European SA studies available, only one is based on a long VOC time series, which is strengthening the novelty and the originality of the current study. Based on 2-years hourly measurements of $C_2$ - $C_7$ NMHCs, Lanz et al. (2008) permitted the identification and characterization of between eight and six emission sources at an urban background site in Zürich (Switzerland) in the years 1993-1994 and 2005-2006. Only measurements from 2005 to 2006 are compared here as they are the most recent observations we have available. Six factors, namely gasoline evaporation, solvents, propane, ethane, wood burning and fuel combustion were determined using the PMF method. This SA study highlighted the importance of vehicular, solvents use, wood burning and gas leakage emissions. The road-traffic-related source included both gasoline evaporation and fuel combustion (motor exhaust) factors. While the first factor is mainly dominated by butanes (iso-/n-) and iso-pentane, the second one is essentially driven by ethane, ethene, propene, benzene and toluene. These two speciation profiles are still consistent with those obtained from this PMF analysis, except for iso-pentane. Considered as a key species of evaporative processes, iso-pentane mostly contributed to the motor vehicle exhaust source (Fig. 8a). It was also identified as one of the main compounds emitted in the highway tunnel experiment (Fig. 2), where measured hydrocarbons were representative of fresh emissions (e.g. fuel combustion). This modeled road-traffic source contributed to 26% of the TVOC mass (13% for gasoline evaporation and 13% for fuel combustion factors), which is in the same order of magnitude than that of our vehicle-related source (25%). The solvents use factor is characterized by pentanes, S-isohexanes (including 2-methylpentane, 3-methylpentane, 2,2-dimethylbutane, 2,3-dimethylbutane) and toluene, in agreement with our solvents use VOC profile which also included oxygenated species (not measured in Lanz et al. (2008)). This industrial factor accounted for 20% of the TVOC mass. This source contribution is comparable to what we obtained from January to November 2010 (20%). The wood burning factor is mainly dominated by ethylene, acetylene, ethane, benzene and contributed to 16% of the TVOC mass for the 2005-2006 sampling period. This finding is fairly in agreement with our annual wood burning contribution. Finally, a natural gas source was also identified and consisted of the combination of two separated factors (“ethane” with “propane”). Its annual contribution is evaluated at 35% of the TVOC mass whereas our mixed natural gas and background source accounted for 23%. No biogenic source was detected for this comparative study. To sum up, average contributions of the road-traffic, solvents use and wood burning sources well matched between this SA study and our modeled results although the input chemical matrix and sampling dates are different.

The importance of these three anthropogenic sources was often reported in other existing urban SA studies from short-term measurements performed in Europe. For instance, Niedojadlo et al. (2007) (Wuppertal, Germany) paid particular attention to solvents use and road-traffic sources contributions using the Chemical Mass Balance (CMB) modeling technique. Main results showed that the road-traffic source dominated total VOC emissions (more than 50% of the total mass) compared to solvents use. In addition, it was considered that the proportion of solvents emissions to TVOC concentrations fell in the range of ~20%
in German cities, which is significantly consistent with Lanz et al. (2008) and with this SA study in Paris.

The consistency in VOC contributions in European urban areas raises the question of their representativity at a larger scale. There are currently many other urban SA studies described in the literature (e.g. Jorquera and Rappenglück, 2004 – Santiago/Chile; Buzcu et al., 2006 – Houston/TX; Brown et al., 2007 – LA; Cai et al., 2010 – Shanghai/China; Morino et al., 2011 – Tokyo/Japan; Yurdakul et al., 2013 – Ankara/Turkey; Zheng et al., 2013 – Mexico). Results of these studies are not detailed here but one common feature for European and global scales is the importance of the road-traffic source (between 30% and 50%). One difference concerns the industrial sector which plays (in the investigated European cities) a lower role than in studied urban areas from other continents.

Governmental regulations and standards to control pollutants emissions and economic developments may differ between European countries and the rest of the world. The location of sampling points (or distances from main sources) and meteorological conditions can strongly affect VOC concentrations and their respective emission sources in the considered urban environments.

