# Principal Component Pursuit for Source Apportionment from Block Missing Data: An Application to NYC PM<sub>2.5</sub> Data

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### Background

Linking identified patterns (e.g. sources) associated with adverse health outcomes could yield:

- Efficient policy/public health regulations
- Targeted interventions

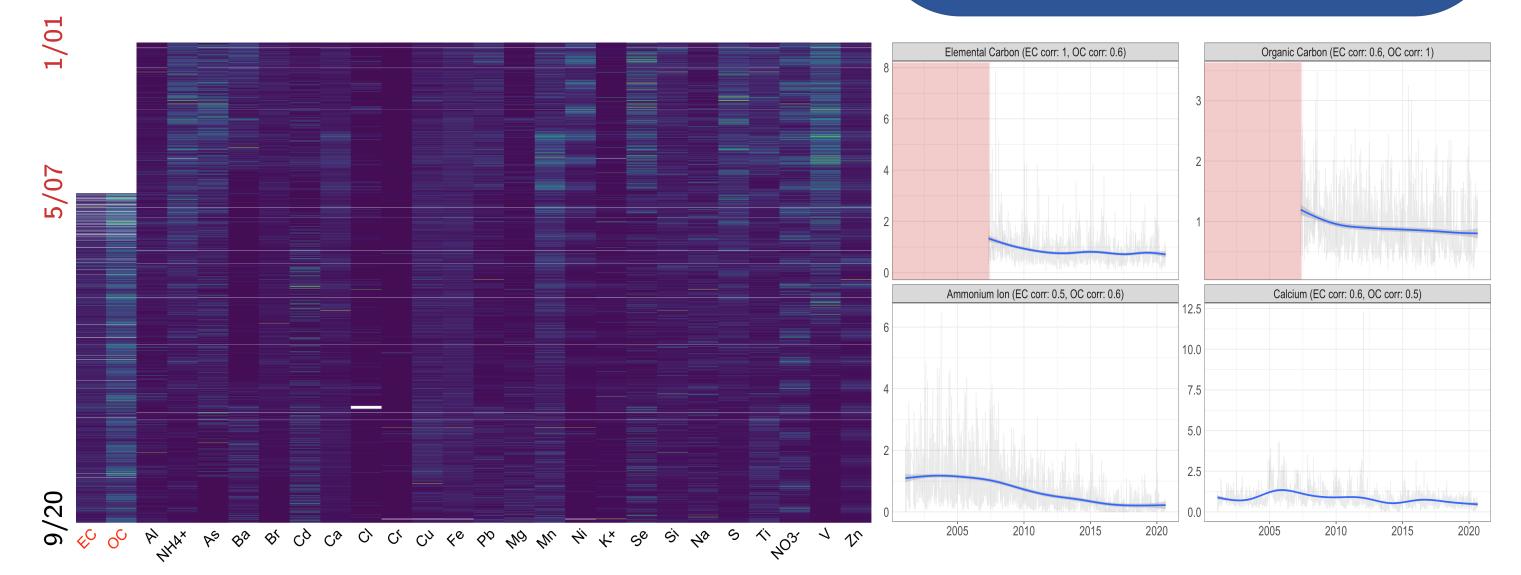
**Elemental Carbon** 

Existing methods limited by:

- Subjective choice of k patterns
- Outliers may affect solution
- No standard for handling structured (block) missingness

Study aim: identify air pollution sources in NYC (2001-2020) & hindcast missing data for two PM<sub>2.5</sub> species (EC, OC)

