

LEVERAGE CROWDSOURCED DATASETS FOR TRAVEL DEMAND ANALYSIS

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SCOPE

For decades, transportation planners and researchers have relied on traditional methodologies and experience-based surveys to analyze transit demand.

[1] Problems of this approach:

- ▶ Mobility Service providers not willing to share their data
- ▶ Users misbehavior on travel surveys

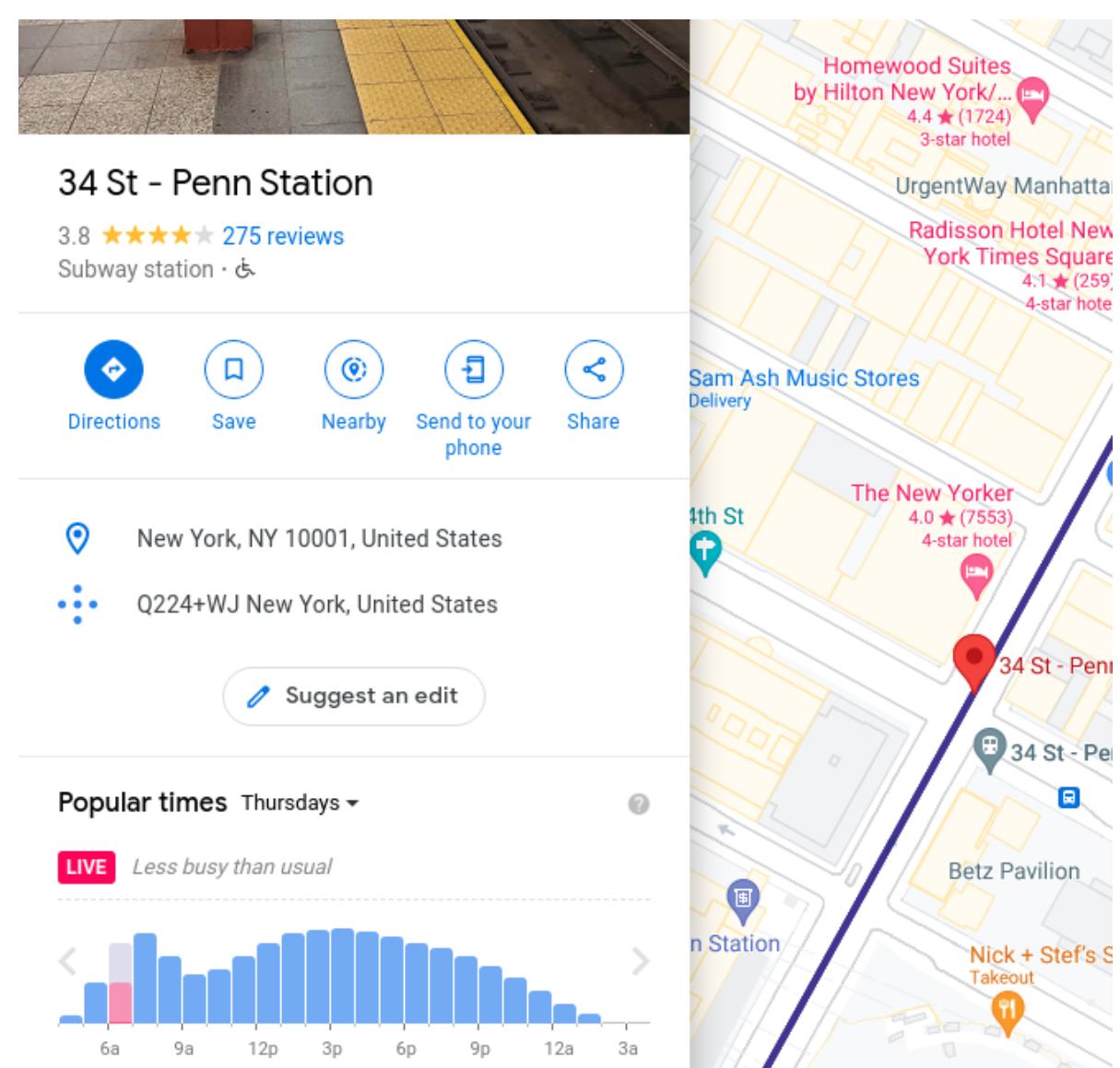
Crowdsourced data-driven approaches may provide novel solutions in this direction.[2]

