

Hotel Therme Zurzach, 11.04.2016 – 15.04.2016

# V. SoFi workshop

## General information



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# General - Logistics

## ➤ **Food**

- Morning coffee break (~10.30)
- Standing lunch (12.30 – 13.30)
- Afternoon coffee break (~15.00)

➔ Social dinner (Wednesday evening)

➤ **Internet (Free Internet, check whiteboard)**

➤ **Restrooms (downstairs)**

# General - Logistics

- **Information / presentations / data / code**

- <http://indico.psi.ch/event/SoFi2016>

- **Current policy with SoFi**

- Collaboration for 2 publications per institute (F. Canonaco, A. Prevot and PSI people supporting your analysis during the workshop)
- Cite the SoFi paper in AMT (Canonaco et al 2013)

# General – Program

Monday (Pre-Workshop)	
Time	Activity
10.30 – 12.30	<ul style="list-style-type: none"><li>➤ <i>General support</i> for communication between SoFi and ME-2, HDF option in Igor</li><li>➤ <i>Theory input</i> on PMF, ME-2, Q-space, robust mode, rotational tools (a-value, fpeak, pulling)</li></ul>
12.30 – 13.30	*****Lunch*****
13.30 – 15.00	<ul style="list-style-type: none"><li>➤ <i>Interactive discussion</i> using ACSM data in SoFi to better visualize the options/features present in SoFi (import raw data, treat data for PMF run, call PMF, import results in igor for SoFi, explore results)</li></ul>
15.00 – 15.30	***Coffee break***
15.30 – 17.00	<ul style="list-style-type: none"><li>➤ <i>Interactive discussion</i> using ACSM data in SoFi to better visualize the options/features present in SoFi (import raw data, treat data for PMF run, call PMF, import results in igor for SoFi, explore results)</li></ul>

# General – Program

Tuesday (Official kick-off)	
Time	Activity
09.00 – 10.30	➤ <i>Theory input</i> on rotational ambiguity, criteria-based approach , propagation of statistical uncertainty, AuRo-SoFi
10.30 – 11.00	***Coffee break***
11.00 – 12.30	<ul style="list-style-type: none"> <li>➤ <i>Theory input</i> on rotational ambiguity, criteria-based approach , propagation of statistical uncertainty, AuRo-SoFi</li> <li>➤ <i>Practical example:</i> Application of SoFi on year-long ACSM data</li> </ul>
12.30 – 13.30	*****Lunch*****
13.30 – 15.00	<ul style="list-style-type: none"> <li>➤ <i>Group discussions:</i> Users treating similar data, e.g. filter-based, offline, UMR-AMS, HR-AMS, combined datasets have the possibility to share gained experience</li> <li>➤ <i>Individual work:</i> participants work on their own data (support provided)</li> </ul>
15.00 – 15.30	***Coffee break***
15.30 – 17.00	<ul style="list-style-type: none"> <li>➤ <i>Group discussions:</i> Users treating similar data, e.g. filter-based, offline, UMR-AMS, HR-AMS, combined datasets have the possibility to share gained experience</li> <li>➤ <i>Individual work:</i> participants work on their own data (support provided)</li> </ul>

# General – Program

Wednesday	
Time	Activity
09.00 – 10.30	➤ <i>Individual work</i> : participants work on their own data (support provided)
10.30 – 11.00	***Coffee break***
11.00 – 12.30	➤ <i>Individual work</i> : participants work on their own data (support provided)
12.30 – 13.30	*****Lunch*****
13.30 – 15.00	➤ <i>Presentations of case studies</i> : source apportionment (SA) studies conducted with SoFi from experienced users (PSI and non-PSI)
15.00 – 15.30	***Coffee break***
15.30 – 17.00	➤ <i>Presentations of case studies</i> : source apportionment (SA) studies conducted with SoFi from experienced users (PSI and non-PSI)
*****Social dinner*****	

# General – Program

Thursday	
<u>Time</u>	<u>Activity</u>
09.00 – 10.30	<ul style="list-style-type: none"><li>➤ <i>Individual work</i>: participants work on their own data (support provided)</li><li>➤ <i>Presentations of participants</i></li></ul>
10.30 – 11.00	***Coffee break***
11.00 – 12.30	<ul style="list-style-type: none"><li>➤ <i>Presentations of participants</i></li></ul> <b>Conclusion of SoFi workshop</b>
12.30 – 13.30	*****Lunch*****
13.30 – 15.00	<b>Start of ACTRIS meeting</b> <ul style="list-style-type: none"><li>➤ <i>ACTRIS-related discussions</i></li></ul>
15.00 – 15.30	***Coffee break***
15.30 – 17.00	<ul style="list-style-type: none"><li>➤ <i>ACTRIS-related discussions</i></li></ul>

Friday	
<u>Time</u>	<u>Activity</u>
09.00 – 17.30	<ul style="list-style-type: none"><li>➤ <i>If wished/needed, further discussion at PSI with PSI people at PSI (please announce this during the workshop)</i></li></ul>

# V. SoFi workshop

## **PMF - general**

### **Key words:**

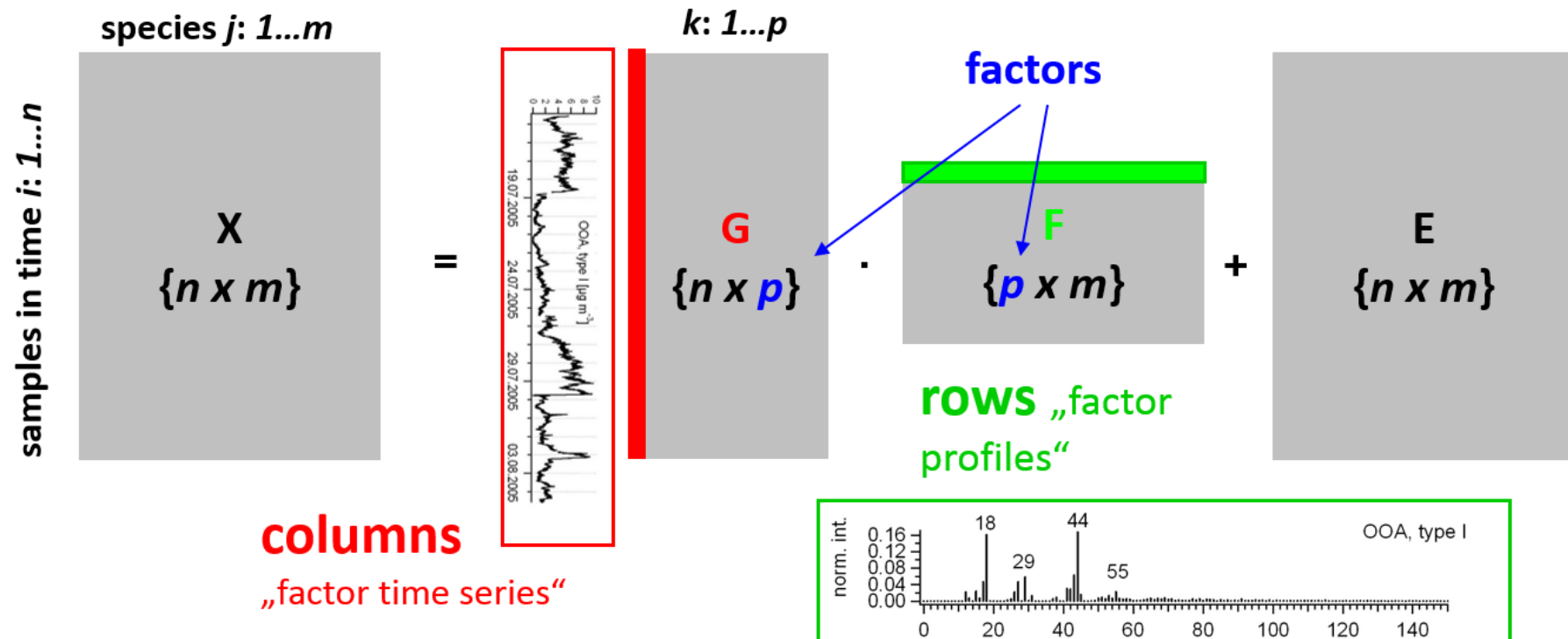
PMF, CMB, PMF2, ME-2, Q-space, robust mode, seed runs, local/global minima, rotational ambiguity / uncertainty



