

Xavier Bacon

xavier.bacon@ens-paris-saclay.fr

<https://baconxavier.github.io/>

Borelli Center

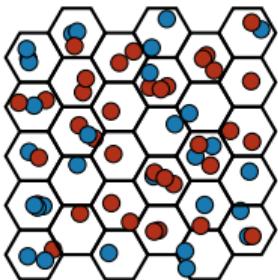
École Normale Supérieure de Paris-Saclay

CCS 2025, September 02, 2025

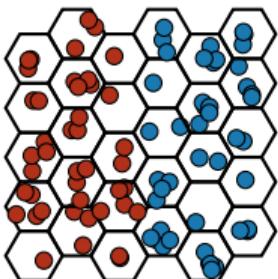
Analysing Spatial Heterogeneity

using Optimal Transport and applied to electoral results

in collaboration with Joseph Touzet (ÉNS Paris-Saclay)



Weak Heterogeneity



High Heterogeneity

Minimal framework

- › A finite set of areal units $\mathcal{X} = \{x_1, x_2, \dots\}$
- › Several groups (chemical or animal species, ethnic groups) distributed across \mathcal{X} :

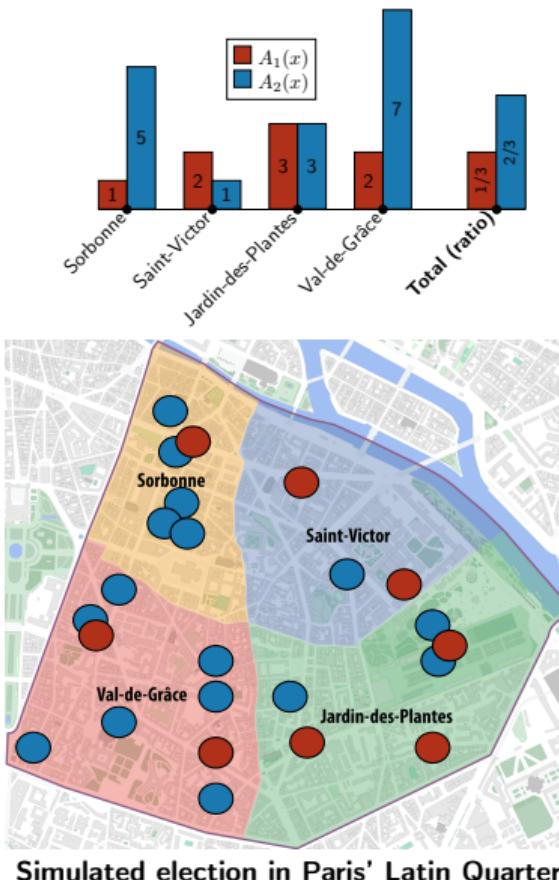
$$\mathbf{A} = \mathbf{A}_1(\mathbf{x}), \mathbf{A}_2(\mathbf{x}), \dots$$

Goals

- › Propose two indices (one global, one local) to measure the **spatial heterogeneity** of these distributions: are they well mixed or not?

Similar works

- › Dissimilarity Index (Duncan & Duncan, 1955), Isolation Index, Entropy-based measures Index (Theil, 1972), Multiscalar Lens Model (Olteanu et al., 2019), complex systems approaches (Agent-Based Modeling *via* Schelling, Network Analysis...).



Electoral data Modelling

- › \mathcal{X} = the set of polling stations (or districts).
- › A_1 = votes cast for the candidate 1,
 A_2 = votes cast for the candidate 2, ...
- › Distributions $\alpha_1, \alpha_2, \dots$ correspond to the votes' ratio for the n -candidate, that is