4 Conclusions

Within the framework of the EU-F7 MEGAPOLI and PRIMEQUAL-FRANCIPOL research programs, a selection of volatile organic compounds (VOCs) were continuously measured in real-time at two background urban sites located in downtown Paris (France) from 15 January to 22 November 2010. Assessed hydrocarbons included alkanes, alkenes-alkynes, isoprene, aromatics and oxygenated compounds (OVOCs). The current study allowed evaluating VOC concentration levels in ambient air and describing their temporal (seasonal and diurnal) time courses over a long period of time in the French megacity. It also showed an innovative methodology to identify, quantify and understand the main VOC emission sources in Paris by combining field experiments (near-field and ambient air measurements) with source-receptor statistical modeling. The modeled factor profiles were interpreted with respect to those obtained from literature and from three near-field experiments (inside a highway tunnel, at a fireplace and from a domestic gas flue) performed within the Paris area. These additional measurements helped better characterizing and/or confirming traffic, wood burning and natural gas-related sources among the existing different source profiles, which can be directly derived from a PMF modeling analysis. These source profile studies therefore allowed to check the representativity and the robustness of our conclusions. This PMF analysis successfully reconstructed at least 88 ± 2 % of the measured total VOC mass.

Among the six identified PMF factors, road-traffic activities appeared to be the main VOC source in Paris with an average contribution of 25% of the TVOC mass at the annual scale. This source both included motor vehicle exhaust (15%) and evaporative sources (10%). For the first time, it was also shown that the residential wood burning source exhibited an important contribution in winter (almost 50%) due to cold temperatures during that season (home heating consumption). The biogenic source also displayed a significant contribution (~30%) in summer mainly due to the weight of oxygenated species in the factor profile. A solvents source was identified and annually contributed to 20% of the total VOC mass. Finally, it was also revealed
that a source mixing natural gas and background (23%) could be highly dependent on air-mass origins (especially during continental-influenced periods) and meteorological conditions (temperatures). It exhibited a higher proportion in springtime (34%, explained by intercontinental imports) and in autumn (25%, partly for home heating consumption reasons).

From this initial source apportionment study, natural gas could not be isolated from background emissions by the PMF method, thus leading to a limitation of this analysis. A further work will aim at constraining the reference speciation profile (obtained from domestic gas flue measurements) in order to evaluate the relative contribution of natural gas emissions. Lastly, the quantitative assessment of the contributions from different modeled sources presented in this study will provide an independent evaluation of the quality and the relevance of the corresponding emission inventories. In particular, the comparison will be very valuable with the updated local emission inventory (provided by the regional air quality network AIRPARIF) as some discrepancies had been pointed out with its previous version.

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References


IARC: Outdoor air pollution a leading environmental cause of cancer deaths, Press Release n°221, 2013.


Table 1. Statistical summaries (µg m$^{-3}$) of selected VOC concentrations measured at urban background sites. Statistics were calculated based on hourly mean data, initially obtained every 30 mn (ethane > isoprene) and every 5 to 10 mn (for aromatics and OVOCs). These measurements were undertaken from 15 January to 22 November 2010 (~ 10 months). A conversion factor is provided here to convert VOC concentrations (µg m$^{-3}$) into (ppb) mixing ratios.

<table>
<thead>
<tr>
<th>Species</th>
<th>Conversion Factor</th>
<th>Minimum</th>
<th>25th Percentile</th>
<th>Median</th>
<th>Mean</th>
<th>75th Percentile</th>
<th>Maximum</th>
<th>σ</th>
</tr>
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<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
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<td>4.56</td>
<td>5.42</td>
<td>26.31</td>
<td>2.26</td>
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<td>0.65</td>
<td>0.77</td>
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Table 2. Comparison of mean concentrations of selected VOCs (measured by PTR-MS) with ambient levels observed in the literature from different urban atmospheres. All average values are reported in ppb.

