

Uncovering Spatiotemporal Dynamics from Non-Trajectory Data

Urban Data Visualization Workshop, IEEE VIS 2018

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Slides: michaeloppermann.com/slides/cityvis2018.pdf



THE UNIVERSITY
OF BRITISH COLUMBIA





Global Heatmap

Heatmap Color

Hot

Blue

Gray

Red

Activity Type

All

Heat Opacity

0%

40%

60%

80%

100%

Layers

Map

Labels

Satellite

[Discover](#) how the heatmap was built.
[Learn](#) how Strava Metro can help your community.

[Learn about heatmap updates.](#)

Fit Leaking: "fitness activities, recorded for personal benefit emit into signals that reveal sensitive and confidential information"

- John Scott-Railton

Bike Sharing

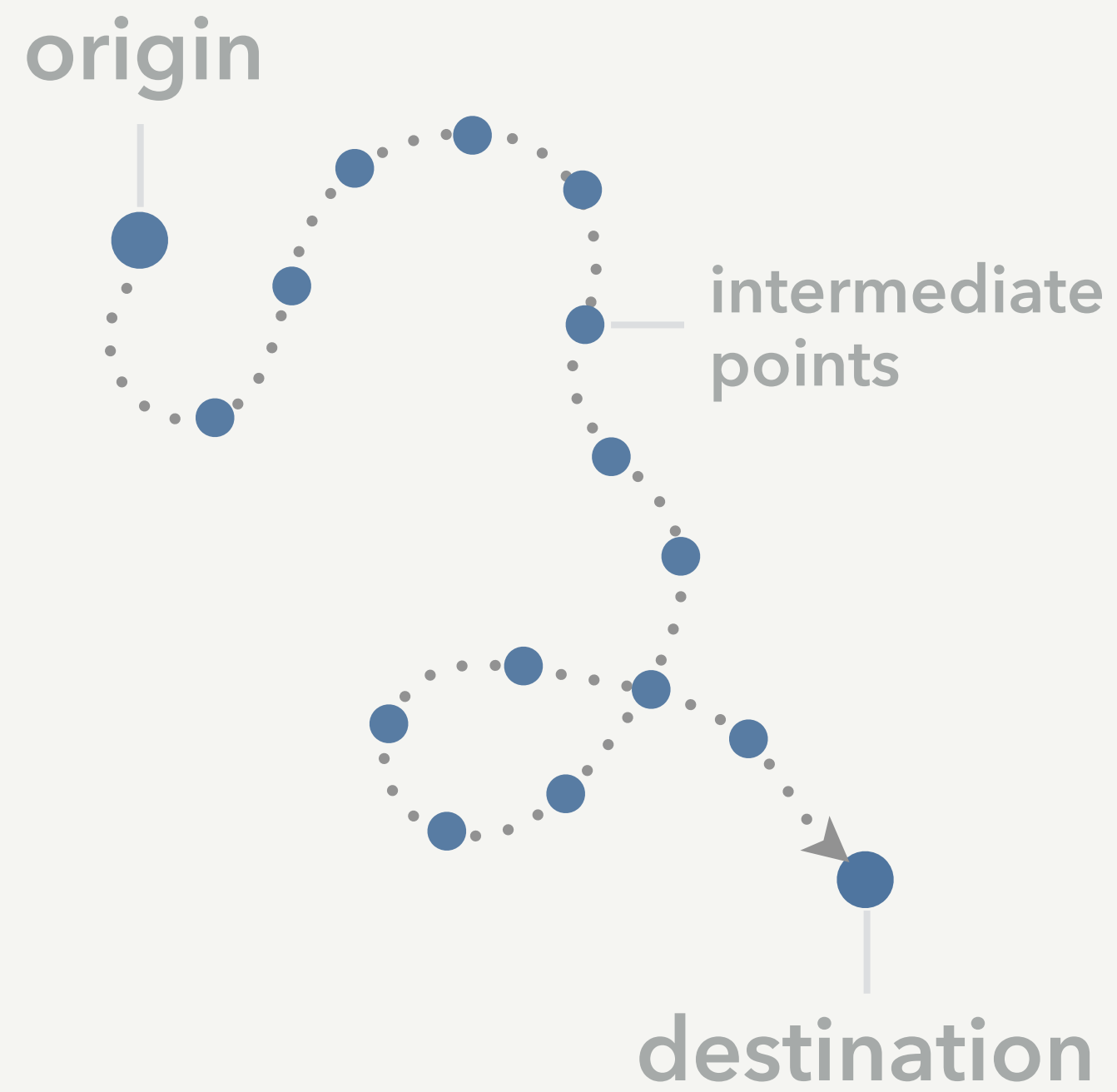


Building Occupancy

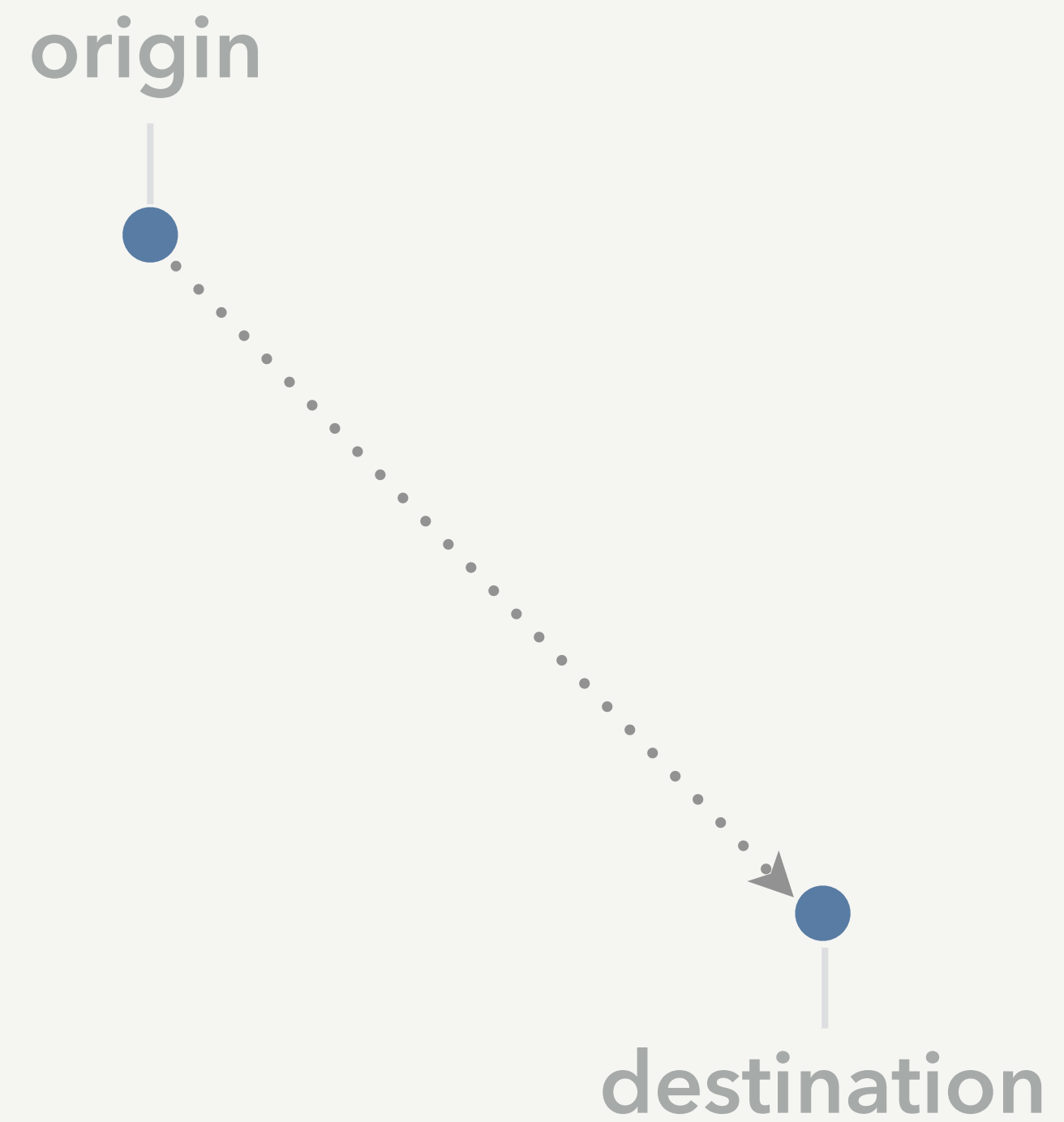


Spatiotemporal data

