

Capturing time in space

—

Investigating equity in transport with multidimensional analyses of accessibility and mobility

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WHO? IN A NUTSHELL



Henrikki
Tenkanen

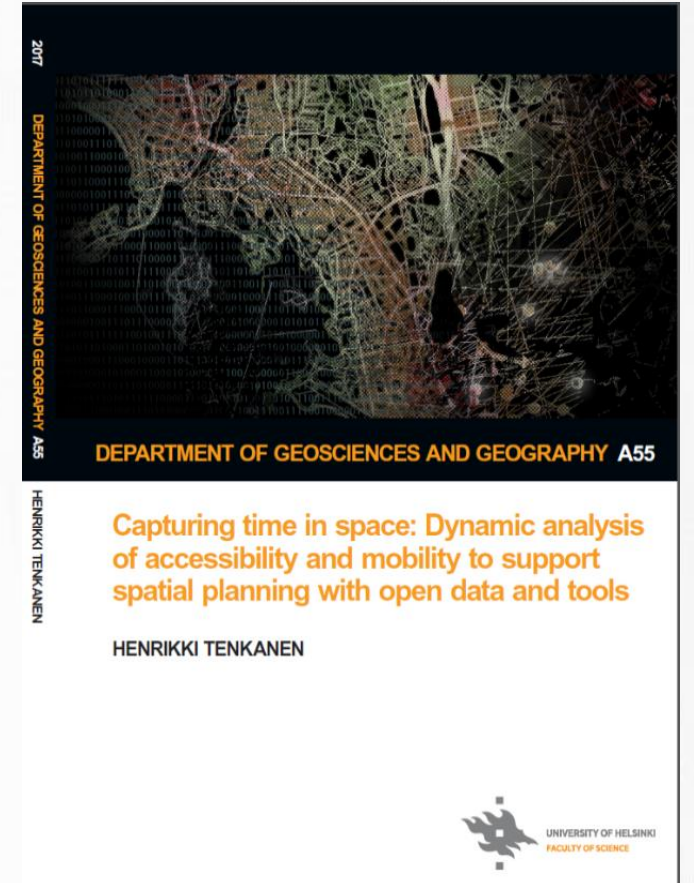
Geographer / geodata scientist

Affiliate academic at Dept. Geography, UCL
running own project (2019 -)

PhD from Uni. Helsinki in 2017

Co-founder of Mapple Analytics Oy

Python GIS educator



GEO PYTHON
GEO PYTHON

geo-python.github.io



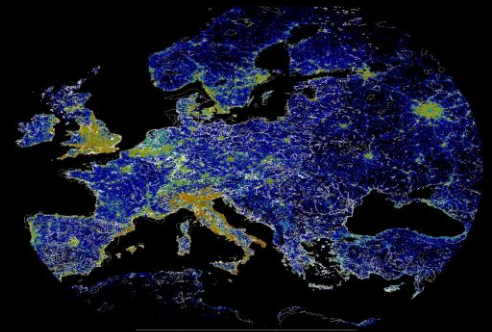
```
print("Hello, world!")
```

**OPEN SOURCE PYTHON COURSES
FOR GEOGRAPHERS
(basics of programming + GIS)**



Automating GIS processes
Automating GIS processes

autogis.github.io





DIGITAL GEOGRAPHY LAB



**Tuuli
Toivonen**



**Enrico
Di Minin**



**Olle
Järv**



**Henrikki
Tenkanen**



**Tuomo
Hiippala**



**Anna
Hausmann**



**Vuokko
Heikinheimo**



**Christoph
Fink**



**Maria
Salonen**



**Gonzalo
Cortés Capano**



**Joel
Jalkanen**



**Johanna
Eklund**



**Jeison
Londoño**



**Elias
Willberg**

Supporting sustainable
spatial planning
using digital data
and novel methods



RESEARCH THEMES



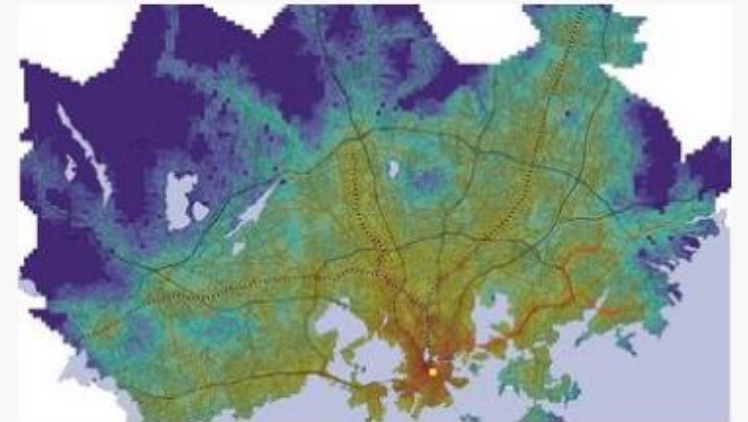
MOBILITY & ACCESSIBILITY

Urban patterns, cross-border flows and global networks of people, goods and information



NATURE CONSERVATION

Conservation planning, nature-based tourism and wildlife trade



NOVEL DATA AND METHODS

Developing open source data and methods for spatio-temporal analysis



OPEN ACCESS

Prospects and challenges for social media data in conservation science

Enrico Di Minin^{1,2*}, Henrikki Tenkanen³ and Tuuli Toivonen^{1,2*}

¹ Finnish Centre of Excellence in Metapopulation Biology, Department of Biosciences, University of Helsinki, Helsinki, Finland, ² School of Life Sciences, University of KwaZulu-Natal, Durban, South Africa, ³ Department of Geosciences and Geography, University of Helsinki, Helsinki, Finland

Social media data have been extensively used in numerous fields of science, but examples of their use in conservation science are still very limited. In this paper, we propose a framework on how social media data could be useful for conservation science and practice. We present the commonly used social media platforms and discuss how their content could provide new data and information for conservation science. Based on this, we discuss how future work in conservation science and practice would benefit from social media data.

Keywords: social media, Twitter, Instagram, Facebook, biodiversity conservation, management, big data, national parks

Article

User-Generated Geographic Information for Visitor Monitoring in a National Park: A Comparison of Social Media Data and Visitor Survey

Vuokko Heikinheimo^{1,*}, Enrico Di Minin^{1,2}, Henrikki Tenkanen¹, Anna Hausmann^{1,2}, Joel Erkkonen³ and Tuuli Toivonen¹

¹ Department of Geosciences and Geography, University of Helsinki, 00014 Helsinki, Finland;

SCIENTIFIC REPORTS

OPEN

Instagram, Flickr, or Twitter: Assessing the usability of social media data for visitor monitoring in protected areas

Received: 24 July 2017
Accepted: 5 December 2017
Published online: 14 December 2017

Henrikki Tenkanen¹, Enrico Di Minin^{1,2},
Marna Herbst¹, Liisa Kajala¹ & Tuuli Toivonen^{1,2}

SCIENTIFIC REPORTS

OPEN

Social media reveal that charismatic species are not the main attractor of ecotourists to sub-Saharan protected areas

Received: 12 December 2016
Accepted: 15 March 2017
Published online: 10 April 2017

Anna Hausmann¹, Tuuli Toivonen¹, Vuokko Heikinheimo¹, Henrikki Tenkanen¹,
Rob Slotow^{1,2,3} & Enrico Di Minin^{1,2}

Charismatic megafauna are arguably considered the primary attractor of ecotourists to sub-Saharan African protected areas. However, the lack of visitation data across the whole continent has thus far prevented the investigation of whether charismatic species are indeed a key attractor of ecotourists to protected areas. Social media data can now be used for this purpose. We mined data from Instagram, and used generalized linear models with site- and country-level deviations to explore which socio-economic, geographical and biological factors explain social media use in sub-Saharan African protected areas. We found that charismatic species richness did not explain social media usage. On the other

Diversity

A framework for investigating illegal wildlife trade on social media with machine learning

Enrico Di Minin^{1,2,3*}, Christoph Fink^{1,2}, Tuomo Hiippala^{1,2,4} and Henrikki Tenkanen^{1,2}

¹ Digital Geography Lab, Department of Geosciences and Geography, University of Helsinki, Helsinki, FI 00014, Finland
² Helsinki Institute of Sustainability Science, University of Helsinki, Helsinki, FI 00014, Finland
³ School of Life Sciences, University of KwaZulu-Natal, KwaZulu-Natal, Durban 4000, South Africa
⁴ Department of Languages, University of Helsinki, Helsinki, FI 00014, Finland

Conservation Letters

A journal of the Society for Conservation Biology

Open Access

LETTER

Social Media Data Can Be Used to Understand Tourists' Preferences for Nature-Based Experiences in Protected Areas

Anna Hausmann¹, Tuuli Toivonen², Rob Slotow^{1,3}, Henrikki Tenkanen², Atte Moilanen⁴, Vuokko Heikinheimo², & Enrico Di Minin^{1,2}

¹ Amarula Elephant Research Programme, School of Life Sciences, University of KwaZulu Natal, Durban 4041, South Africa
² Department of Geosciences and Geography, University of Helsinki, FI 00014 Helsinki, Finland
³ Department of Genetics, Evolution and Environment, University College, London, WC1E 6BT, UK
⁴ Finnish Centre of Excellence in Metapopulation Research, Department of Biosciences, University of Helsinki, 00014 Helsinki, Finland

Keywords

Biodiversity conservation, conservation marketing, ecotourism, Flickr, Instagram, social media.

Correspondence

Anna Hausmann, Amarula Elephant Research Programme, School of Life Sciences, University of KwaZulu Natal, Durban 4041, South Africa. Tel: +358 41 7088225. E-mail: anna.hausmann87@gmail.com

Received

21 June 2016

Revised

14 December 2016

Accepted

15 January 2017

Abstract

Can social media data be used as an alternative to understand tourists' preferences for nature-based experiences? We explored this by comparing preferences for nature-based experiences from a traditional survey conducted in Kruger National Park with preferences revealed by social media content. We found that charismatic species were found to be the favorite group, both in the survey and in social media content. However, Flickr was found to better reflect less-charismatic biodiversity. Our findings suggest that social media data can be used as a cost-efficient way to explore, and monitor, preferences for biodiversity and human

correspondence

Machine learning for tracking illegal wildlife trade on social media

To the Editor — Illegal trade in wildlife is booming on e-commerce platforms, the 'dark web' and social media^{1,2}. The ease of

techniques to social media data allows human behaviour to be investigated on an unprecedented scale. Yet these techniques

species by their songs and calls, as well as interrogating the image stream. Natural language processing can be used to infer



ACCESSIBILITY ...

Car, bus or bike?



Sustainable cities?



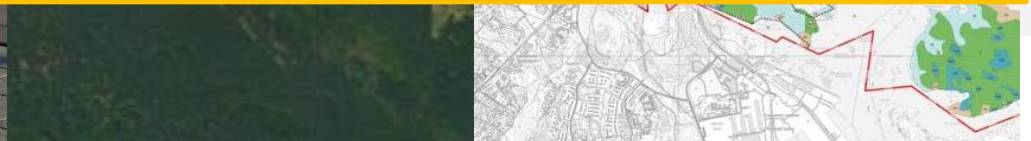
Carbon footprint?



Landcover change?



...binds together questions about traffic, landuse, housing, trade, and the environment!

