

Human mobility analysis by collaborative radio landscape observation

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Information Science

Outline

① Motivation

- Human Mobility Patterns
- Previous Work

② Mobility Analysis

- Personal Mobility Maps
- Proximity Graphs

③ Results

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- Study the mobility patterns of humans.
- Use their own smart phones to collect the data.
- Utilize the information to **create reports about their mobility.**

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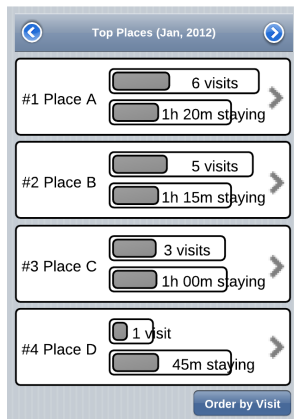
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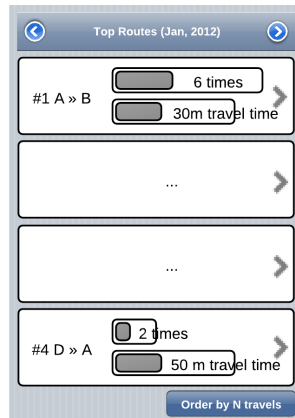
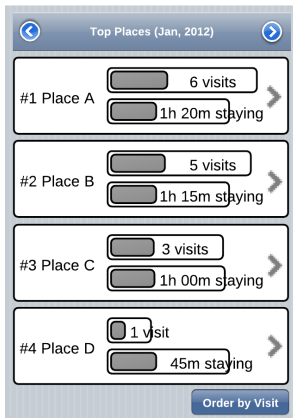
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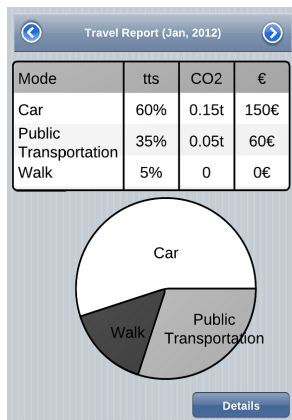


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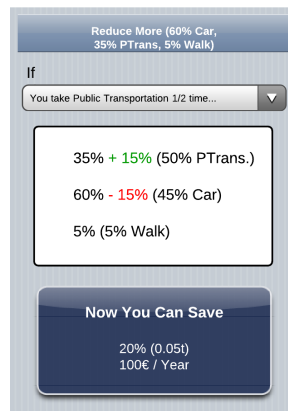
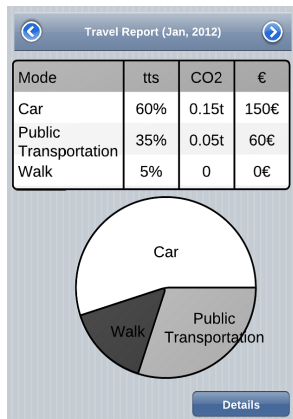


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Transportation Systems Efficiency.



- Greener approach to transportation.
- Help users save money and help the environment.

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What Is Known from the Literature?

- ① Humans follow reproducible patterns with deep regularity.
- ② It is possible to discover daily-life routines of users.
- ③ Individuals spend around 87% of their time indoors.
- ④ Increasing popularity on the use of wireless networks.
- ⑤ Several indoor localization strategies exist.

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- 1 Use GSM and WiFi networks to infer the mobility patterns in a symbolic space, without necessarily knowing the geographic coordinates of places.
- 2 Infer the logical space structure from the radio landscape and use it to provide spatial context to the mobility patterns.

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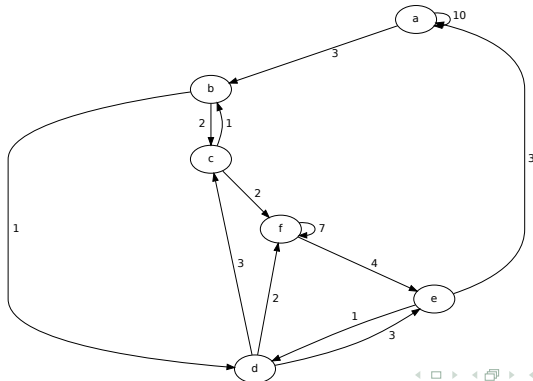
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Representation of Mobility.

Challenges

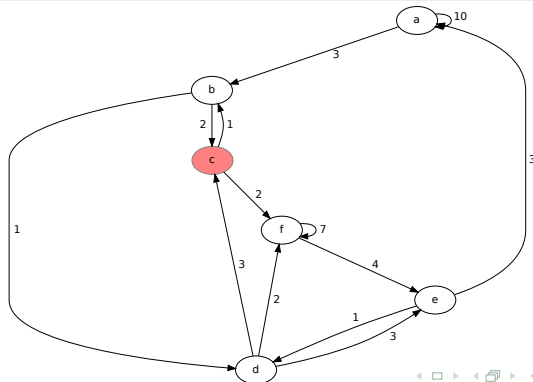
- 1 Place learning.
- 2 Trip detection.