# Model – Positive Matrix Factorization (PMF)

## ➤ Bilinear factor analytic algorithm

$$X_{measured} = \hat{X}_{model} + E_{model}$$



Paatero 1994

# Model – Positive Matrix Factorization (PMF)

## ➤ Least-squares problem

$$Q = \sum_{i=1}^m \sum_{j=1}^n \left( \frac{e_{ij}}{\sigma_{ij}} \right)^2$$

$e_{ij}$ : difference (measured – model)  
 $\sigma_{ij}$ : uncertainty (statistical error)

## ➤ Q will be minimized with respect to all model variables

- ME-2 starts the conjugate gradient algorithm for solving this task

## ➤ Goal

- Factor solution must be environmentally reasonable
- Unstructured residuals over time (ts, diurnals, etc.) and over profile (variables)

*Paatero 1999*

# Model - Q-space

## ➤ **Real case**

- ACSM data with 100 variables for 1000 scans, four factors, unconstrained
- $G\{n \times p\}$ ,  $F\{p \times m\} \rightarrow G\{1000 \times 4\}$ ,  $F\{4 \times 100\}$ , there are 4400 model variables
  - ➔  $Q(4400 \text{ model variables})$ , multidimensional Q-space

## ➤ **Simplified case**

- Simply the real case with two model variables
  - ➔  $Q(2 \text{ model variables})$ , three dimensional Q-space

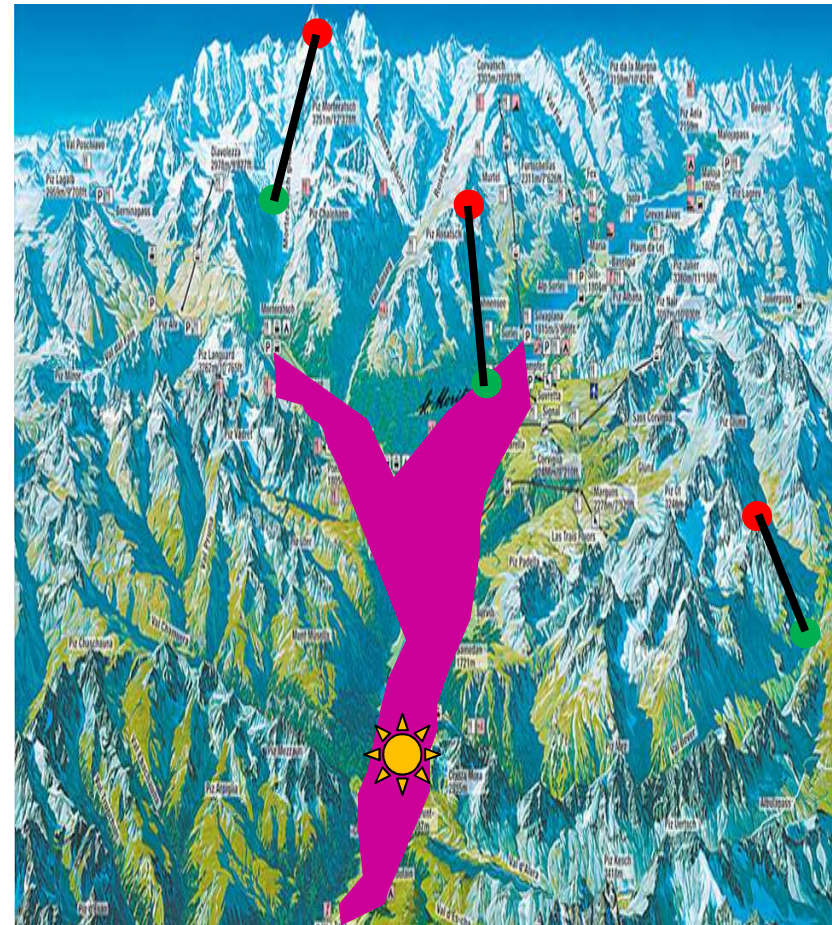
# Model - Q-space

- ❑ Q(2 model variables) similar to the height  $h(x,y)$  in the map
- ❑ PMF is performed through the conjugate gradient algorithm minimizing Q based on the starting conditions, following the steepest descent (from red to the green)
- ❑ Goal is to find the smallest possible Q-value (global minimum) (violet area) together with the best solution ☀

Search for this minimum based on different starting values (seed run)

- ❑ There are many points on the map, for which  $h(x,y)$  is equal → rotational ambiguity

Explore the rotational ambiguity with proper techniques (fpeak, ind. fpeak, a-value, pulling)