PREDICTION OF LOCAL BUSINESSES ATTRACTIVENESS

We leverage crowdsensed data to enforce highly-accurate estimation of passengers demand at subway stations, this information can reach low level of aggregation not easily accessible for traditional transit dataset.

DATASET USED

Google Popular Times(GPT)



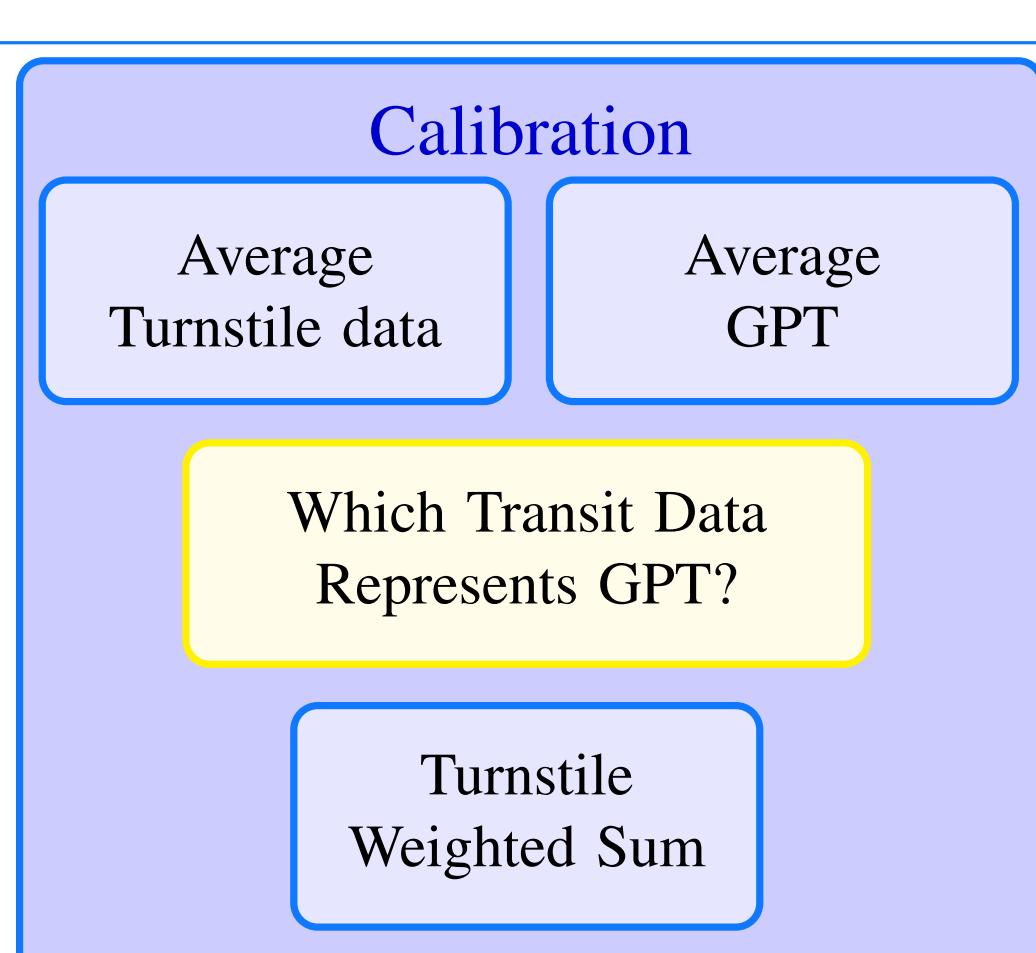
- ▶ How busy is a certain Business
- ▶ Between 0-100
- ▶ Live Value of Popularity

