## Harmonized methodology for the source apportionment of PM in France using EPA-PMF5.0 with constraints

# Presented by Dalia SALAMEH



## Presentation of the SOURCES project

- ✓ Context, objectives
- ✓ Spatial distribution and characteristics of the sampling sites
- ✓ PM source apportionment methodology
  - Selection of input variables
  - Uncertainty estimation of the variables
  - > PMF5.0 with **constraints**: target species in factor profiles

## Ongoing works (innovative methods)

- ✓ Analysis of **new tracers** (e.g. nitrocatechols, cellulose, BSOA,...)
- Coupled methodologies (with PMF): N isotopes; <sup>14</sup>C; online AE-33; back trajectories; oxidative potential (OP) of PM

## **SOURCES** project

## Objectives of SOURCES project (2015-2018)

### Determination of a standard and harmonized methodology for quantifying PM sources at different French urban environments using EPA-PMF5.0 with constraints



- PM sampling sites (n=20):
- 12 urban (yellow mark)
- 2 traffic: Roubaix, Strasbourg
- 3 rural: Revin, Peyrusse, OPE ANDRA
- 3 alpine valleys: Passy, Marnaz, Chamonix
- Generally, 24h PM samples were collected every third day (at least 120 filters/year)

#### ✓ Detailed PM chemical speciation:

- OC and EC
- lons (Cl<sup>-</sup>, NO<sub>3</sub><sup>-</sup>, SO<sub>4</sub><sup>2-</sup>, Na<sup>+</sup>, NH<sub>4</sub><sup>+</sup>, K<sup>+</sup>, Mg<sup>2+</sup>, Ca<sup>2+</sup>)
- Metals/trace elements (Al, Ca, Fe, K, As, Ba, Cd, Co, Cu, La, Mn, Mo, Ni, Pb, Rb, Sb, Sr, V)
- Common organic markers: levoglucosan, mannosan, galactosan, arabitol, mannitol, sorbitol, MSA, oxalate

## **SOURCES** project

## Objectives of SOURCES project (2015-2018)

### Determination of a standard and harmonized methodology for quantifying PM sources at different French urban environments using EPA-PMF5.0 with constraints

**1.** Homogeneous and harmonized PMF pre-treatment of PM comprehensive chemical dataset (OC, EC, ions, metals, and organic markers) established at various urban environments in France:

- Selection of input variables
- Estimation of the uncertainties

**2.** Integration of an homogeneous and minimal set of specific chemical constraints to the factor profiles based on external knowledge:

- Improve separation of correlating sources
- "Cleaner" source profiles and better estimation of their contributions

**3.** Geographical origin of main PM sources (PSCF approach: associating PMF temporal contributions with air mass back trajectories)

**4.** Integration of the resolved source profiles with constrained PMF approach at the different studied sites into SPECIEUROPE database

## **1. Selection of input variables**

- Classic PMF inputs: OC, EC, and inorganic components, i.e. metals (Al, Ca, Fe, Ti, V, Ni, Cu, Zn, As, Rb, Pb, Cd, Sn, Sb), and major ions (NH<sub>4</sub><sup>+</sup>, NO<sub>3</sub><sup>-</sup>, SO<sub>4</sub><sup>2-</sup>, Cl<sup>-</sup>, Na<sup>+</sup>, K<sup>+</sup>, Mg<sup>2+</sup>)
- Very few studies involving organic markers
- **Extensive PMF input data matrix**: Classic PMF inputs + additional **organic markers**:
  - Levoglucosan (biomass burning)
  - **Polyols** (sum of arabitol, mannitol, sorbitol; primary soil biogenic)
  - **MSA** (marine biogenic, phytoplankton?)
  - **Oxalate** (secondary organic indicator)
  - **PAH** (combustion processes)
  - **Hopanes** (fossil fuel combustion, e.g. vehicular emissions)
  - Lignin pyrolysis products (vanillin, coniferaldehyde, vanillic acid...)

Input variables were selected based on the percentage of values above the detection limit (DL) and the signal-to-noise (S/N) ratios (focus on common species)

### 2. Estimation of the uncertainties (literature)

- No standardized methodology is supplied for the treatment of uncertainties
- Commonly used methods (JRC report, 2013):
  - Polissar et al. (1998) set the uncertainty of values below the detection limit to 5/6 of the detection limit, while the uncertainty of missing values is set at four times the geometric mean
  - Gianini et al. (2012) and adapted from Anttila et al. (1995). It uses the detection limit (DL, twice of the standard deviation of the field blanks) and the coefficient of variation (CV, standard deviation of repeated analysis divided by the mean value of the repeated analysis).

