

Goal

- Does controlling mobility control the spread of an epidemic?
- Simplification: Does the foresight of mobility help us better predict the spread of an epidemic?

Data Sources

- SafeGraph Inc. point of interest (POI) visitor patterns data
- Google Mobility Report LA County
- 2010 census block group data
- LA Times confirmed cases
- LA County Public Health cumulative infection

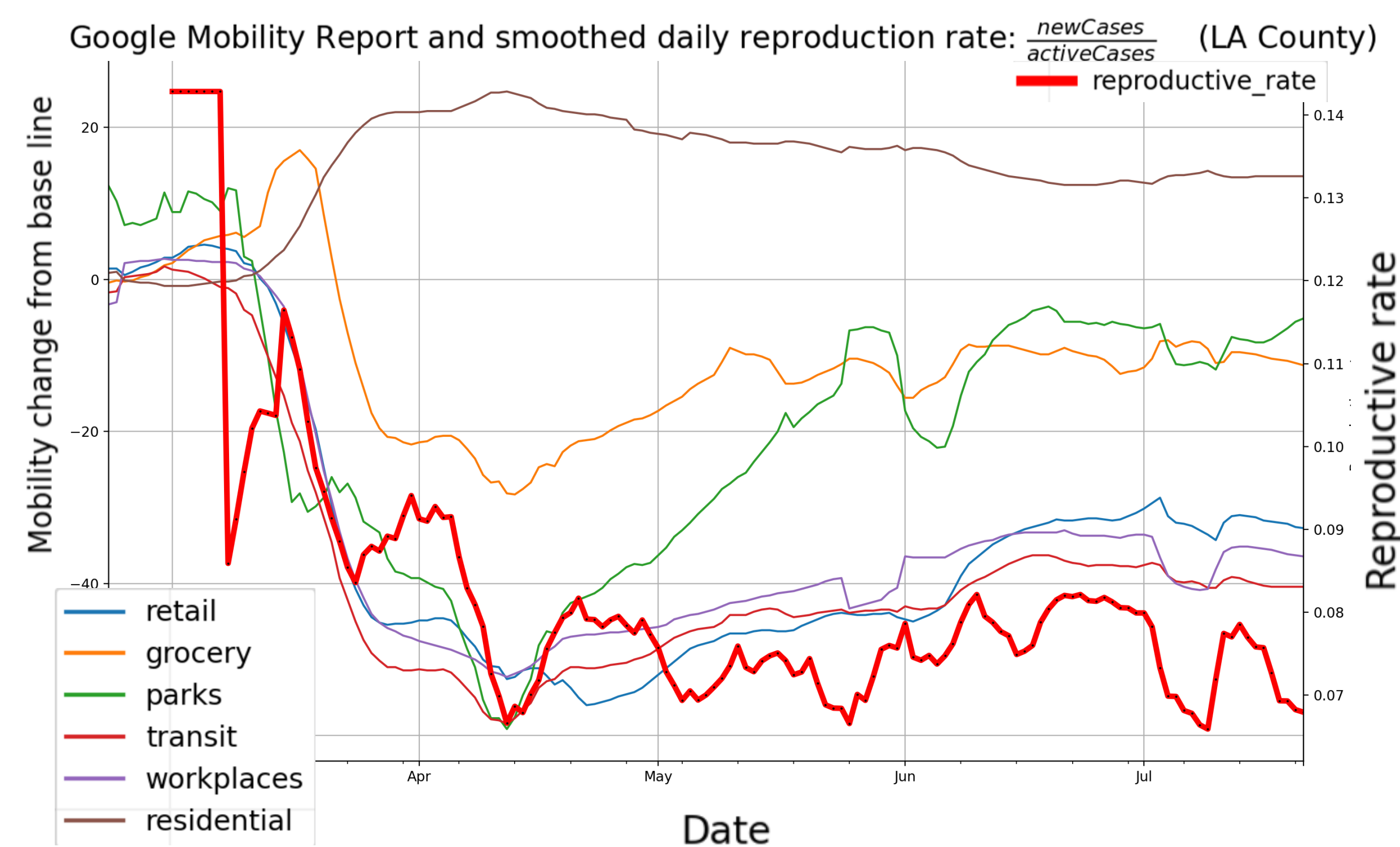


Figure 1: Google Mobility Report LA County data and COVID-19 reproduction rate.