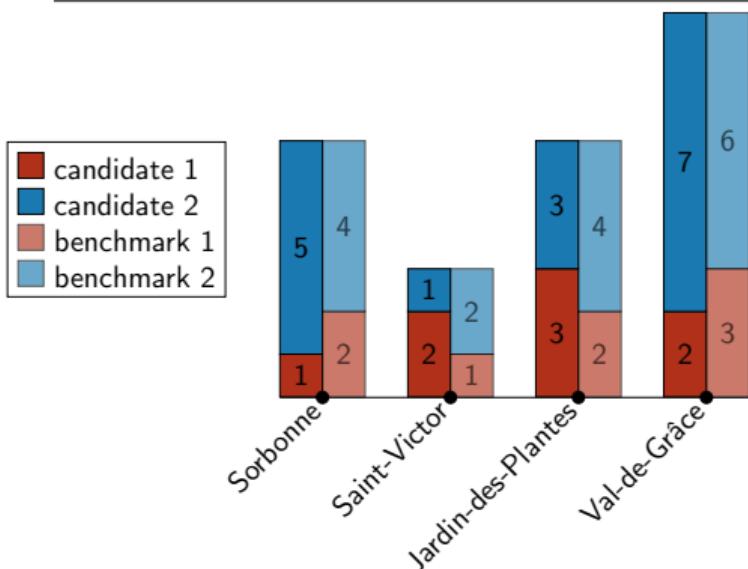
$$\alpha_n(x) = \frac{A_n(x)}{\sum_y A_n(y)}$$

- › Distance-cost matrix $C = C(x, y)$: straight-line distance between two polling stations.

Global OT-Index for the n -candidate

$$D_n = \mathbf{OT}_C(\alpha_n, \bar{\alpha})$$

Candidates' scores and their respective benchmarks



- › **Benchmark $\bar{\alpha}$:** represents the *perfect mixed* distribution of votes:

$$\bar{\alpha}(x) = \frac{\sum_n A_n(x)}{\sum_y^n \sum A_n(y)}.$$

- › **Optimal Transport (OT):** a tool to compute a distance between two distributions.
- › For the overall election:

$$D = \sum_n p_n D_n$$

where p_n is the n -candidate's score.

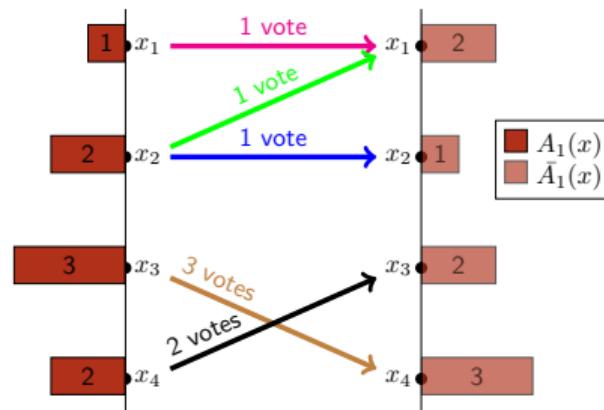
Optimal Transport between α_n and $\bar{\alpha}$

$$D_n = \text{OT}_C(\alpha_n, \bar{\alpha})$$

$$= \min \left\{ \langle C | \gamma_n \rangle : \gamma_n \cdot \mathbb{1} = \alpha_n, \gamma_n^T \cdot \mathbb{1} = \bar{\alpha} \right\}$$

- » Thanks to $C = C(x, y)$ which defines the " cost to move one vote from x to y ", D_n captured the geographic properties of \mathcal{X} .
- » γ_n dispatches votes from one polling station to another.
- » **D_n is a geographical index:** spatial arrangements of areal units is taken into account.

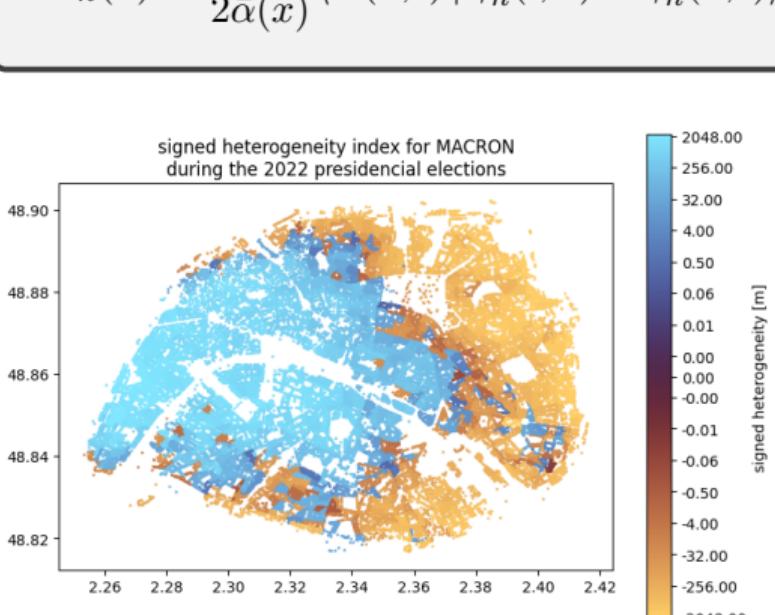
$$\gamma_n = \frac{1}{8} \cdot \begin{bmatrix} 1 & 0 & 0 & 0 \\ 1 & 1 & 0 & 0 \\ 0 & 0 & 0 & 3 \\ 0 & 0 & 2 & 0 \end{bmatrix}$$