PCP+Nystrom

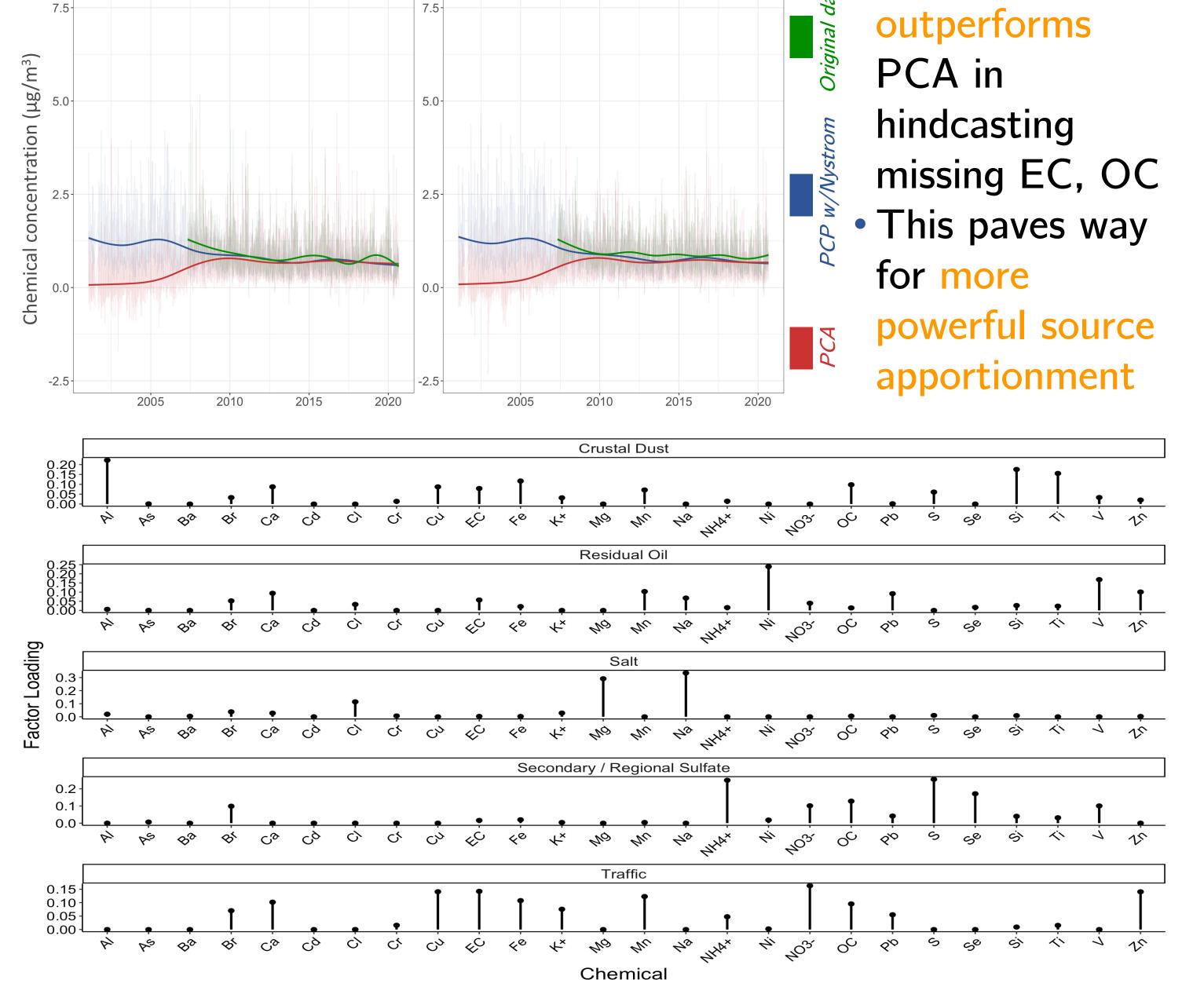


EPA AQS PM2.5 data: NYC, 2001 - 2020

- 26 PM<sub>2.5</sub> chemical species measured across 2,378 days
- EC, OC missing all measurements from '01 '07 (2.6% of overall mixture)

## Results

Organic Carbon



 PCP identified 5 sources of PM<sub>2.5</sub> pollution: crustal dust, residual oil, salt, secondary sulfate, & traffic, as well as 3 single-constituent components (not shown): As, Ba, & Cd.

#### Methods

Principal Component Pursuit (PCP)

- Convex optimization algorithm from computer vision
- Dimension reduction by decomposing mixture into:



Original: environmental mixture



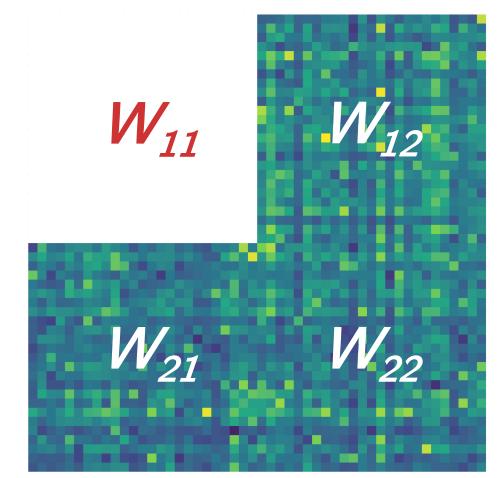
Low-rank (L): consistent exposure patterns



Sparse (S): outlying exposure events

Objective Fn: RRMC : 
$$\min_{L,S} |\mathcal{I}_{(L) \leq r} + \eta ||S||_0 + ||L + S - D||_F^2$$

PCP & block missingness: how to reconstruct  $W_{11}$ ?