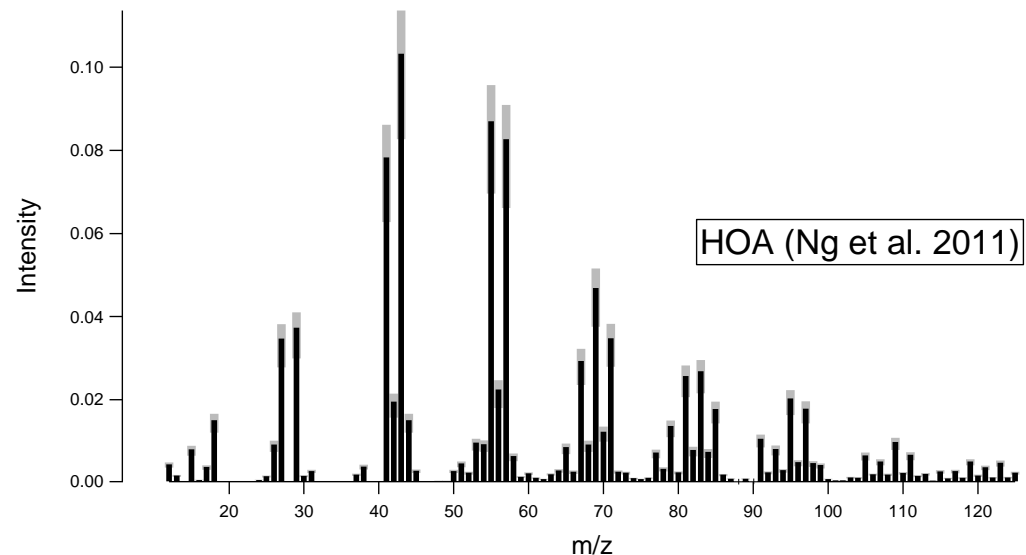
# Solution space – a-value

## ➤ Assess rotational ambiguity

### ■ a-value technique

- Full Q-space can potentially be investigated
- Advantage: easy to perform and computationally inexpensive
- Disadvantage: Sensitivity analysis on the constrained model variables

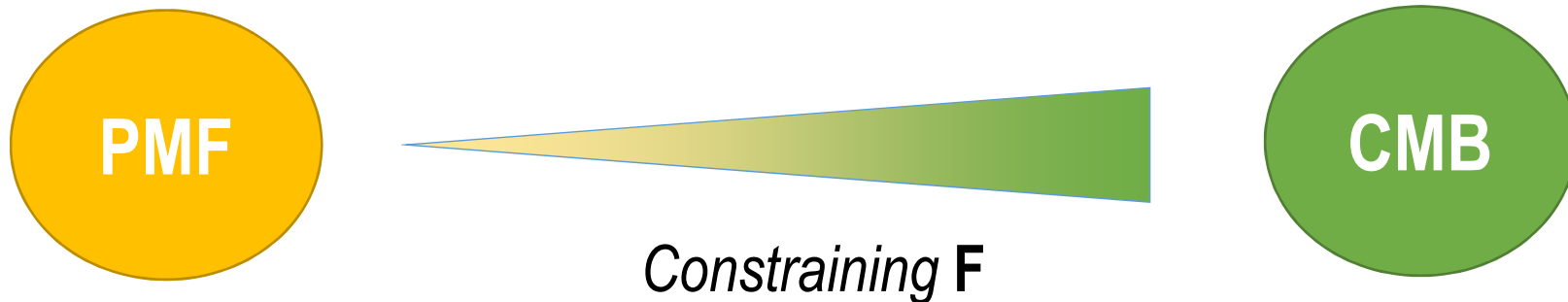
$$f_{p,j,\text{solution}} = f_{p,j} \pm a \cdot f_{p,j}$$



*Paatero 1999/2008*

# Model – PMF/CMB/solvers

## ➤ PMF / CMB (chemical mass balance) approach



## ➤ Solvers

Solver	Unconstrained	Constrained	Communication
PMF2 / PMF3	X	only to zero	Limited
ME-2	X	X	All quantities easily accessible

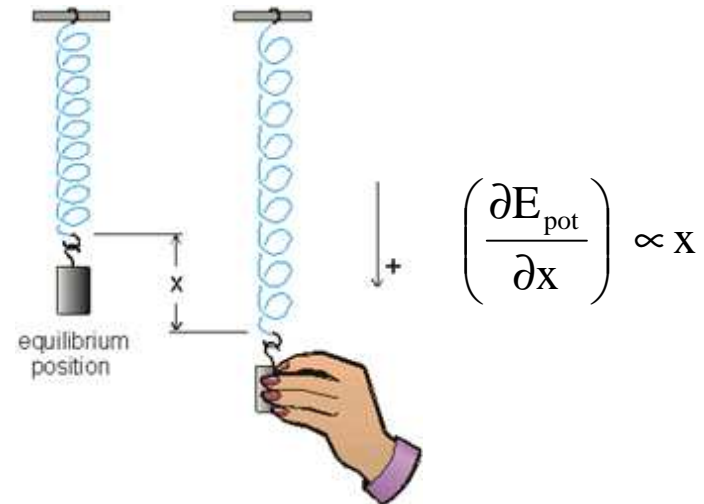
*Paatero 1999*

# Model – Positive Matrix Factorization (PMF)

## ➤ Advantages

- Values in **G** & **F** are non-negative
- Factors represent sources / processes
- PMF algorithm scales with the residual

$$\left( \frac{\partial Q_{ij}}{\partial e_{ij}} \right) \propto e_{ij}$$



*Paatero 1994*

# Model – robust mode

## ➤ **PMF run (non-robust mode)**

- Computational power is proportional to the residual (in theory ideal)
- Outliers, e.g. transient sources, wrong nb. of factors, electronic recording issues, etc. violate this relation and PMF could spend more time, reducing “wrong” residuals

## ➤ **PMF run (robust mode)**

- Allow for this dependency only in a certain range and damp afterwards (robust mode, default value = 4)

$$\text{if } \left| \frac{e_{ij}}{\sigma_{ij}} \right| \leq 4 \Rightarrow \left( \frac{\partial Q_{ij}}{\partial e_{ij}} \right) \propto e_{ij} \quad \text{else } \left| \frac{e_{ij}}{\sigma_{ij}} \right| > 4 \Rightarrow \left( \frac{\partial Q_{ij}}{\partial e_{ij}} \right) \propto 4$$

*Paatero 1997*



# Model – Positive Matrix Factorization (PMF)

## ➤ Disadvantages

- Assess number of factors
- Constant factor profiles (mass spectra)
- Uncertainties are not fully defined, minimal Q-value is not necessarily the best solution
  - ➔ Investigate the solution space even for slightly higher Q-values (few %)
- Bilinear factor analytic models suffer from rotational ambiguity

$$\mathbf{X}_{\text{model}} = \mathbf{G} \cdot \mathbf{F} = \mathbf{G} \cdot \mathbf{T} \cdot \mathbf{T}^{-1} \cdot \mathbf{F} = \mathbf{G}' \cdot \mathbf{F}'$$

- ➔ Investigate the solution space

*Paatero 1994/97*

# Model – Positive Matrix Factorization (PMF)

- **Weight Q by  $Q_{\text{exp}}$ , the remaining degrees of freedom**

$$Q_{\text{exp}} = n \cdot m - p \cdot (n + m) \sim n \cdot m$$

- If all residuals were similar as their  $\sigma$ 's,  $Q / Q_{\text{exp}} \sim 1$
- Monitor  $Q / Q_{\text{exp}}$  values → Too high values may indicate systematic problems of the PMF solution
- Monitor the changes of  $Q/Q_{\text{exp}}$  over various model runs

# V. SoFi workshop

## Tutorial – SoFi

### **Learning goal**

- Learn how to prepare the data for a PMF run in SoFi
- Learn how to import and look at various PMF results