A Local OT-Index

Local OT-Index

$$\omega_n(x) = \frac{1}{2\bar{\alpha}(x)} \langle C(x, \cdot) | \gamma_n^*(\cdot, x) - \gamma_n^*(x, \cdot) \rangle$$



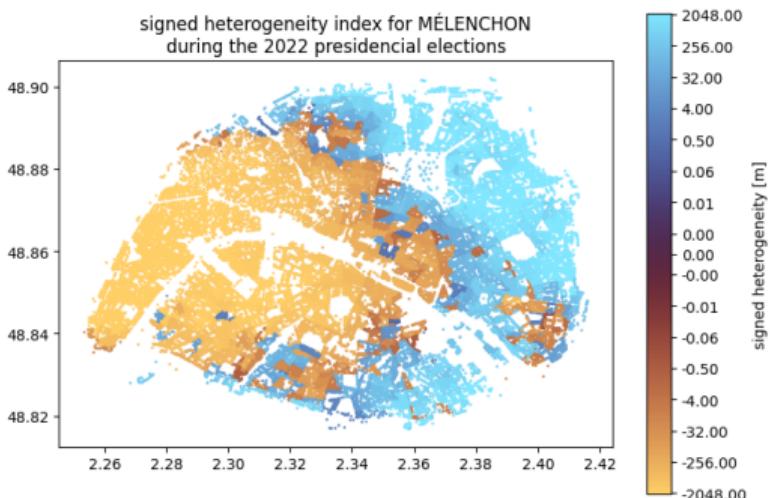
- Optimal transference plans γ_n^* defines a flow between pairs of polling stations.
- Readily interpretable:**
 C in m $\implies \omega_n$ and D_n in m
- French 2022 Presidential results of 4 candidates (among 12) (*Paris area*):

| Candidate | OT-Index D_n | Score (%) |
|-----------|----------------|-----------|
| Macron | 519 meters | 35,49 |
| Mélenchon | 980 meters | 29,93 |
| Zemmour | 1005 meters | 8,2 |
| Le Pen | 295 meters | 5,39 |

A Local OT-Index

Local OT-Index

$$\omega_n(x) = \frac{1}{2\bar{\alpha}(x)} \langle C(x, \cdot) | \gamma_n^*(\cdot, x) - \gamma_n^*(x, \cdot) \rangle$$



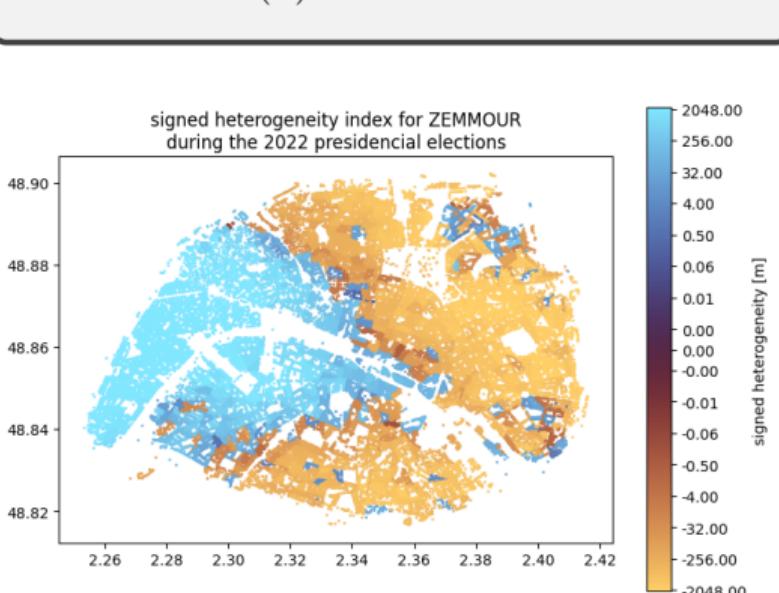
- › Optimal transference plans γ_n^* defines a flow between pairs of polling stations.
- › **Readily interpretable:**
 C in m $\implies \omega_n$ and D_n in m
- › French 2022 Presidential results of 4 candidates (among 12) (*Paris area*):

| Candidate | OT-Index D_n | Score (%) |
|-----------|----------------|-----------|
| Macron | 519 meters | 35,49 |
| Mélenchon | 980 meters | 29,93 |
| Zemmour | 1005 meters | 8,2 |
| Le Pen | 295 meters | 5,39 |