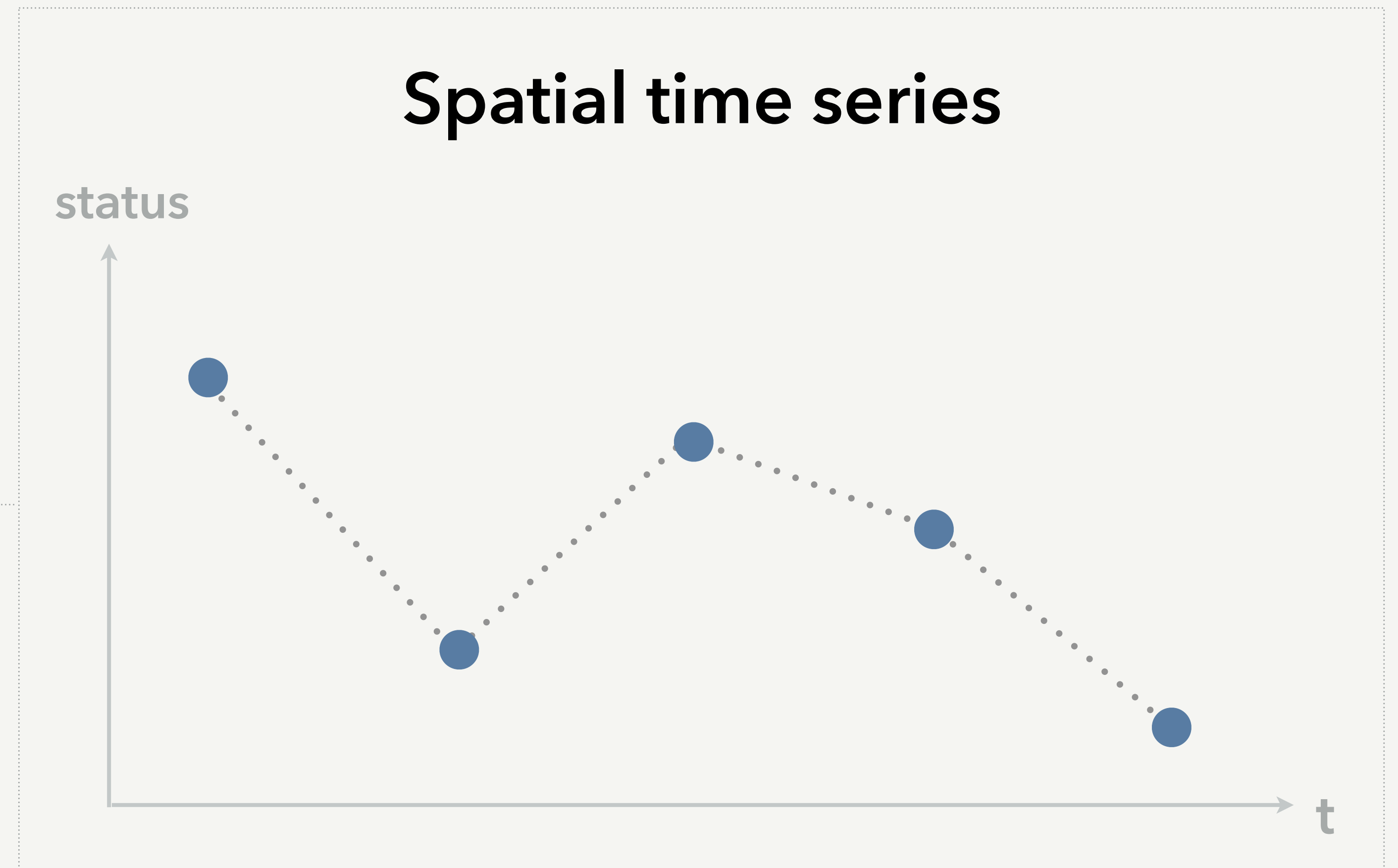
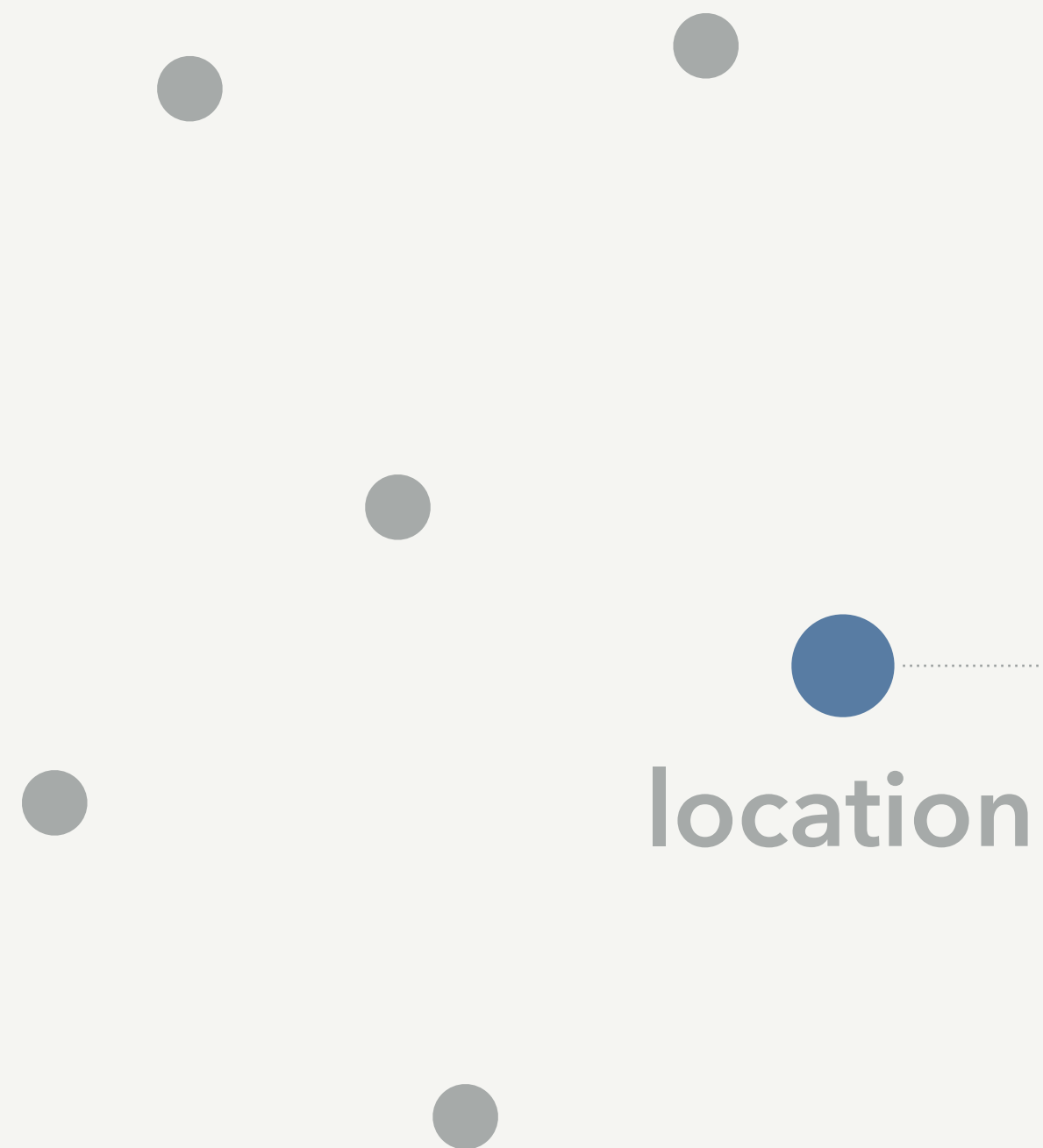
Trajectories



Origin–destination data



Spatial events / location-based counts



Bike Sharing Atlas

- Analyzed bike sharing in Vienna
(bike sharing operator as collaborator)
- Traditional station-based networks
- Expanded from one to many cities
- Recorded station fill levels from >400 networks (21,500 stations)
- 15 min interval over 1.5 years
- Additional user groups: *politicians, city planners, general public, etc.*



M. Sedlmair



T. Möller

[Bike Sharing Atlas: Visual Analysis of Bike-Sharing Networks.

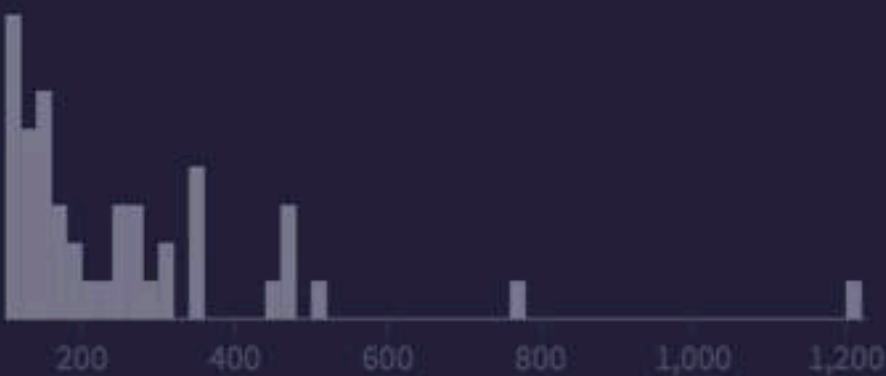
Oppermann, Möller, Sedlmair. International Journal of Transportation (IJT), 6(1): 1-14, 2018]

Networks

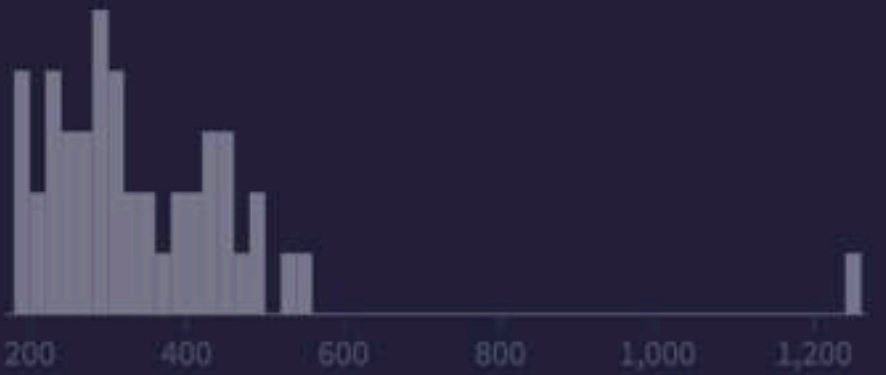
Explore bike-sharing networks worldwide.