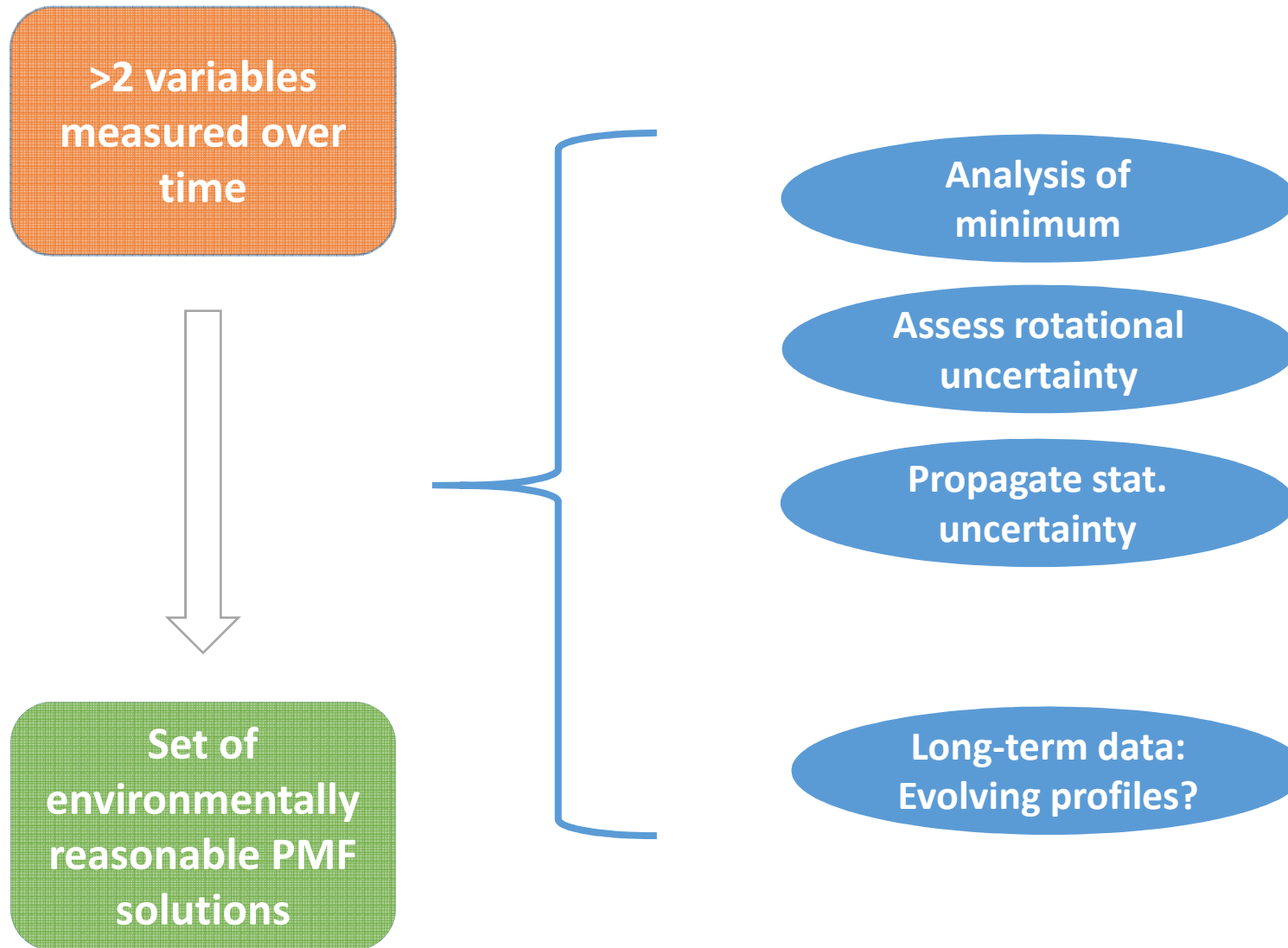
# V. SoFi workshop

## **PMF - advanced**

### **Key words:**

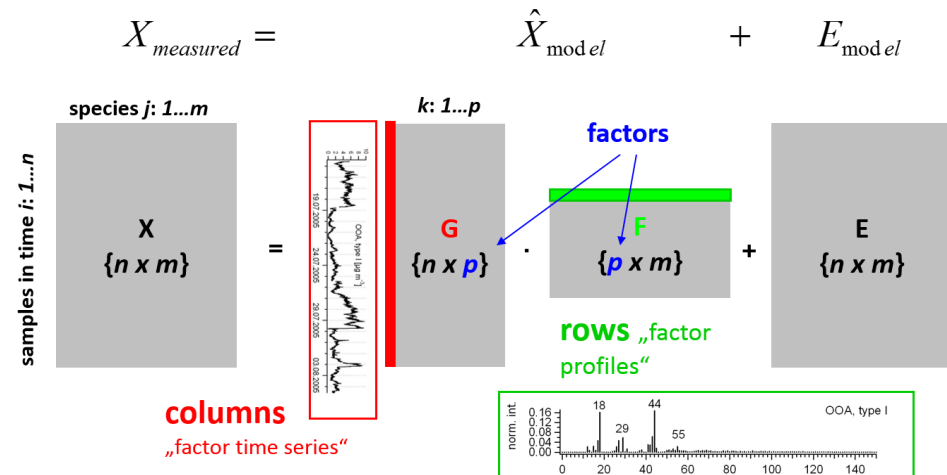
ME-2, validation of PMF solution, exploration of solution space (f<sub>peak</sub>, a-value, CMB-like approach), propagation of statistical uncertainty, AuRo-SoFi

# Model – Positive Matrix Factorization (PMF)



# Solution space – search for global minimum

## ➤ Bilinear factor model (PMF)



## ➤ Least-squares problem

$$Q = \sum_{i=1}^m \sum_{j=1}^n \left( \frac{e_{ij}}{\sigma_{ij}} \right)^2$$

$e_{ij}$ : difference (measured – model)  
 $\sigma_{ij}$ : uncertainty (statistical error)

## ➤ Seed runs

- Initialize PMF run with random values for the unconstrained model variables
- Search for the PMF solution(s) with the smallest possible Q-value (global minimum)
  - ➔ compare rotated solutions to this Q-value

Paatero 1994

# Solution space – rotational ambiguity

## ➤ **PMF solutions suffer from rotational ambiguity**

$$\mathbf{X}_{\text{model}} = \mathbf{G} \cdot \mathbf{F} = \mathbf{G} \cdot \mathbf{T} \cdot \mathbf{T}^{-1} \cdot \mathbf{F} = \mathbf{G}' \cdot \mathbf{F}'$$

## ➤ **Assess rotational ambiguity**

- Vary the model variables (f<sub>peak</sub>, individual f<sub>peak</sub>, a-value, CMB-like, pulling) and monitor the change of the PMF solution with various parameters:
  - I. Q-value
  - II. Residual (global / key variables)
  - III. Weighted residual (global / key variables)
  - IV. Shape of factor profile(s)
  - V. Time series / diurnal correlation with external tracers