Representation of Mobility.

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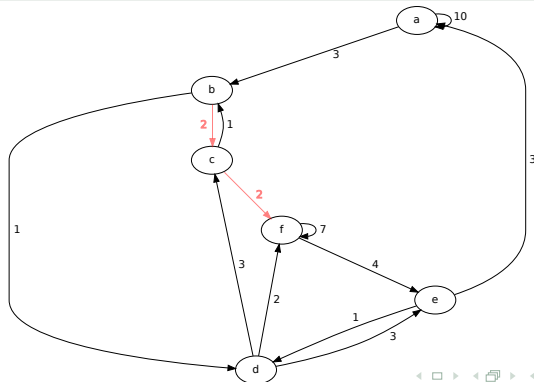
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Creating the Mobility Maps.

$$O = \{(mID_1, t_1, BSs_1), \dots (mID_{Ns}, t_{Ns}, BSs_{Ns})\}$$

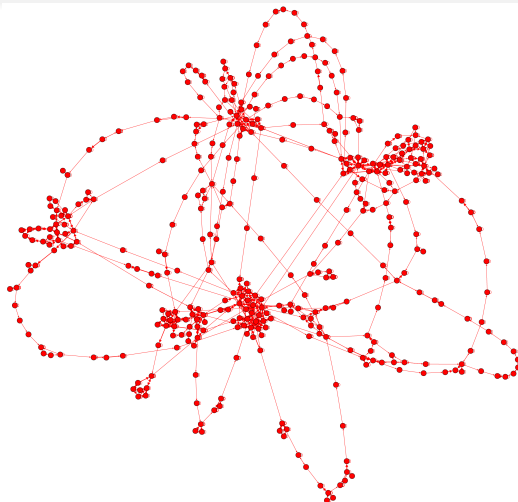
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Example of a Real Personal Mobility Map.

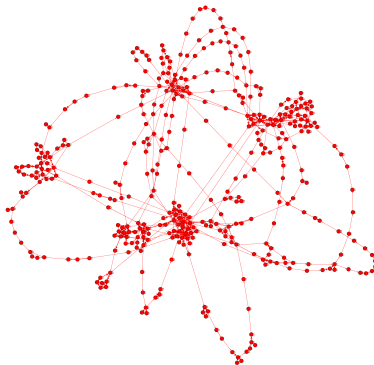


93.629 observations obtained during 22 days. 404 nodes and 1.183 edges.

Mobility Maps.

The Good and the Bad

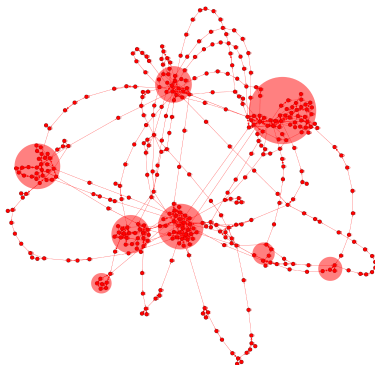
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- Lack of spatial context.



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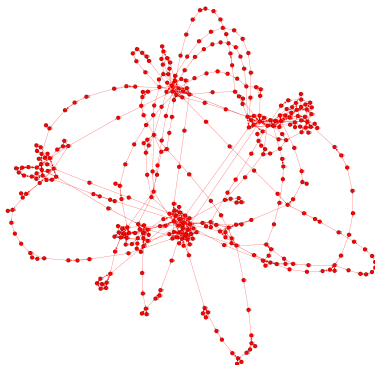
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Adding Spatial Context with WiFi.

$$G_p = \{N_p, E_p\}$$

- Nodes N_p are all the APs in all the observations:

$$N_p = \bigcup_{i,j=1}^{N_s, Nbs_i} bsID_{i,j}$$

- Edges E_p only between simultaneously visible APs.