## 2. Estimation of the uncertainties (LGGE, Waked et al., 2014)

- Previous tests for the uncertainties assessments were performed at LGGE by Waked et al. (2014) (Lens dataset 2011-2012)

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Uncertainty calculation methodology	Gianini methodology for all the species	Polissar methodology for all the species	Gianini methodology for all the species except for trace & metal elements where the relative uncertainty was used instead of CV	Polissar methodology for all the species except for trace & metal elements where the relative uncertainty was used instead of CV
Number of Factors	9	9	9	9
Scaled residuals	For many species, scaled residuals were not within the range of $-3$ and $+3$ the standard deviation	For many species, scaled residuals were not within the range of $-3$ and $+3$ the standard deviation	Set between $-3$ and $+3$ the standard deviation	Set between $-3$ and $+3$ the standard deviation
Bootstrap results	For some factors, just 58 % of the runs are correlated between base run and the bootstrap runs	For some factors, just 66 % of the runs are correlated between base run and the bootstrap runs	Good correlation between factors: more than 92 % of the runs are correlated between the base run and the bootstrap runs	Good correlation between factors: more than 92 % of the runs are correlated between the base run and the bootstrap runs

Simulation with the Gianini methodology and the relative uncertainty for trace elements was selected

## 2. Estimation of the uncertainties (SOURCES project)

- **Trial and error tests** to define a common methodology for the estimation of data uncertainty of all the species (OC, EC, ions, metals/trace elements and organic markers)
- **Current tests** for the optimization of uncertainty estimation using Gianini methodology, 3 sites were chosen: urban, traffic and rural

$$\sqrt{(DL)^2 + (x_{ij} \times CV)^2 + (x_{ij} \times \boldsymbol{a})^2}$$

- <u>Objective</u>: Define a variation range for the a coefficient depending on the type of analysis (a=0.03 by default)
- Evaluation of different statistical parameters (best model fit):
  - ✓ Signal-to-Noise (S/N) ratios
  - ✓ Variation of  $Q_{true}$ -to- $Q_{robust}$  ratios
  - ✓ Coefficients of determination (R<sup>2</sup>)
  - ✓ Bootstrap and DISP results
  - ✓ Interpretability of the obtained factor profiles
  - ✓ Distribution of scaled residuals

### 3. EPA-PMF5.0 with constraints

- PMF results are generally affected by co-linearity induced by processes other than coemissions (e.g. seasonality, meteorological parameters), providing mixed factors
- To minimize the influence of mixing between factors, additional environmentally and meaningful chemical constraints can be imposed in the factor profiles

### **Objective:**

- Define and apply a set of **minimal constraints** that are able to provide optimal results across the different studied sites
- Generally, the use of constraints allows obtaining:
  - Better separation of the factors with more "cleaner" source profiles
  - Better estimation of the source contributions
  - Better bootstrap results

### 3. EPA-PMF5.0 with constraints

# Minimum set of specific and plausible chemical constraints imposed to elements in factor profiles mostly identified in recent PMF studies in France

(e.g. CAMERA,	, Part'Aera,	Decombio,	etc.,)
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PMF factors	Levo /Manno	Lignin products	PAH	Hopanes	Polyols	MSA	EC	CI/Na, K/Na, Ca/Na, Mg/Na (0.119)
Biomass burning	Pull up maximally	Pull up maximally						
Vehicularexhaust	= 0	= 0		Pull up maximally				
Biogenic emissions	= 0	= 0	= 0	= 0	Pull up maximally		Pull down maximally	
Marine biogenic	= 0	= 0	= 0	= 0	Pull down maximally	Pull up maximally	Pull down maximally	
HFO combustion	= 0	= 0			= 0	= 0		
Industries								
Coal burning					= 0	= 0		
Seasalt								Ratios sea salt
Aged seasalt								
Mineral dust								
Road dust								
Nitrate rich								
Ammonium rich								
Secondary organics/ Secondary processes								

#### EPA-PMF5.0 with constraints 3.

Minimum set of specific and plausible chemical constraints imposed to elements in factor profiles mostly identified in recent PMF studies in France

(e.g. CAMERA, Part'Aera, Decombio, etc.,)

#### Sampling site: Lens (2011-2012), IE

100 **Biogenic emissions** Constrained run Base run 80 % of species 60 40 20 0 NO3 NH4 S04 ₽ Z **PM10** C29 C33  $\overline{O}$ Levoglucosan Hopane S-Homohopane R-Homohopane C31 sum Polyols mannosan Norhopane ဗ  $\Sigma$ Polyols: pull up maximally EC: pull down maximally

Biogenic emissions factor: Constrained vs. base run

- Bootstrap from 95 to 100%
- Contribution of this factor increased from 13 to 16%