*Paatero 2008*

# Solution space – global fpeak

## ➤ Assess rotational ambiguity

- Global fpeak ( $\phi$ ) technique
  - All rotations are performed at the same time
  - Advantage: easy to perform
  - Disadvantage: rotations cannot always be fully predicted, lower estimate of the rotational uncertainty
  - Example:

$$\bar{G} = GT \text{ and } \bar{F} = T^{-1}F \quad T_{\text{fpeak, p=3}} = \begin{bmatrix} 1 & \phi & \phi \\ \phi & 1 & \phi \\ \phi & \phi & 1 \end{bmatrix}$$

*Paatero 2008*



# Solution space – individual fpeak

## ➤ Assess rotational ambiguity

- Individual fpeak ( $\phi$ ) technique
  - All rotations are performed at the same time
  - Advantage: easy to perform
  - Disadvantage: rotations cannot always be fully predicted, lower estimate of the rotational uncertainty
  - Example:

$$\bar{G} = GT \text{ and } \bar{F} = T^{-1}F \quad T_{\rho=3} = \begin{bmatrix} 1 & 0 & \phi \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

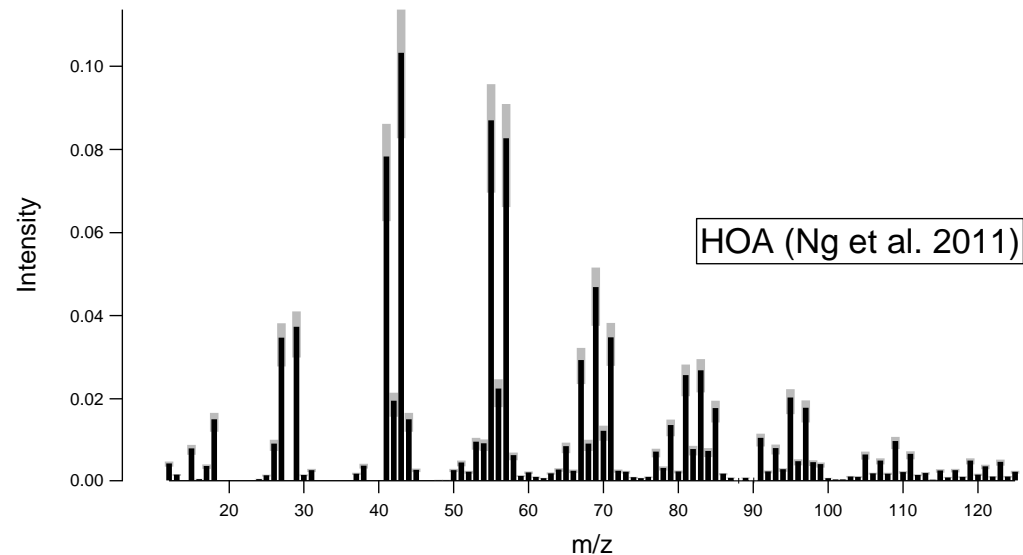
*Paatero 2008*

# Solution space – a-value

## ➤ Assess rotational ambiguity

- a-value technique
  - Full Q-space can potentially be investigated
  - Advantage: easy to perform and computationally inexpensive
  - Disadvantage: Sensitivity analysis on the constrained model variables

$$f_{p,j,\text{solution}} = f_{p,j} \pm a \cdot f_{p,j}$$



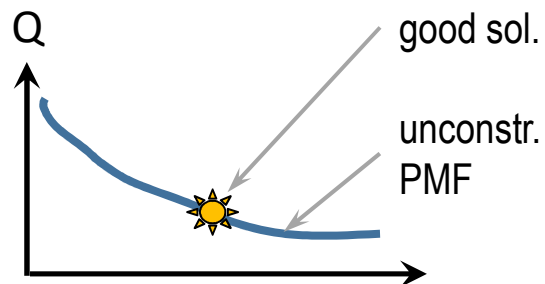
*Paatero 1999/2008*

# Solution space – a-value

## ➤ Assess rotational ambiguity

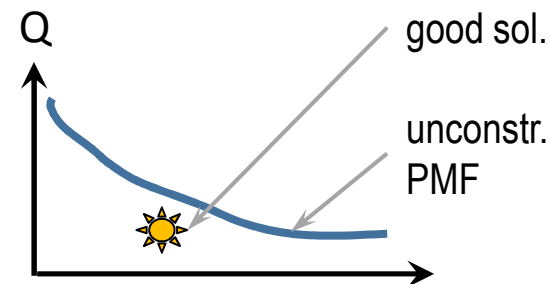
- a-value technique
  - Full Q-space can potentially be investigated
  - Advantage: easy to perform and computationally inexpensive
  - Disadvantage: Sensitivity analysis on the constrained model variables

### Good case



Sensitivity analysis performed on the constrained anchor meets/finds the good solution

### Bad case

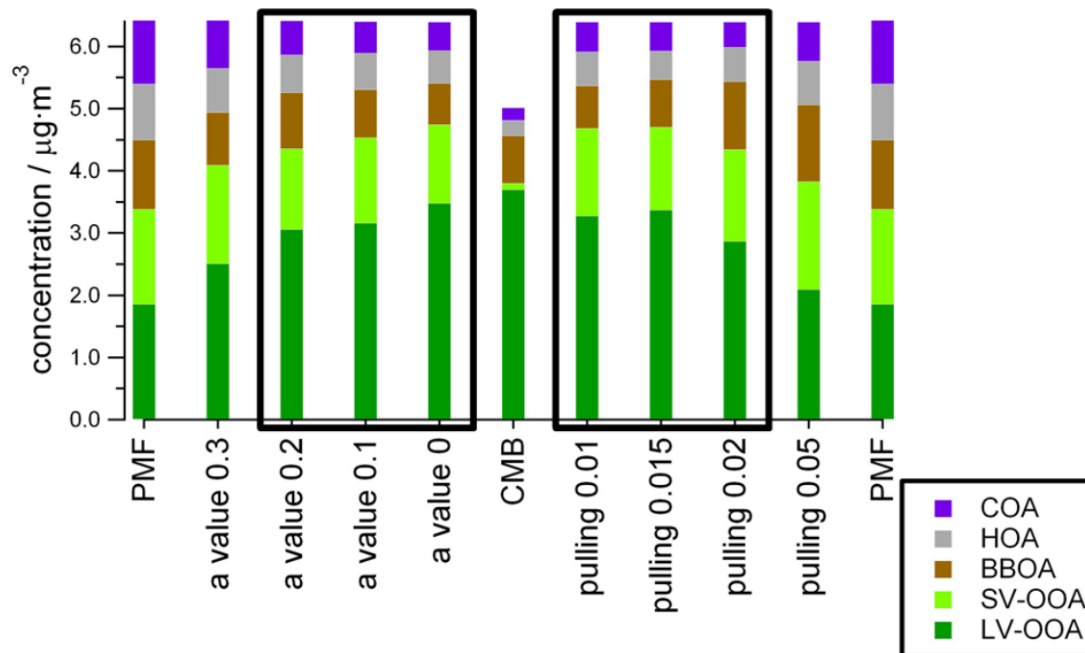


Sensitivity analysis performed on the constrained anchor does not find the good solution  
➔ change factor profile (**AMS Spectral Database**)  
(<http://cires.colorado.edu/jimenez-group/AMSsd/>)