46/467 NETWORKS

Number of stations



Ø Distance to nearest station [m]



Population (urban area)



Number of stations 100 1226 Ø Distance to nearest station 0m 7268m

Sort by Number of stations

Paris, FR

Velib | 1226 Stations

London, GB

Santander Cycles | 774 Stations

New York, NY, US

Citi Bike | 511 Stations

Chicago, IL, US

Divvy | 475 Stations

Barcelona, ES

Bicing | 465 Stations

Montreal, QC, CA

Bixi | 461 Stations

Mexico City, MX

EcoBici | 444 Stations

Washington, DC, US

Capital BikeShare | 357 Stations

Lyon, FR

Vélo'V | 346 Stations

Bruxelles, BE

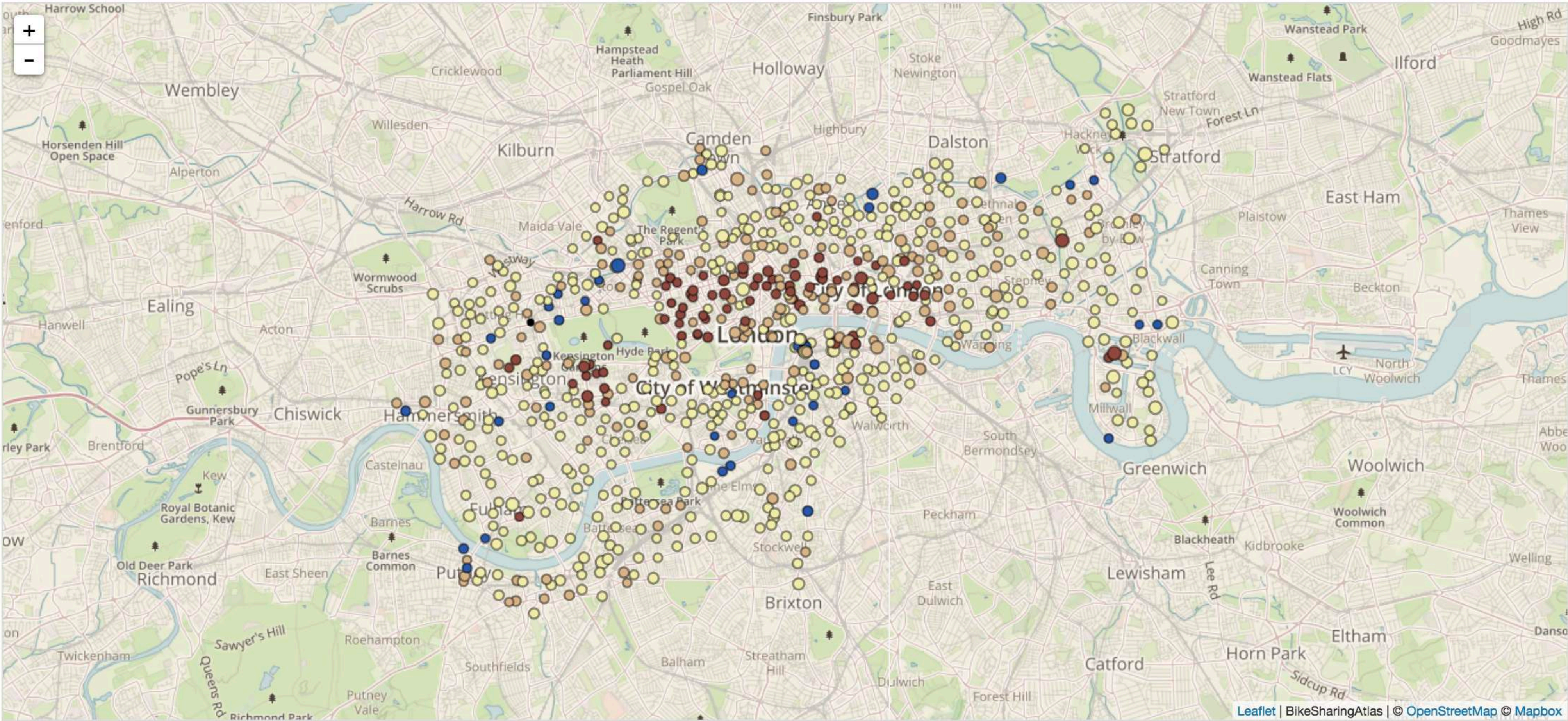
villo | 343 Stations

London

Capital of United Kingdom | 8.67mio inhabitants

- Stations
- Route Planner
- Price
- Availability
- Network Activity
- Trips

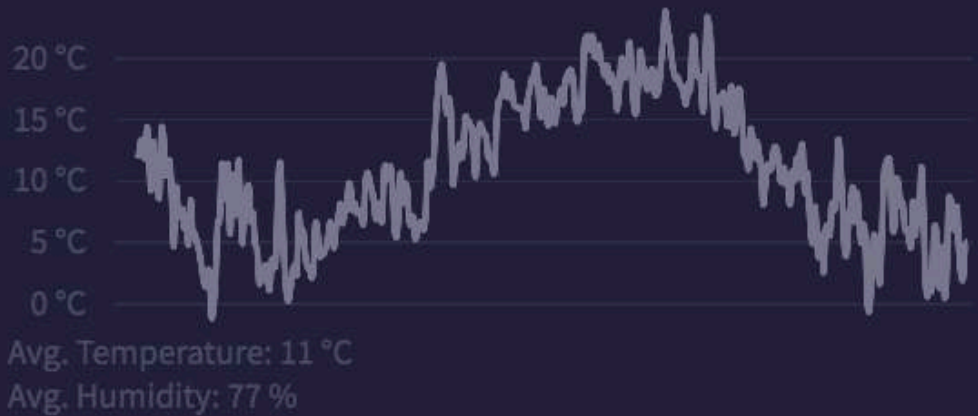
Everything ok Almost full or empty (15%) Full Empty



Elevation Profile



Weather



Similar Networks

- Paris, FR
- New York, NY, US
- Chicago, IL, US



London

Capital of United Kingdom | 8.67mio inhabitants

Stations Route Planner Price Availability Network Activity Trips

SANTANDER CYCLES (774 STATIONS)

Elevation Profile



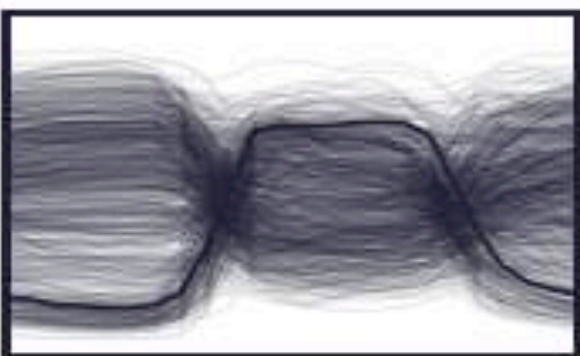
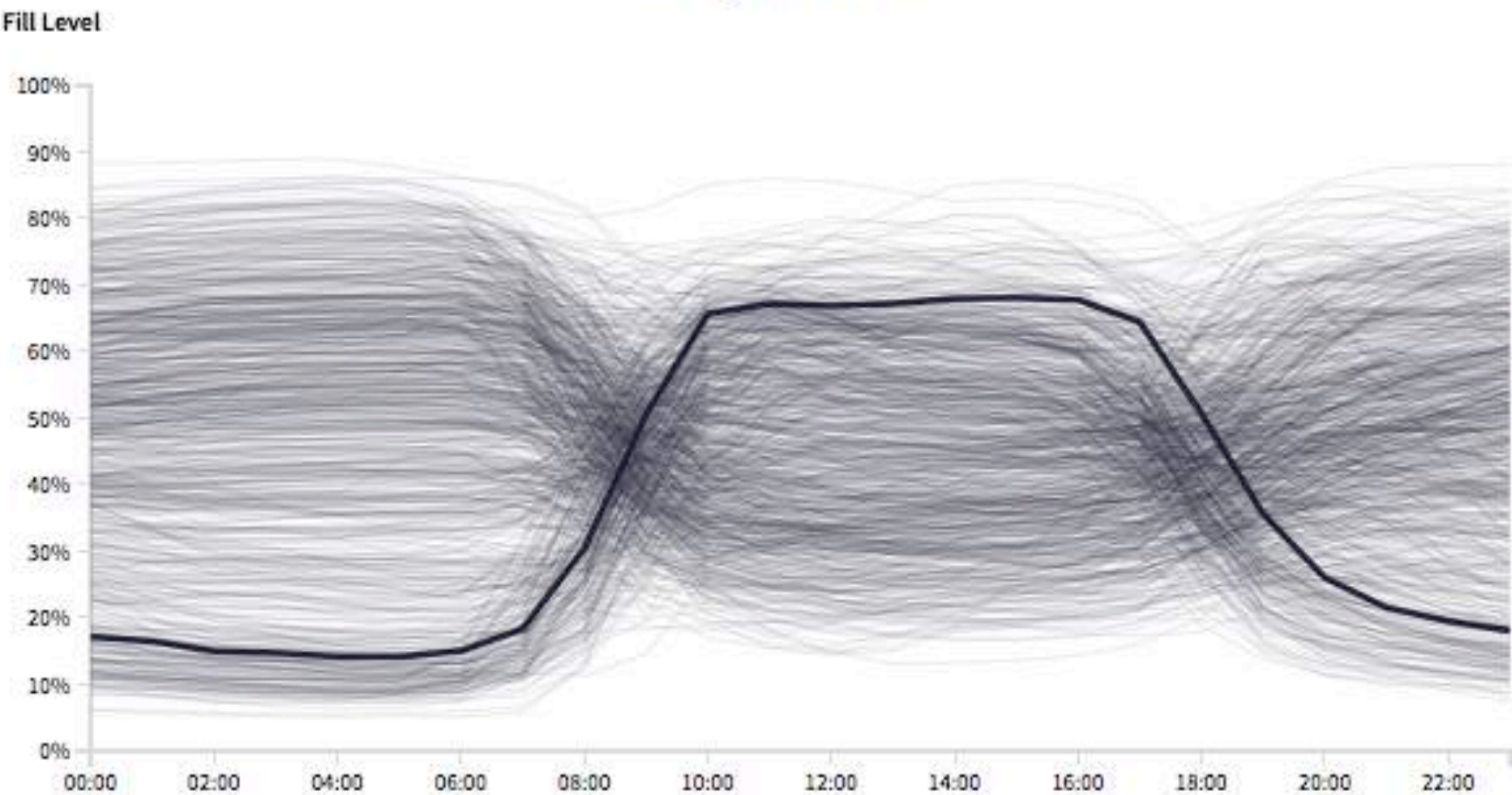
Weather