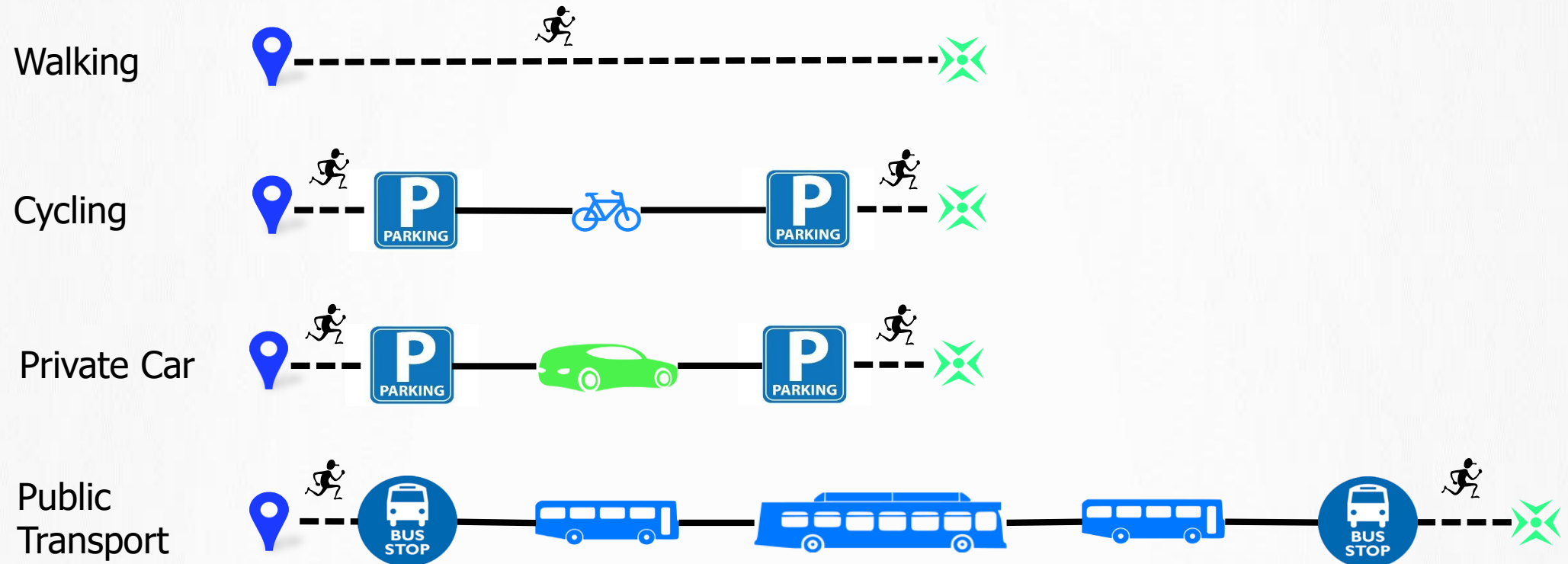




NEED FOR COMPARABLE MEASURES OF ACCESSIBILITY



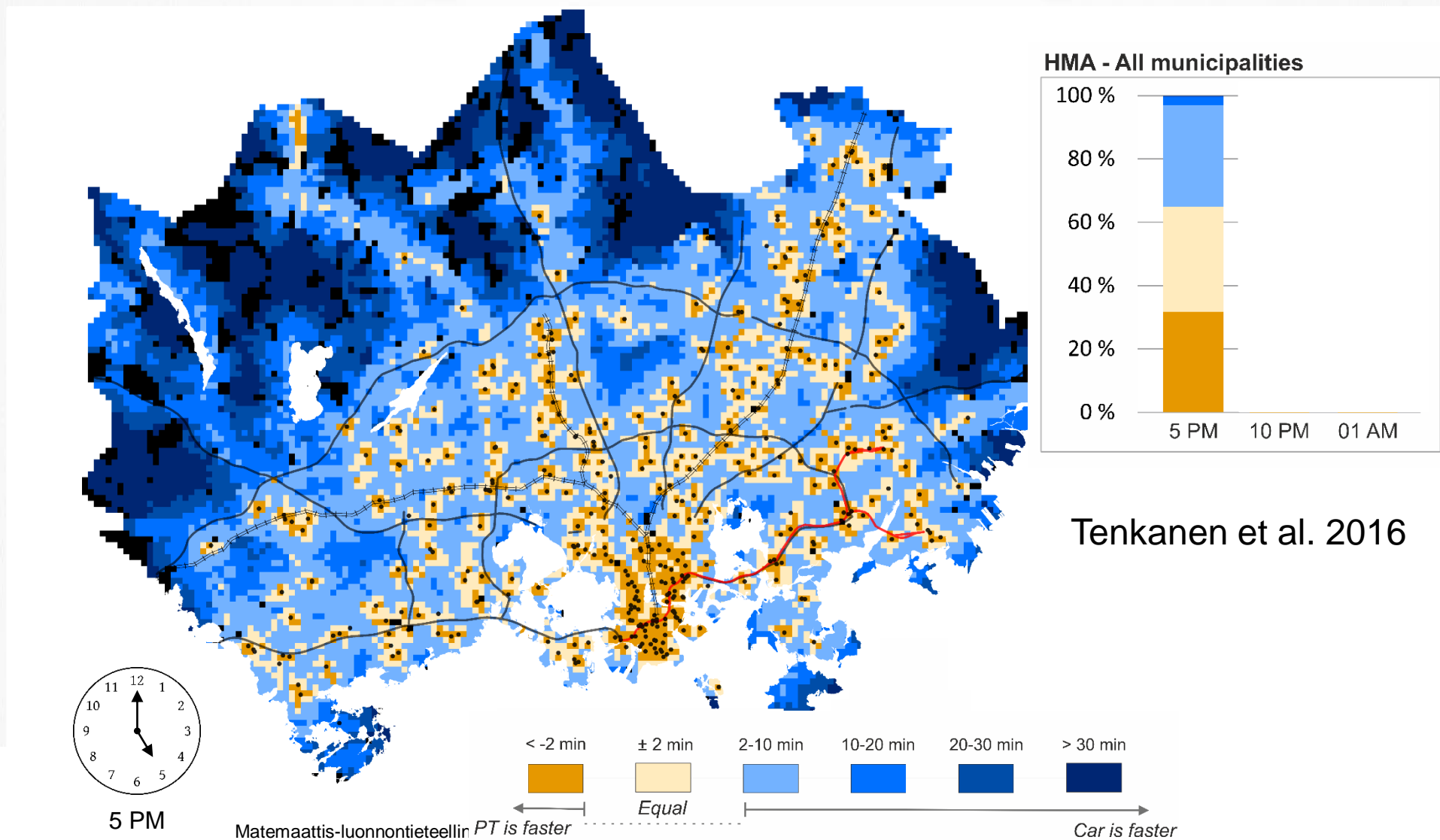
COMPARABLE MEASURES WITH DOOR-TO-DOOR APPROACH



Salonen & Toivonen (2013)



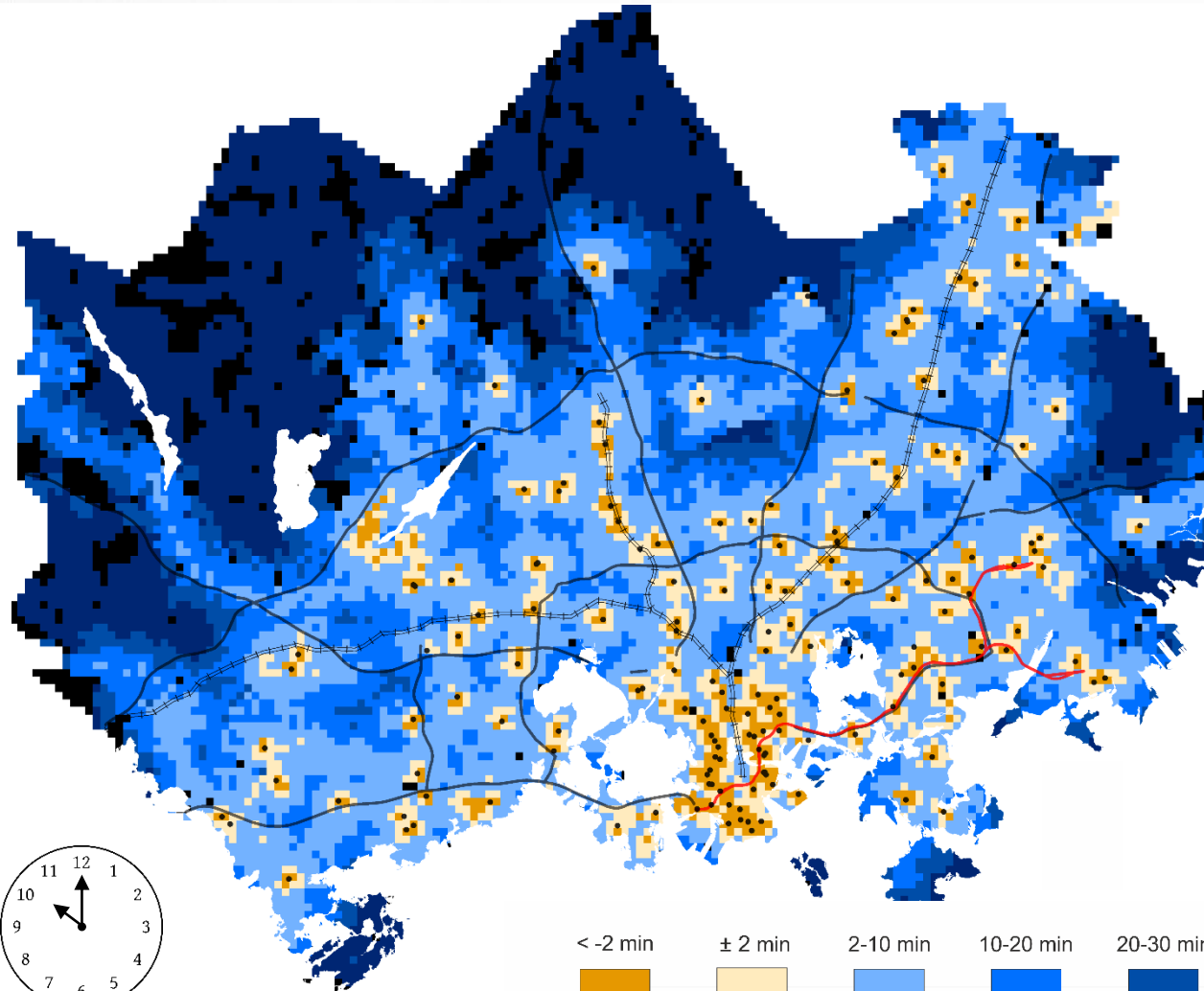
TRIP TO CLOSEST FOOD STORE AT 5 PM DIFFERENCE BETWEEN TRANSIT AND CAR



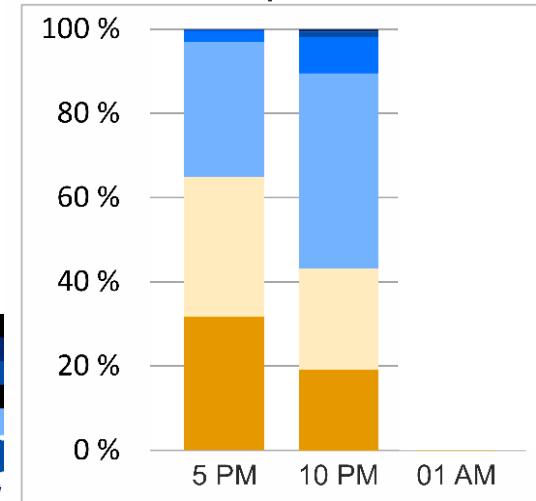
Tenkanen et al. 2016



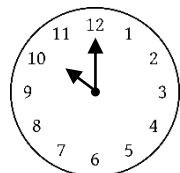
TRIP TO CLOSEST FOOD STORE AT 10 PM DIFFERENCE BETWEEN TRANSIT AND CAR



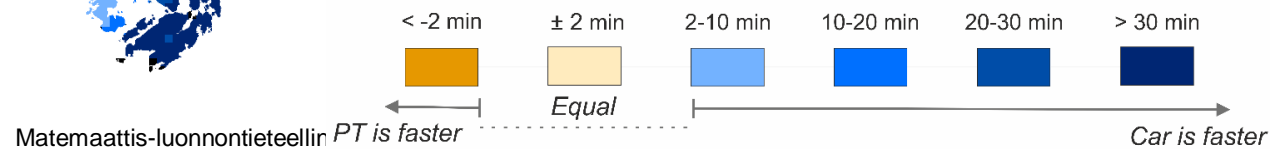
HMA - All municipalities



Tenkanen et al. 2016

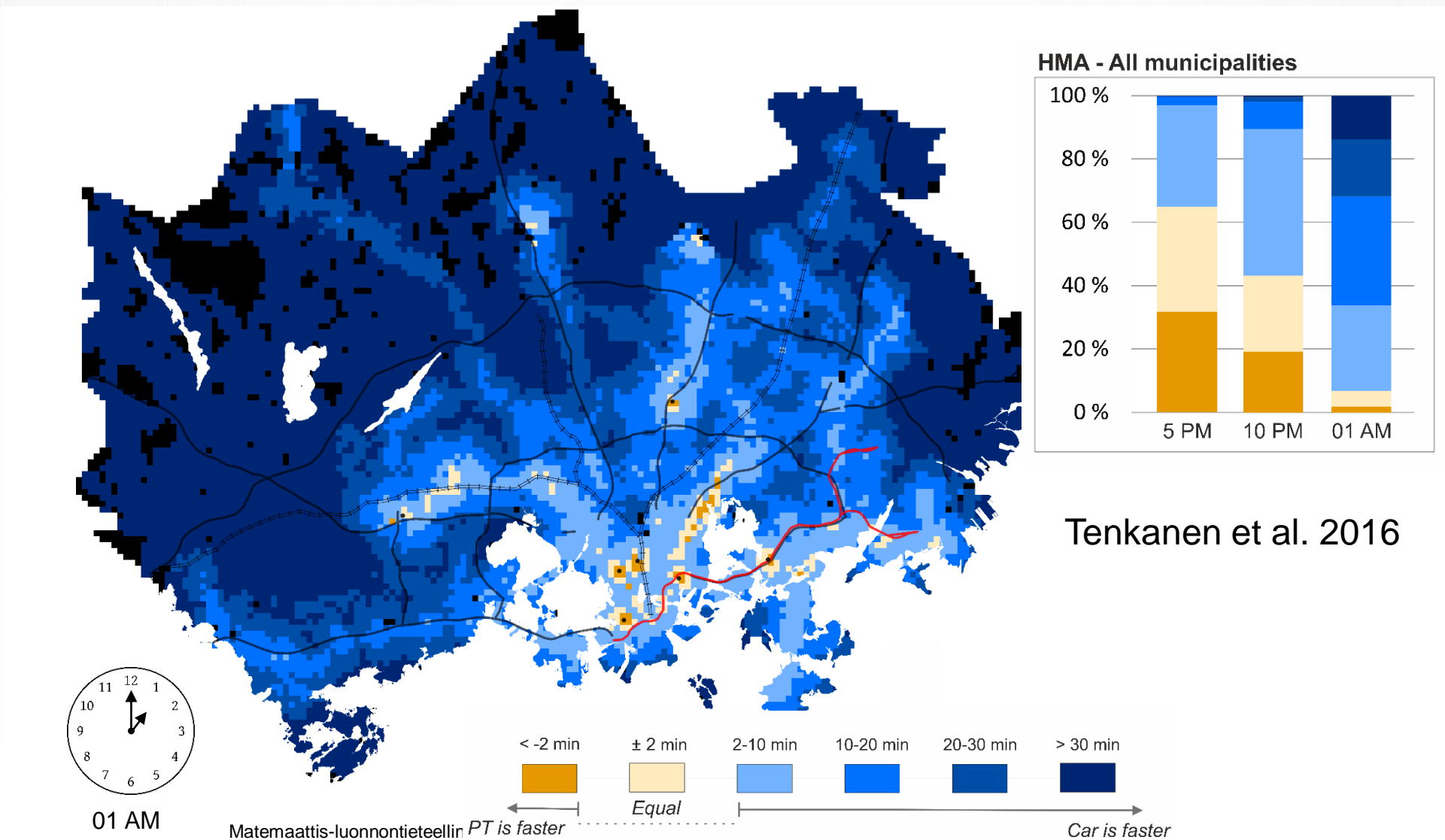


10 PM





TRIP TO CLOSEST FOOD STORE AT 01 AM DIFFERENCE BETWEEN TRANSIT AND CAR

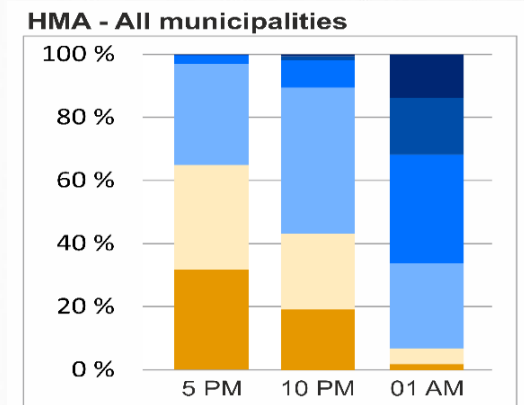
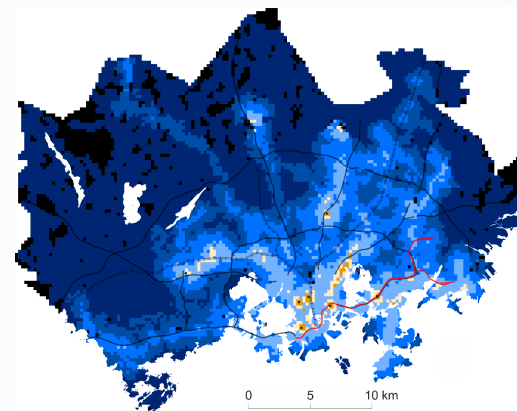
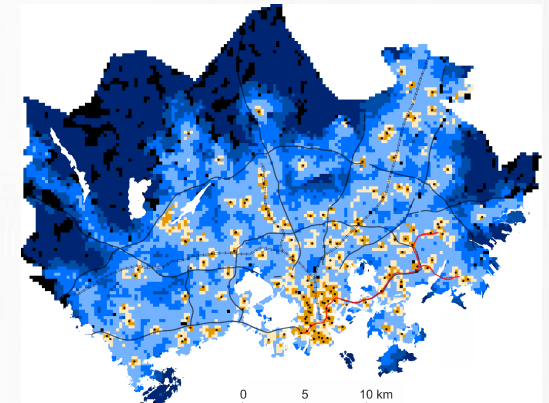
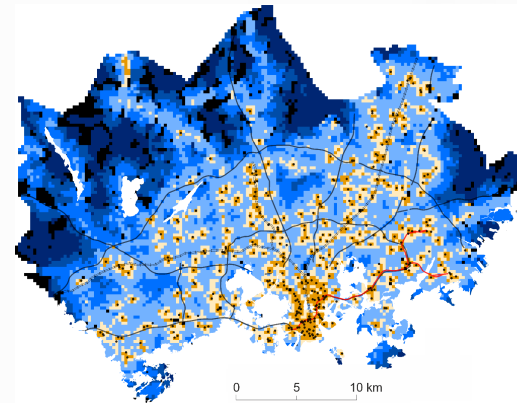


Tenkanen et al. 2016



The competitiveness of specific travel mode is time dependent

The problem in here?