# Solution space – a-value

## ➤ ACSM Zurich winter 2011

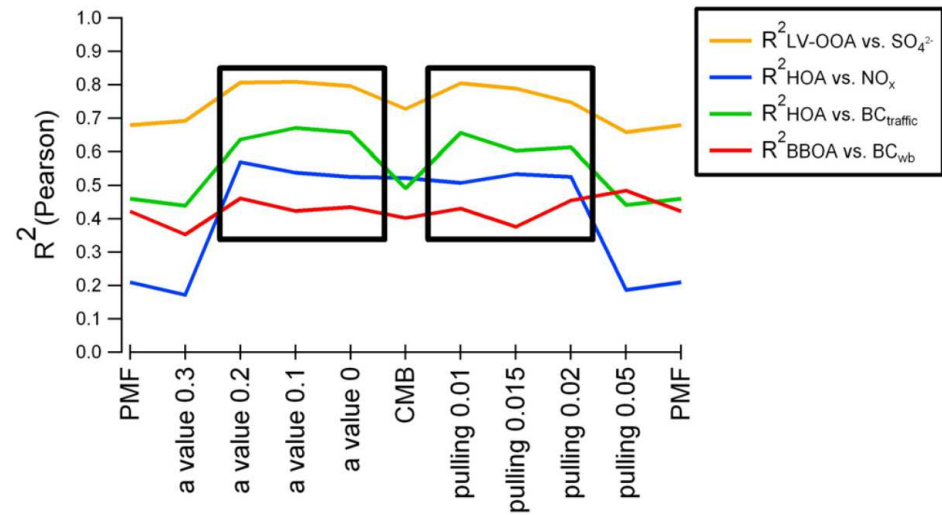
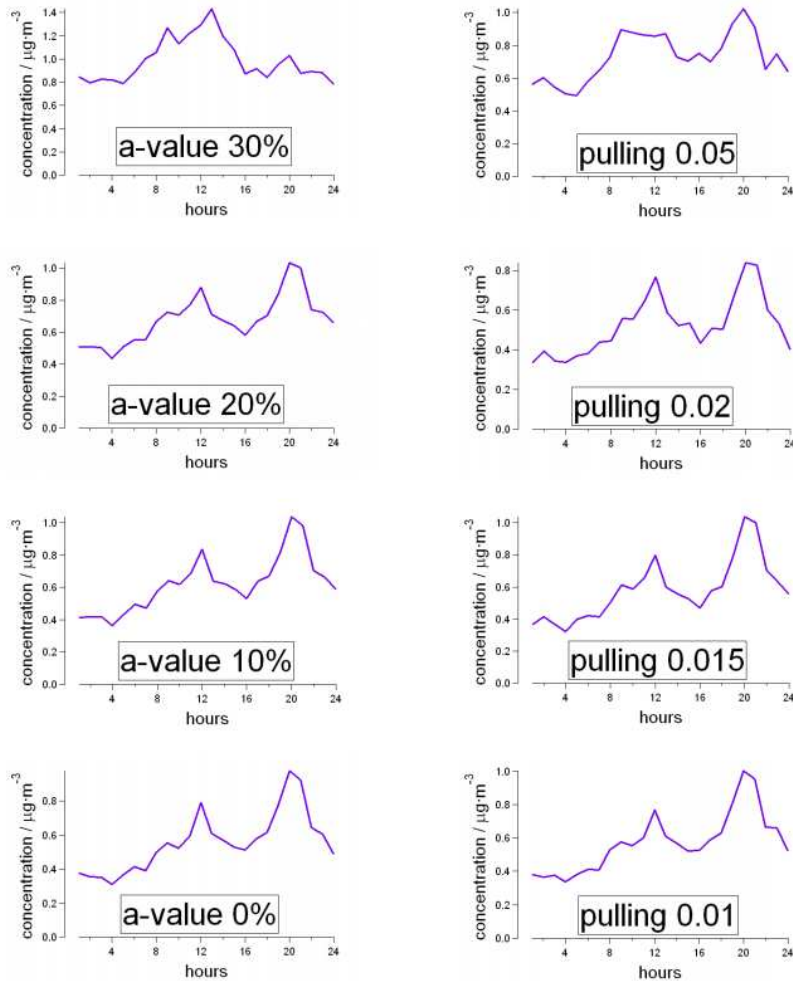
- Employed anchors (HOA, COA, BBOA) meet reasonable solutions for a-value range 0 – 0.2



*Canonaco et al. 2013*

# Solution space – a-value

## ➤ ACSM Zurich winter 2011

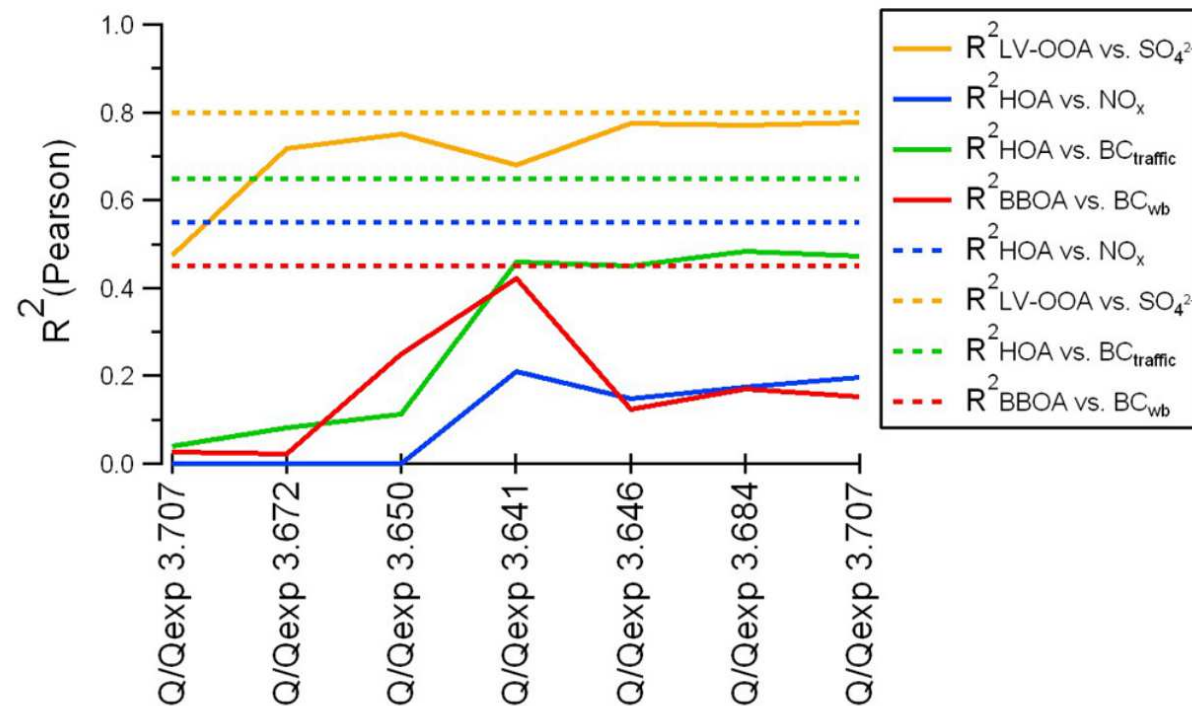


Canonaco et al. 2013

# Solution space – comparison to fpeak

## ➤ ACSM Zurich winter 2011

- Pure fpeak analysis (solid lines) do not reproduce the reasonable PMF solutions (dashed lines)