Turnstile Data

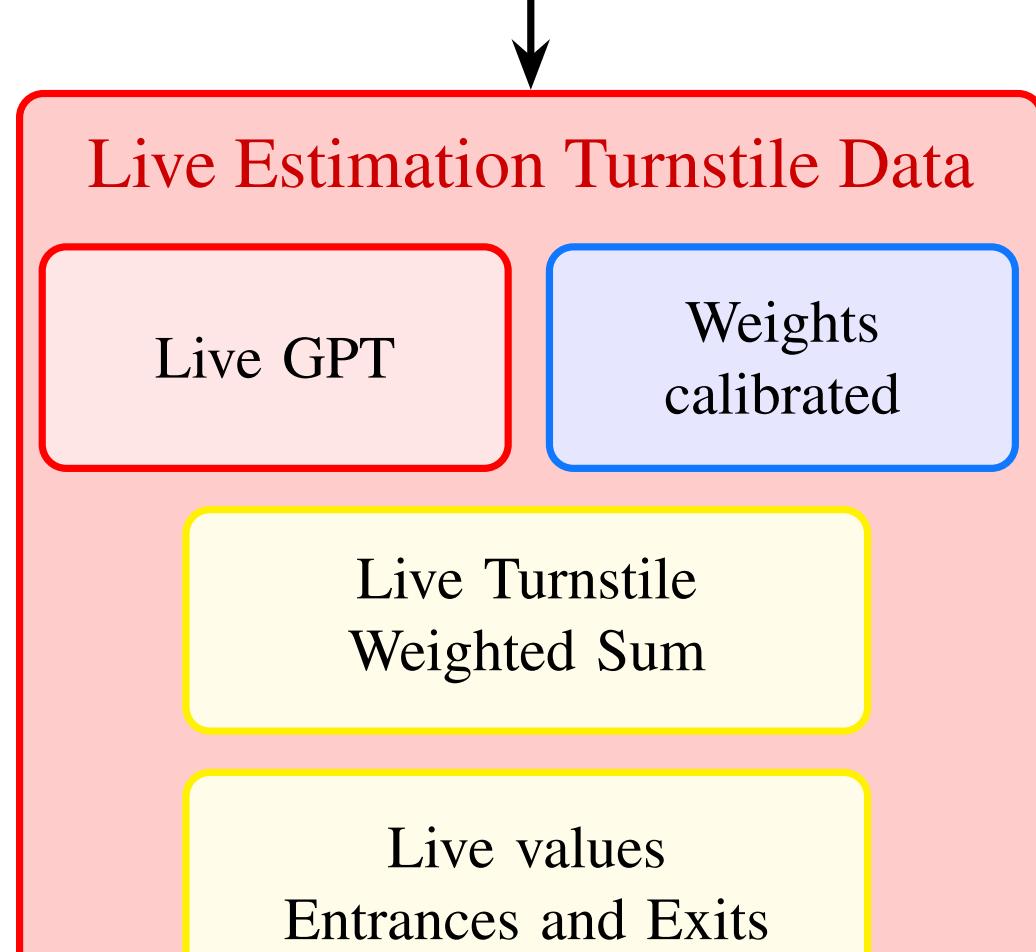


- ▶ Manhattan metro stations
- ▶ Entrances and Exits
- ▶ Granularity every 4 hours