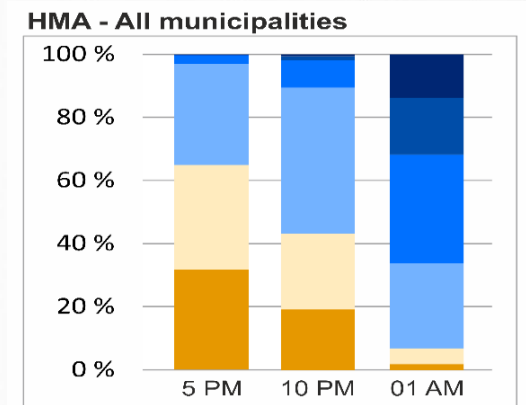
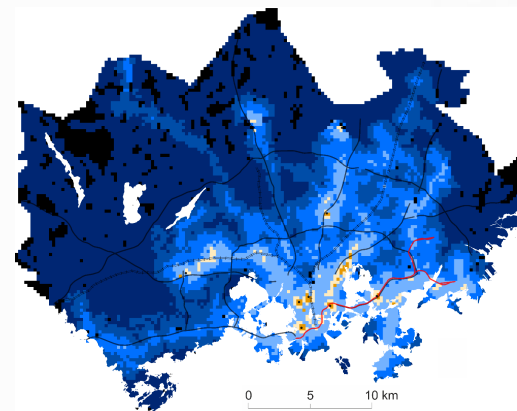
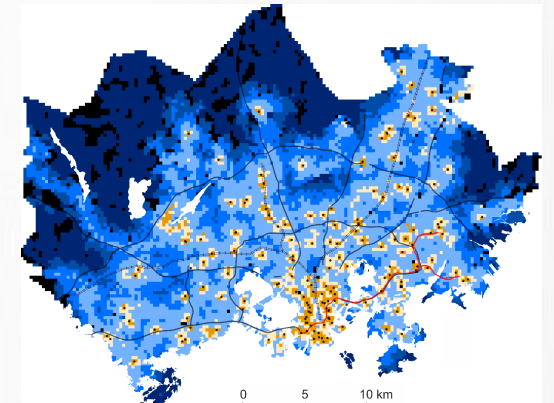
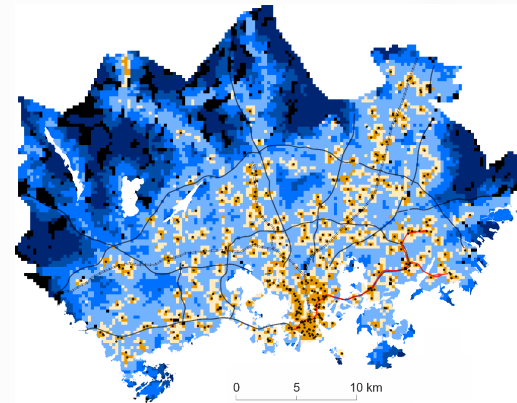


The competitiveness of specific travel mode is time dependent

The problem in here?



We assume that people are where they sleep.





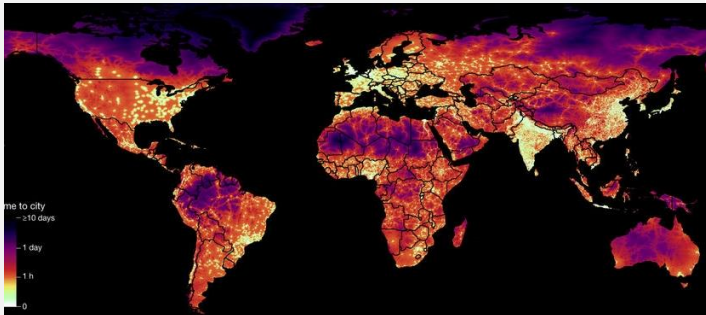
HOW TO CONDUCT “FULLY DYNAMIC” ANALYSES OF ACCESSIBILITY?



TIME IN ACCESSIBILITY MODELLING?

Location-based models

- ❑ Focus in understanding large- and regional scale patterns
- ❑ Mostly analyzed in static terms



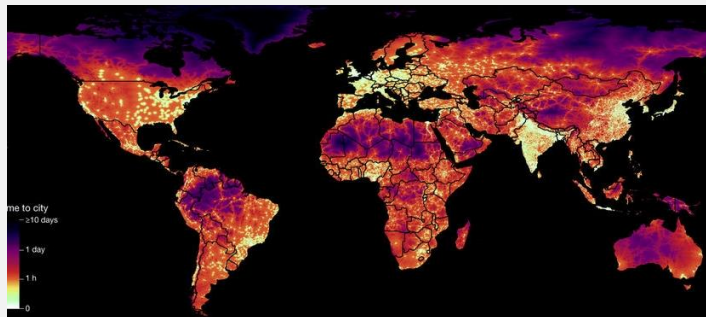
Weiss et al. (2018). *Nature*.



TIME IN ACCESSIBILITY MODELLING?

Location-based models

- ❑ Focus in understanding large- and regional scale patterns
- ❑ Mostly analyzed in static terms

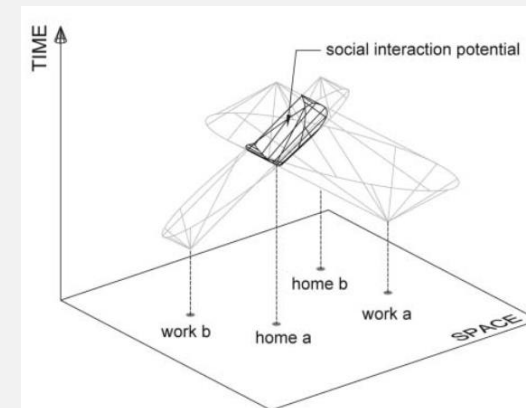


Weiss et al. (2018). *Nature*.

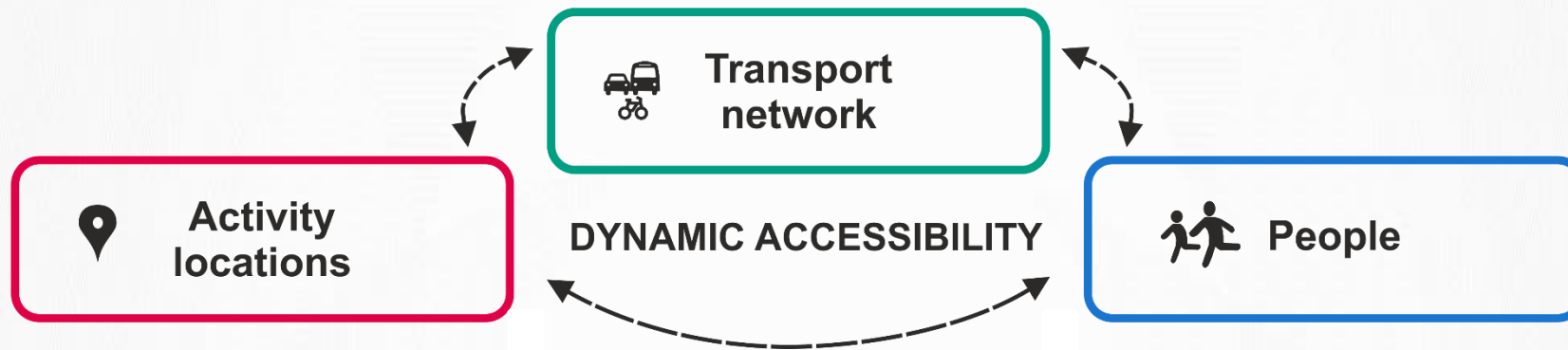


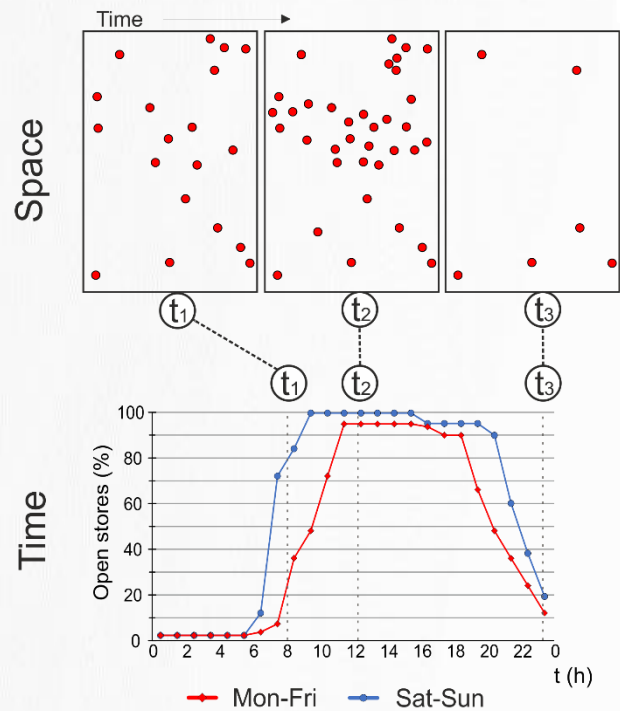
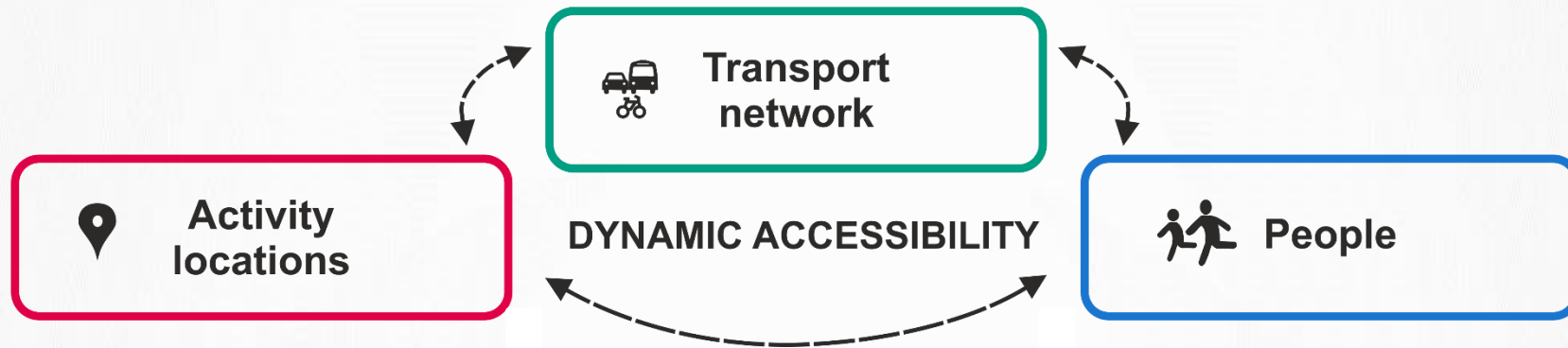
Person-based models

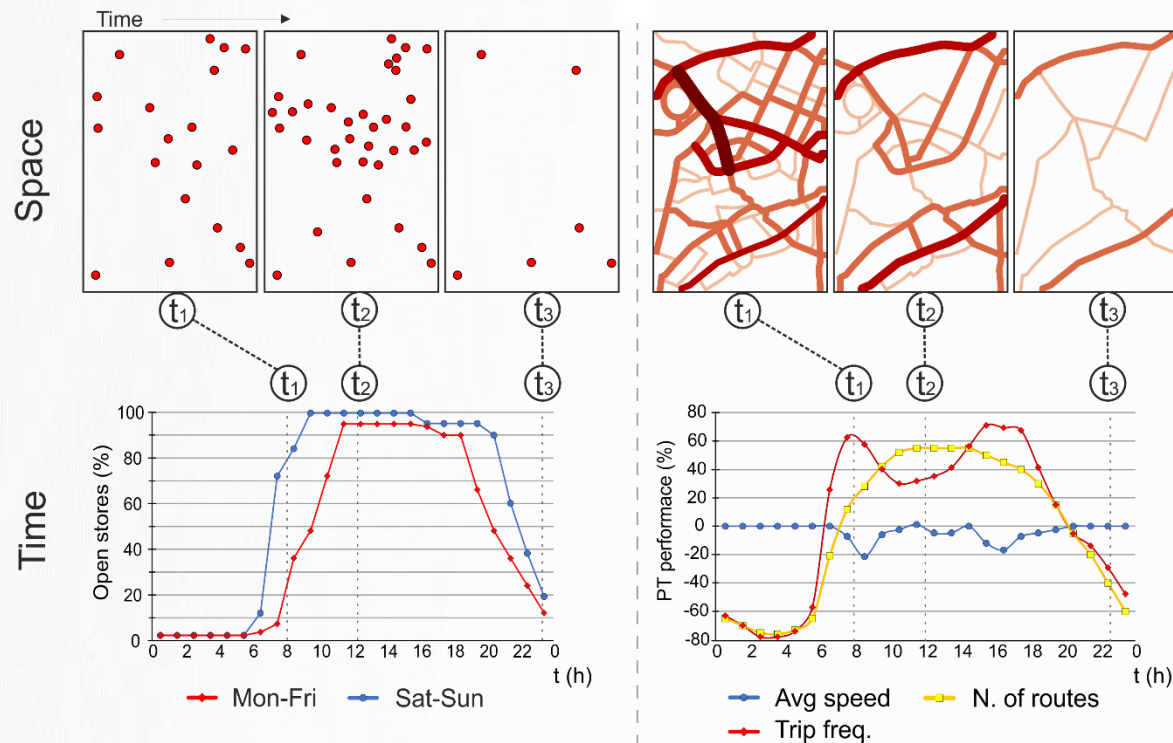
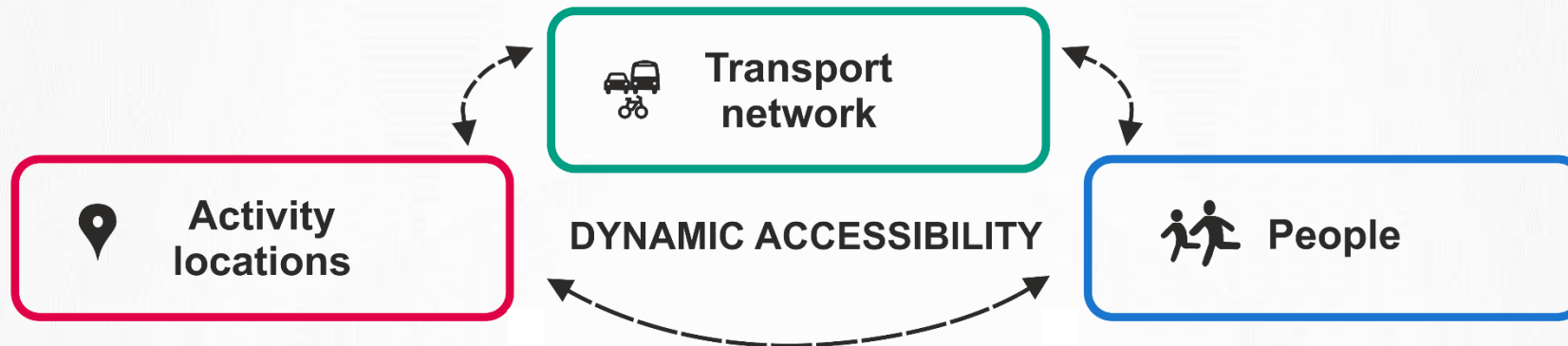
- ❑ Focus in understanding small(er) scale patterns
- ❑ Time has been incorporated already for long (Hägerstrand 1970; Miller 1991; Kwan 1998)