Canonaco et al. 2013

# Solution space – a-value space investigation

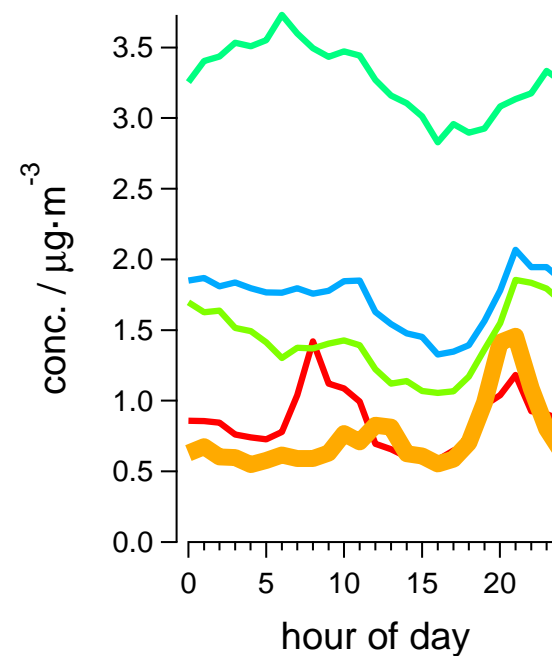
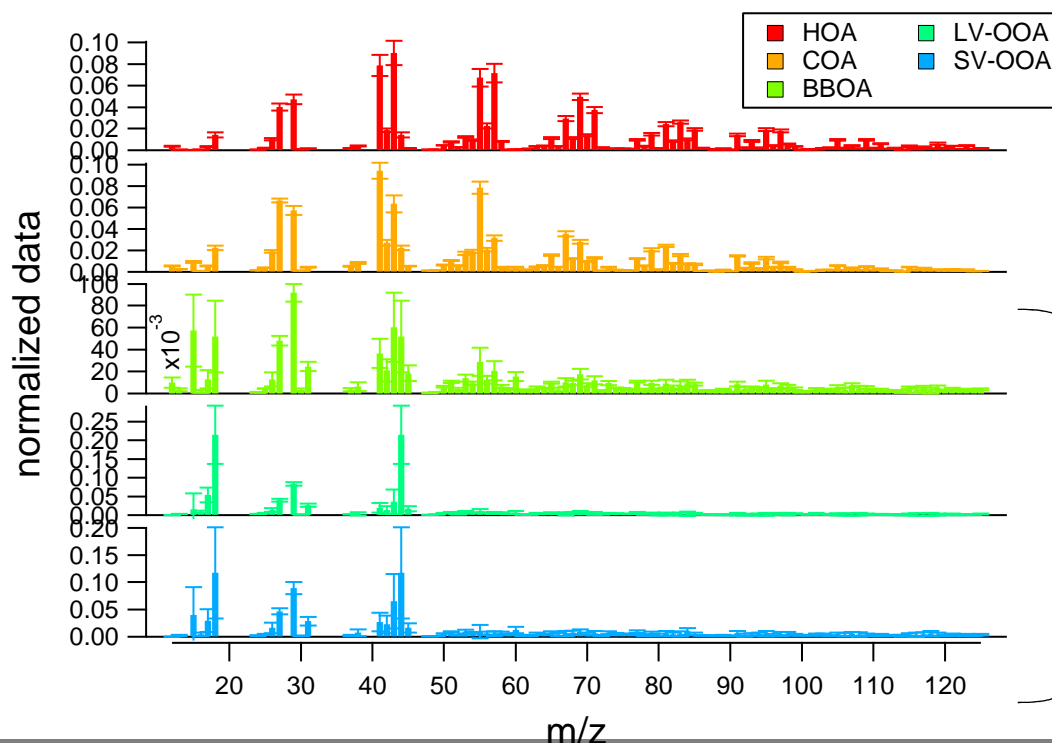
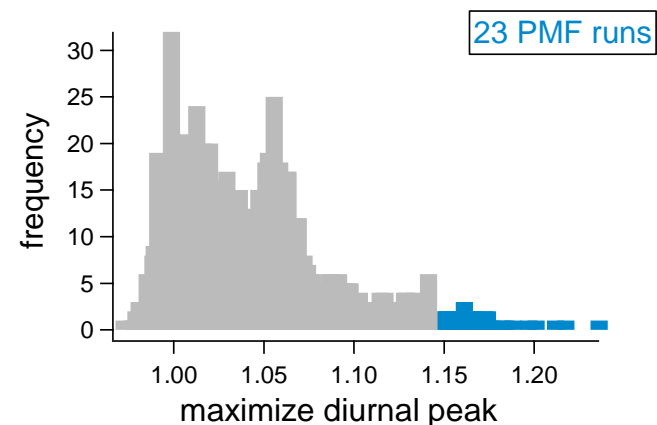
## ➤ Issue for a-value space for two and more constrained vectors

- Analyzing systematically all solutions becomes difficult
- Possible alternative approach
  - ➔ reorder all PMF solutions based on a list of possible criteria
  - ➔ dynamic change of the criteria/weight and inspection of the PMF solutions
- Package (criteria-based approach) ready to be shared for testing (free license for one year)
- More details presented later this morning (Yuliya Sosedova)

*Canonaco et al. 2013*

# Solution space – a-value space investigation

factors	criteria
constrained HOA	ts-correlation $\text{NO}_x$
constrained COA	<b>maximize diurnal peak at noon</b>
constrained BBOA	ts-correlation to $\text{BC}_{\text{wb}}$
LV-OOA	ts-correlation to sulfate
SV-OOA	ts-correlation to nitrate



Canonaco et al., in prep.