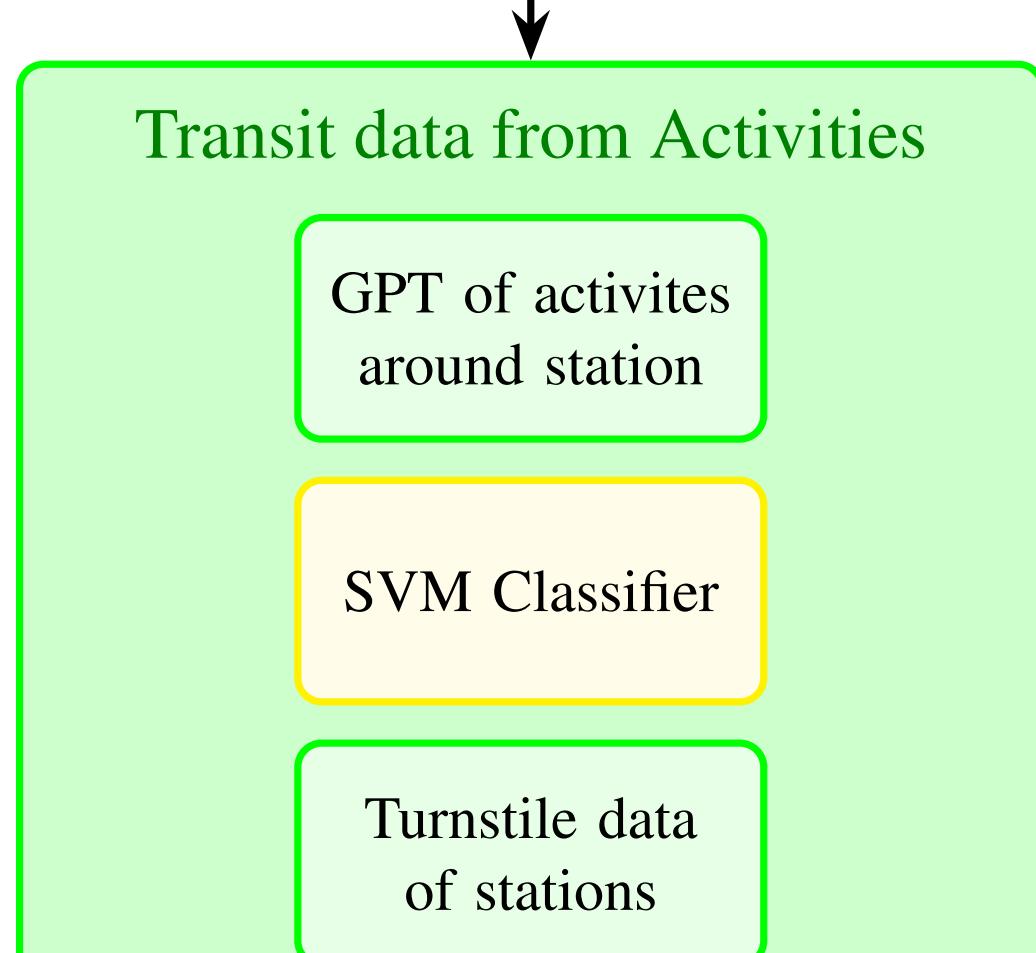
METHODOLOGY



- ▶ We compare GPT of Stations and Turnstile data of previous month
- ▶ We compute a weighted sum $wsum = a \cdot \text{Entrances} + b \cdot \text{Exits}$
- ▶ We find the optimal a and b that minimize the error between $wsum$ and GPT