$$E_p = \left\{ \bigcup_i (bsID_{i,j}, bsID_{i,m}) : j \neq m \right\}$$

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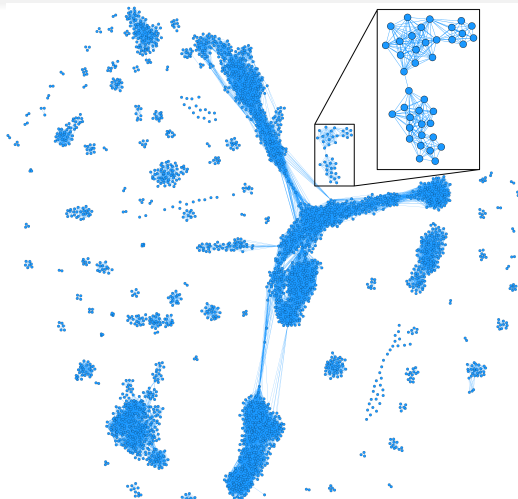
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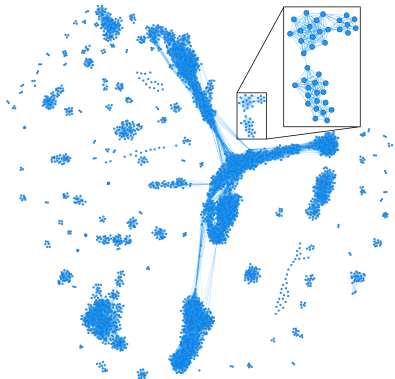


93.629 observations over 22 days by two users. 2.607 nodes and 41.470 edges.

WiFi Proximity Maps.

The Good and the Bad

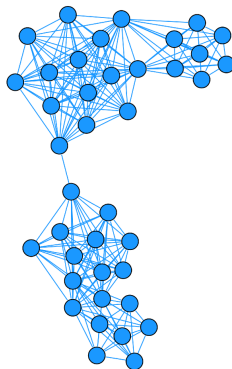
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- A number of disconnected subgraphs exists.
- Long range is not easily captured.



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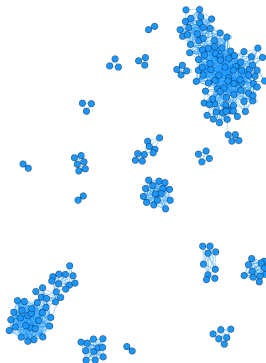
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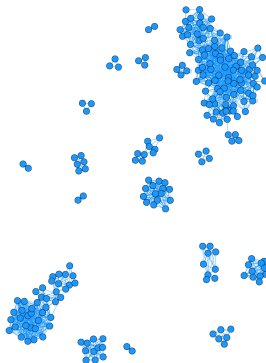
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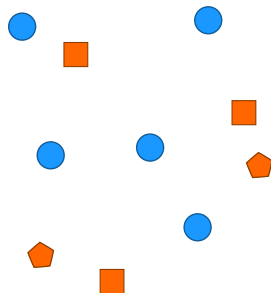
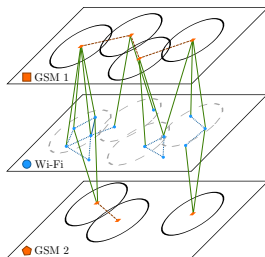
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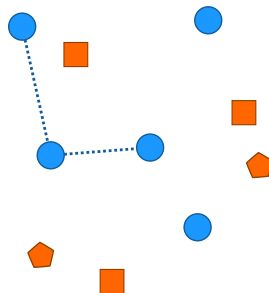
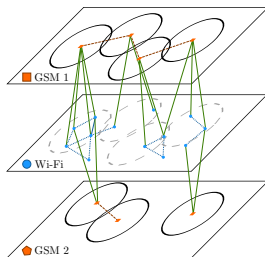
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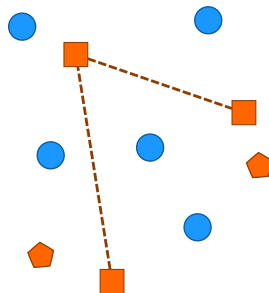
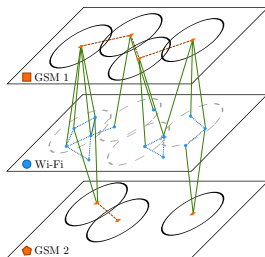
WiFi/GSM Proximity Maps.



