

Source localization on graphs via l1-recovery and spectral graph theory

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Goal

Locate the **source(s)** of an outbreak from a **single snapshot**.

Examples:

- Disease spread
- Diffusion processes (wind, heat, etc.)
- Rumors in social networks

Graph Signal Processing

Data lies on graph $G = (\mathcal{V}, \mathcal{E}, w)$

Normalized graph Laplacian:

$$\mathcal{L} = I - D^{-1/2} W D^{-1/2} = U \Lambda U^T$$

Graph filtering:

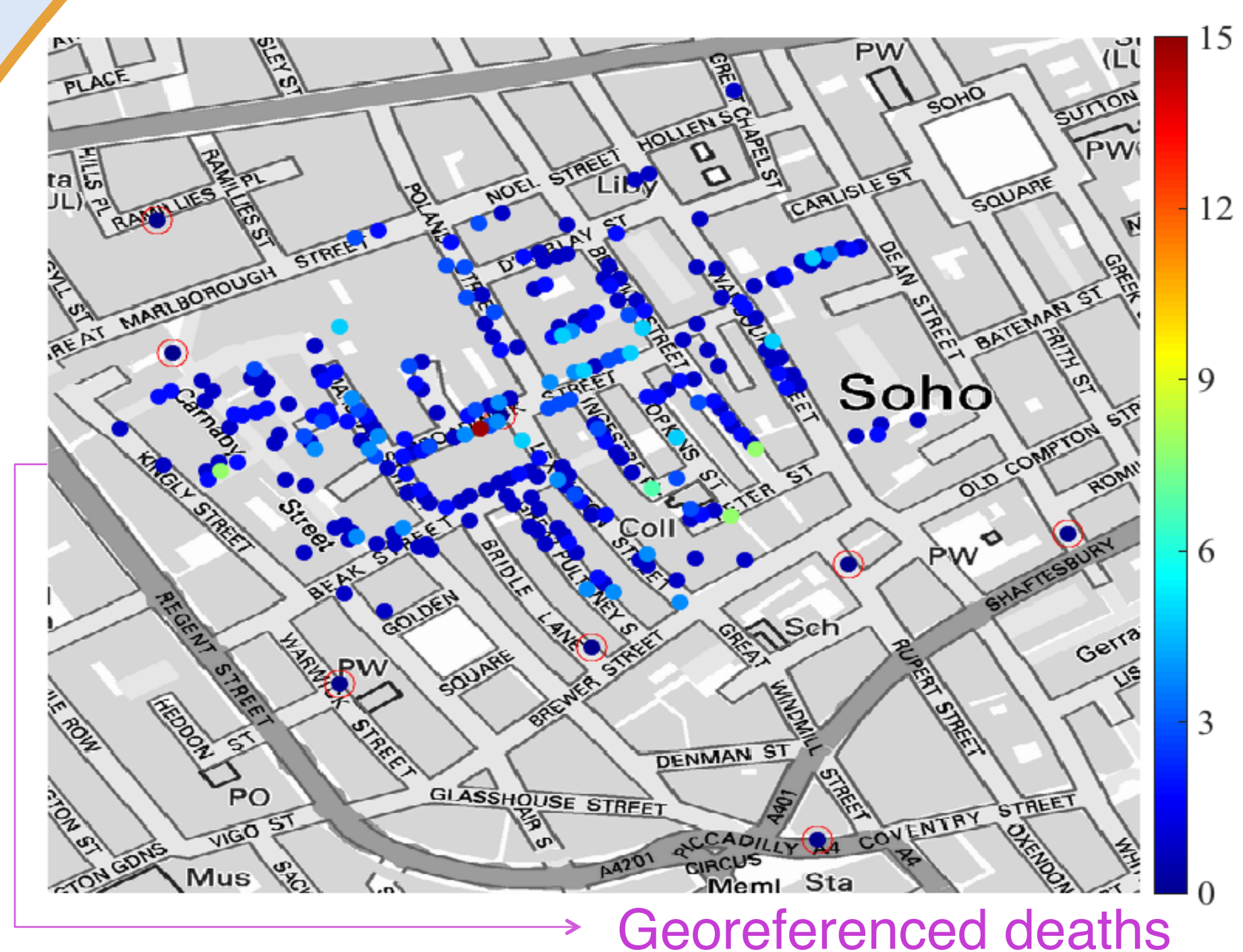
$$b = g(\mathcal{L})x = Ax$$

Example: **heat kernel**

$$g_\theta(\mathcal{L}) = \exp(-\theta \mathcal{L})$$

Real data 1

Cholera outbreak, Soho, London (1854)