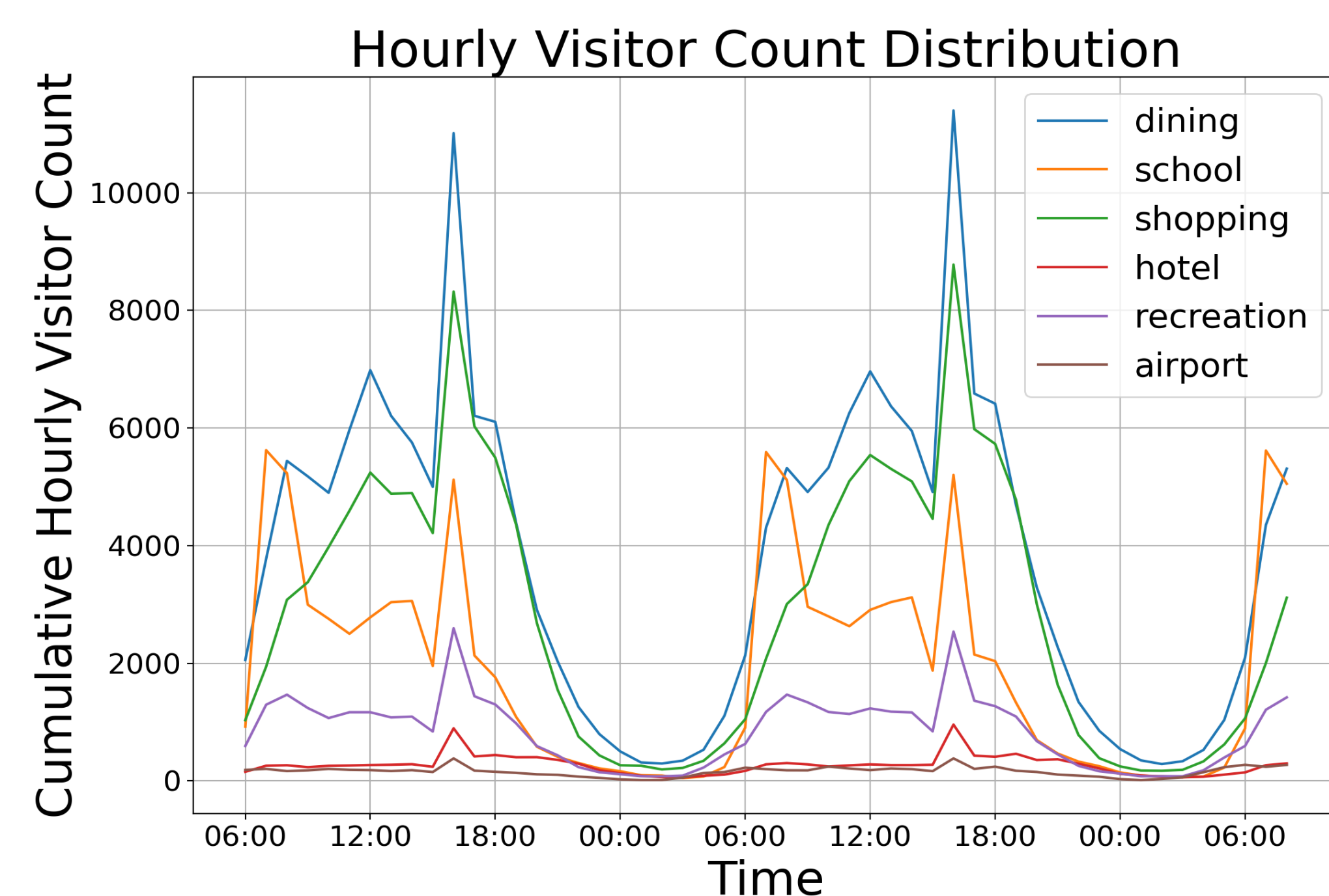


Figure 2: A snapshot of hourly visitor patterns in SafeGraph data

Methodology

Two branches of variations of the SIKJ α model that incorporate

- Inter-region Mobility (original SIKJ α formula)
- Visitor-pattern Mobility. (extended formula, see equation 1)

$$\Delta I_d = \sum_{i=0}^{M-1} \left(\sum_{j=0}^{k_i-1} \beta_{ij} \cdot \text{sus}(start_{d,i,j}) \cdot \Delta I(start_{d,i,j}, end_{d,i,j}) \cdot \text{mobavg}_i(start_{d,i,j}, end_{d,i,j}) \right)_{(1)}$$

- **M**: number of mobility features
- **sus(d)**: susceptible population on date d
- **$\Delta I(s, t)$** : increase in infection from date s to date t.
- **mobavg $_i(s, t)$** : date s to t's average mobility score for i^{th} mobility feature
- **start $_{d,i,j}$ and end $_{d,i,j}$** : starting and ending date of j^{th} time window for i^{th} mobility feature.