<table>
<thead>
<tr>
<th>VOCs measured by PTR-MS (m/z)</th>
<th>Paris(^a)((P))</th>
<th>Paris(^b)</th>
<th>Barcelona(^c,1)</th>
<th>London(^d)((P))</th>
<th>Mohali(^e)((P))</th>
<th>Mexico City(^f)((P))</th>
<th>Beijing(^g)((P))</th>
<th>Houston(^h)((P))</th>
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<tr>
<td></td>
<td>Jan.-Nov. (Spring)</td>
<td>Spring</td>
<td>Winter</td>
<td>October</td>
<td>May</td>
<td>March</td>
<td>August</td>
<td>Aug.-Sept.</td>
</tr>
<tr>
<td>Methanol (33.0)</td>
<td>4.5 (6.6)</td>
<td>5.9</td>
<td>NA</td>
<td>NA (3.3)</td>
<td>38 (5.3)</td>
<td>NA (1.6)</td>
<td>11.7 (2.8 )</td>
<td>10.8 (3.9)</td>
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<td>Acetonitrile (42.0)</td>
<td>0.7 (1.2)</td>
<td>0.4</td>
<td>0.2–0.5</td>
<td>0.3 (0.2)</td>
<td>1.4 (0.5)</td>
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<td>Acetaldehyde (45.0)</td>
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<td>0.8–1.7</td>
<td>3.6 (1.5)</td>
<td>6.7 (1.7)</td>
<td>3.0–12.0 (1.1)</td>
<td>3.6 (1.1)</td>
<td>3.4 (1.5)</td>
</tr>
<tr>
<td>Acetone (59.0)</td>
<td>2.1 (2.5)</td>
<td>3.0</td>
<td>1.1–1.6</td>
<td>1.6 (2.2)</td>
<td>5.9 (2.1)</td>
<td>NA (1.7)</td>
<td>4.4 (1.6)</td>
<td>4.0 (2.1)</td>
</tr>
<tr>
<td>MVK + MACR</td>
<td>0.2 (0.2)</td>
<td>0.3</td>
<td>0.07–0.12</td>
<td>NA (0.2)</td>
<td>NA (0.1)</td>
<td>NA (0.1)</td>
<td>0.3–0.6 (0.2)</td>
<td>0.8 (0.2)</td>
</tr>
<tr>
<td>+ ISOPOOHs (71.0)</td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
<td></td>
</tr>
<tr>
<td>Benzene (79.0)</td>
<td>0.3 (0.3)</td>
<td>0.4</td>
<td>0.2–0.6</td>
<td>0.1 (0.4)</td>
<td>1.7 (0.3)</td>
<td>NA (0.3)</td>
<td>NA (0.2)</td>
<td>0.6 (0.3)</td>
</tr>
<tr>
<td>Toluene (93.0)</td>
<td>0.9 (0.9)</td>
<td>1.4</td>
<td>0.8–2.7</td>
<td>1.9 (0.9)</td>
<td>2.7 (0.8)</td>
<td>3.0–28.0 (0.6)</td>
<td>1.0–4.0 (0.5)</td>
<td>0.8 (0.9)</td>
</tr>
<tr>
<td>Xylenes + C(_8) (107.0)</td>
<td>0.6 (0.6)</td>
<td>1.3</td>
<td>0.9–3.4</td>
<td>0.2 (0.7)</td>
<td>2.0 (0.7)</td>
<td>NA (0.4)</td>
<td>NA (0.4)</td>
<td>0.6 (0.6)</td>
</tr>
</tbody>
</table>

\(^a\) This study (Values in brackets from VOC measurements performed during the same sampling period of the other urban studies are given for comparison).

\(^b\) Gros et al. (2011)

\(^c\) Seco et al. (2013)

\(^d\) Langford et al. (2010)

\(^e\) Sinha et al. (2014)

\(^f\) Fortner et al. (2009) – Values estimated from graphs

\(^g\) Shao et al. (2009)

\(^h\) Karl et al. (2003)

\(^1\) A full comparison was not possible because no data was available between 16 February and 24 March 2010.

NA – Non Available data.
Figure 1. Maps of Paris and Île-de-France region. The picture A shows the location of the two main sampling sites in downtown Paris. The white and red stars locate the position of the LHVP laboratory and the AIRPARIF site, respectively. The picture B shows the terrace roof of LHVP.
**Figure 2.** Average highway tunnel profile (in mass contribution, %) assessed from traffic peaks concentrations and subtracted from nighttime values. Red, blue, light/dark green and purple bars correspond to alkanes, alkenes-alkynes, isoprene/aromatics and oxygenated species, respectively.
Figure 3. Average VOC fingerprint (in mass contribution, %) from domestic biomass burning obtained during the fireplace experiment. Red, blue, light/dark green and purple bars correspond to alkanes, alkenes-alkynes, isoprene/aromatics and oxygenated species, respectively.
Figure 4. Average chemical composition of natural gas used in Paris. *Other VOCs include heavier alkanes (e.g. cyclopentane/hexane, dimethyl butanes) and butenes in lower proportions. Whiskers correspond to error bars ($1\sigma$).
Figure 5. Average trajectories obtained after clustering analysis and the relative proportion of clusters (%) over the year and per season.