### Measurements of new tracers

#### Cellulose ( $C_6H_{10}O_5$ )<sub>n</sub> (LGGE, Picot P)

- Following the procedure of Kunit and Puxbaum (1996):
  - Double enzymatic hydrolysis
  - Analysis of glucose (=analyte) by HPLC-PAD
- Grenoble, University campus (May-July 2015)
  - OC: 4.9 µgC/m<sup>3</sup> (2.42-8.08 µgC/m<sup>3</sup>)
  - Cellulose: 108 ngC/m<sup>3</sup> (25-447 ngC/m<sup>3</sup>)  $\rightarrow \rightarrow 2\%$  of OC on avg.
- Analysis of samples from Estonia (collaboration with PSI)



- Biomass burning SOA tracer (*m*-cresol)
- GC/MS after derivatization



#### Alpine valleys:

- Lanslebourg (winter): 26 ng/m<sup>3</sup> (0-85 ng/m<sup>3</sup>)
- Passy (winter): 20 ng/m<sup>3</sup> (0-76 ng/m<sup>3</sup>)



MBTCA: 3-methyl-1,2,3-butanetricarboxylic acid

- Formed from the oxidation of  $\alpha$ -pinene
- More highly functionalized than traditional SOA markers such as pinonic and pinic acid
- Muller et al. (2012): MBTCA explains about 10% of the newly formed SOA mass (experimental yield about 0.6 %)

### Coupled methodologies in development

- > N isotopes with PMF (LGGE, S Weber et J Savarino)
- Use of Nitrogen isotope ratios to elucidate the primary sources of ammoniac

### 1. Monte Carlo simulation (stochastic model)

Determination of NH4<sub>bio</sub>; NH4<sub>agr</sub>; and NH4<sub>veh</sub>

### 2. PMF analysis

Input data matrix combining PM chemical measurements (e.g. OC; EC; ions; metals) and isotopy (NH4<sub>bio</sub>; NH4<sub>agr</sub>; and NH4<sub>veh</sub>)



Sampling site : OPE ANDRA, rural site, 2013 PM and N isotopes data



- Ammonium fractions were relevantly apportioned to their corresponding sources, i.e. nitrate rich (90% of NH4<sup>+</sup><sub>agr</sub>); biomass (70% of NH4<sup>+</sup><sub>bio</sub>); and Industry/traffic (50% of NH4<sup>+</sup><sub>veh</sub> mass)
- Total ammonium concentrations were well reconstructed by the model

### Other coupled methodologies in development

### Radiocarbon <sup>14</sup>C (E Bard, CEREGE) > Online AE-33 measurements and PMF



- ✓ High correlation coefficients for Passy (0.995) and Chamonix (0.989)
- ✓ Non fossil fraction represents between 80-90% of total carbon for winter samples ↔↔ dominance of biomass burning
- ✓ Coupling <sup>14</sup>C with ME-2 (*PhD A Sylvestre, LCE Marseille*)

#### FAIRMODE technical meeting, 27-29 June 2016

#### Chamonix, Alpine site, 2013-2014

#### Use of BCff and BCwb variables instead of EC



 Relevant geochemical distribution of BCwb and BCff

- Other coupled methodologies in development
  - PSCF (Potential Source Contribution Function) approach

- PSCF represents the probability that an air parcel may be responsible for high concentrations measured at the receptor site
- Allows geographical identification of potential emission areas by associating temporal contributions of PMF factors with back trajectories (Ashbaugh et al., 1985)



### Other coupled methodologies in development

> Measurements of the oxidative potential (OP) of PM (LGGE, Uzu G)

OP: capacity of Particulate Matter (PM) to oxidize target molecules - Different assays exist for measuring OP

- Trying to relate the variability of OP to chemistry and emission sources
- OP already measured on 5 yearly series of SOURCES (PhD A Calas, LGGE)



#### Results from Passy (A-cellular assays; ascorbate acid depletion (AA))

Calas et al., 2016 (in prep)

- PM10 concentrations vary by a factor of 4 (annual avg.)
- OP AA vary by a factor of 12 (annual avg.)

## **Conclusions, Open questions**

#### **Objective:**

Determination of a standard and harmonized methodology for quantifying PM sources at different urban French environments using EPA-PMF5.0 with constraints



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- O Favez (INERIS)

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- F Chevrier (PhD LGGE / LCME)
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#### New tracers

- Nitrocatechols :
  - JL Besombes (LCME)
- Cellulose and MBTCA :
  - V Jacob (LGGE)
  - B Golly (LGGE)

### **Oxidative potential of PM**

- Funding : CNRS LEFE
- G Uzu (LTHE / LGGE)
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# Thank you for your attention











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