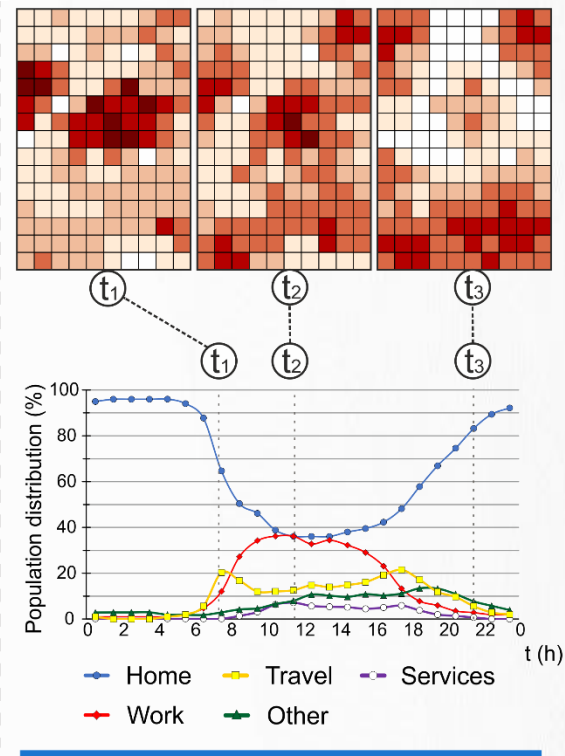
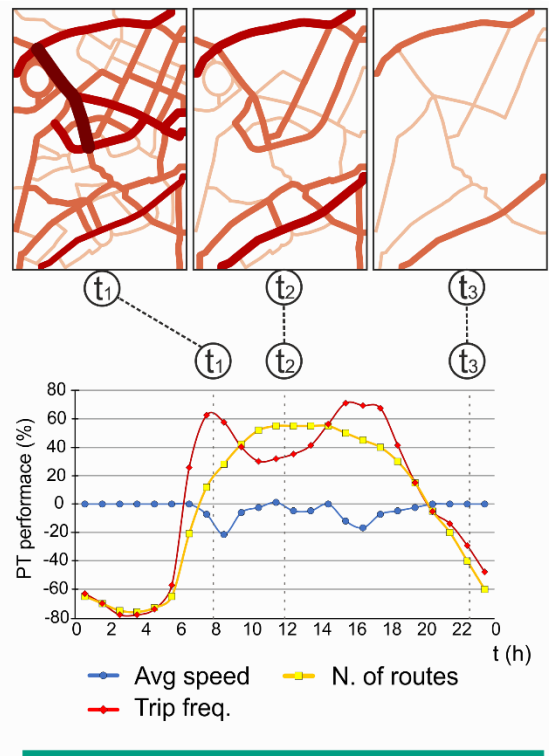
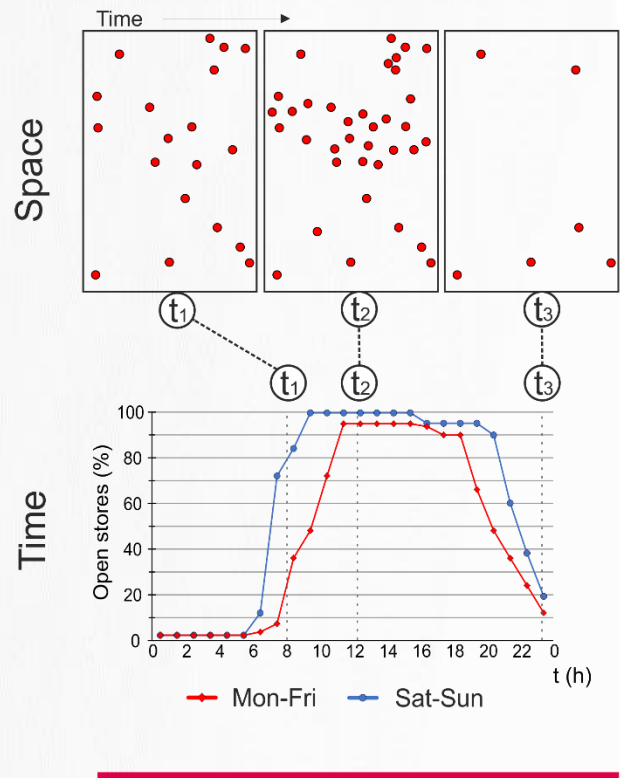
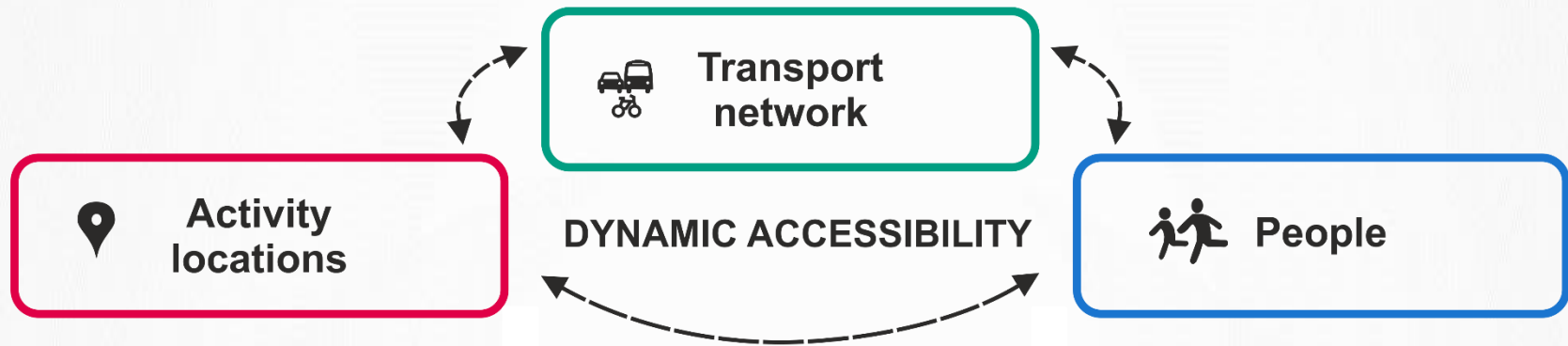


Farber et al. (2013). *Ann. of AAG*.



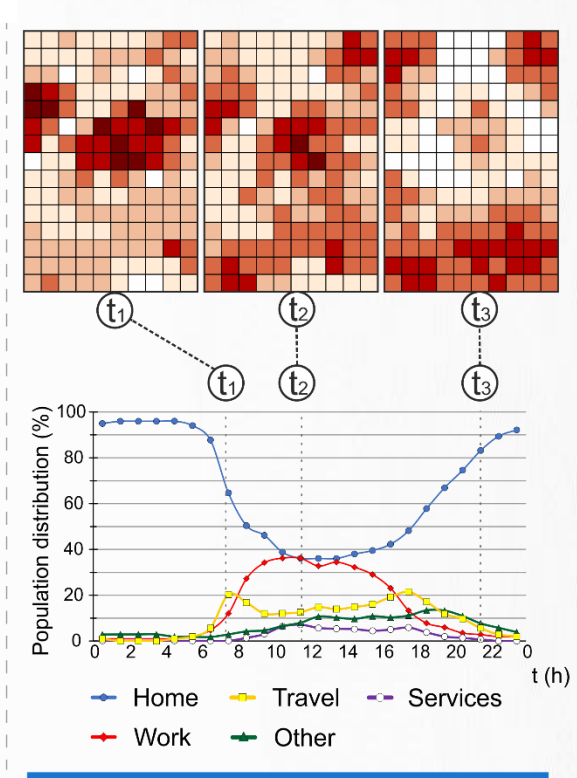
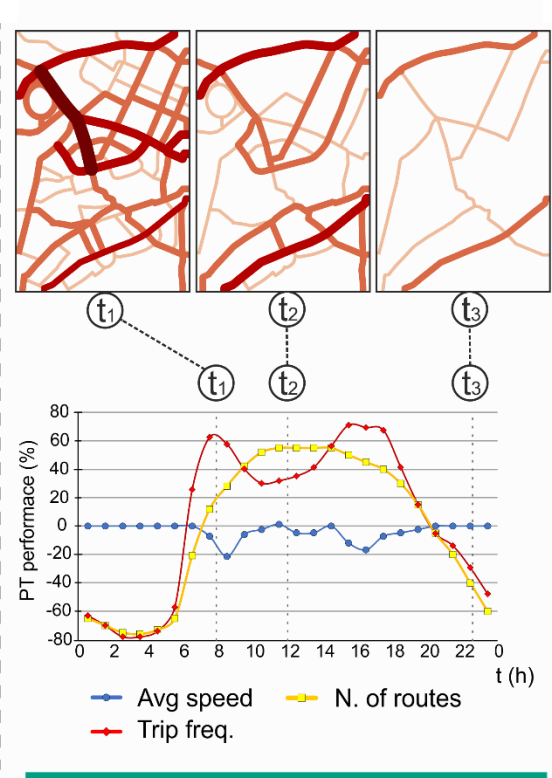
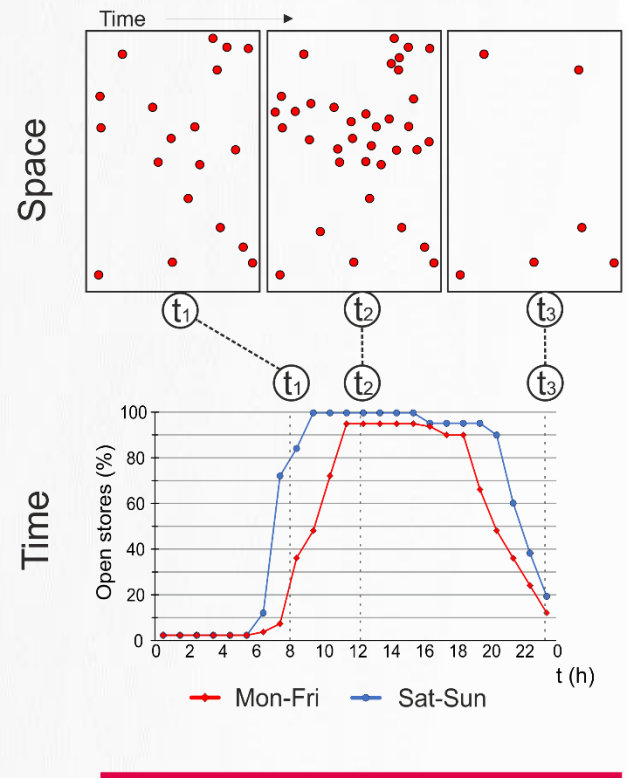
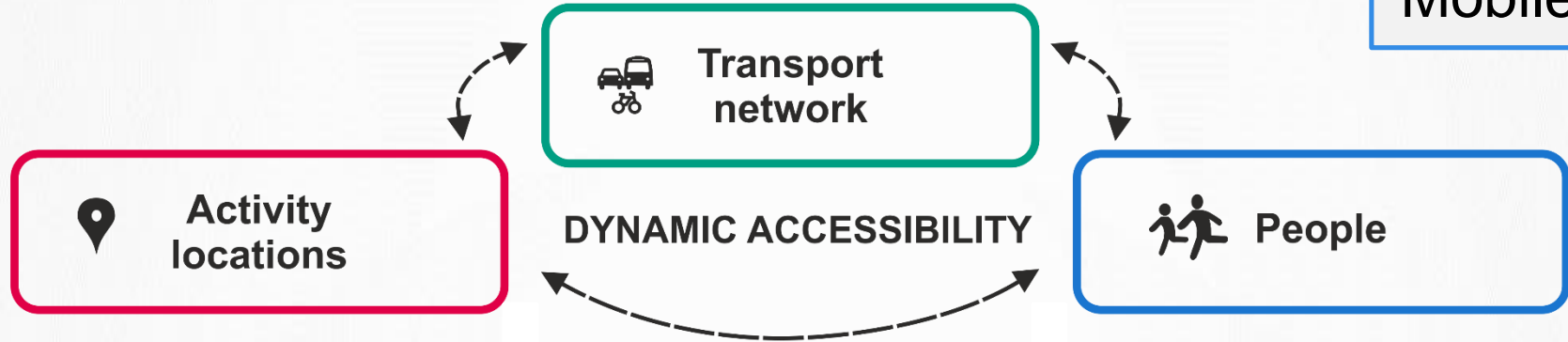








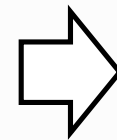
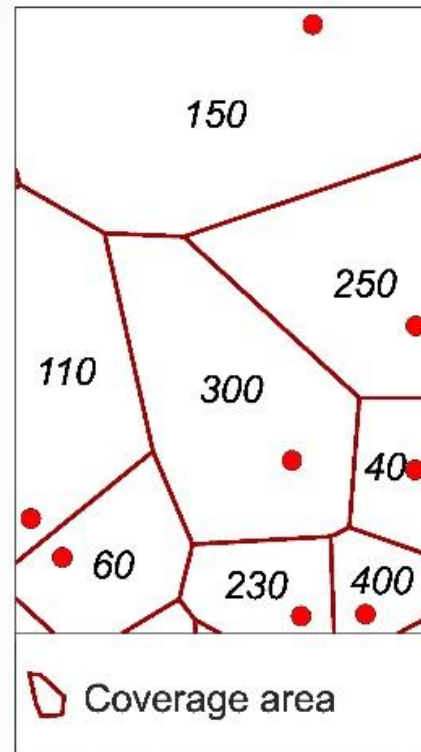
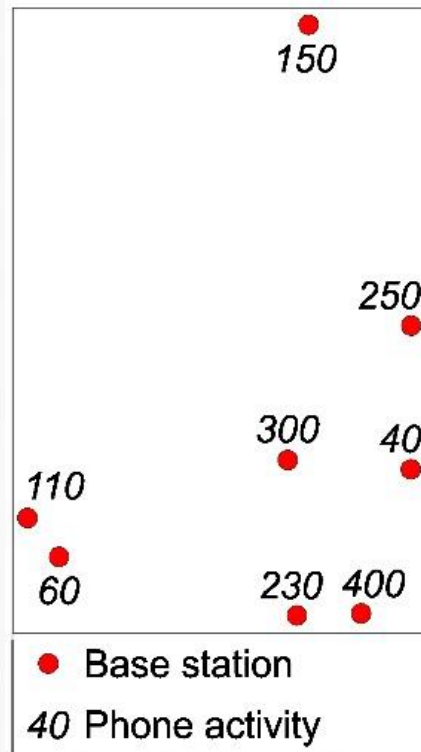
Mobile phone data



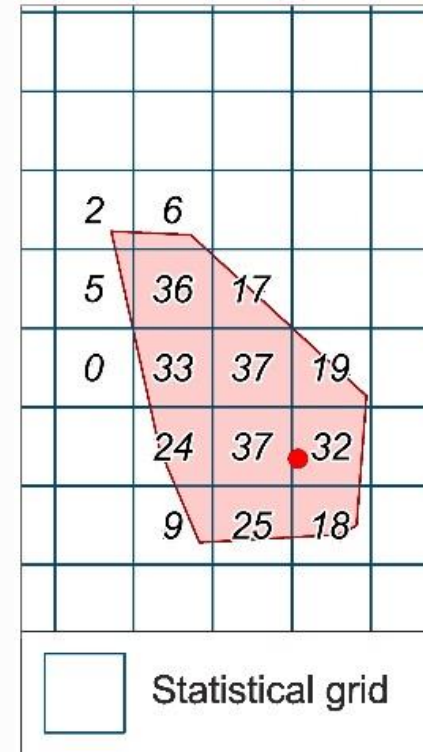


HOW TO HARMONIZE MOBILE PHONE DATA?