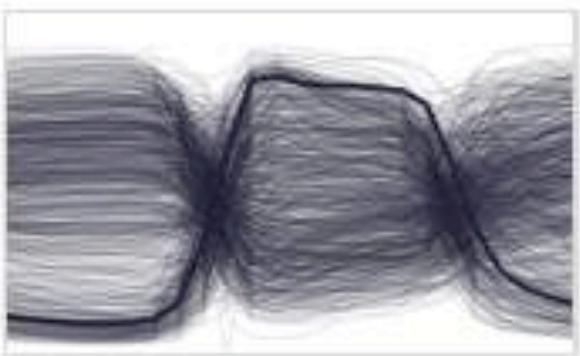
Similar Networks

- Paris, FR
- New York, NY, US
- Chicago, IL, US

Average utilization



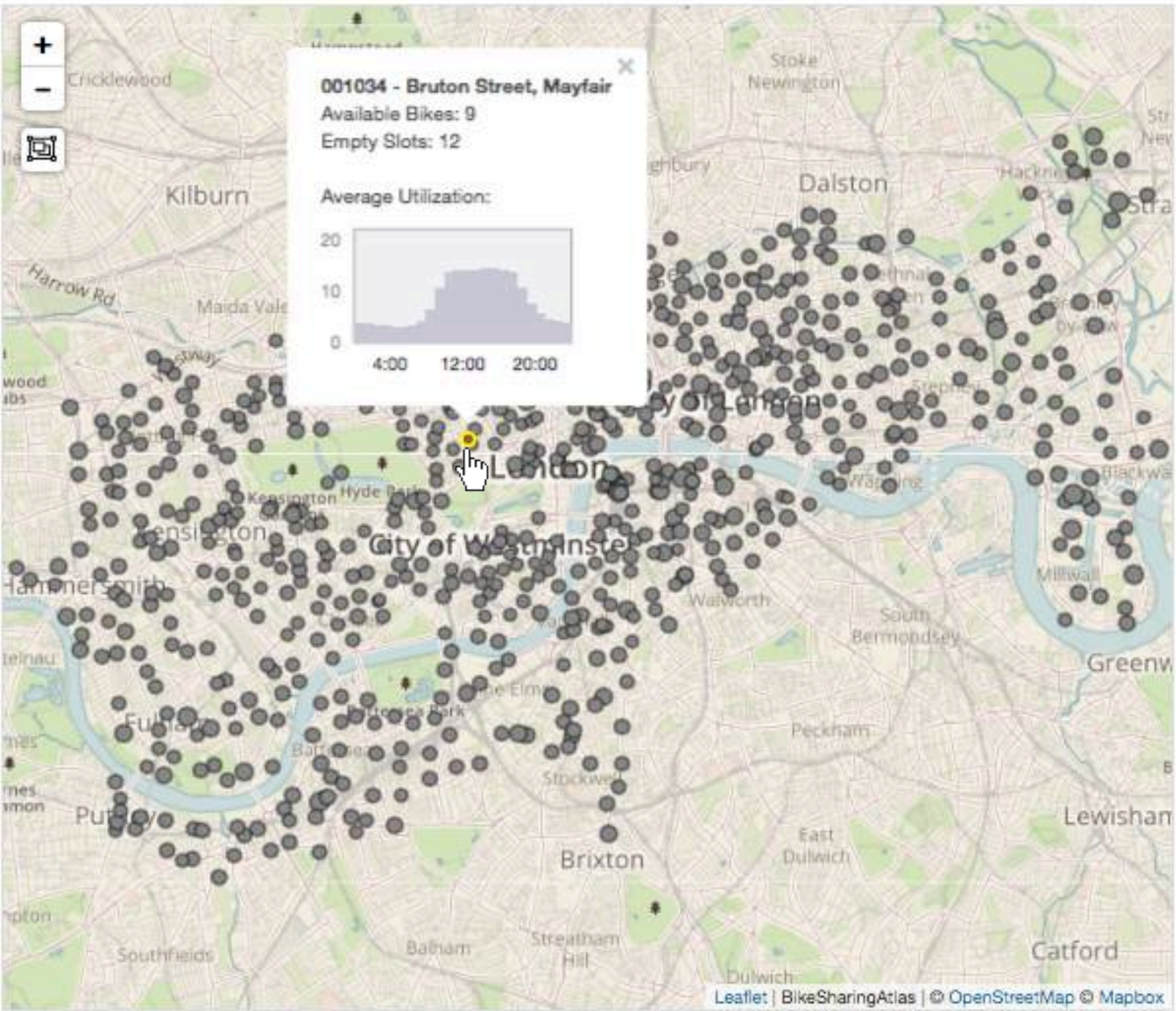
ALL



WEEK



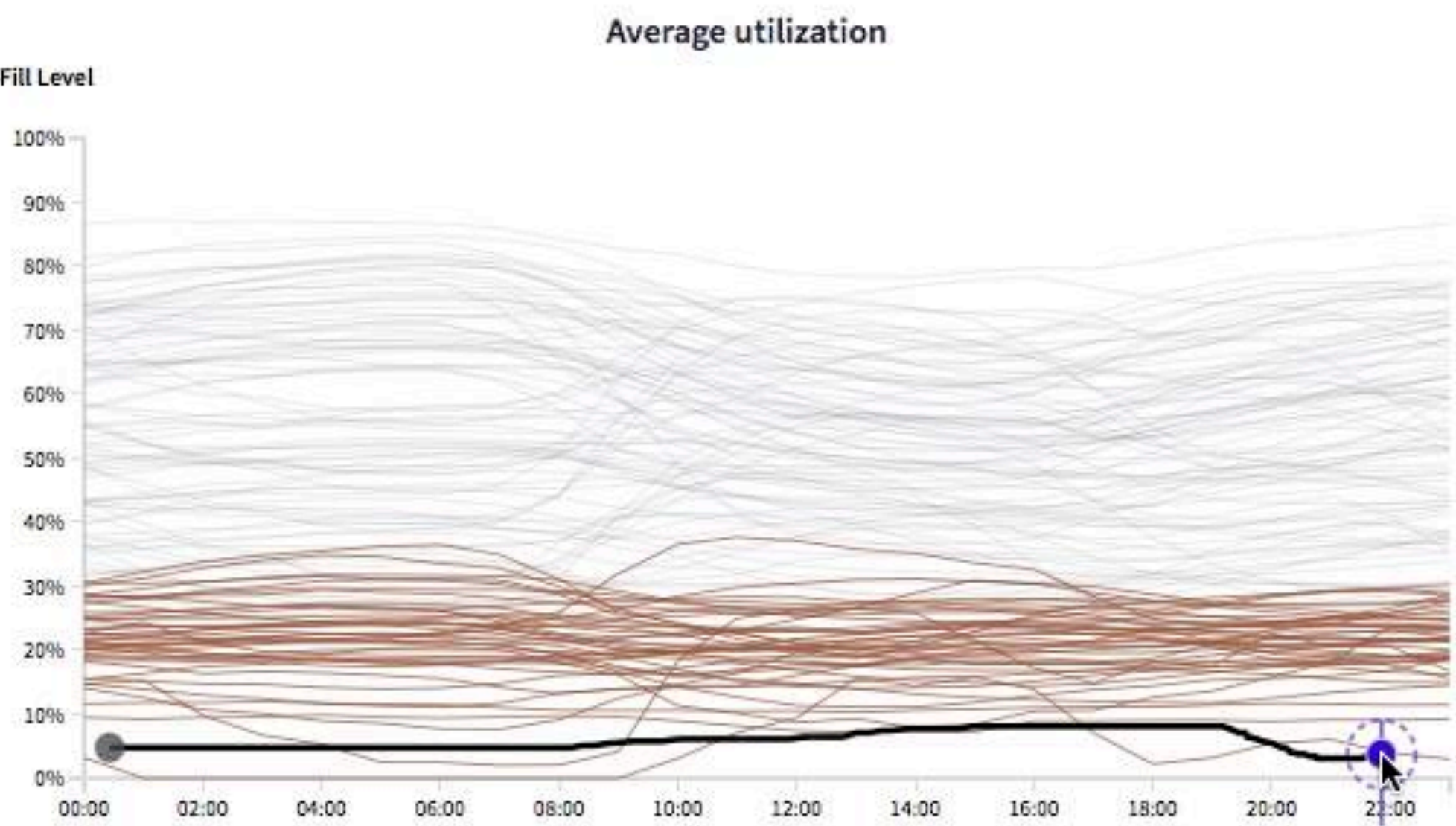
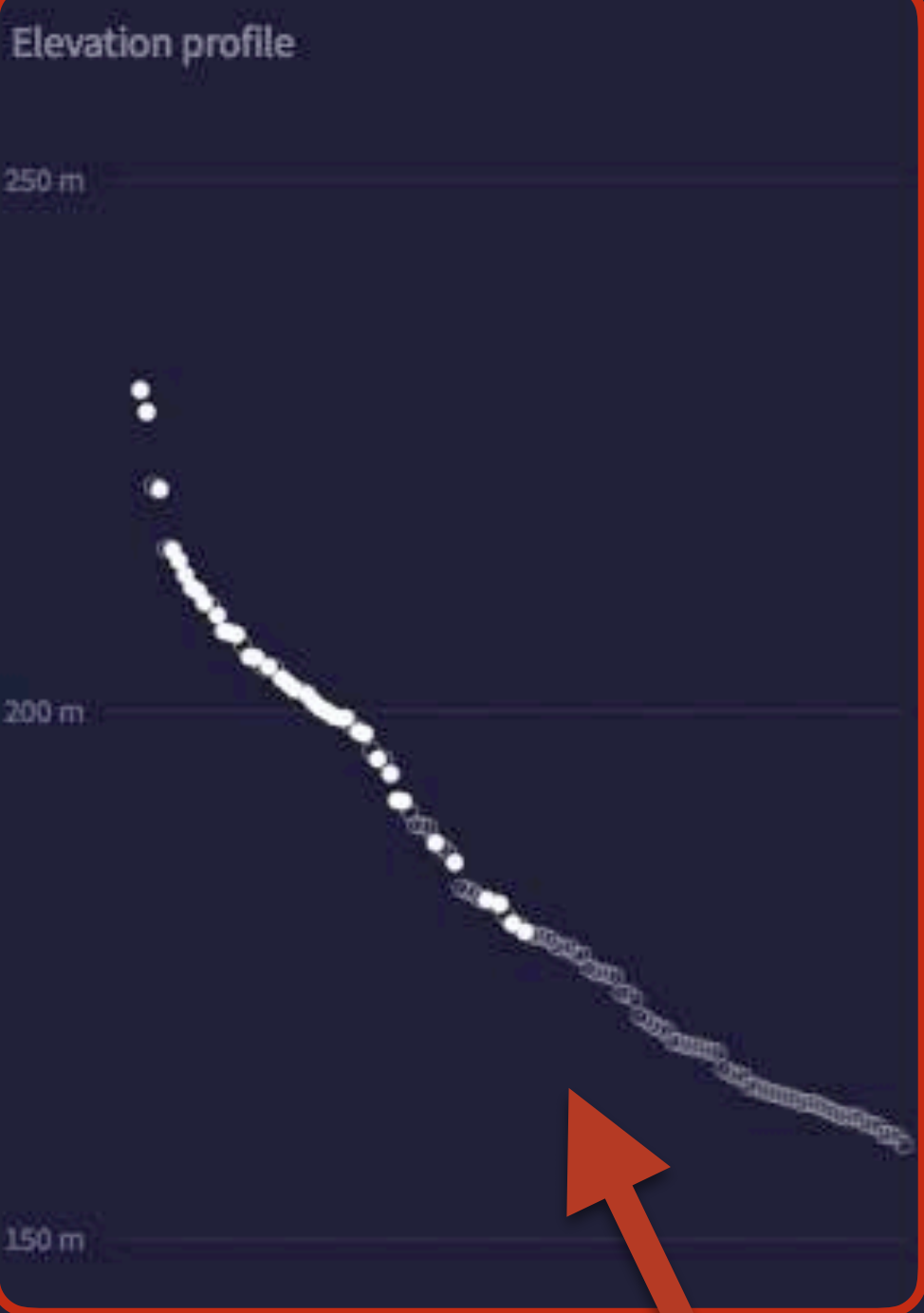
WEEKEND