<table>
<thead>
<tr>
<th>Cluster CL#</th>
<th>Air mass origins</th>
<th>Year (J-N)</th>
<th>Winter</th>
<th>Spring</th>
<th>Summer</th>
<th>Autumn</th>
</tr>
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<tr>
<td>CL#1</td>
<td>LRT (West)</td>
<td>29 %</td>
<td>24 %</td>
<td>25 %</td>
<td>32 %</td>
<td>32 %</td>
</tr>
<tr>
<td>CL#2</td>
<td>SRT (West)</td>
<td>33 %</td>
<td>29 %</td>
<td>25 %</td>
<td>41 %</td>
<td>33 %</td>
</tr>
<tr>
<td>CL#3</td>
<td>Continental influences (NE)</td>
<td>26 %</td>
<td>34 %</td>
<td>39 %</td>
<td>19 %</td>
<td>22 %</td>
</tr>
<tr>
<td>CL#4</td>
<td>North</td>
<td>11 %</td>
<td>13 %</td>
<td>11 %</td>
<td>8 %</td>
<td>12 %</td>
</tr>
</tbody>
</table>

*LRT: Long Range Transport
SRT: Short Range Transport*
Figure 6. (Left) Monthly box and whisker plots of NO (A) and CO (B) expressed in µg m\(^{-3}\) and ppb, respectively. Solid lines represent the median concentration and the box shows the InterQuartile Range (IQR). The bottom and top of the box depict the 25th (the first quartile) and the 75th (the third quartile) percentile. The ends of the whiskers correspond to the lowest and highest data still within 1.5 times the IQR of Q1 and Q3, respectively.

(Right) Diurnal variations of NO and CO averaged over the whole sampling period. Time is given as Local Time. Lines correspond to hourly means and shaded areas indicate the 95% confidence intervals of the mean.
**Figure 7.** (Left) Monthly box and whisker plots of benzene (A), toluene (B), methanol (C), acetaldehyde (D), acetone (E) and MVK + MACR + ISOPOOHs (F) expressed as µg m$^{-3}$. Solid lines represent the median concentration and the box shows the InterQuartile Range (IQR). The bottom and top of the box depict the 25th (the first quartile) and the 75th (the third quartile) percentile. The ends of the whiskers correspond to the lowest and highest data still within 1.5 times the IQR of Q1 and Q3, respectively. (Right) Diurnal variations of (O)VOCs averaged over the whole sampling period. Time is given as Local Time. Lines correspond to hourly means and shaded areas indicate the 95% confidence intervals of the mean.
WINTER
(January – February)

SUMMER
(June – July – August)
Figure 8. Source composition profiles of the 6-factor PMF solution. The concentrations (µg m$^{-3}$) and the percent of each species apportioned to the factor are displayed as a pale blue bar and a color box, respectively. (a) F1 – Motor Vehicle Exhaust; (b) F2 – Evaporative sources; (c) F3 – Wood Burning; (d) F4 – Biogenic; (e) F5 – Solvents use; (f) F6 – Natural Gas and Background.
Figure 9. Comparison of speciated profiles issued from the highway tunnel experiment and PMF simulations (F1 - Motor Vehicle Exhaust). The species contributions are expressed in %. NF = Near-Field.
Figure 10. (Left) Monthly box and whisker plots of modeled sources from the 6-factor solution. Concentration levels are expressed in µg m\(^{-3}\). Solid lines represent the median concentrations and the box shows the InterQuartile Range (IQR). The bottom and top of the box depict the 25th (the first quartile) and the 75th (the third quartile) percentile. The ends of the whiskers correspond to the lowest and highest data still within 1.5 times the IQR of Q1 and Q3, respectively. (Right) Diurnal variations of the resolved PMF factors. Time is given in Local Time. Lines correspond to hourly means and shaded areas indicate the 95% confidence intervals of the mean.
Figure 11. Variations of monthly averaged contributions of the six modeled VOC sources (expressed in %); (Top) - Average predicted VOC concentration levels per month (µg m$^{-3}$); (Bottom) - Completeness of the data per month (%).
Figure 12. Relative and absolute contributions of reactivity of each PMF factors (% and s\(^{-1}\), respectively).
Figure 13. Relative contributions to the TVOC mass of 7 and 6 PMF sources identified from 25 May to 14 June 2007 (Gaimoz et al., 2011 - left pie chart) and 2010 (this study - right pie chart), respectively.
Appendix A: Application of the Positive Matrix Factorization (PMF) approach in source apportionment of VOCs in Paris.