Uneven spatial configuration

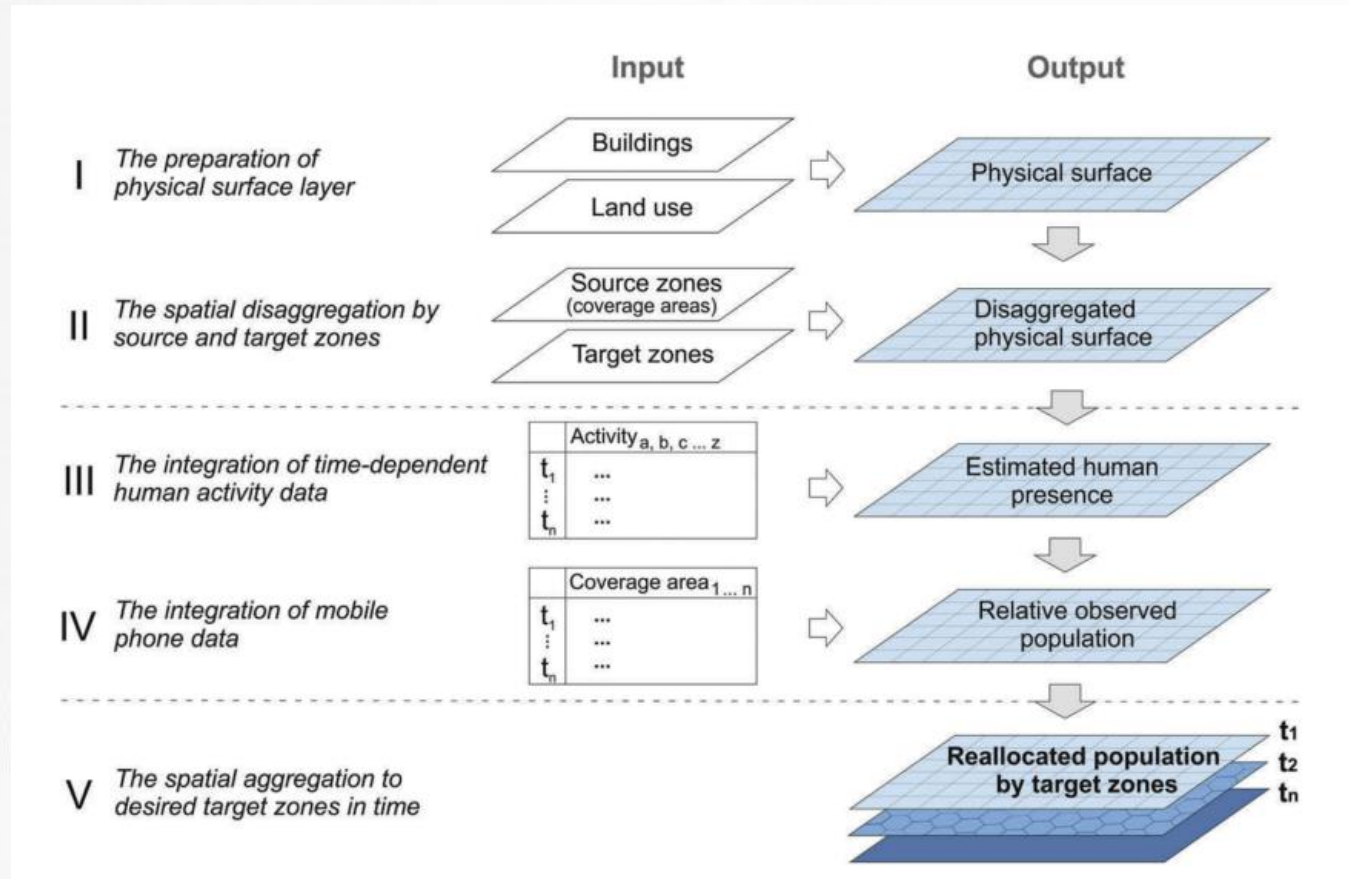


Harmonized standardized grid





Improving accuracy of mobile phone based data



Data fusion and harmonization model increase the quality of the data

(Järvi, Tenkanen & Toivonen 2017, IJGS)

A wide-angle, high-angle photograph of a busy city street on a sunny day. In the foreground, a paved pedestrian crossing with white stripes is visible. A yellow and green tram is moving along the street. Pedestrians are walking in various directions. In the background, there are modern and classical buildings, trees, and a clear blue sky. A semi-transparent grey box is overlaid on the center of the image, containing white text.

24H CITY

How much time influences accessibility?

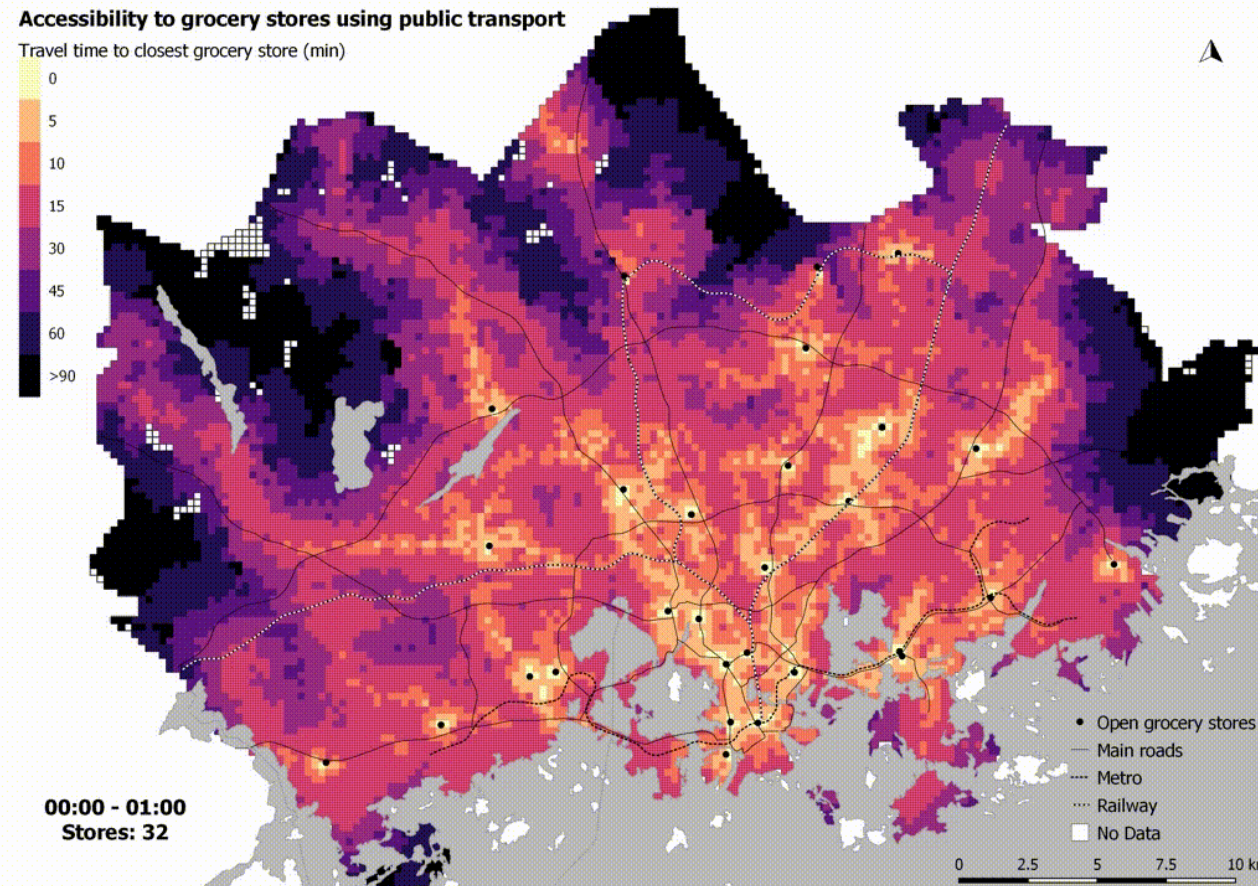
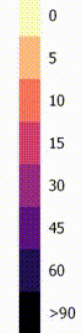
People – Transport System – Activity locations



24H ACCESSIBILITY

Accessibility to grocery stores using public transport

Travel time to closest grocery store (min)



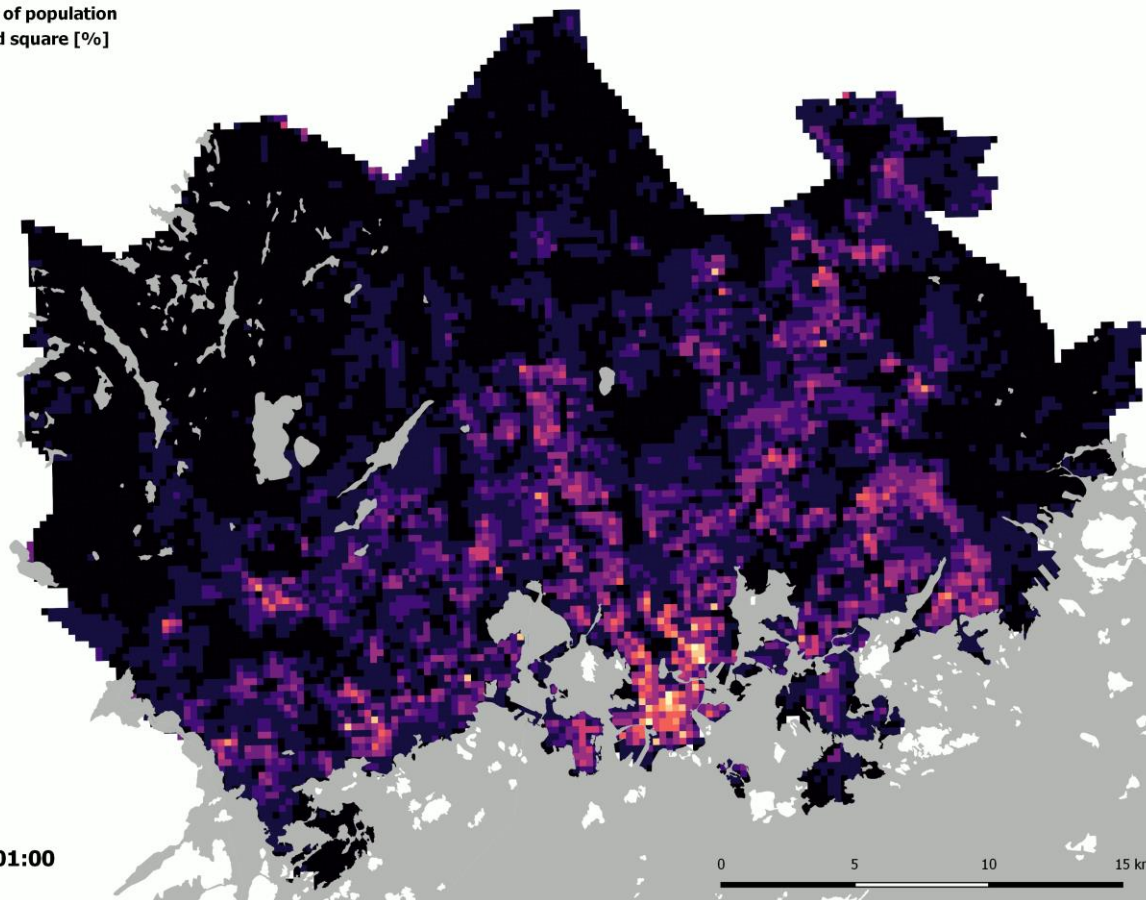
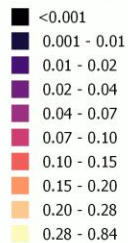
00:00 - 01:00
Stores: 32

Travel times to
closest open grocery
store at each hour
of the day – Jan 2018



24H MOBILITY

Relative share of population
in a 250 m grid square [%]



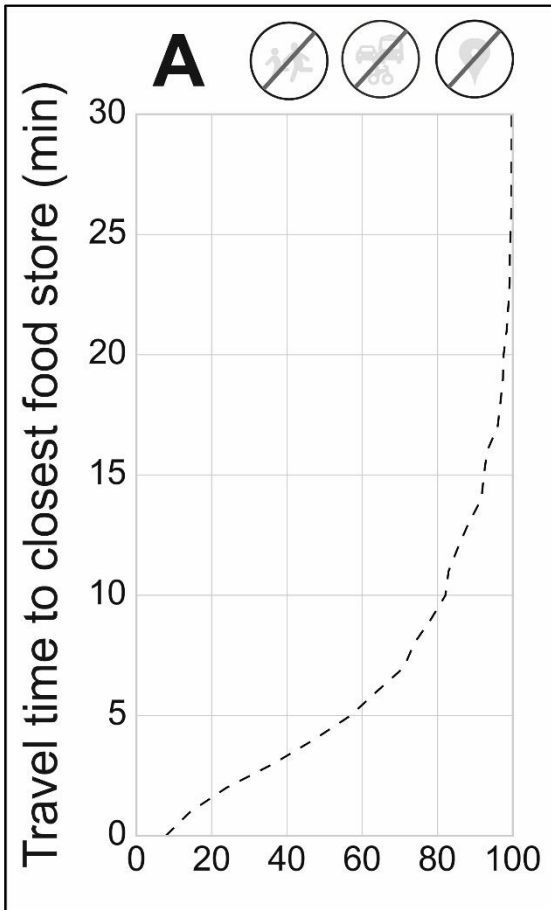
00:00-01:00

0 5 10 15 km

Relative distribution of
people at each hour of
the day

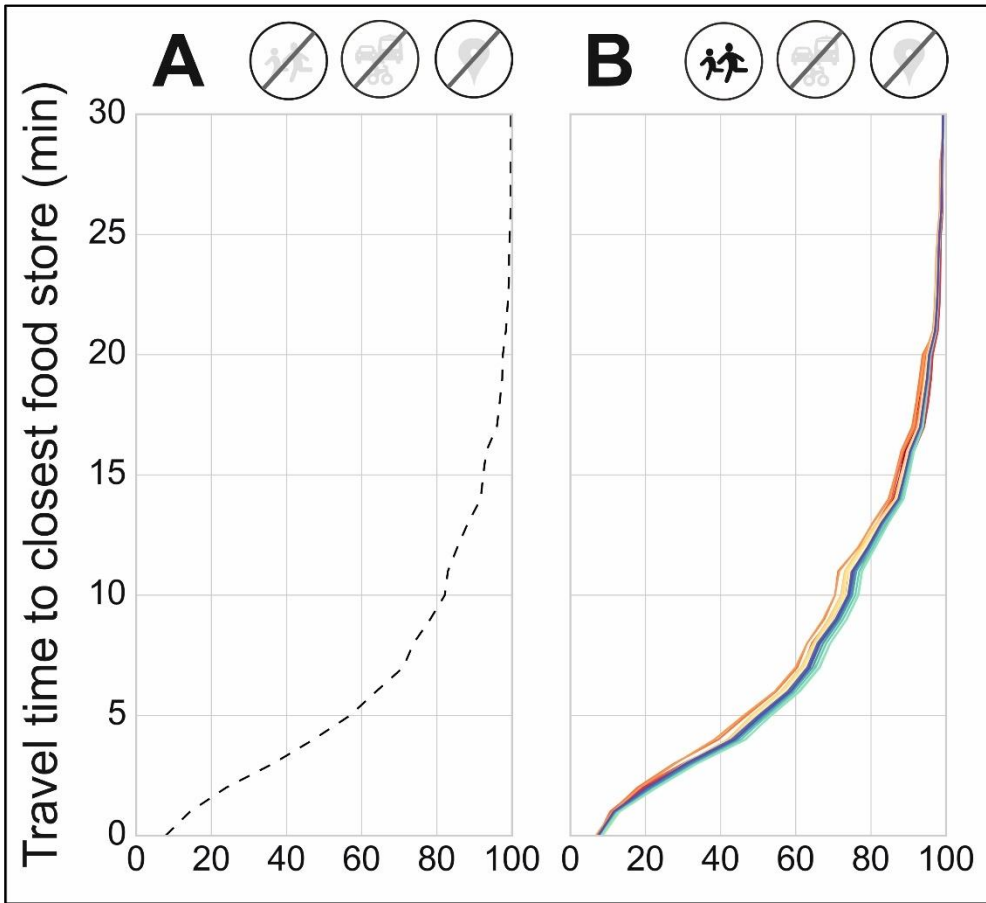
(based on model by:
Järv, Tenkanen & Toivonen 2017)