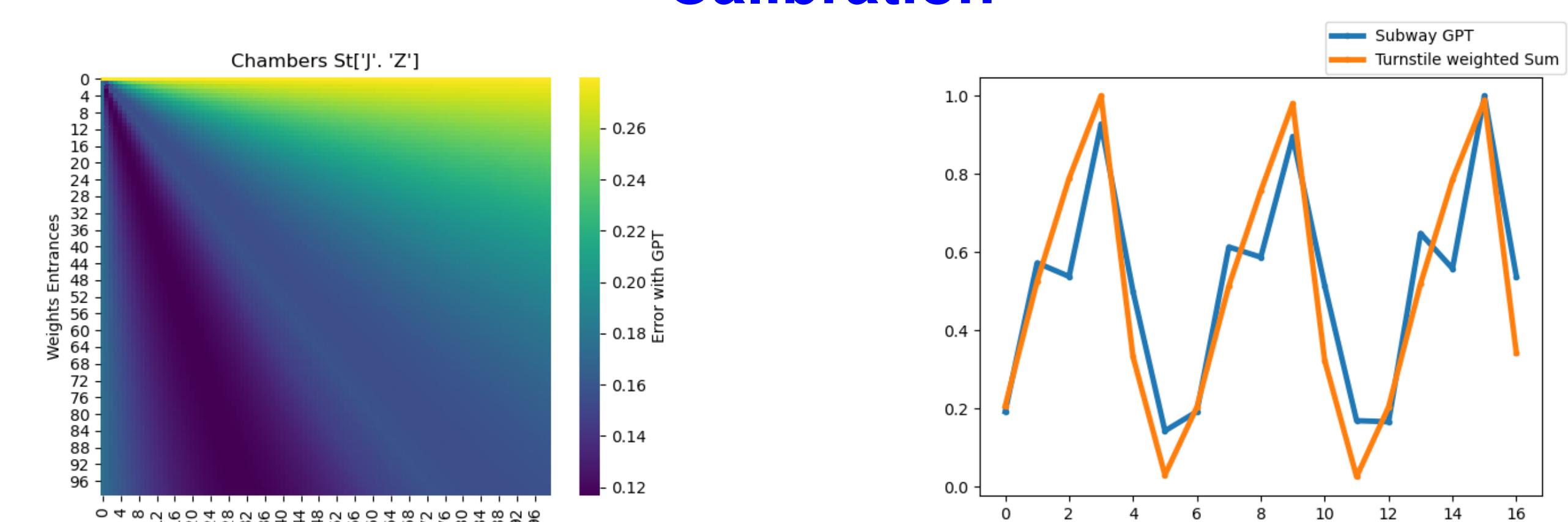
- ▶ We apply a and b calibrated before on Live GPT of the station
- ▶ We compare such data with the turnstile data of the same week
- ▶ Using the proportion computed on calibration phase, we reconstruct the real amounts of Entrances and Exits



- ▶ We create a profile of the GPT around the station (Fingerprint)
- ▶ We Use a SVM classifier with:
 - **Inputs:** Intervals of the fingerprint of the area of the station
 - **Outputs:** The transit demand at the station on the corresponding interval

