





ON THE DESIGN OF PUBLIC TRANSPORT FOR EQUALITY OF ACCESSIBILITY

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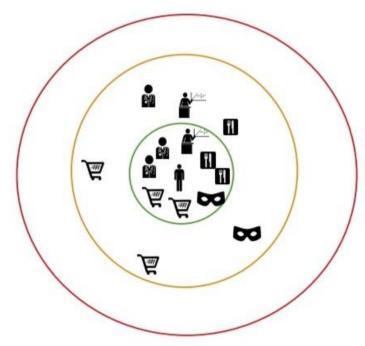


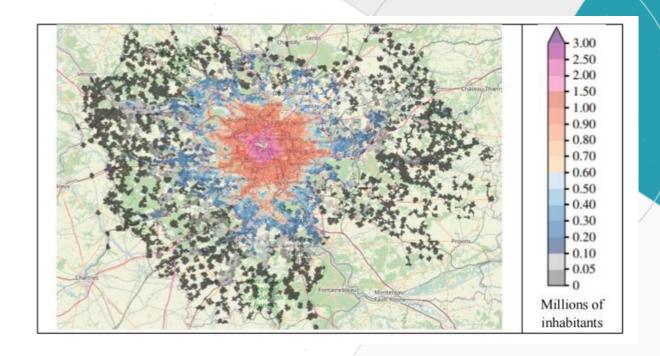


INTRODUCTION



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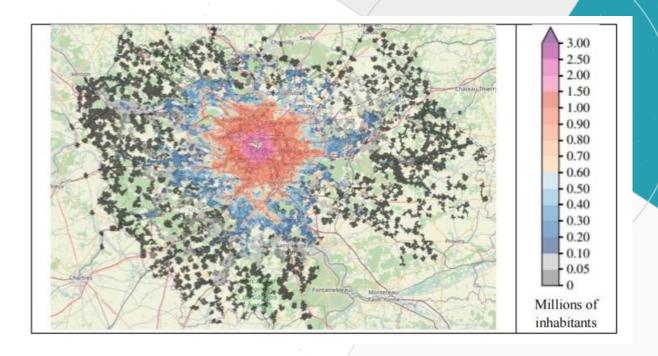
Accessibility: ease of reaching surrounding opportunities.

Inequality: suburbs suffer from poor accessibility from public transit.



INTRODUCTION





Accessibility: ease of reaching surrounding opportunities.

Inequality: suburbs suffer from poor accessibility from public transit.

Consequence: car-dependency → pollution

Need for designing more equitable public



INTRODUCTION



OBJECTIVE

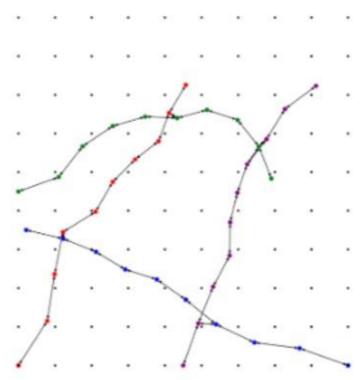
Improve equality while preserving PT efficiency by serving or skipping PT stops







METHODOLOGY



An example of PT network: 4 PT lines containing 40 stops and 100 centroids

PUBLIC TRANSPORT (PT) AND ACCESSIBILITY

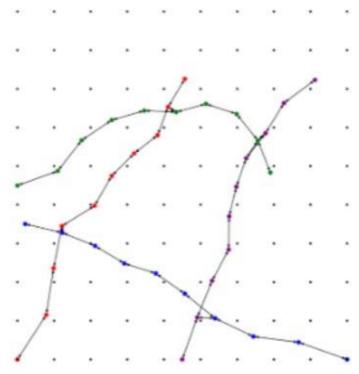
1.G: PT network

2.V: set of centroids

3.S: set of bus stops



IP PARIS



An example of PT network: 4 PT lines containing 40 stops and 100 centroids

PUBLIC TRANSPORT (PT) AND ACCESSIBILITY

1.G: PT network

2.V: set of centroids

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4.The accessibility of centroid $v \in V$,

$$acc(v) = \sum_{u \in V} \frac{X(u)}{T(v, u)}$$

T(v, u): shortest time to go from v to u,

X(u): amount of the opportunities of the u.