# Solution space – statistical error prop.

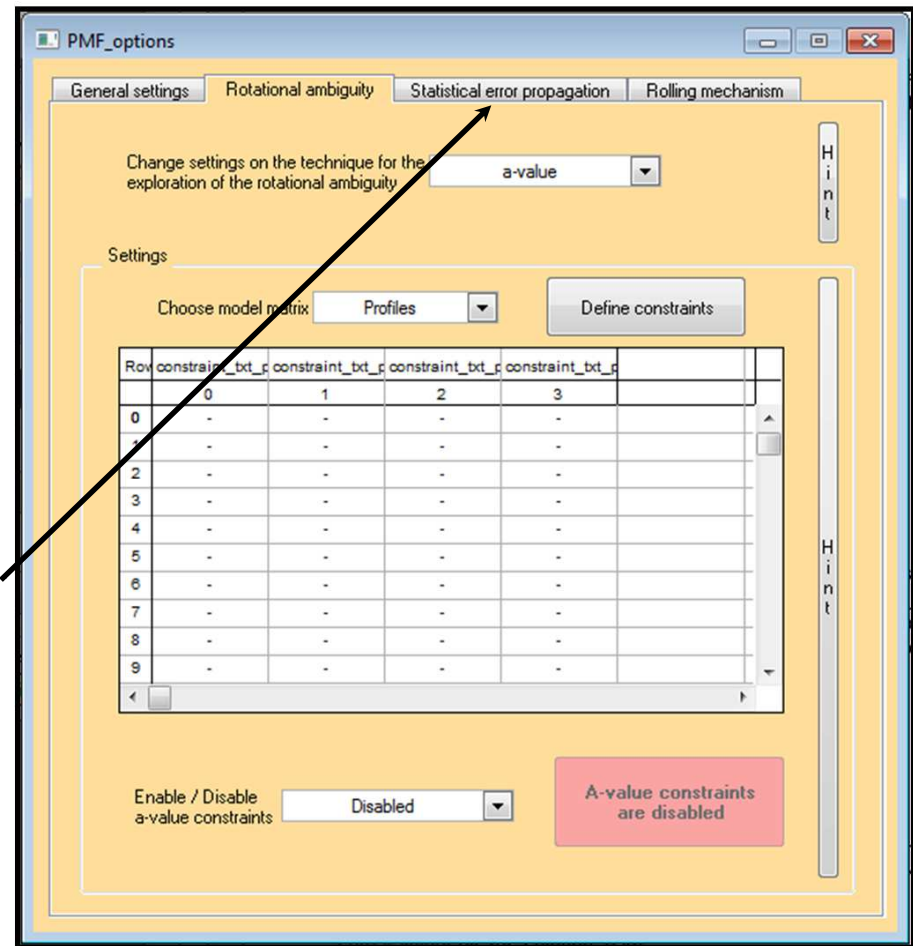
- **Propagate the statistical uncertainty to the PMF result**
  - Monte Carlo method (noise insertion)  
vary the PMF input within the statistical error and call PMF
  - Bootstrap method (resampling strategy)  
resample data with identical underlying sources and call PMF

# Solution space – statistical error prop.

## ➤ Propagate the statistical uncertainty to the PMF result

Approach should be performed in addition to the exploration of the rotational ambiguity)

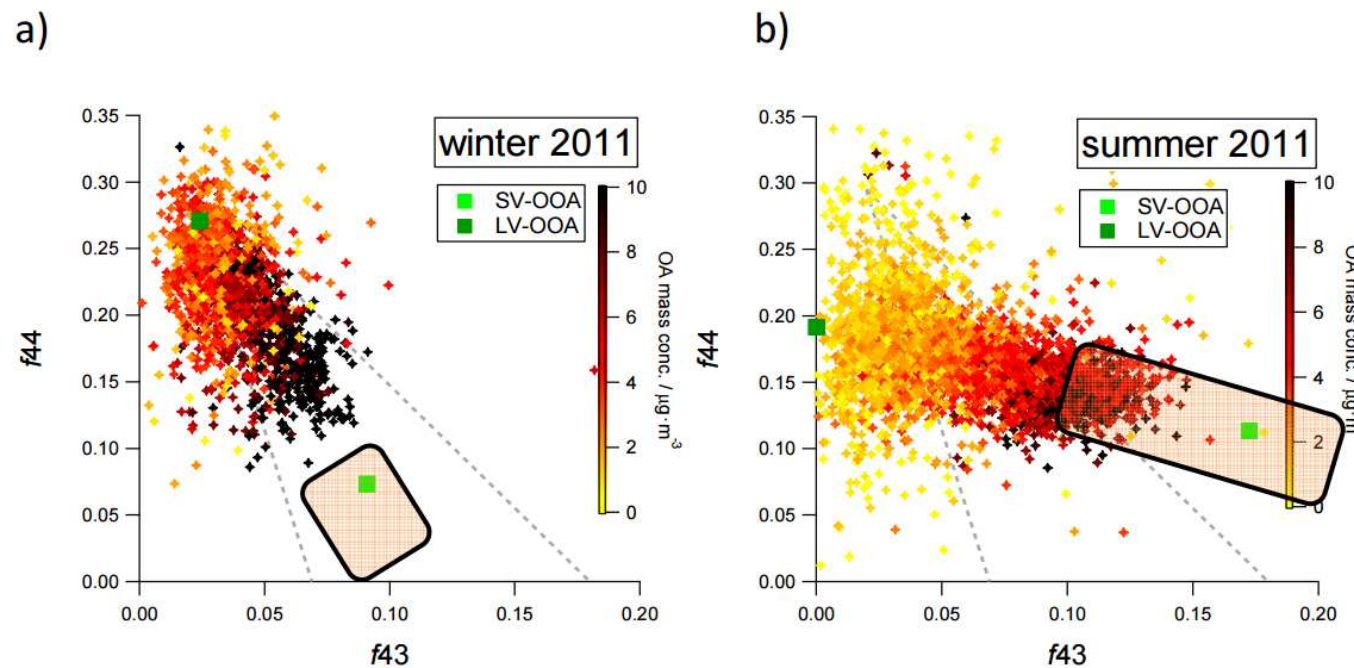
- I. Select approach
- II. Enable
- III. Run PMF (e.g. 10-1000x more runs)
- IV. Analyze runs already containing the statistical uncertainty



# Solution space – AuRo-SoFi

## ➤ ACSM Zurich: winter and summer data 2011

- SOA f44/f43 vary over the seasons
  - Running PMF over the entire year would average this out
- ➔ Run PMF season/month-wise (manually) / apply AuRo-SoFi (automatic)



*Canonaco et al. 2015*

# Solution space – AuRo-SoFi

## ➤ **AuRo-SoFi algorithm**

- Run PMF using a small frame, e.g. two weeks/one month of data (Assumption: source is constant over this period)
- Optimize PMF solution based on criteria defined in advance based on manual pretests  
➔ Automatic part
- Shift PMF frame forward and rerun PMF
- Repeats for small shifts (daily shift compared to length of PMF frame) is facsimile of the bootstrap technique and hence partially propagates the statistical uncertainty  
➔ Rolling part

## ➔ **AuRo – SoFi algorithm**

- More details presented on Wednesday afternoon (Yuliya Sosedova)

*Canonaco et al. in prep., Sosedova et al., in prep.*