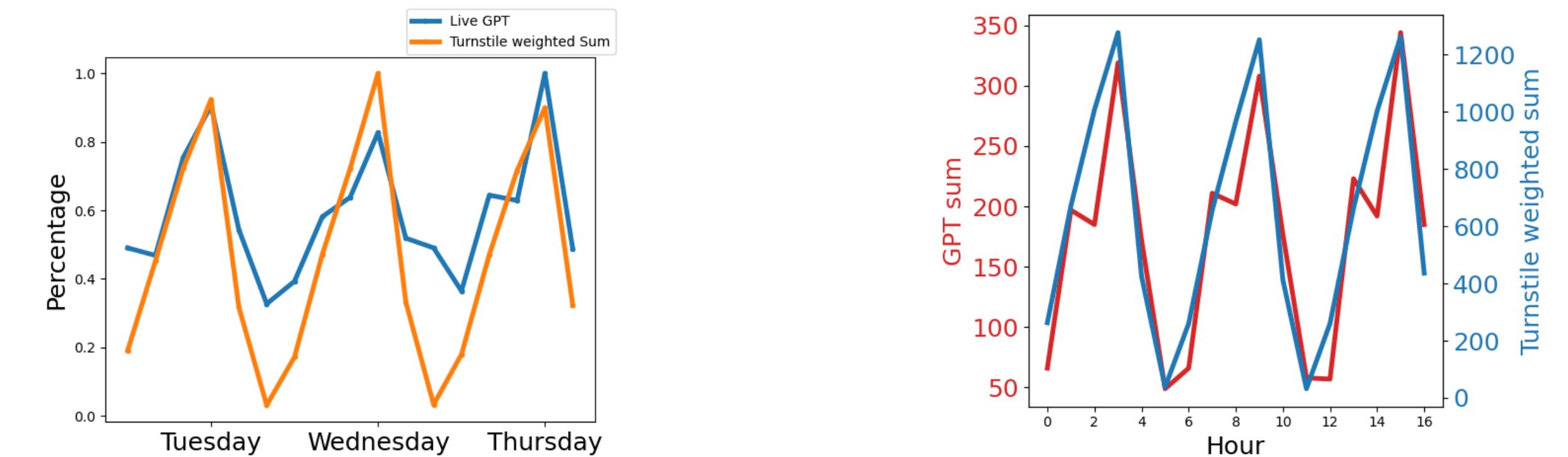
RESULTS

Calibration



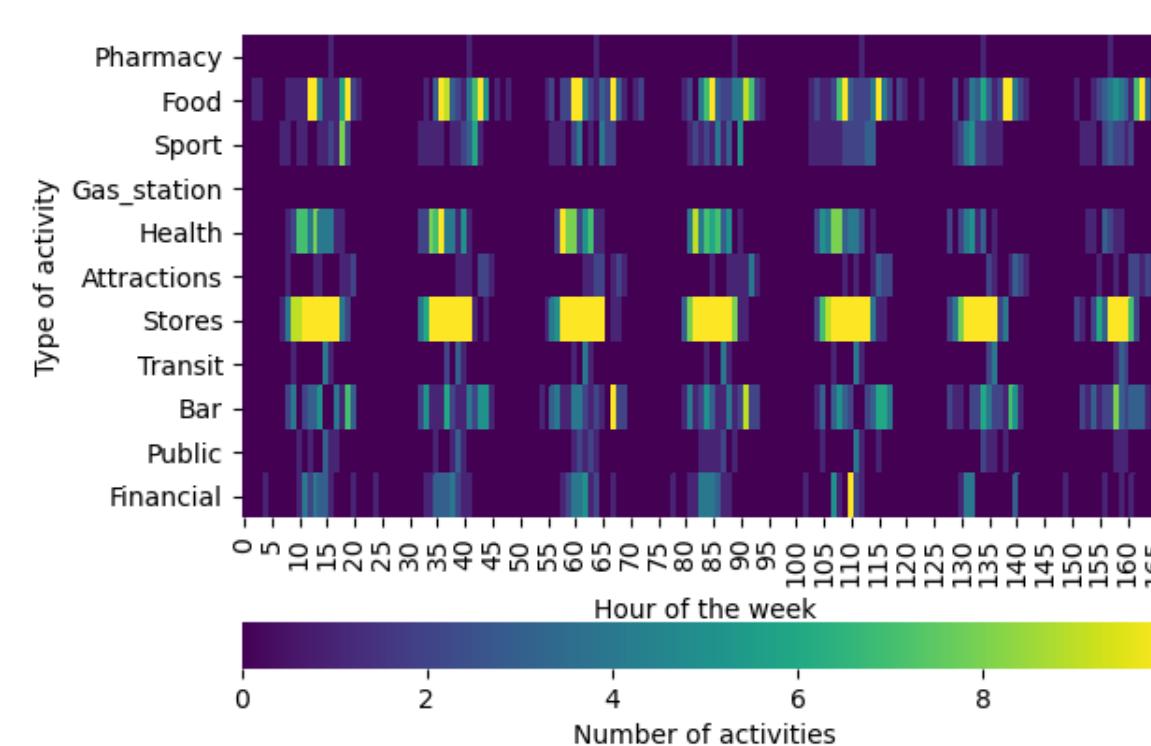
- ▶ We calibrate the weights of $wsum$ for each station
- ▶ The majority of stations have higher weights for entrances than exits (passengers entering the station have to wait the subway)
- ▶ $wsum$ is a good representation of GPT of stations