Georeferenced deaths

Error Measure

$$e(x, x^*) = \sum_{i \in \mathcal{A}} \frac{\sum_{j \in \mathcal{N}_i} |x^*(j)| h(i, j)}{\sum_{j \in \mathcal{N}_i} |x^*(j)|}$$

“Center of mass” around every true source

Problem

Assumption: observation is **graph-low-pass** filtering of the **sparse** source signal.

$$(x^*, \theta^*) = \arg \min_{x, \theta} \left\{ \gamma \|x\|_1 + \frac{\alpha}{2} \|A_\theta x - b\|_2^2 \right\}$$

Non-convex.

Alternating scheme:

$$x_{k+1} = \arg \min_x E(x, \theta_k) \text{ (FISTA)}$$

$$\theta_{k+1} = \arg \min_\theta E(x_{k+1}, \theta) \text{ (Newton)}$$

Graph construction

k-Nearest Neighbors

$$W_{ij} = \exp\left(\frac{-d(i, j)^2}{\sigma^2}\right)$$

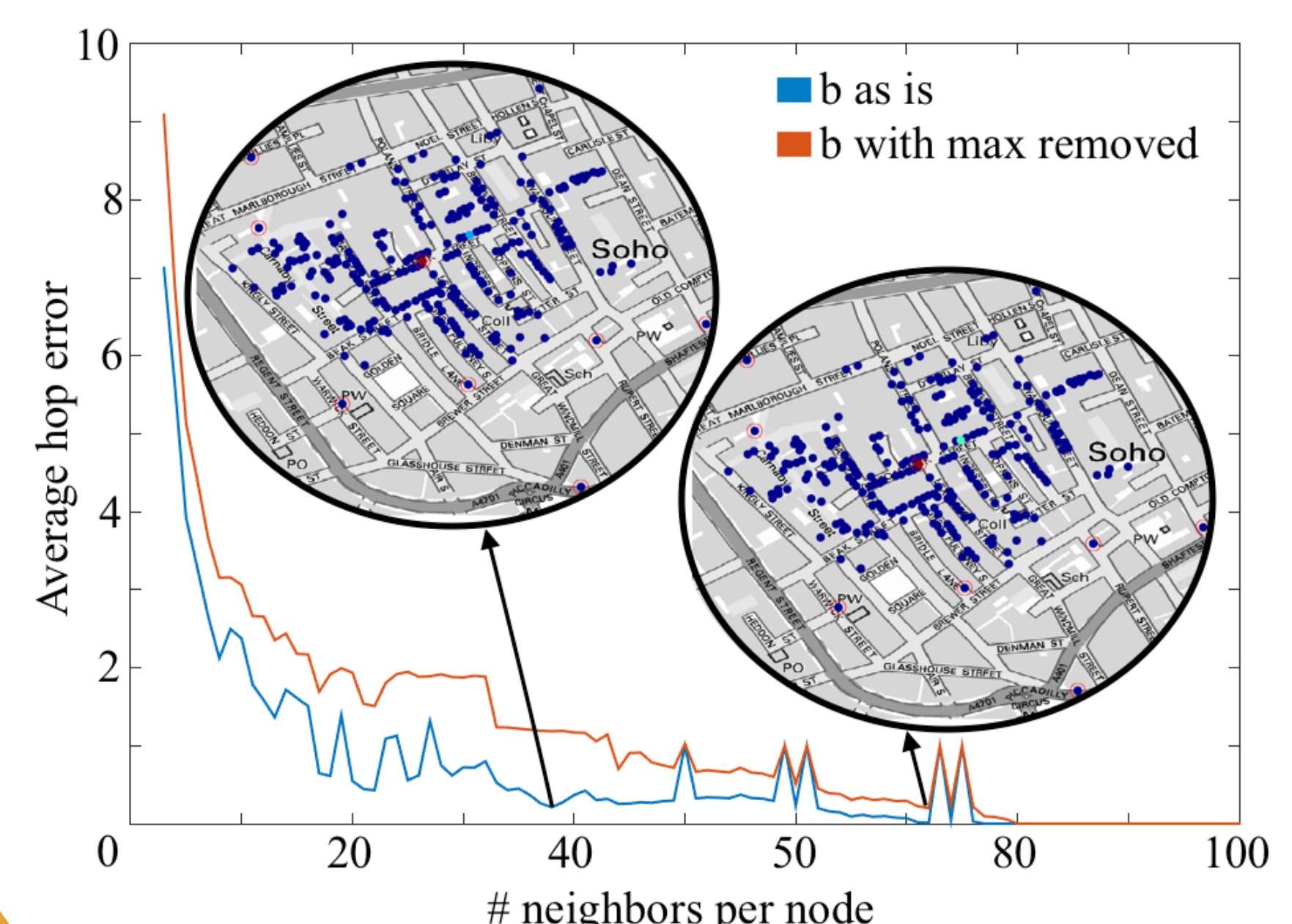
Control data

Simulations:

- ✓ Noise
- ✓ Source separation
- ✗ Bad recovery if snapshot far away in time

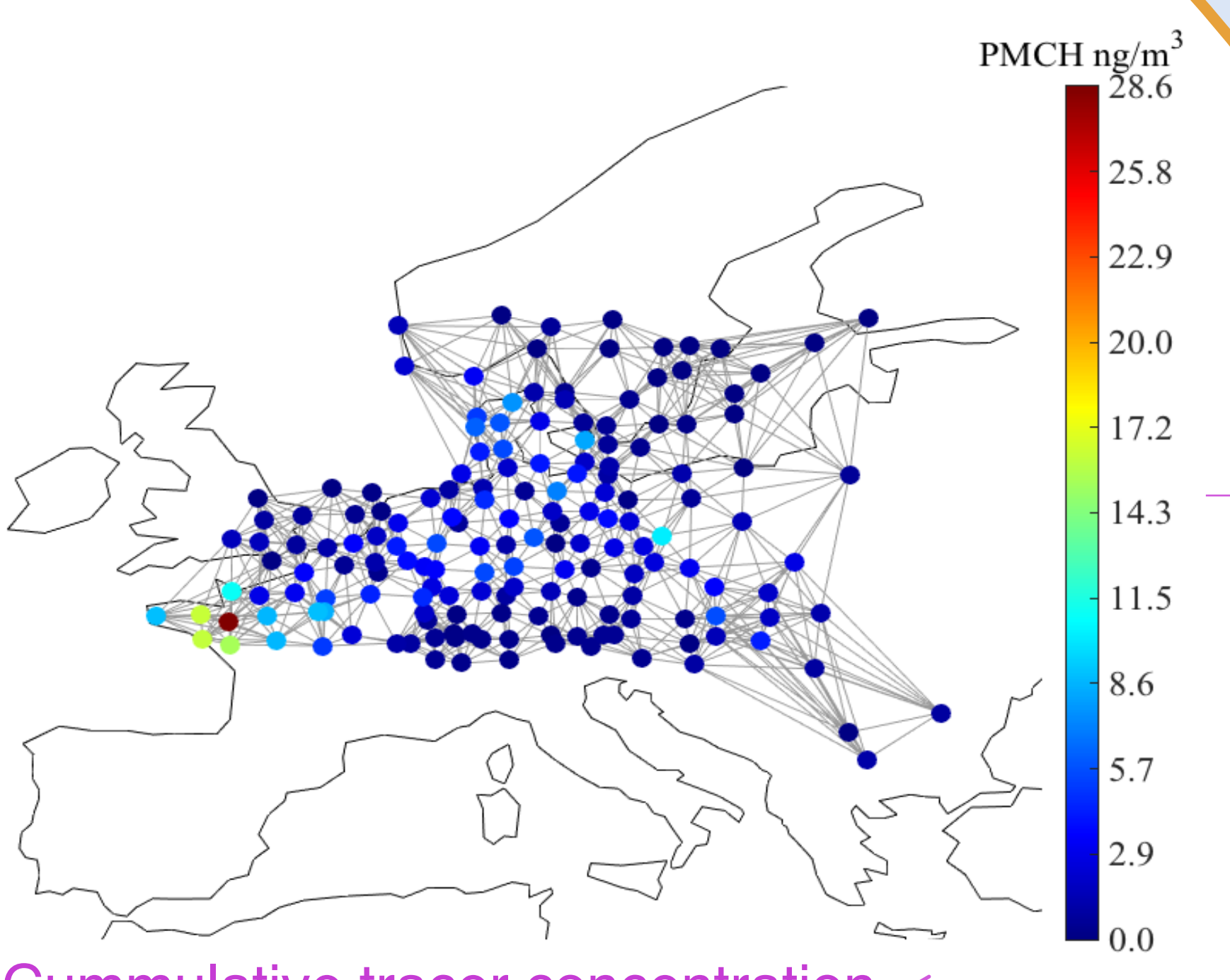
Summary

- ✓ Linear model + sparsity: it works!
- ✓ General framework
- ✓ Robust to noise and source separation
- ✗ Initialization plays big role
- ✗ Graph construction as well
- ✗ Improvement if diffusion model is application-specific?