MSc thesis work of Claudia Bergroth



Share of population reaching the closest grocery store

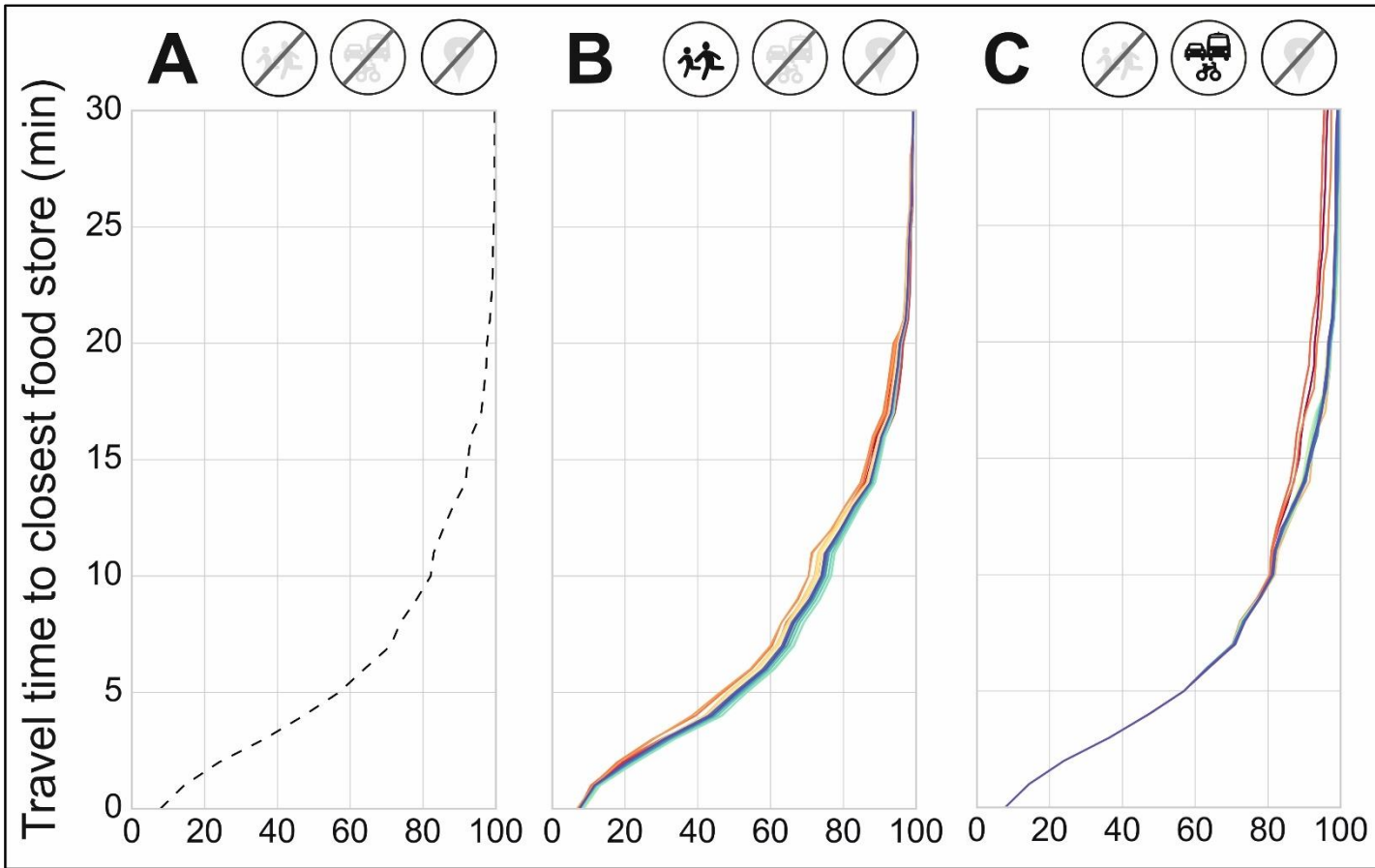
A Fully atemporal model



Share of population reaching the closest grocery store

A Fully atemporal model

B Only people dynamic

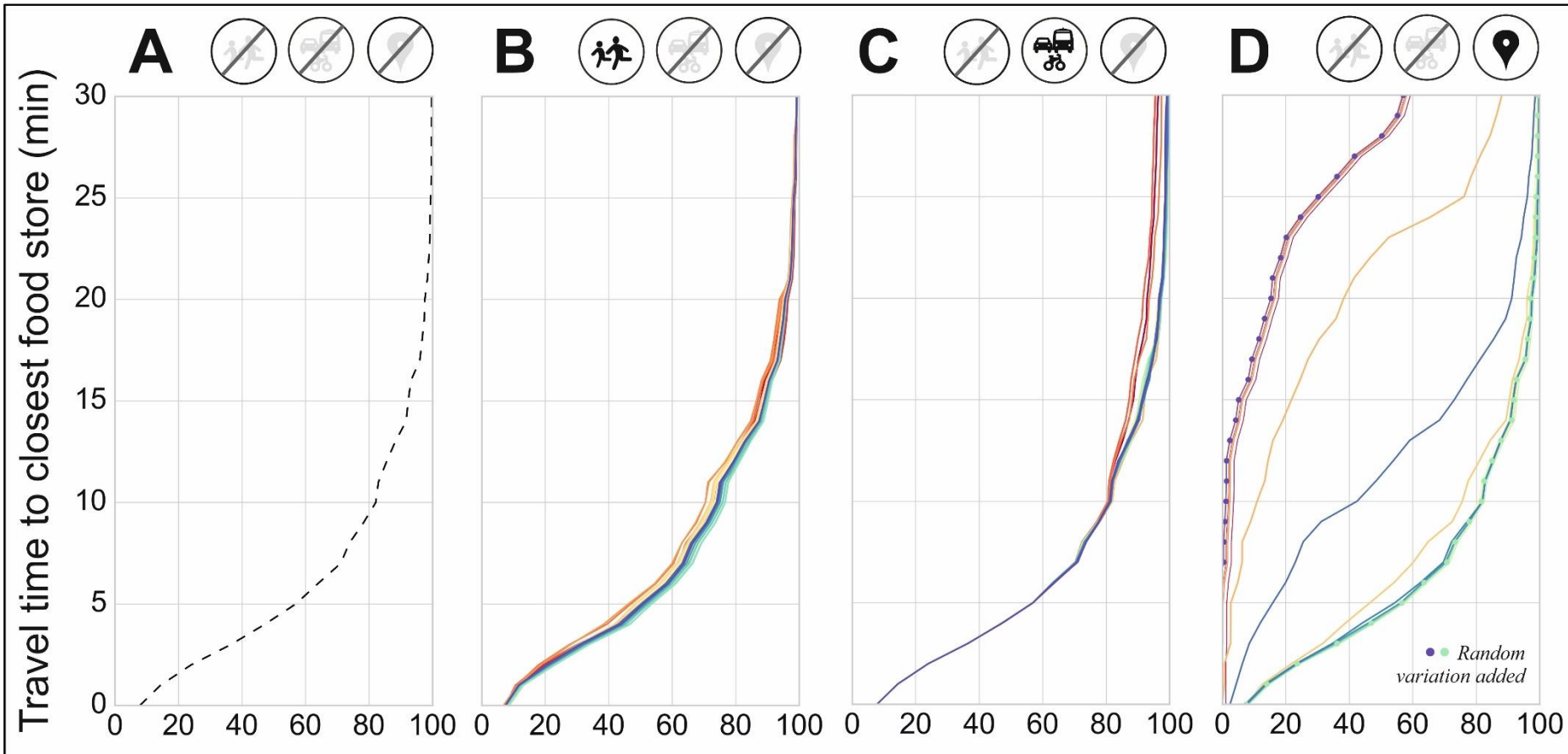


Share of population reaching the closest grocery store

A Fully atemporal model

B Only people dynamic

C Only transport network dynamic



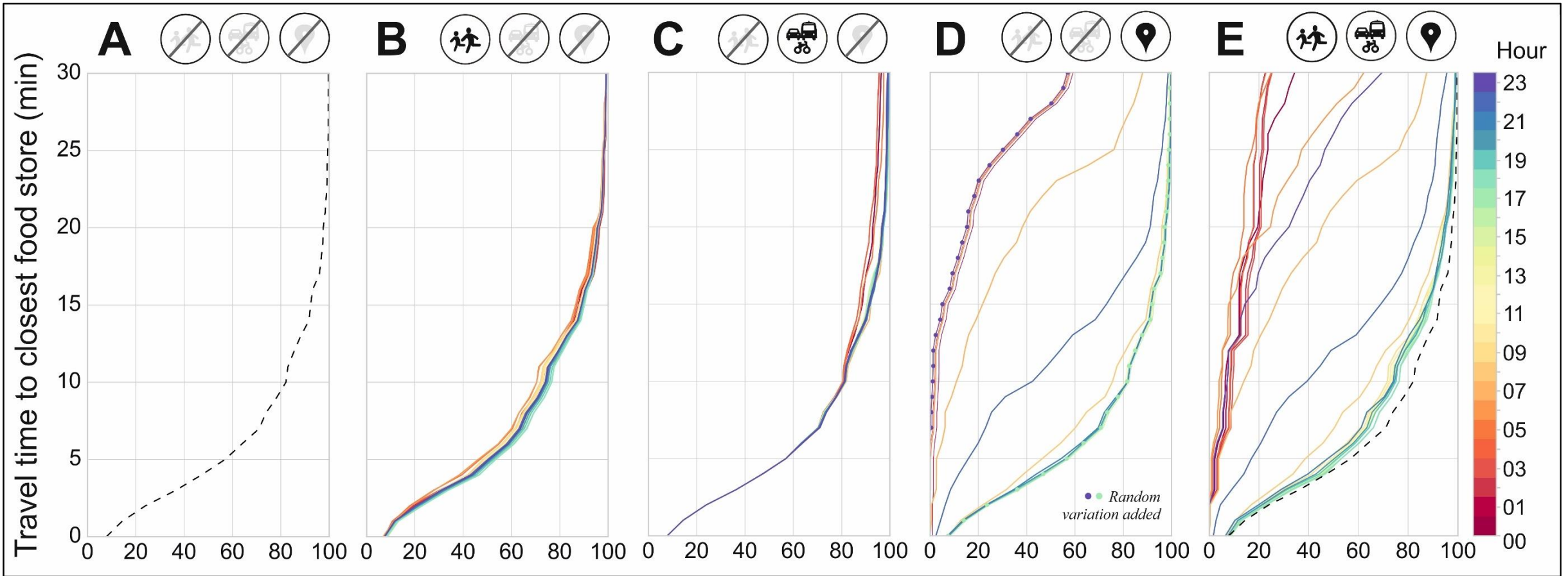
Share of population reaching the closest grocery store