Live Estimation Turnstile data through Live GPT (Transit)



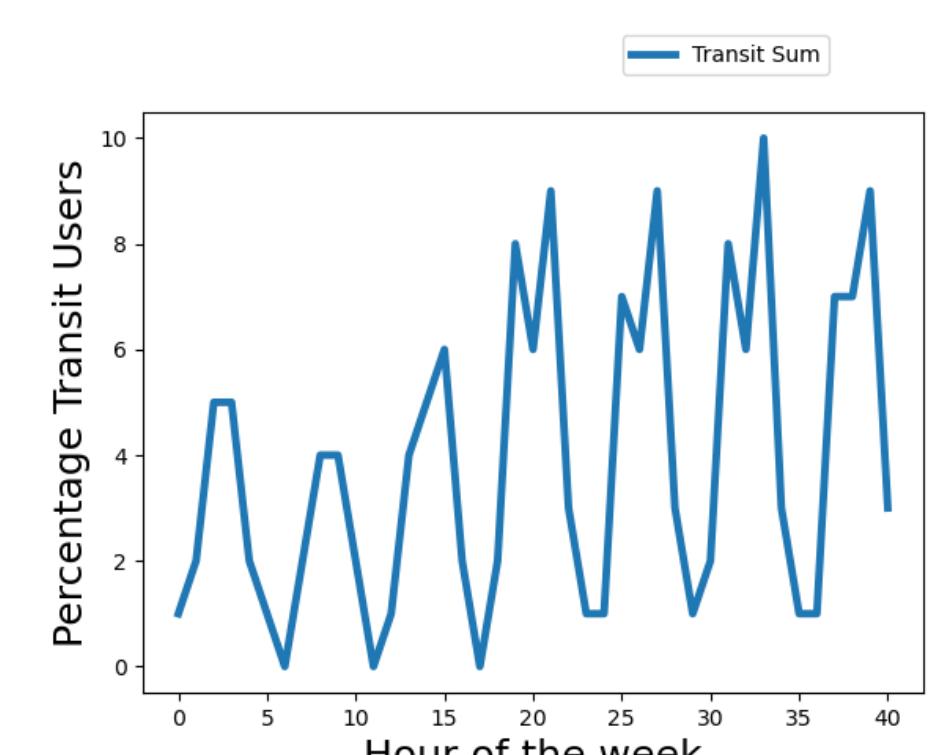
- ▶ First we try to replicate $wsum$ on Live GPT of a specific week
- ▶ We obtain a similar profile but with a lower accuracy than calibration phase
- ▶ Then using the live GPT we reconstruct the real number of passengers using the proportion between GPT and Turnstile computed during calibration

Estimation Turnstile data through GPT of Local businesses



- ▶ The heatmap represents the behavior of activities around the station
- ▶ Lighter colors mean higher popularity for a specific category
- ▶ We use slices of this heatmap as Input for a SVM Classifier

- ▶ We extract the turnstile behavior on the interval of the heatmap
- ▶ We use slices of this data as output class for SVM



- ▶ The SVM estimate the Turnstile data with a high accuracy

ACKNOWLEDGEMENT

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REFERENCES

- [1] Marie-Pier Pelletier, Martin Trépanier, and Catherine Morency. "Smart card data use in public transit: A literature review". In: *Transportation Research Part C: Emerging Technologies* 19.4 (2011), pp. 557–568.
- [2] Andrea Capponi et al. "Crowdsensed Data Learning-Driven Prediction of Local Businesses Attractiveness in Smart Cities". In: *IEEE Symposium on Computers and Communications (ISCC)*, Barcelona, Spain, 2019.