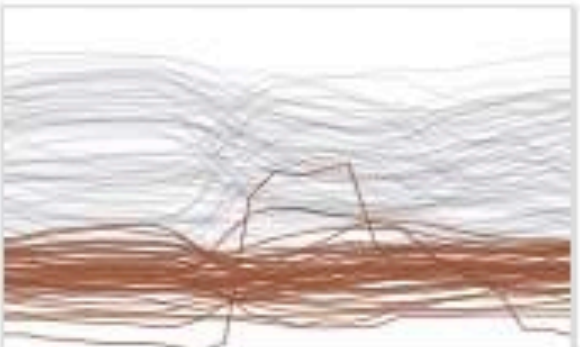


Vienna
Capital of Austria | 1.74mio inhabitants

Stations Route Planner Price Availability Network Activity Trips



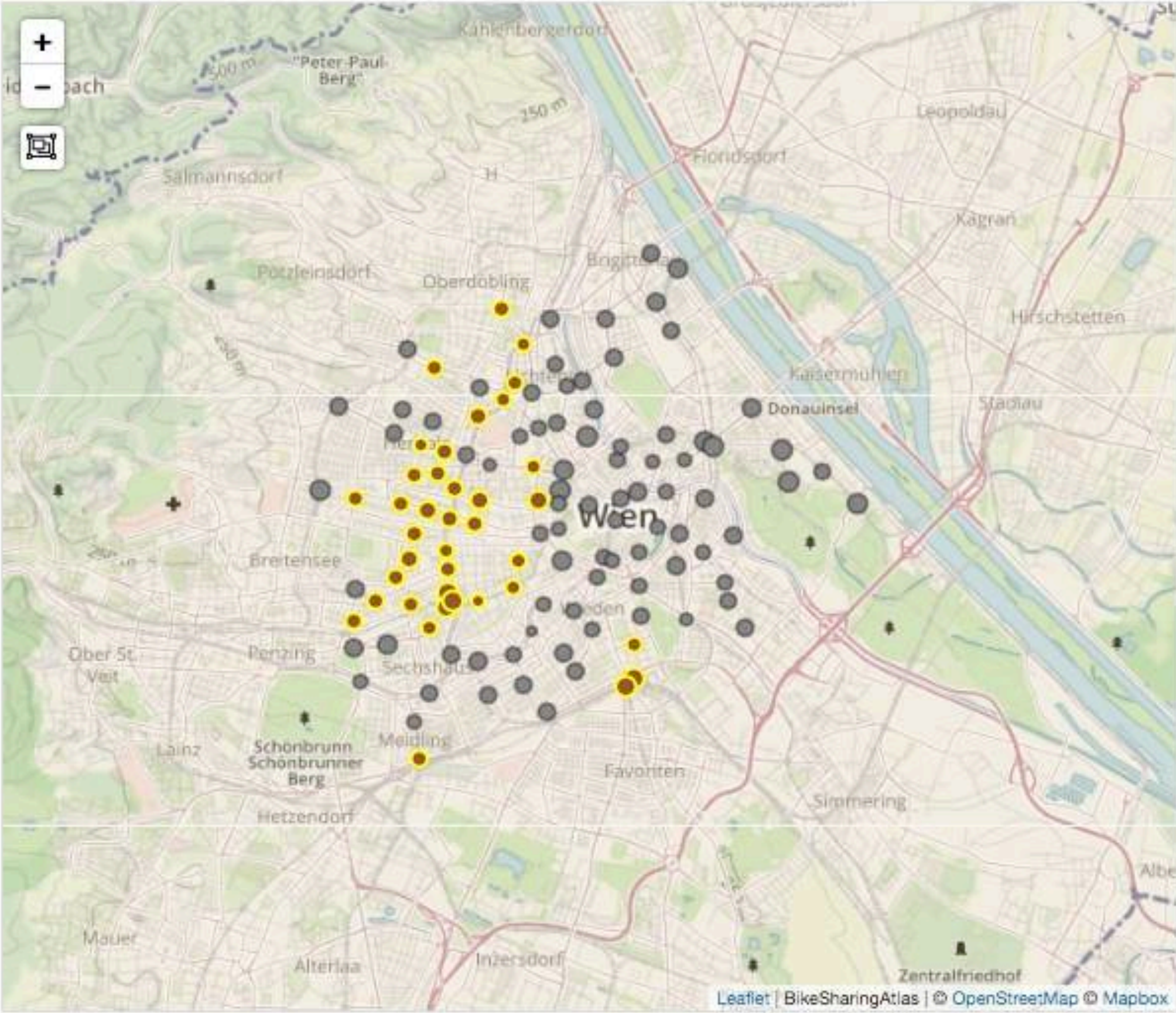
ALL



WEEK



WEEKEND

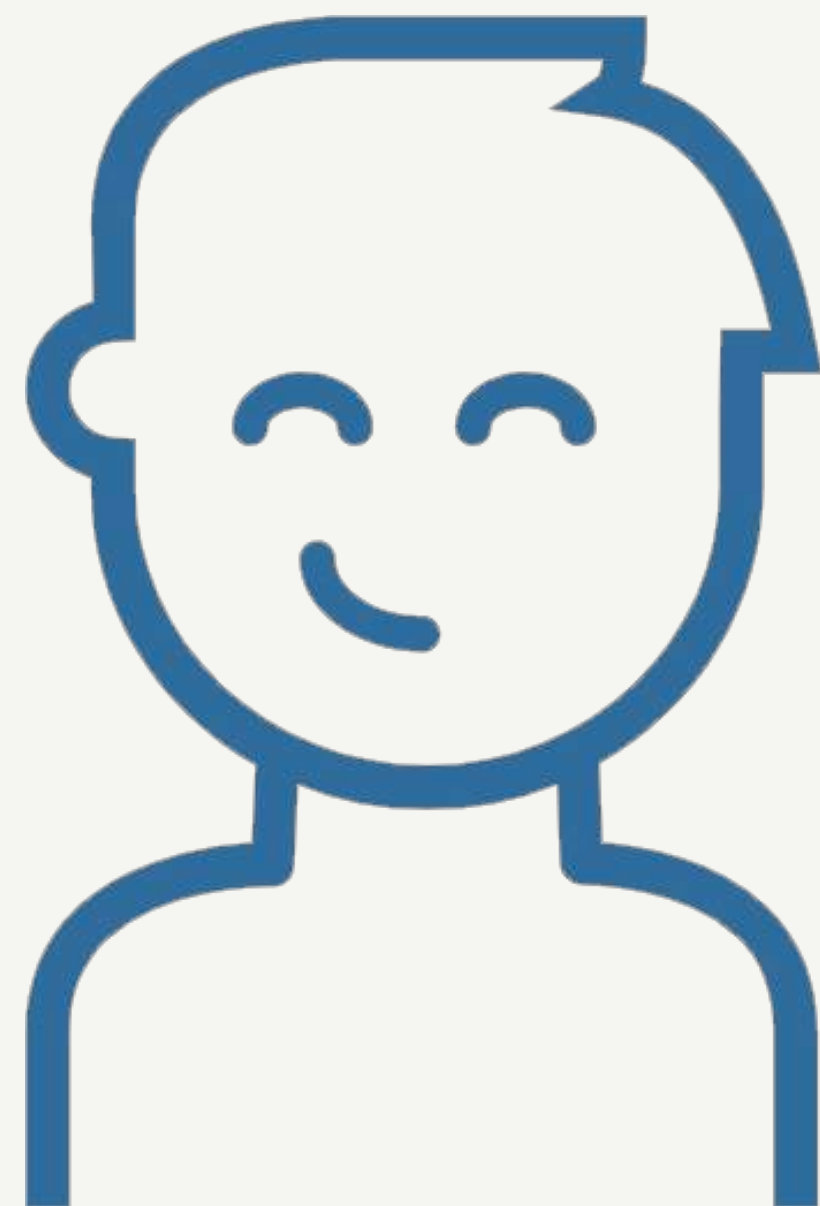


Ocupado: Visual analytics for occupancy applications

- Visual and predictive decision-support tools for facilities management
- Stakeholders:
 - ▶ Custodial services
 - ▶ Building manager
 - ▶ Facilities planner
 - ▶ Energy manager