A Fully atemporal model

B Only people dynamic

C Only transport network dynamic

D Only services dynamic



Share of population reaching the closest grocery store

**FROM ~10
UP TO 70 %
DIFFERENCE!!**

A Fully atemporal model

B Only people dynamic

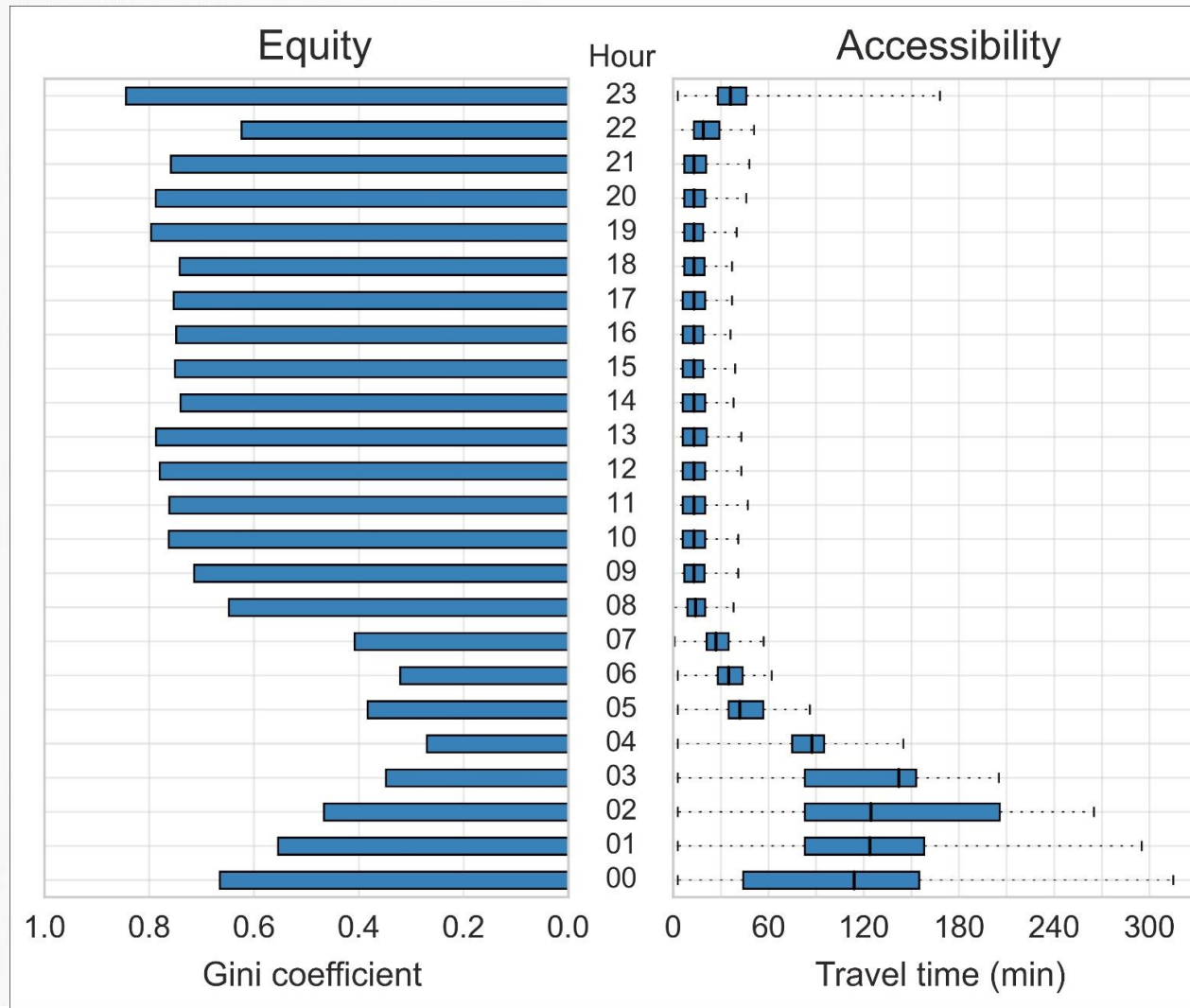
C Only transport network dynamic

D Only services dynamic

E All components dynamic



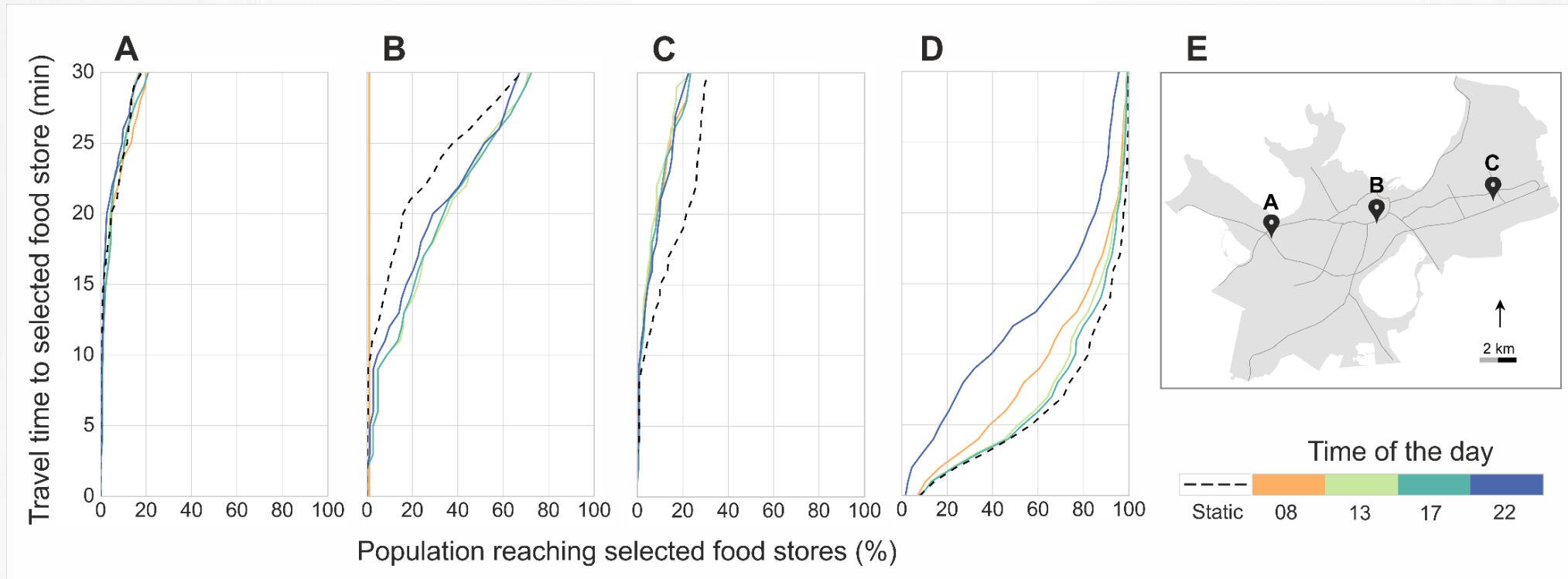
Effect of temporal variation in accessibility on equity?



Spatial equity on access to grocery stores varies in time



The significance of dynamism is network / location specific



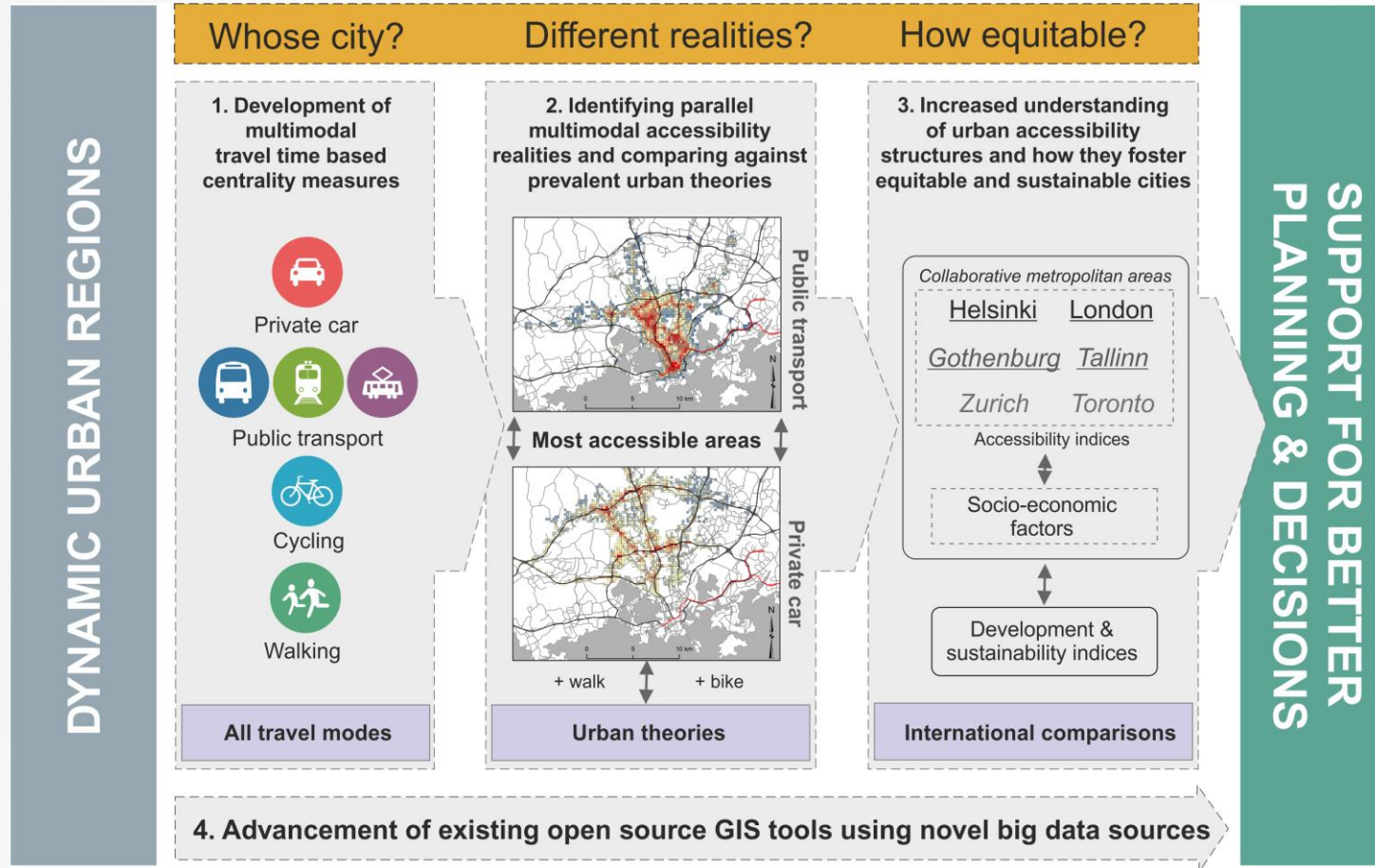


ANALYZING MODAL EQUITY IN TRANSPORT USING BIG DATA?

Project running 2019 -



PROJECT / STUDY DESIGN



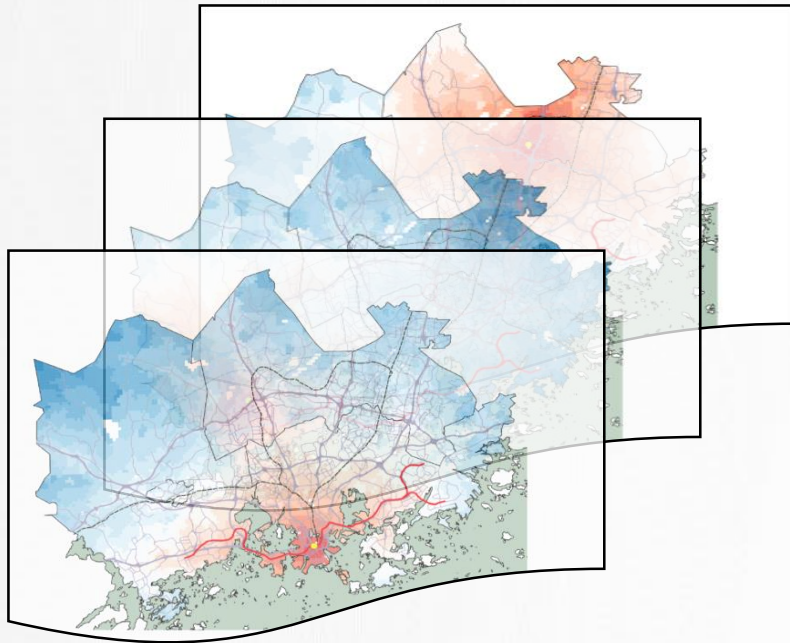
Car vs Public transport

- ❑ Do the "realities of accessibility" differ in cities based on mode of transport?
- ❑ Where are the "city centers" of a city? → Centrality
- ❑ Influence on modal equity



METHODS – MULTIMODAL CENTRALITY

INPUT



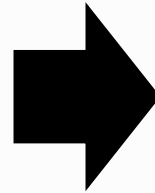
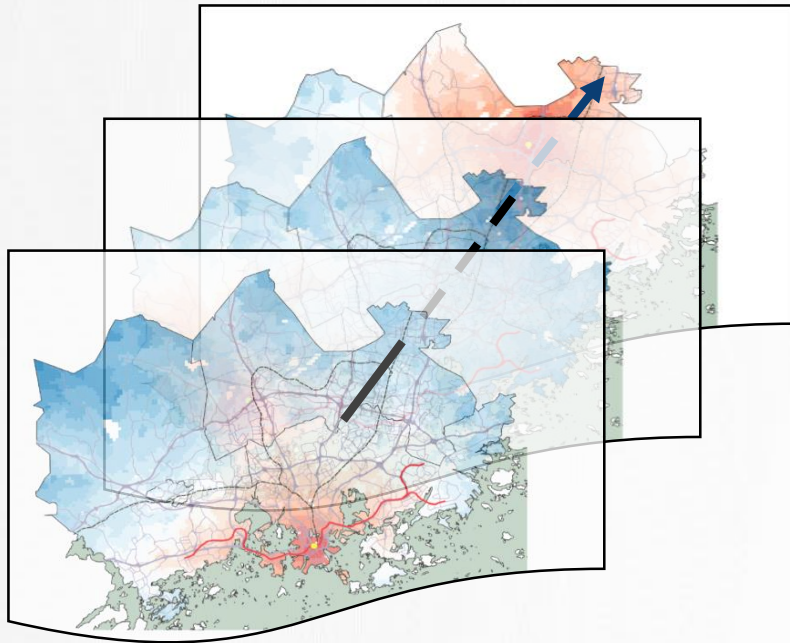
As input we have travel time layers to all locations in the study region



METHODS – MULTIMODAL CENTRALITY

INPUT

ANALYSIS



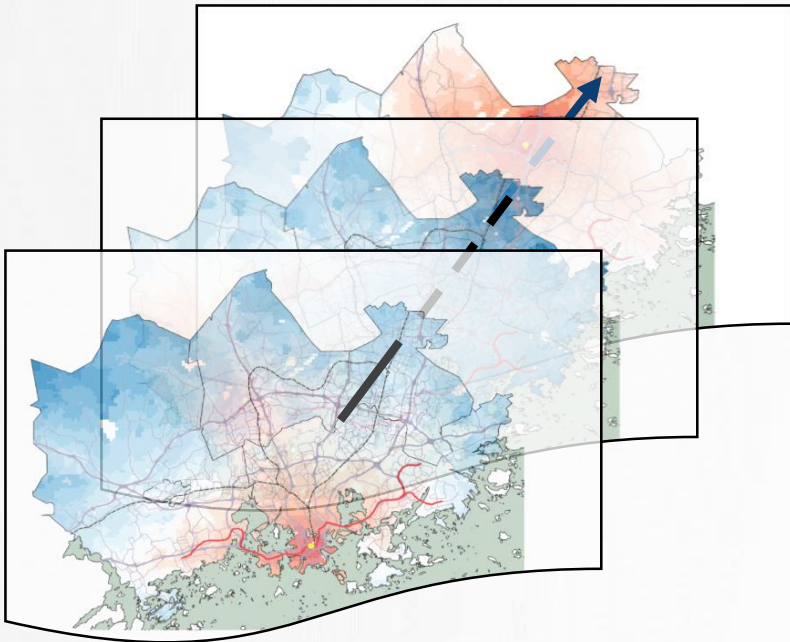
As input we have travel time layers to all locations in the study region.

Calculation of **Median Travel Times** as a **cross-section** of all layers with different travel modes using door-to-door approach



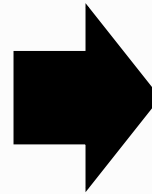
METHODS – MULTIMODAL CENTRALITY

INPUT



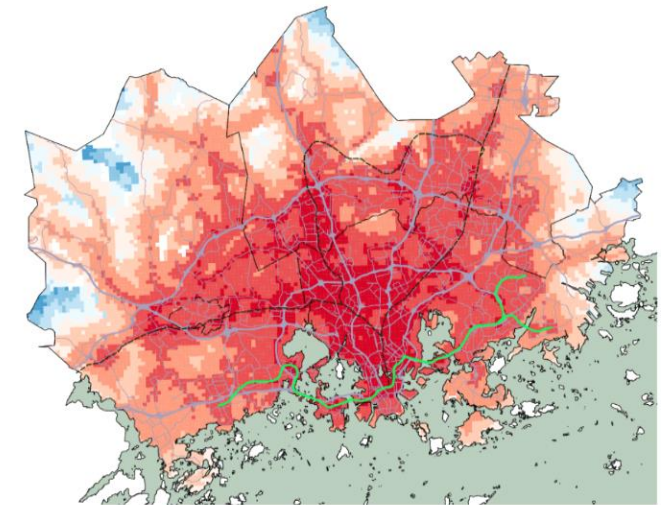
As input we have travel time layers to all locations in the study region

ANALYSIS



Calculation of **Median Travel Times as a cross-section** of all layers with different travel modes using door-to-door approach

RESULT (centrality layer)