A Local OT-Index

Local OT-Index

$$\omega_n(x) = \frac{1}{2\bar{\alpha}(x)} \langle C(x, \cdot) | \gamma_n^*(\cdot, x) - \gamma_n^*(x, \cdot) \rangle$$



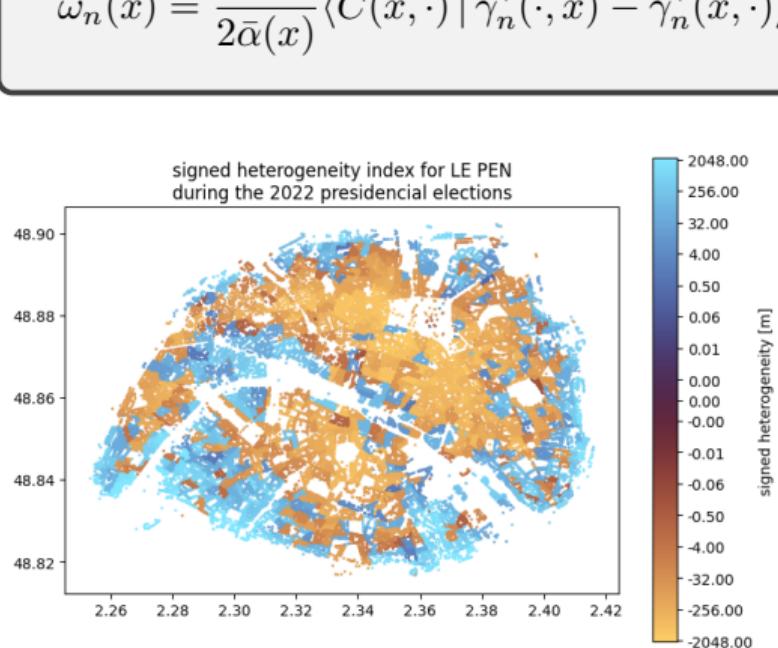
- Optimal transference plans γ_n^* defines a flow between pairs of polling stations.
- Readily interpretable:**
 C in m $\implies \omega_n$ and D_n in m
- French 2022 Presidential results of 4 candidates (among 12) (*Paris area*):

| Candidate | OT-Index D_n | Score (%) |
|-----------|----------------|-----------|
| Macron | 519 meters | 35,49 |
| Mélenchon | 980 meters | 29,93 |
| Zemmour | 1005 meters | 8,2 |
| Le Pen | 295 meters | 5,39 |

A Local OT-Index

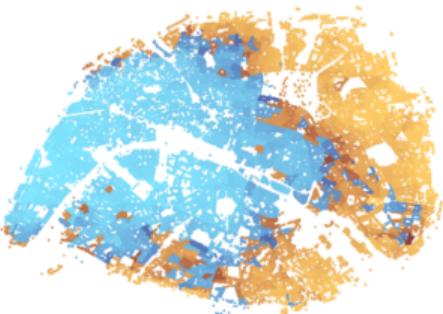
Local OT-Index

$$\omega_n(x) = \frac{1}{2\bar{\alpha}(x)} \langle C(x, \cdot) | \gamma_n^*(\cdot, x) - \gamma_n^*(x, \cdot) \rangle$$



- Optimal transference plans γ_n^* defines a flow between pairs of polling stations.
- Readily interpretable:**
 C in m $\implies \omega_n$ and D_n in m
- French 2022 Presidential results of 4 candidates (among 12) (*Paris area*):

| Candidate | OT-Index D_n | Score (%) |
|-----------|----------------|-----------|
| Macron | 519 meters | 35,49 |
| Mélenchon | 980 meters | 29,93 |
| Zemmour | 1005 meters | 8,2 |
| Le Pen | 295 meters | 5,39 |



- Future works: focus on special costs: concave (induces a few large displacements) or convex (induces a lot of small displacements).
- Code available:
<https://github.com/jolatechno/ot-heterogeneity>
-

Thank you for your attention!