T. Munzner



**Which regions are
busy/quiet now?**



**Which regions were
heavily used and are
empty now?**



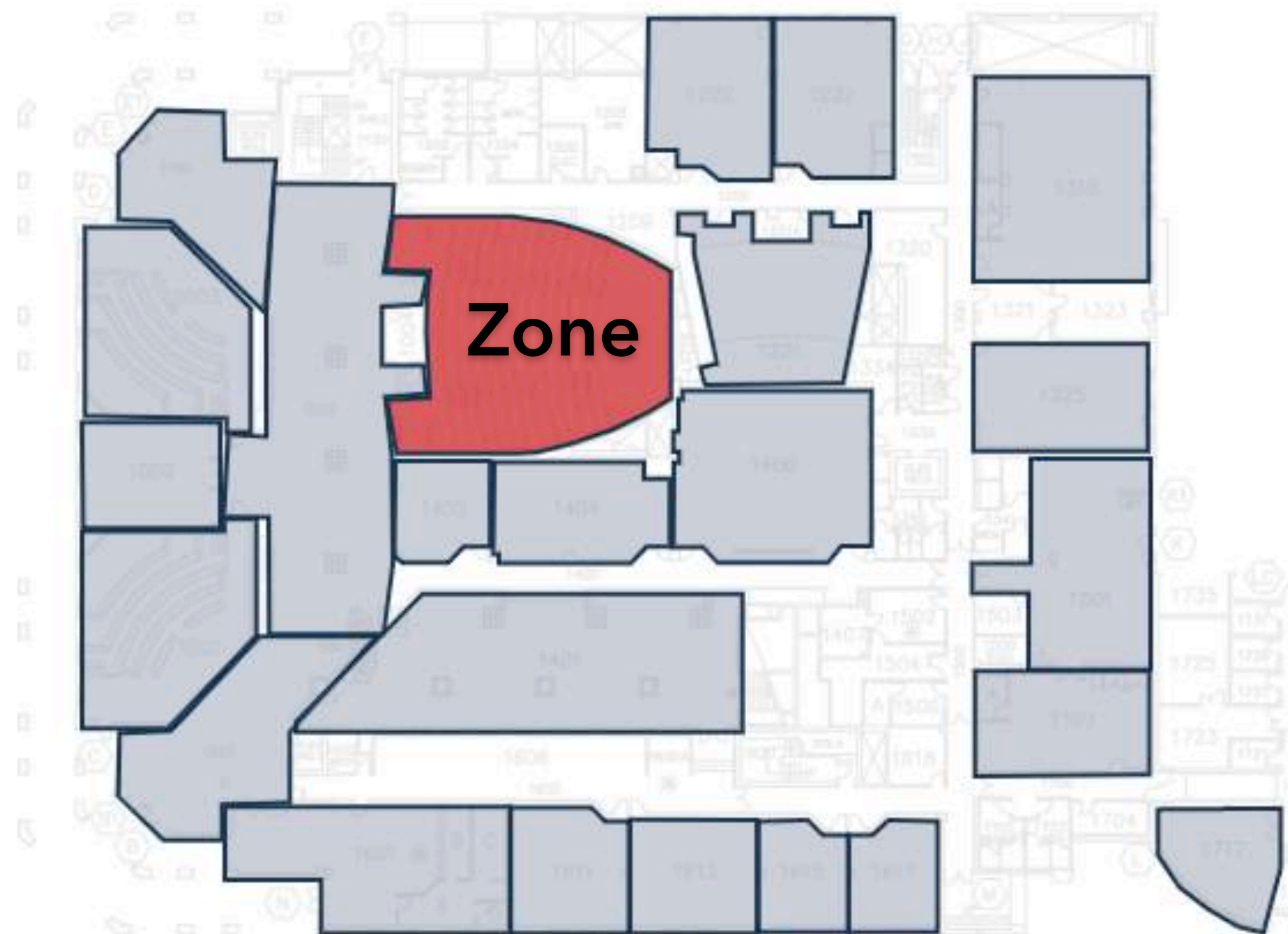
**What is the typical
usage pattern of
this region?**

during
holidays?

weekdays vs.
weekends?

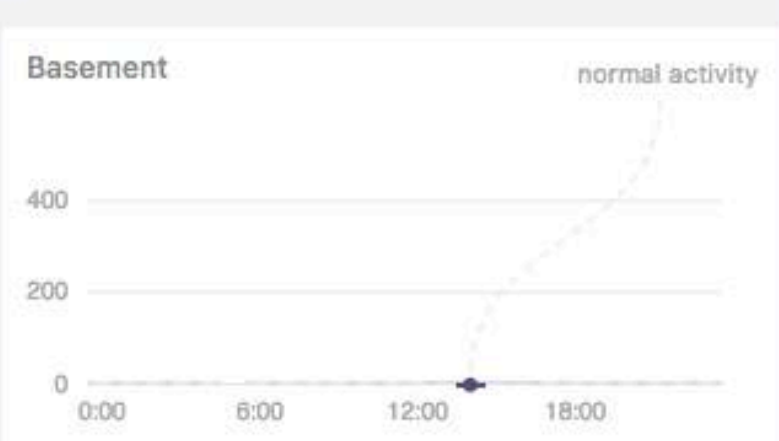
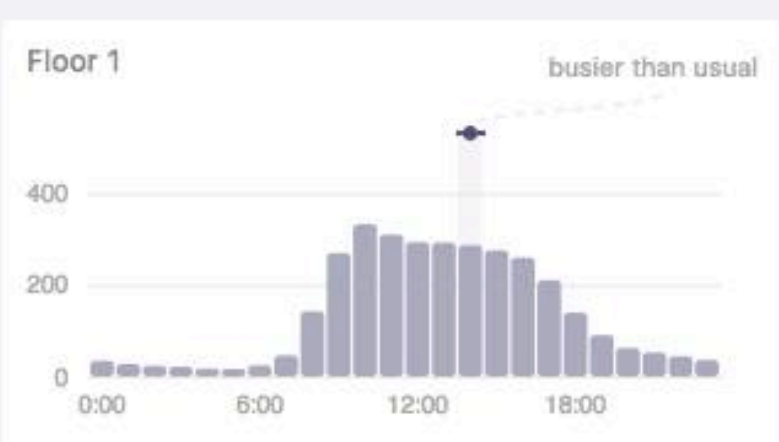
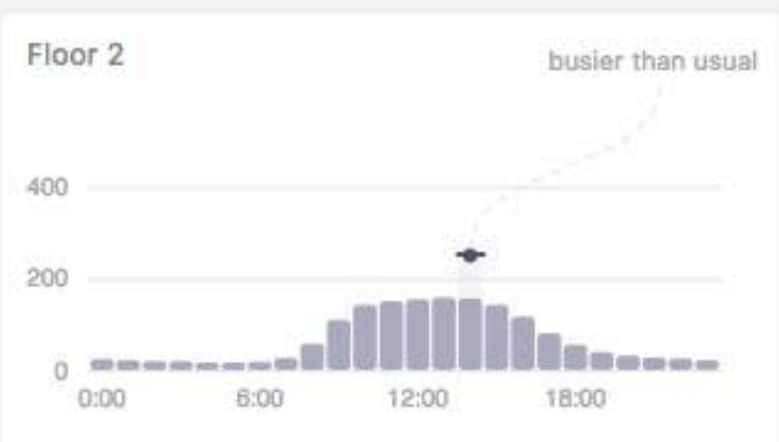
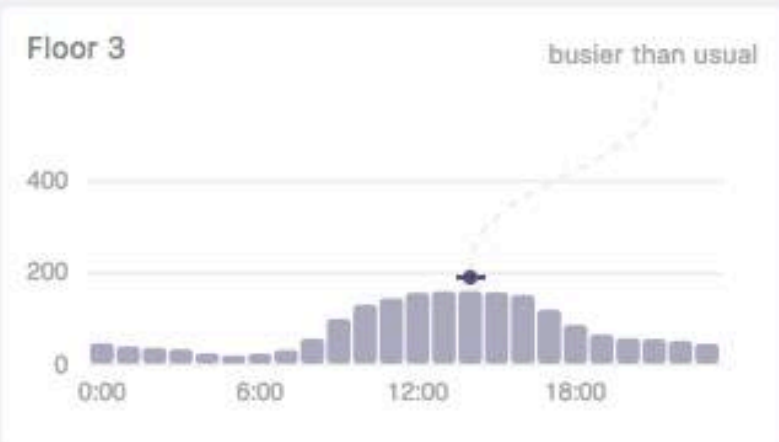
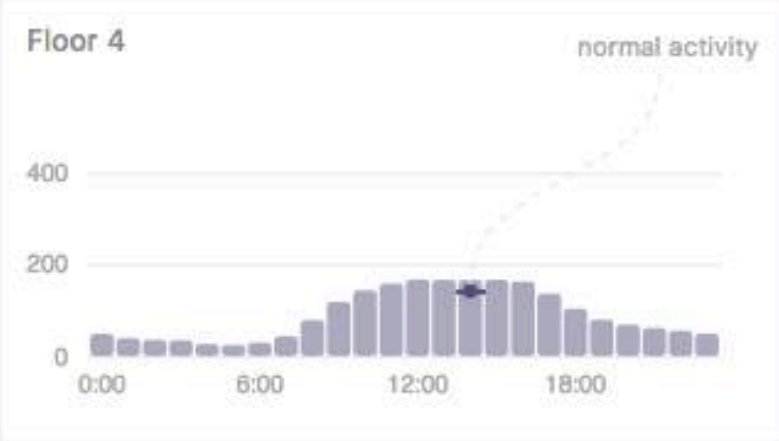
according to
(custodial) shifts?