Mixture w/missing block W<sub>11</sub>

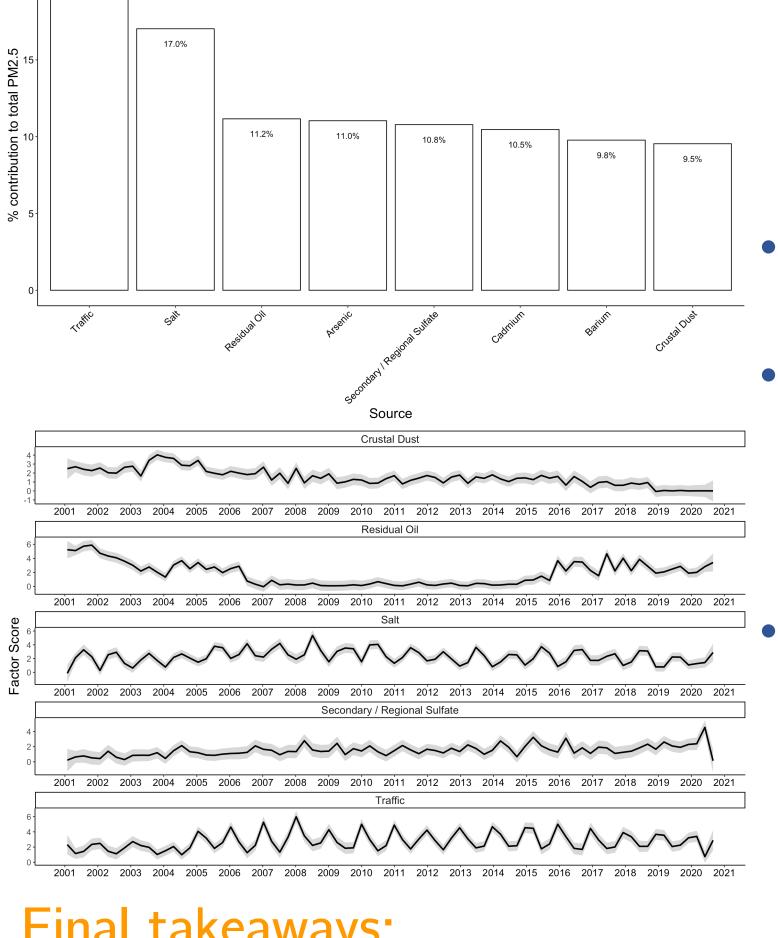
PCP's Nystrom extension:

$$W = \begin{bmatrix} W_{11} & W_{12} \\ W_{21} & W_{22} \end{bmatrix}$$

$$\widehat{\boldsymbol{W}} = \begin{bmatrix} \boldsymbol{W}_{12} [\mathcal{P}_r(\boldsymbol{W}_{22})]^{\dagger} \boldsymbol{W}_{21} & \boldsymbol{W}_{12} \\ \boldsymbol{W}_{21} & \boldsymbol{W}_{22} \end{bmatrix}$$

Main idea: the missing block is reconstructed from observed data

#### Conclusions



- Traffic contributed most to total PM<sub>2.5</sub> concentrations (20.2%) across study period
- Traffic peaked weekdays
- Traffic peaked in winters, including during block missing period
- Spikes observed in K<sup>+</sup> concentrations around each fourth of July (including during missing period)

#### Final takeaways:

- Nystrom extension improves recovery of missing block
  - Exact recovery in no-noise conditions
  - As noise increases, harder to recover missing block
  - Main assumption: Missing block characterized by same patterns governing observed blocks
- PCP w/Nystrom offers reliable pattern recognition allowing researchers to leverage more data
- Sources can be used in subsequent health models