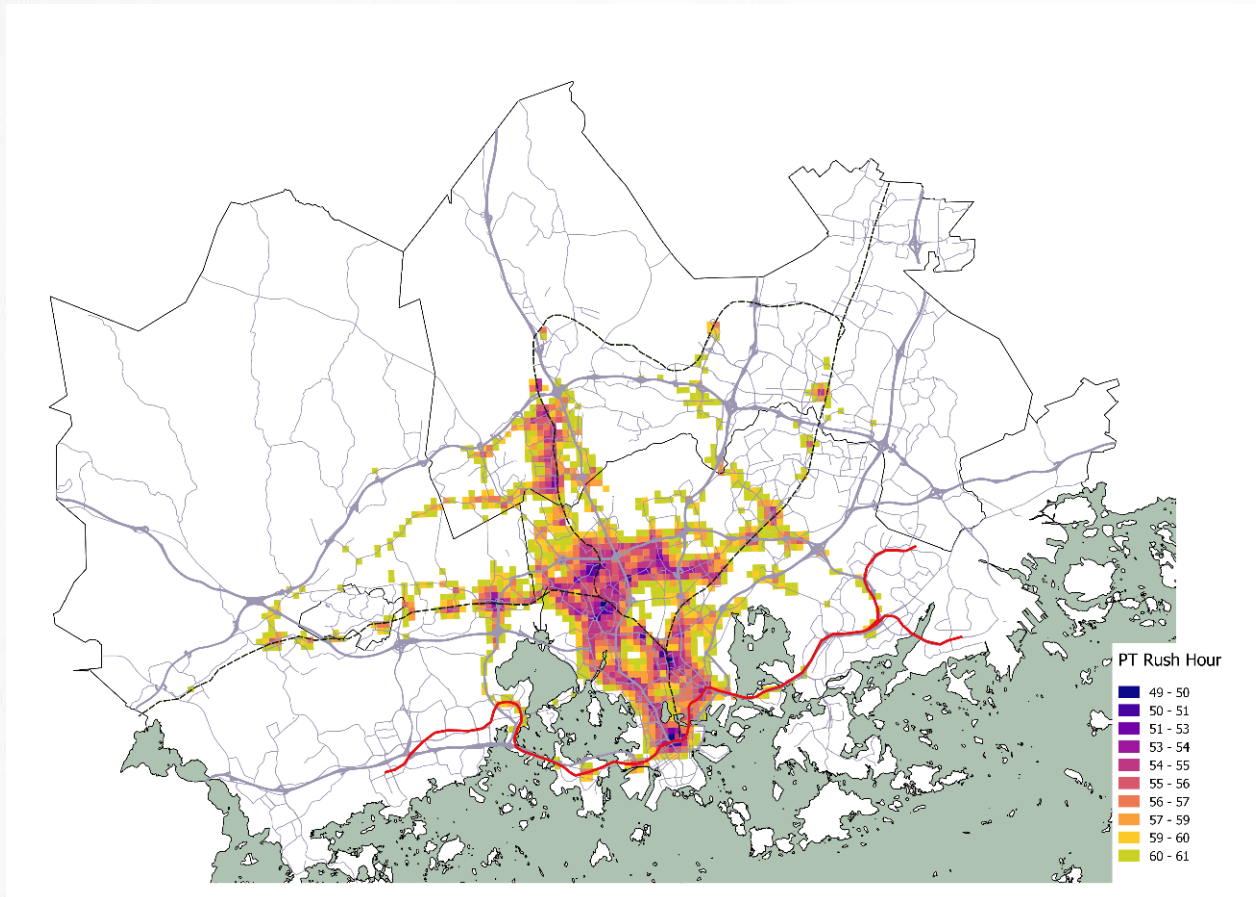
The **more red** the **more central**/lower median time

The **more blue** the **less central**/higher median time



CENTRALITY & MODAL EQUITY

Best 10 % by public transport

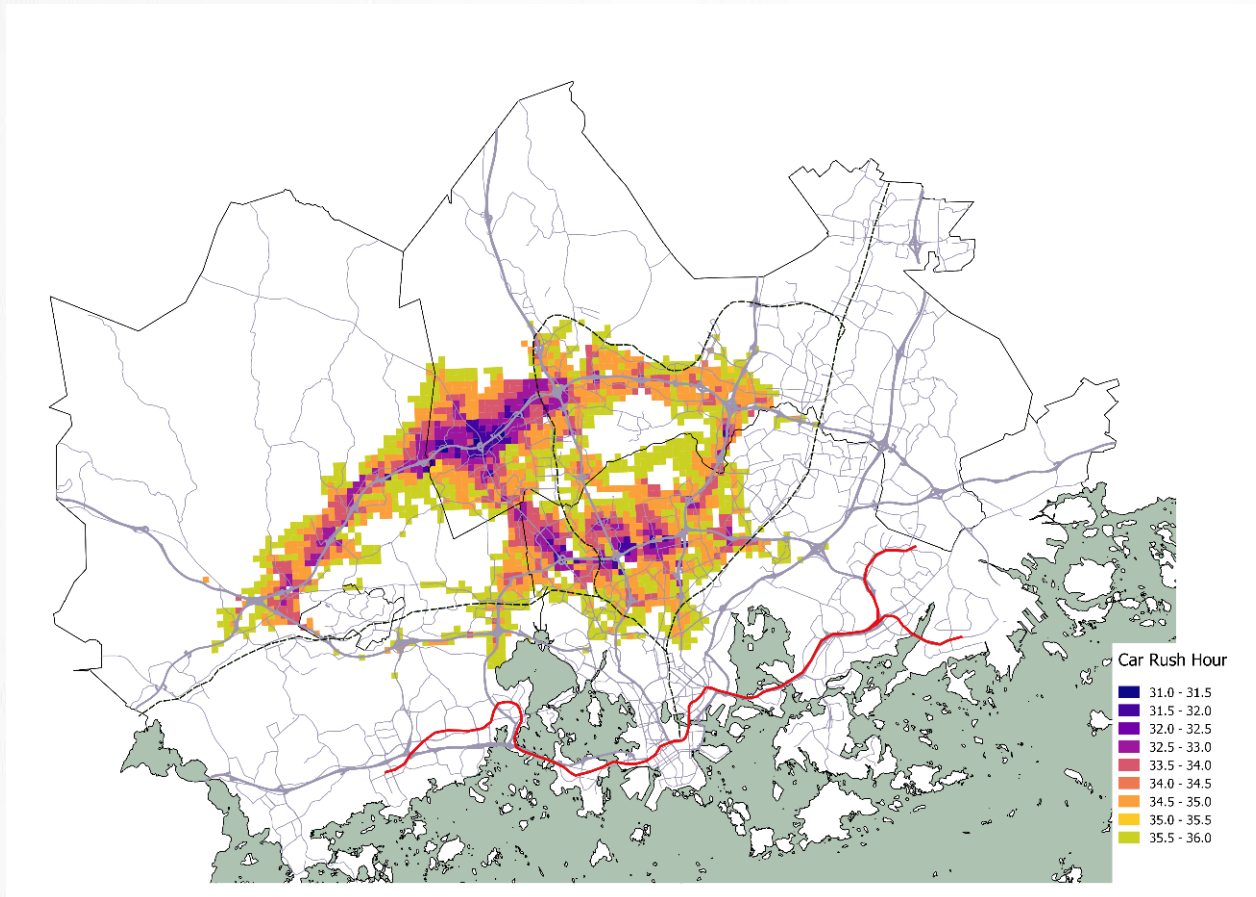


Most central areas in accessibility by different travel modes differ significantly



CENTRALITY & MODAL EQUITY

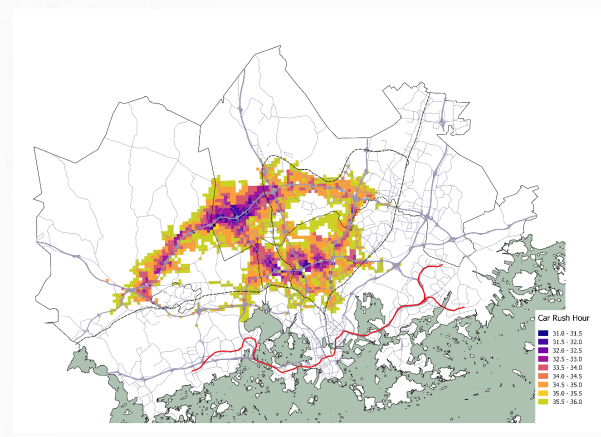
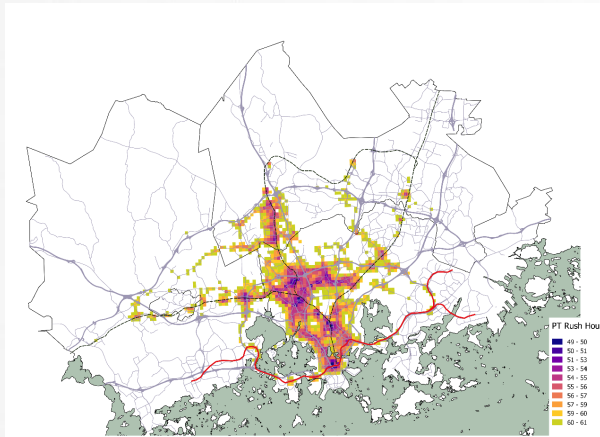
Best 10 % by car



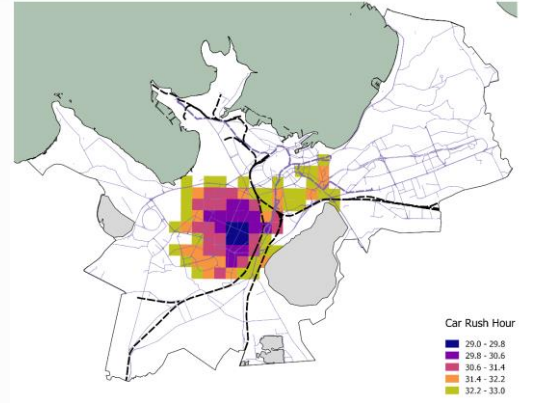
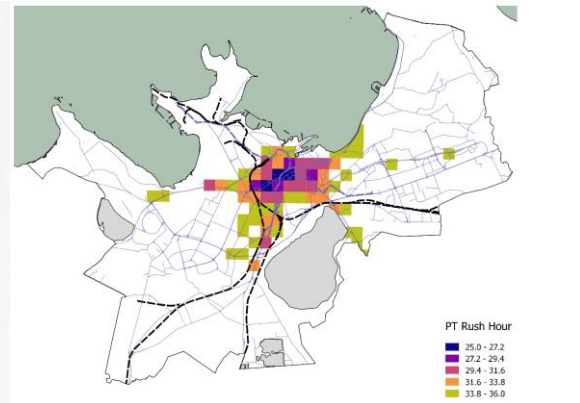
Most central areas in accessibility by different travel modes differ significantly



CENTRALITY & MODAL EQUITY



HELSINKI



TALLINN

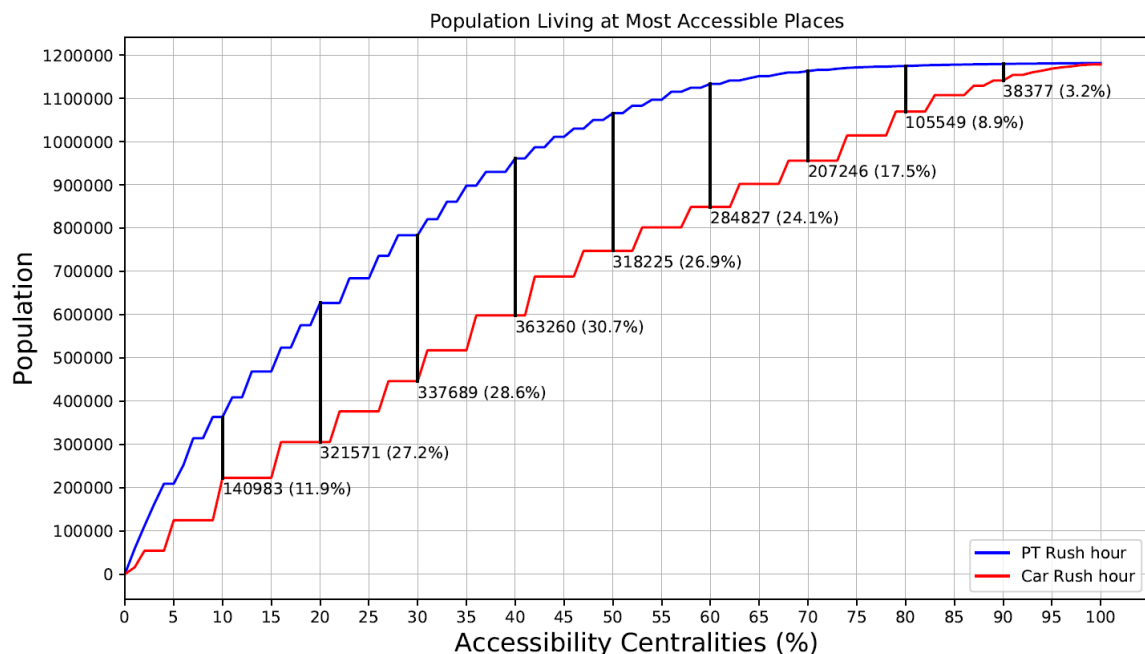
PT

CAR

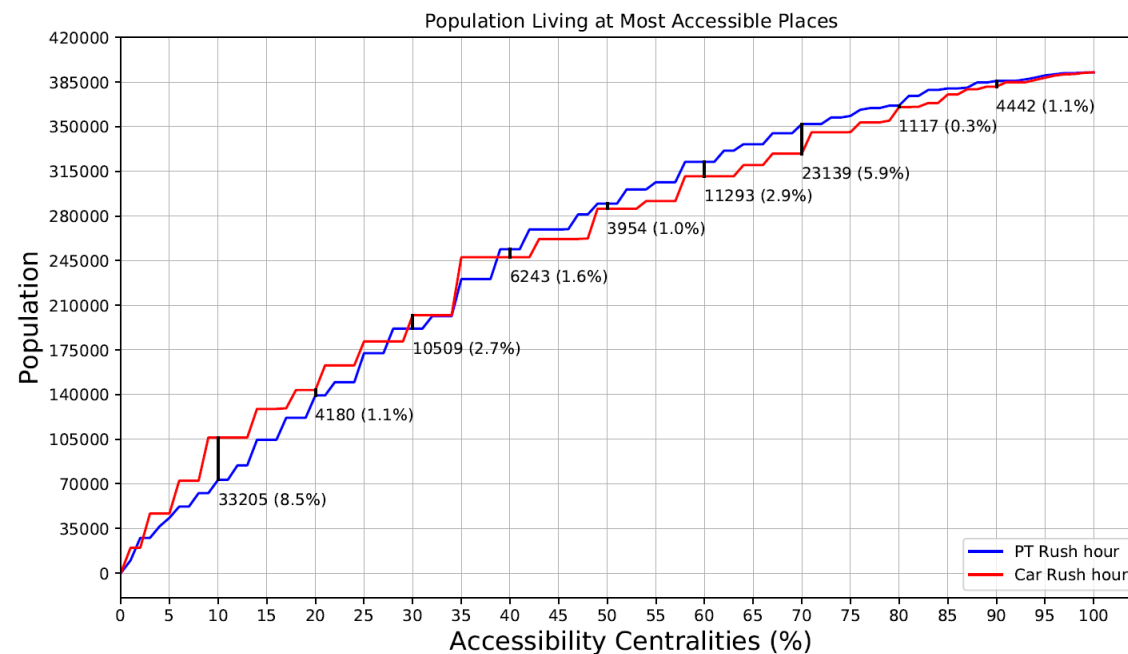
Most central areas in accessibility by different travel modes differ significantly



HOW MUCH OF THE POPULATION IS CONCENTRATED IN CAR CENTRIC VS PUBLIC TRANSPORT CENTRIC AREAS?



HELSINKI



TALLINN



CONCLUSIONS

- ❑ Current accessibility and mobility analyses can provide valuable information about the functioning of cities from various perspectives
- ❑ Various big data sources and open data has made this possible!
- ❑ Time really makes a difference! Both in accessibility and mobility
- ❑ Travel mode matters – All modes should be considered when doing decisions



INTERESTED TO LEARN AND COLLABORATE

Experience in:

- Accessability and mobility research
- Social media research
- Big data analytics

Would need help with:

- Urban geography
- Sociology

Would be interested to learn:

- Machine learning in transportation research?
- Agent based modelling?

CALL ME! 😊



Kiitos! Thank you!

Any questions?

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