Data acquisition & processing

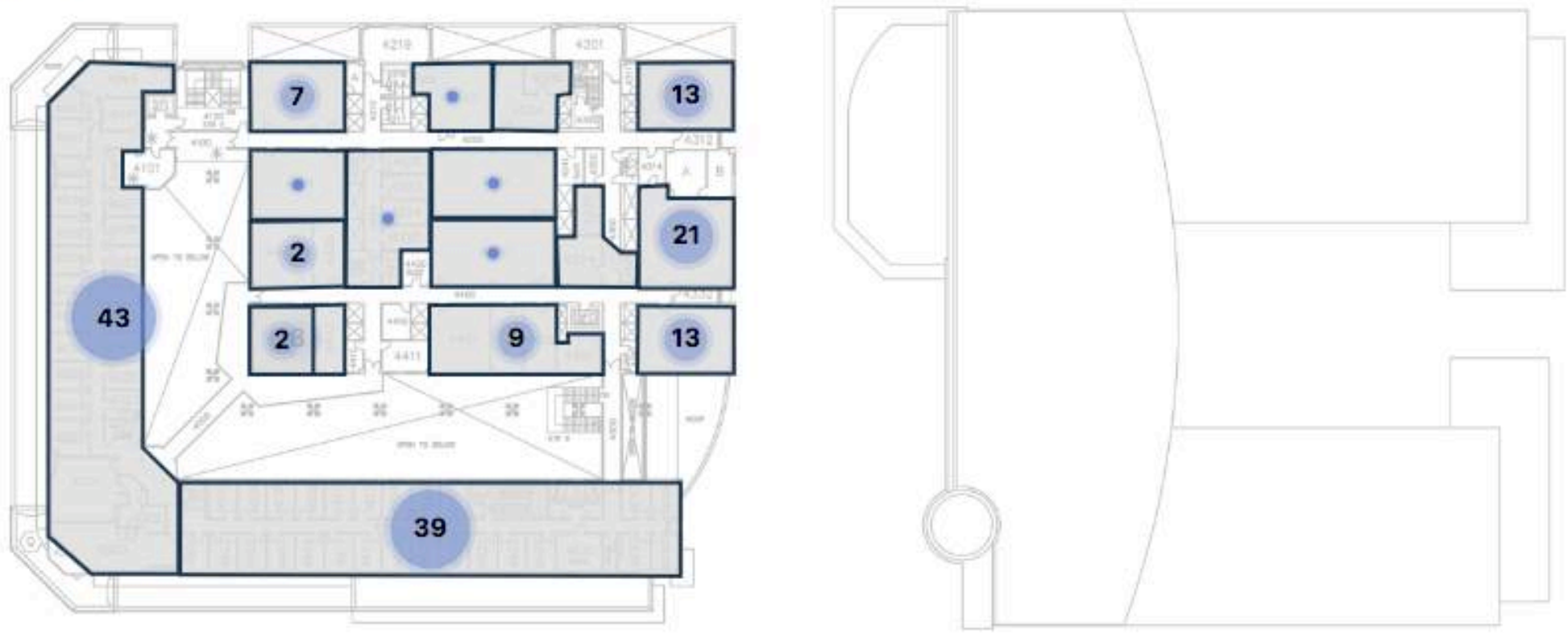


- Estimate building occupancy by using Wi-Fi connections (*in collaboration with start-up: Sensible Building Science*)
- No movements or individual devices are recorded
- Only zone-based counts every 5 min
→ *spatial time series*
- Rough proxy depends on Wi-Fi coverage; good spatial precision if zone large enough