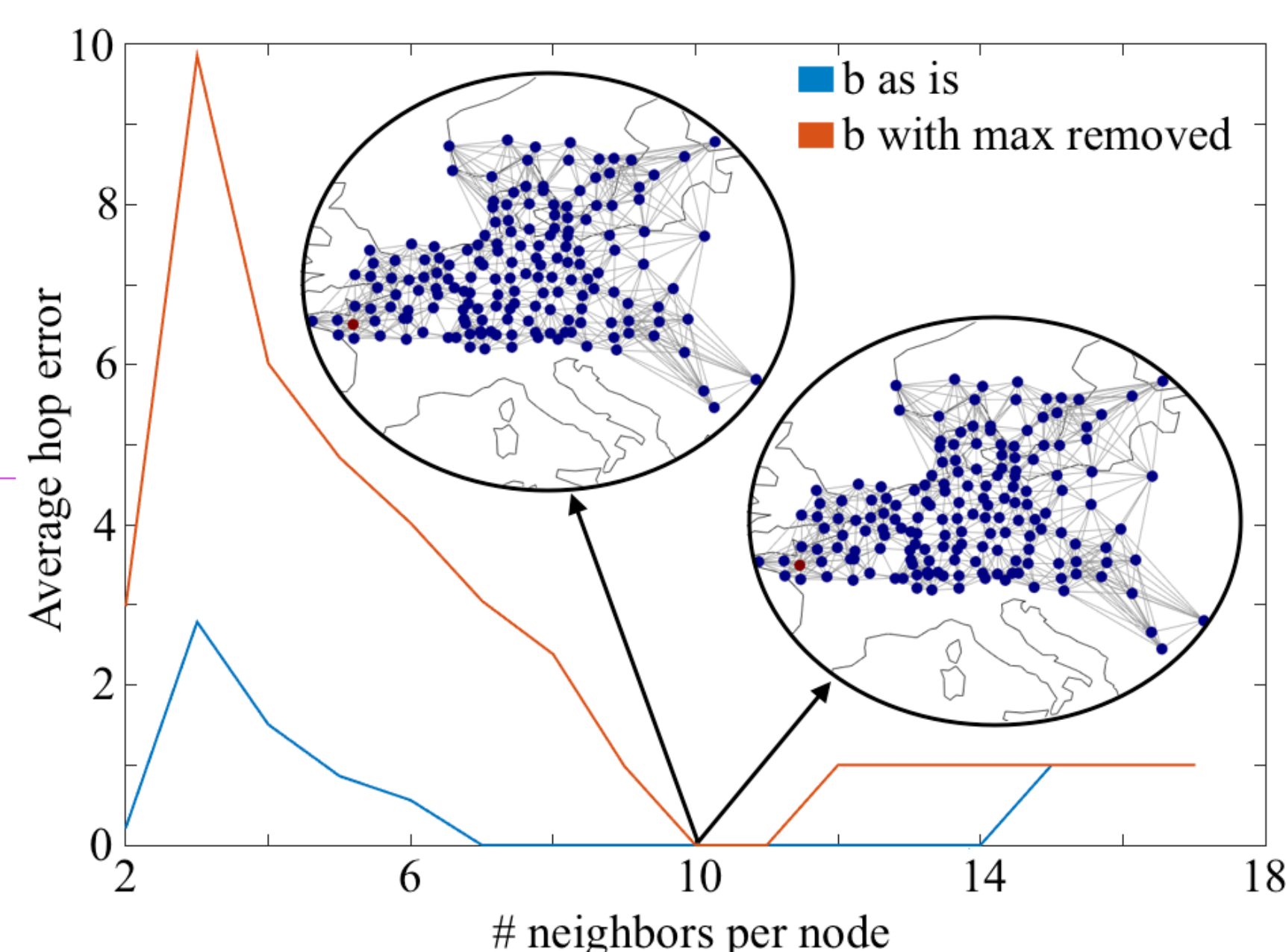


Recovery error as a function of graph density

Euclidean distance



Cummulative tracer concentration



Recovery error as a function of graph density