A1 Data preparation

Initially, the EPA PMF 5.0 model requires two input datasets: one with the chemical species atmospheric concentrations for each observation point and another with either uncertainties values or parameters for calculating the associated uncertainty.

The initial chemical dataset contains a selection of 19 hydrocarbon species and masses (for a detailed overview, see list of compounds in the Sub-section 2.4.1) measured from 15 January to 22 November 2010. Non methane hydrocarbons (NMHCs) and aromatic/oxygenated species (OVOCs) were respectively monitored with GC-FID and PTR-MS instruments belonging to different partners involved during the MEGAPOLI and FRANCIPOL intensive field campaigns. Unfortunately, no intercomparison between these instruments was possible because there was approximately a one-month delay between both experiments. However, preliminary PMF modeling simulations were performed using only the FRANCIPOL dataset (24 March – 22 November). The results have shown similar source profiles (see Section S1), as those already described in this paper. Consequently, two datasets (corresponding to MEGAPOLI and FRANCIPOL ones, respectively) were considered to form a single one and use it as an input unified database for the final PMF analysis.

The uncertainty dataset was built upon the equation-based method described by Norris et al. (2014). It requires both Method Detection Limit (MDL, here in µg m\(^{-3}\)) and the analytical uncertainty (\(u\), here in %) for each considered species. Two sets of MDLs were used, one for each measurement campaign. Slight differences among species MDLs were found for n-hexane, aromatics, acetaldehyde and MVK+MACR+ISOPOOHs between both experiments. Of these VOCs, MDLs from the FRANCIPOL campaign were chosen for representativity reasons (as the corresponding dataset represents ~ 88 % of the total data matrix) to keep consistency in uncertainty calculations. The analytical uncertainties were respectively estimated at 15 % and 20 % and kept constant over the experiments.

The PMF uncertainty (\(\sigma\)) is therefore calculated as follows:

\[
\begin{align*}
&\text{If } X_{ij} \leq \text{MDL}, \forall j; \quad X_{ij} = \frac{\text{MDL}}{2} \quad \text{and} \quad \sigma_{ij} = \frac{5}{6} \times \text{MDL} \\
&\text{If } X_{ij} \geq \text{MDL}, \forall j; \quad X_{ij} \text{ does not change and } \sigma_{ij} = \sqrt{(\text{ErrorFraction} \times X_{ij})^2 + (0.5 \times \text{MDL})^2}
\end{align*}
\]  

(A1)  

(A2)

MDLs and analytical uncertainties (\(u\)) for each VOC are reported in Table A.1.
A2 Estimation of the number of PMF factors ($p$)

The accurate number of PMF factors ($p$ values) in models must be ultimately estimated by the user using several exploratory means. Specific parameters were used to determine the appropriate $p$ value such as the assessment of $Q$ values, scaled residuals, predicted versus observed concentrations interpretation and the physical meaning of factor profiles.

Eight different modeling conditions were examined with $p$ values ranging from 3 to 10, each simulation being randomly conducted 20 times. The reviewing of the IS (the maximum individual standard deviation) parameter highlighted a slope failure for $p = 5$ whereas the IM (the maximum individual column mean) indicator reported another slope failure for $p = 6$. Choosing less factors, $p < 6$, concatenated three source profiles (attributed to solvents use, natural gas and background emissions, respectively) into a factor, whereas choosing $p = 6$ allowed splitting one of them. Opting for $p > 6$ did not provide any supplemental physical meaningfulness to existing profiles. The investigation of $r^2$ from ten modeled solutions also reported a slope failure for $p = 6$. In addition, only $Q_{true} / Q_{expected}$ value for $p = 6$ was closer than 1.0 (e.g. 0.94 in comparison with 1.12 for $p = 5$ and 0.8 for $p = 7$), thus suggesting that the 6-factor configuration is supposed to be the most optimum solution for this PMF analysis. Finally, this configuration was investigated over all the details. Usually, PMF identifies the best solution by the lowest $Q_{robust}$ value (e.g. the minimum $Q$). Within this analysis, its corresponding PMF solution was not considered due to a lack of physical significance for one factor profile (e.g. solvents). Therefore, another PMF solution closest to the selected $Q_{robust}$ value was subsequently examined and chosen in terms of interpretability and fitting scores.