trend typical day



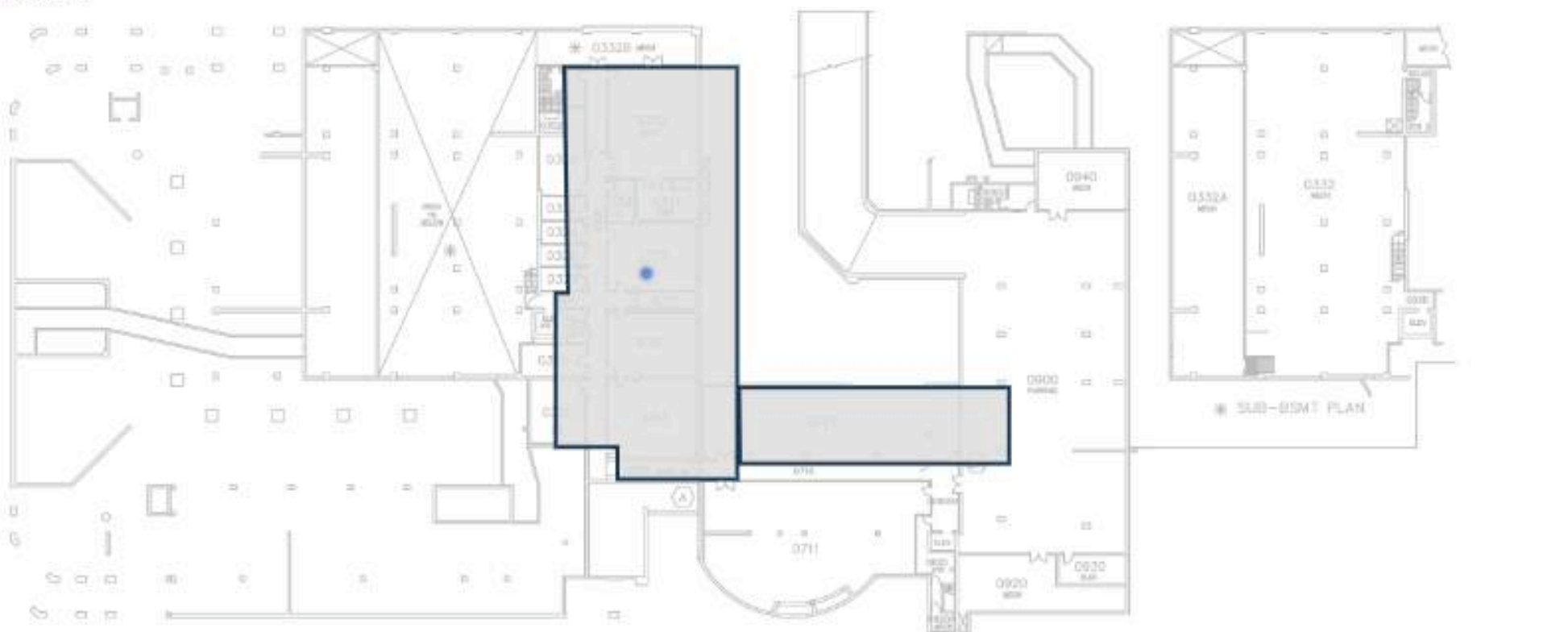
Floor 4



Floor 2



Floor -1



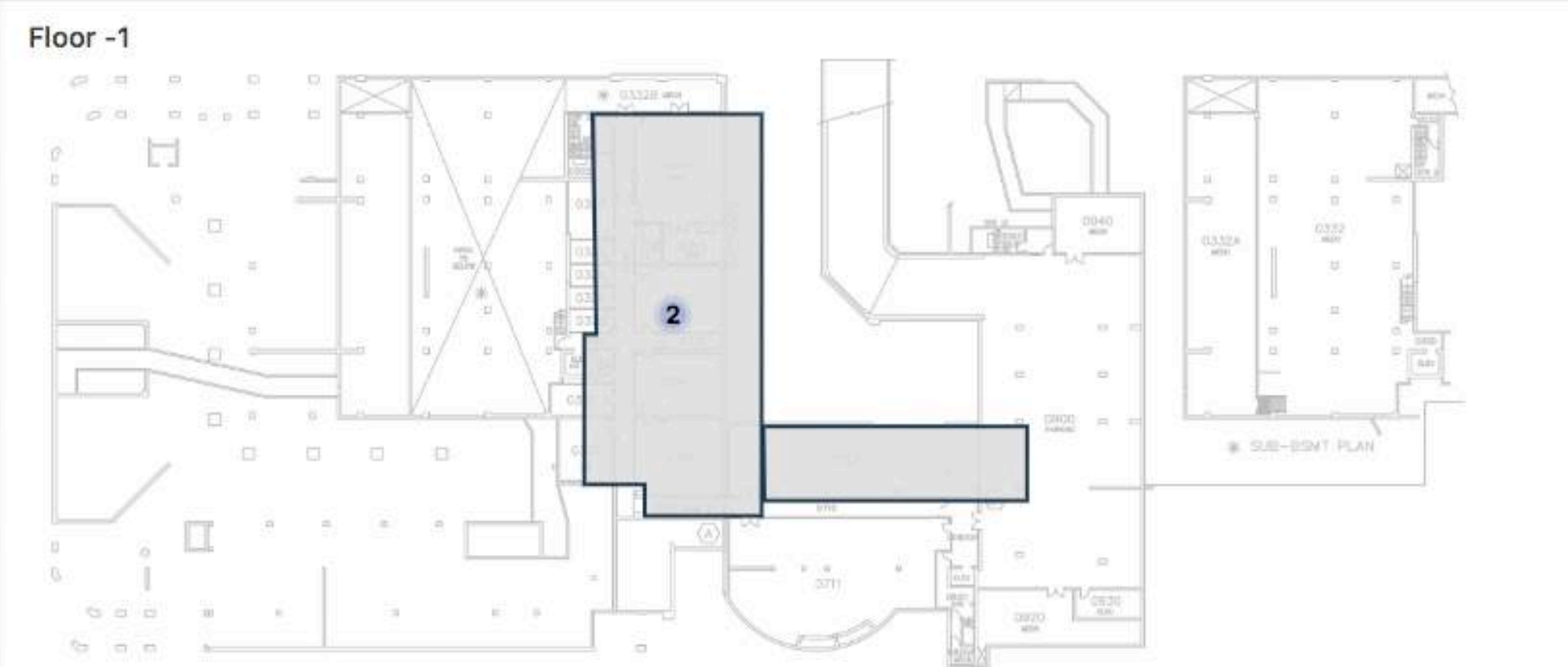
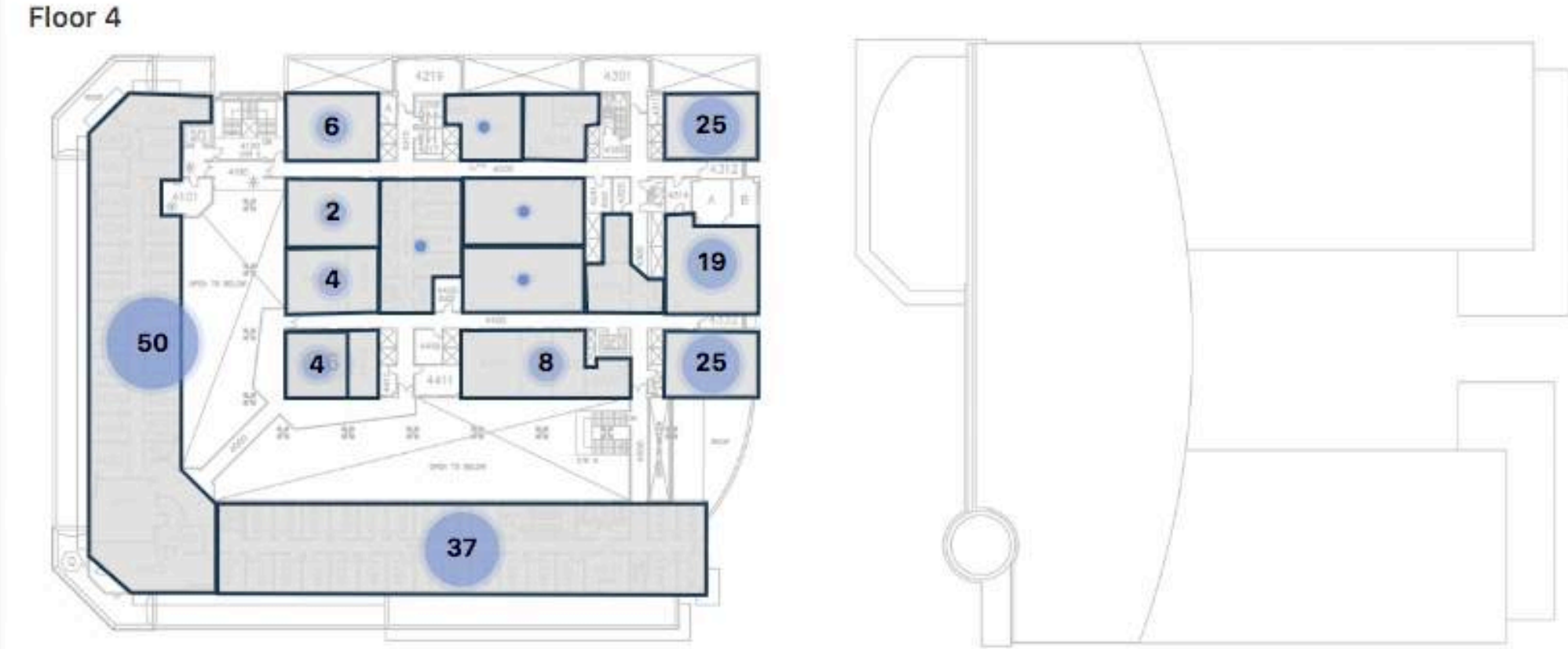
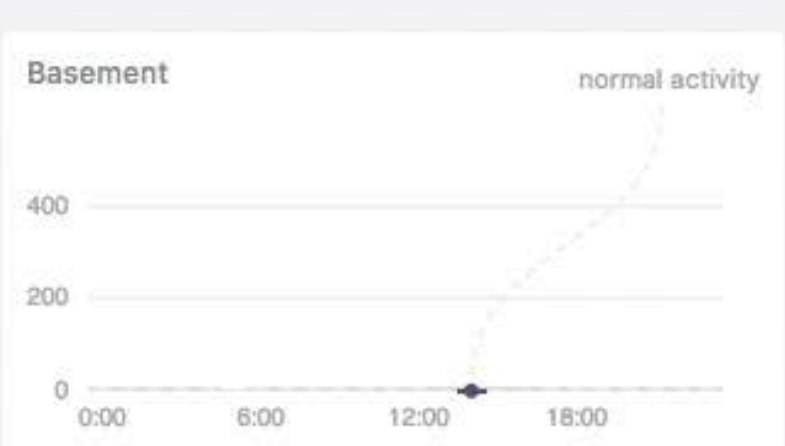
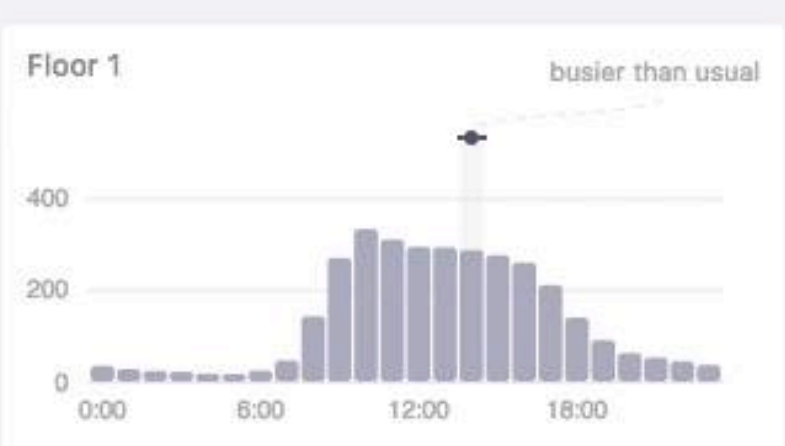
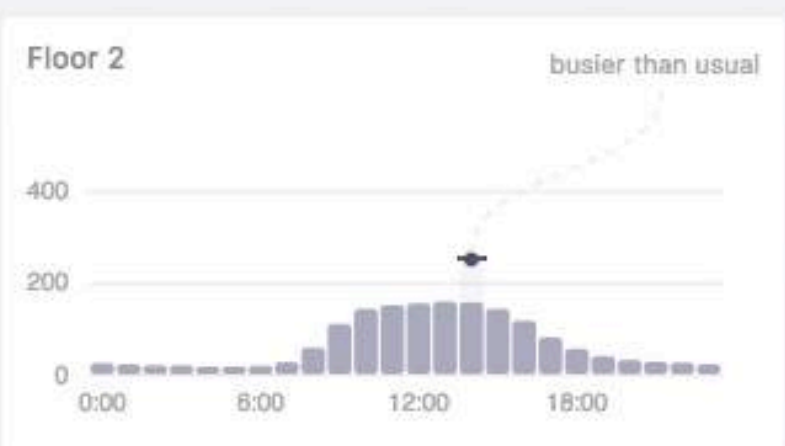
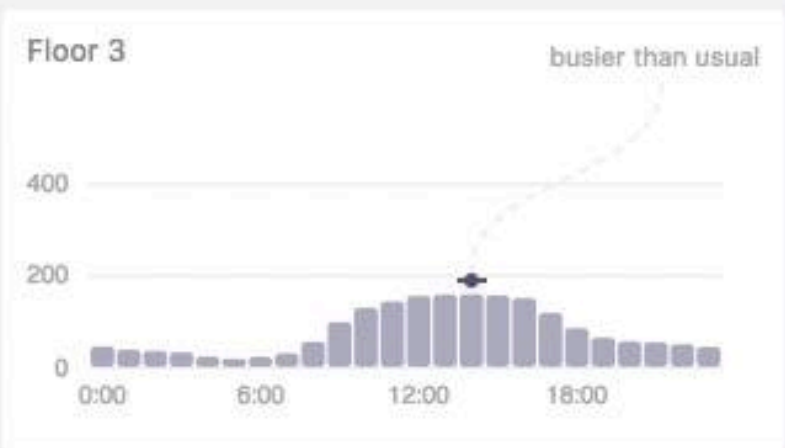
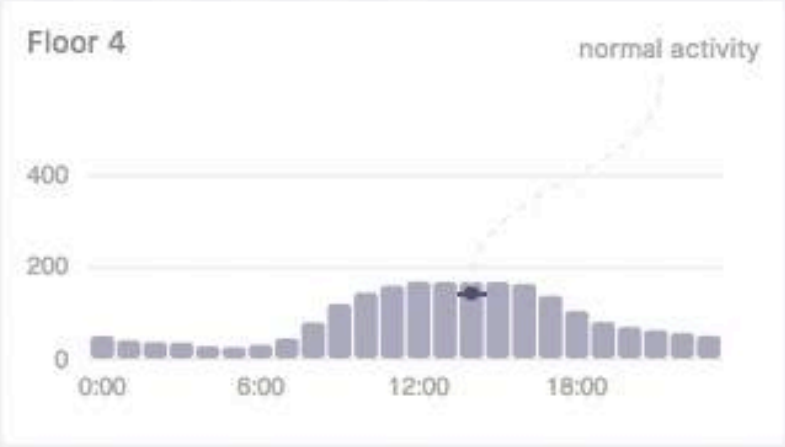
Floor 3



Floor 1



trend typical day



Implications of using non-trajectory data

- Location-based counts are often easily accessible (especially live data)
- Many analysis questions can be answered without trajectories (privacy & ethics)
- Flexible aggregation and normalization
- Use similar techniques that have been proposed for time-oriented data + new types of visual encodings

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 @oppermann_m

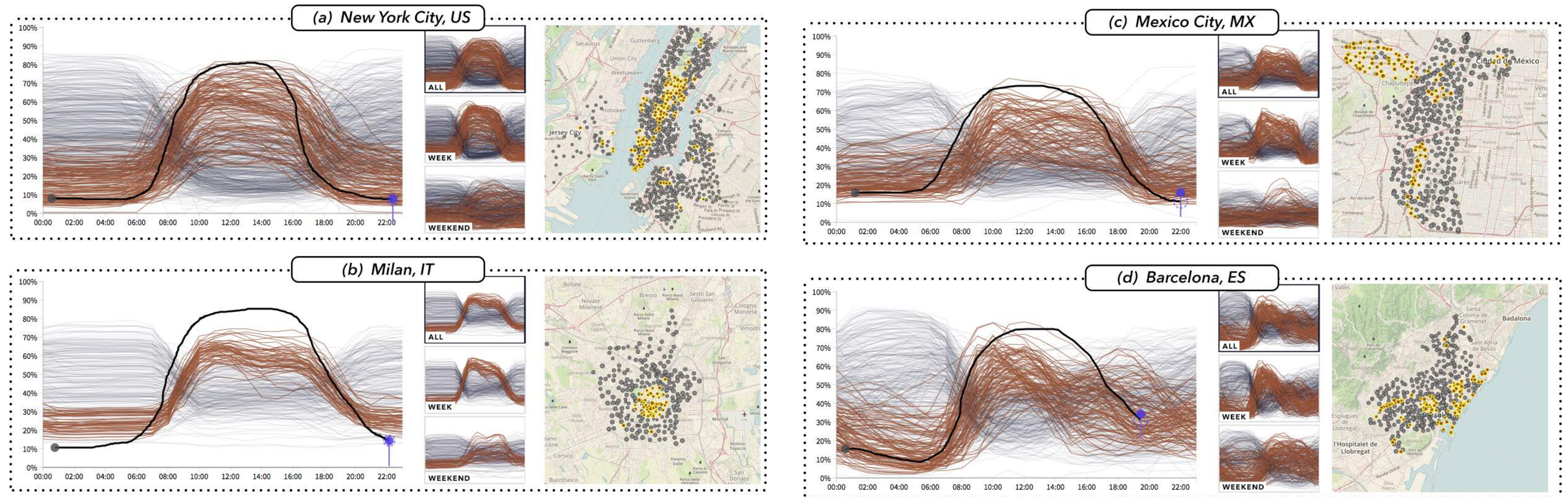


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DESIGNING
for PEOPLE

Commuting behaviours worldwide



Commuting behaviours worldwide