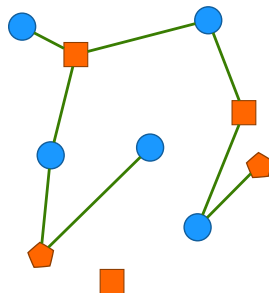
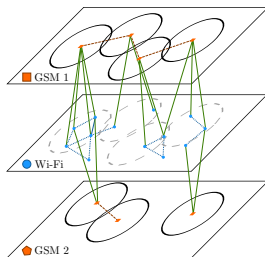
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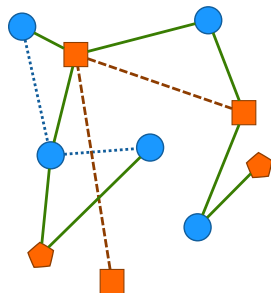
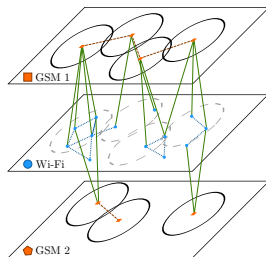
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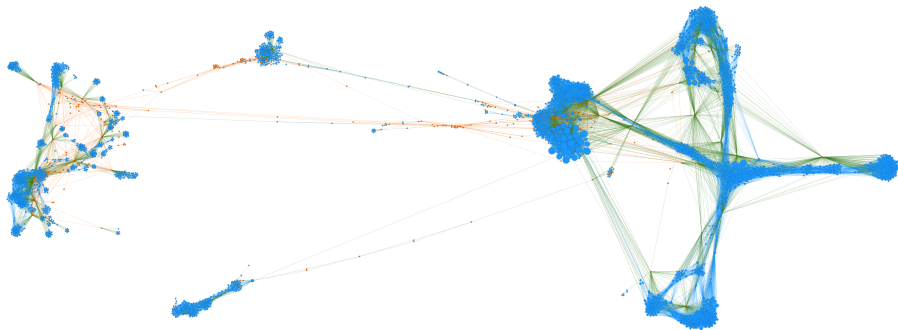
WiFi/GSM Proximity Maps.



WiFi/GSM Proximity Maps.



Example of WiFi/GSM Proximity Map



93.629 WiFi observations over 22 days by two users. 2.607 WiFi nodes and 41.470 edges.

346.150 GSM observations over 22 days by two users. 353 GSM nodes and 644 GSM edges. 6.289 mixed edges.

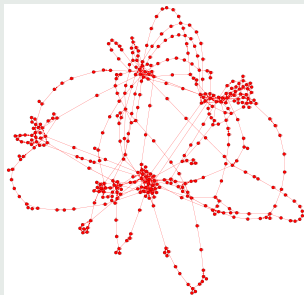
Putting Everything Together

No Spatial Content

Spatial Relations

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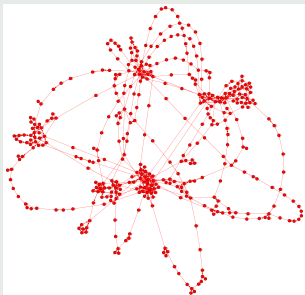
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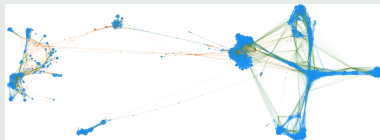
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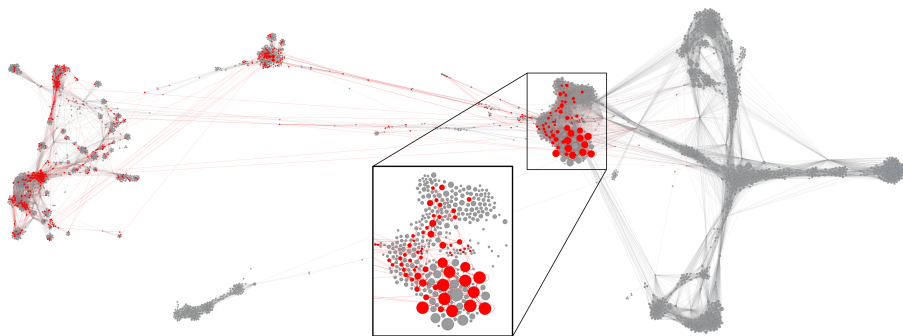
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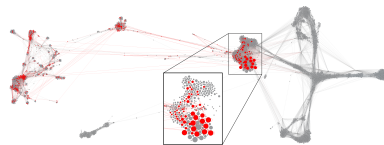


Personal Map overlaid on the GSM/WiFi proximity map



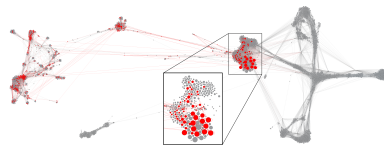
What We Learnt

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- Clear view of trips among places.
- Popularity of places.



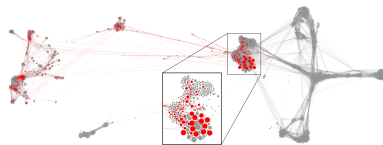
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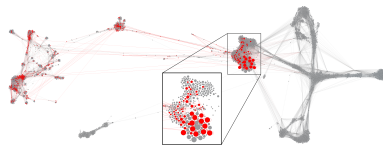
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Summary

- The combined analysis of mobility and proximity maps is much more informative than mobility alone.
- Merging the Proximity Graphs of multiple users, Proximity Maps with larger detail can be built.
- Multiple spatial proxies (WiFi/GSM) complement each other to greatly improve the accuracy of the spatial representation.
- Outlook
 - Deal with mobile APs.
 - Suitable metric for the edges of the Proximity Graph.

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