A3 Robustness of PMF results

Further technical and mathematical indicators regarding to the 6-factor configuration are reported here to assess the robustness and the quality of the final PMF solution. Firstly, the ratio between $Q_{robust}$ and $Q_{true}$ reached around 1.0, thus indicating that the modeled results were not biased by peak events. Almost 100% of the scaled residuals were within $\pm 3\sigma$ and were normally distributed for all species. In addition, the Kolmogorov-Smirnoff (KS) test granted a KS $p$-value very close to zero, thus illustrating a statistically significant test with a $\alpha$ risk of 5%. The correlation between total VOC reconstructed concentrations from all the factors with total VOC observed concentrations is depicted in Figure A.1. With $R^2$ very close to 0.9, almost all variance in the total concentration of the 19 VOCs can be explained by the PMF model.

Almost all the chemical species also displayed good determination coefficients ($r^2$ higher than 0.6 for 15 compounds) between predicted and observed concentrations, with the exception of propane and $n$-hexane showing a fairly reasonable coefficient between 0.5 and 0.6 (due to their 31% and 36% missing values, respectively). Isoprene and acetonitrile exhibited bad $r^2$ values (0.29 and 0.06, respectively) due to either a relatively high number of missing values or a “weak” additional error, for which sample uncertainties were tripled. Slopes were close to 1.0 for most species (higher than 0.6 for 17 VOCs), excepted for isoprene (0.5) and acetonitrile (0.02). The limitations of the PMF model to simulate isoprene and acetonitrile have therefore been kept in mind within the reconstructed results description and discussions.
A4 Estimation of model prediction uncertainties

PMF output uncertainties can be estimated using the Error Estimation options starting with DISP (dQ-controlled displacement of factor elements) and processing to BS (Classical bootstrap). These two uncertainty methods are designed to provide key information on the stability and the precision of the chosen PMF solution (Paatero et al., 2014).

The DISP, Base Model Displacement Error Estimation, assesses the rotational ambiguity of the PMF solution by exploring intervals (minimum and maximum) of source profile values. During the DISP, a minimum Q value is newly calculated (based on the adjustment up and down in factor profile values) and compared with the unadjusted solution Q value. The difference between the initial Q value and the modified Q value (so-called dQ) should be lower than dQmax value, for which four levels (values: 4, 8, 15 and 25) were taken into account. For each dQmax value, 120 intervals were estimated. The DISP analysis results were considered validated: no error could be detected and no drop of Q was observed. As no swap occurred, the PMF solution was considered sufficiently robust to be used.

The BS, Base Model Bootstrap Error Estimation, is also used to evaluate the reproducibility of the PMF solution, with a specific focus on the original sub-matrix F. A further description on the bootstrapping technique is presented in Norris et al. (2014) and in Paatero et al. (2014). A base model bootstrap method was then carried out, executing one hundred iterations, using a random seed, a block size of 874 samples (calculated according to the methodology of Politis and White (2004)) and a minimum Pearson correlation coefficient (R-Value) of 0.6. All factors were well reproduced through this technique over at least 88% of runs, thus indicating that BS uncertainties can be interpreted and the number of factors may be appropriate. Consequently, 12% of runs were redistributed into the different existing factors. No runs were unmapped. Finally, around 91% of species with the base run profile value were identified within the InterQuartile Range (IQR, e.g. 25th - 75th percentile of bootstrap runs) for all factors considered.