An example of PT network: 4 PT lines containing 40 stops and 100 centroids

PUBLIC TRANSPORT (PT) AND ACCESSIBILITY

 $V^{\%}$: set of centroids with the worst accessibility. The accessibility of graph G as:

$$Acc(G; m) = \frac{1}{|V^{\%}|} \sum_{v \in V^{\%}} acc(v)$$

The Atkison inequality index is

$$Atk(G) = 1 - \frac{1}{\overline{acc}(G)} \cdot \left(\frac{1}{K} \sum_{i=1}^{k} y_i^{-1}\right)^{-1}$$







RANDOM SEARCH OPTIMIZATION ALGORITHM

- 1.Ef-Opt: maximize the average accessibility
- 2.Eq-Opt: maximize the accessibility of the worst centroids

Algorithm 1: Random search optimization algorithm

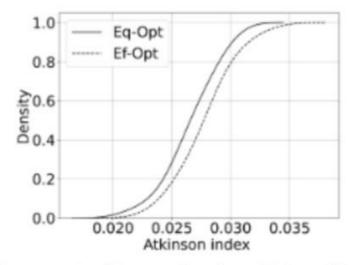
- 1: **Input** Public transport graph G with stops S.
 - Parameter m of the accessibility formula (2).
- 2: For search instance $i \leftarrow 1$ to n:
- 3: Initialize $\mathcal{G}_0 \leftarrow \mathcal{G}$ and $\mathcal{S}_0 \leftarrow \mathcal{S}$.
- 4: For step $t \leftarrow 0$ to ∞ until termination condition:
- 5: Select a random stop $s_t \in S_t$ and deactivate it.
- 6: Set $S_{t+1} \leftarrow S_t \setminus \{s_t\}$ and let G_{t+1} the resulting PT graph.
- 7: Compute the new accessibility: $Acc(\mathcal{G}_{t+1}; m)$
- 8: EndFor
- 9: Record $\mathcal{G}^i = \arg\min_{\tau=0}^{t+1} Acc(\mathcal{G}_{\tau}, m)$.
- 10: EndFor
- 11: **Return** PT graph $\mathcal{G}^* = \arg\min_{i=1}^n \mathcal{G}^i$.







Performance on the entire data set



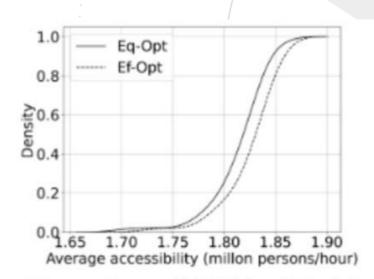


Figure 2: Cumulative Distribution Functions (CDFs) of Atkinson index and average accessibility after Eq-Opt and Ef-Opt



Performance on each graph

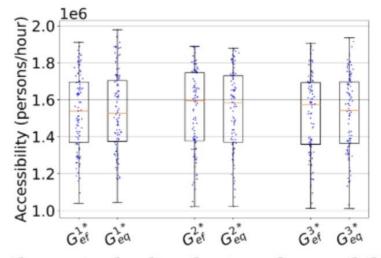


Figure 3: Change in the distribution of accessibility across centroids in three exemplary graphs.

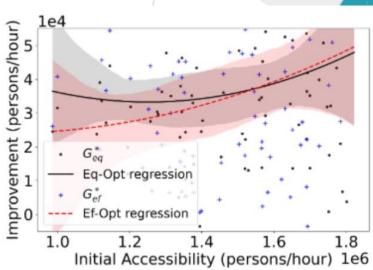


Figure 4: Change in accessibility on one exemplary graph.





The difference between selection strategies

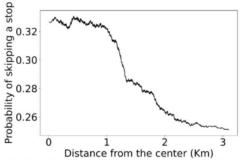
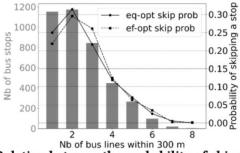


Figure 5: Relation between the probability of skipping a stop and distance from the center by Eq-Opt



Nb of bus lines within 300 m

Figure 6: Relation between the probability of skipping a stop and the Nb of bus lines nearby

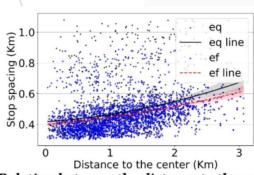


Figure 7: Relation between the distance to the nearest stop and the distance to the center





FUTURE WORK





FUTURE WORK

- 1. Apply optimization to real PT network data (GTFS)
- 2. Consider the deployment of shared mobility
- 3. Use Deep-Q learning to select stops