Finally, the rotational ambiguity of this 6-factor PMF configuration was also investigated using the Fpeak parameter. Different Fpeak values from -5 to 5 were used to generate a more realistic PMF solution. The results from the nonzero Fpeak values were generally consistent with the runs associated with the zero Fpeak value (e.g. Base Model run), thus illustrating a low rotational ambiguity of the final PMF solution.
<table>
<thead>
<tr>
<th>Species</th>
<th>MDL–MEGAPOLI</th>
<th>MDL–FRANCIPOI</th>
<th>$u$</th>
</tr>
</thead>
<tbody>
<tr>
<td></td>
<td>$\mu g m^{-3}$ (ppb)</td>
<td>$\mu g m^{-3}$ (ppb)</td>
<td>(%)</td>
</tr>
<tr>
<td>Ethane*</td>
<td>0.025 (0.020)</td>
<td>0.024 (0.019)</td>
<td>15</td>
</tr>
<tr>
<td>Ethylene*</td>
<td>0.023 (0.020)</td>
<td>0.024 (0.021)</td>
<td>15</td>
</tr>
<tr>
<td>Propane*</td>
<td>0.037 (0.020)</td>
<td>0.024 (0.013)</td>
<td>15</td>
</tr>
<tr>
<td>Propene*</td>
<td>0.035 (0.020)</td>
<td>0.024 (0.014)</td>
<td>15</td>
</tr>
<tr>
<td>Iso-butane*</td>
<td>0.048 (0.020)</td>
<td>0.024 (0.014)</td>
<td>15</td>
</tr>
<tr>
<td>N-butane*</td>
<td>0.048 (0.020)</td>
<td>0.024 (0.010)</td>
<td>15</td>
</tr>
<tr>
<td>Acetylene*</td>
<td>0.022 (0.020)</td>
<td>0.024 (0.022)</td>
<td>15</td>
</tr>
<tr>
<td>Iso-pentane*</td>
<td>0.060 (0.020)</td>
<td>0.024 (0.008)</td>
<td>15</td>
</tr>
<tr>
<td>N-pentane*</td>
<td>0.060 (0.020)</td>
<td>0.024 (0.008)</td>
<td>15</td>
</tr>
<tr>
<td>N-hexane*</td>
<td>0.013 (0.004)</td>
<td>0.013 (0.004)</td>
<td>15</td>
</tr>
<tr>
<td>Isoprene*</td>
<td>0.024 (0.008)</td>
<td>0.024 (0.008)</td>
<td>20</td>
</tr>
<tr>
<td>Benzene**</td>
<td>0.071 (0.022)</td>
<td>0.071 (0.022)</td>
<td>20</td>
</tr>
<tr>
<td>Toluene**</td>
<td>0.240 (0.063)</td>
<td>0.240 (0.063)</td>
<td>20</td>
</tr>
<tr>
<td>Xylenes + C$_8$**</td>
<td>0.259 (0.059)</td>
<td>0.259 (0.059)</td>
<td>20</td>
</tr>
<tr>
<td>Methanol**</td>
<td>0.317 (0.238)</td>
<td>0.330 (0.248)</td>
<td>20</td>
</tr>
<tr>
<td>Acetonitrile**</td>
<td>0.068 (0.040)</td>
<td>0.084 (0.049)</td>
<td>20</td>
</tr>
<tr>
<td>Acetaldehyde**</td>
<td>0.167 (0.091)</td>
<td>0.167 (0.091)</td>
<td>20</td>
</tr>
<tr>
<td>Acetone**</td>
<td>0.092 (0.038)</td>
<td>0.118 (0.049)</td>
<td>20</td>
</tr>
<tr>
<td>MVK$^1$ + MACR$^2$ + ISOPOOHs$^3$**</td>
<td>0.020 (0.007)</td>
<td>0.020 (0.007)</td>
<td>20</td>
</tr>
<tr>
<td>$\Sigma$ VOC</td>
<td>1.629 (0.749)</td>
<td>1.542 (0.723)</td>
<td>20</td>
</tr>
</tbody>
</table>

1 MVK = Methylvinylketone
2 MACR = Methacrolein
3 ISOPOOHs = Isoprene hydroxy hydroperoxides (Rivera-Rios et al., 2014)

* Hydrocarbons measured using a GC-FID by LSCE (MEGAPOLI, LHVP) and AIRPARIF (FRANCIPOI, “Les Halles” subway station).
** Masses measured using a PTR-MS by LCP (MEGAPOLI, LHVP) and LSCE (FRANCIPOI, LHVP).
Figure A.1. Agreement between total predicted and observed VOC concentrations based on the 6-factor PMF solution.

\[
\begin{align*}
    a &= 0.933 \pm 0.004 \\
    b &= 2.033 \pm 0.162 \\
    R &= 0.95 \\
    R^2 &= 0.91
\end{align*}
\]