We also performed manual and agglomerative hierarchical clustering on the SafeGraph visitor patterns data to build distinct mobility features based on the nature of POI locations.

Results

- Inclusion of mobility features consistently produce lower errors
- Micro-level mobility features outperform macro-level ones.
- The underlying pattern need not be explicitly identified by location types

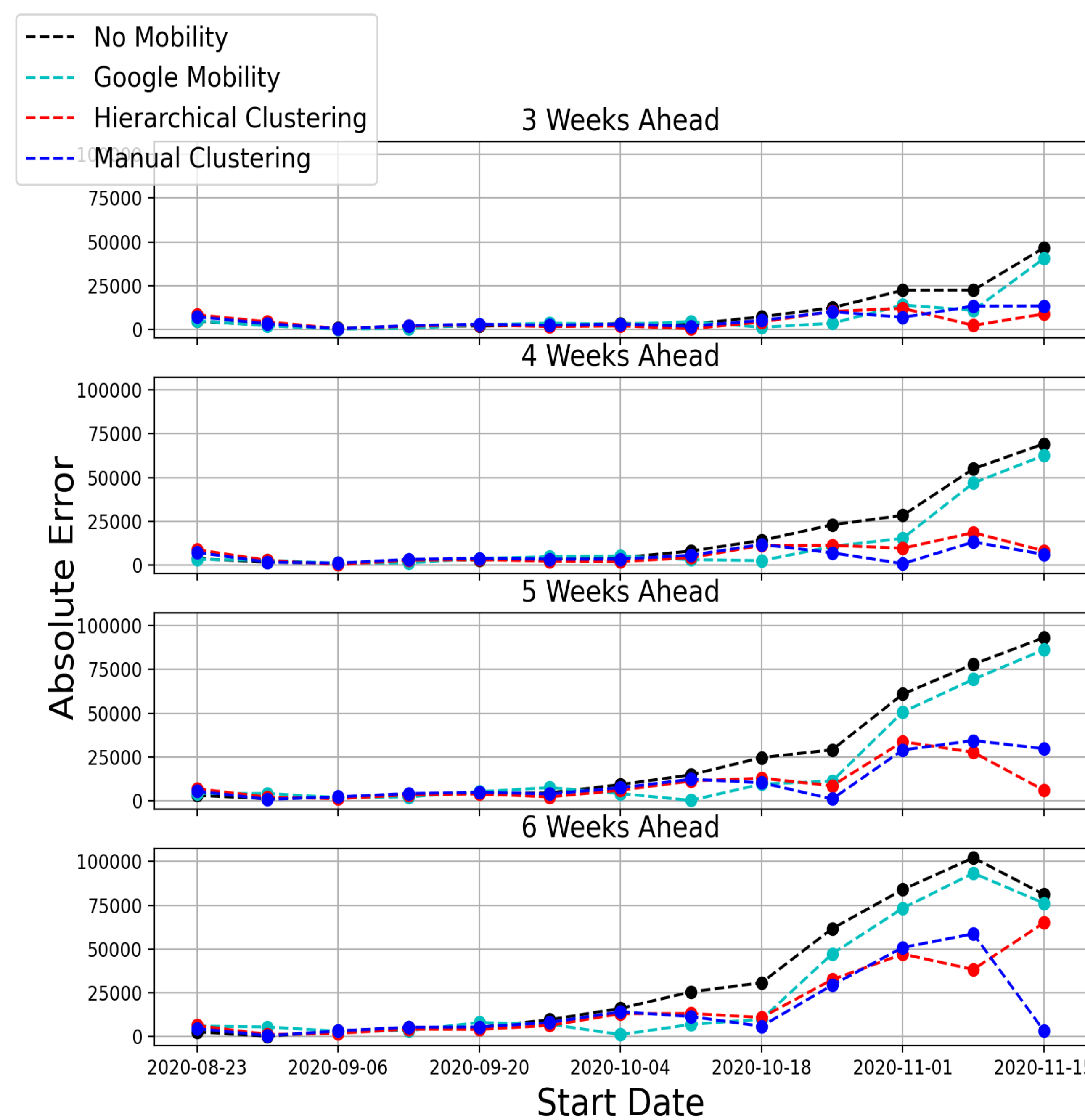


Figure 3: Absolute error of all the visitor pattern mobility models

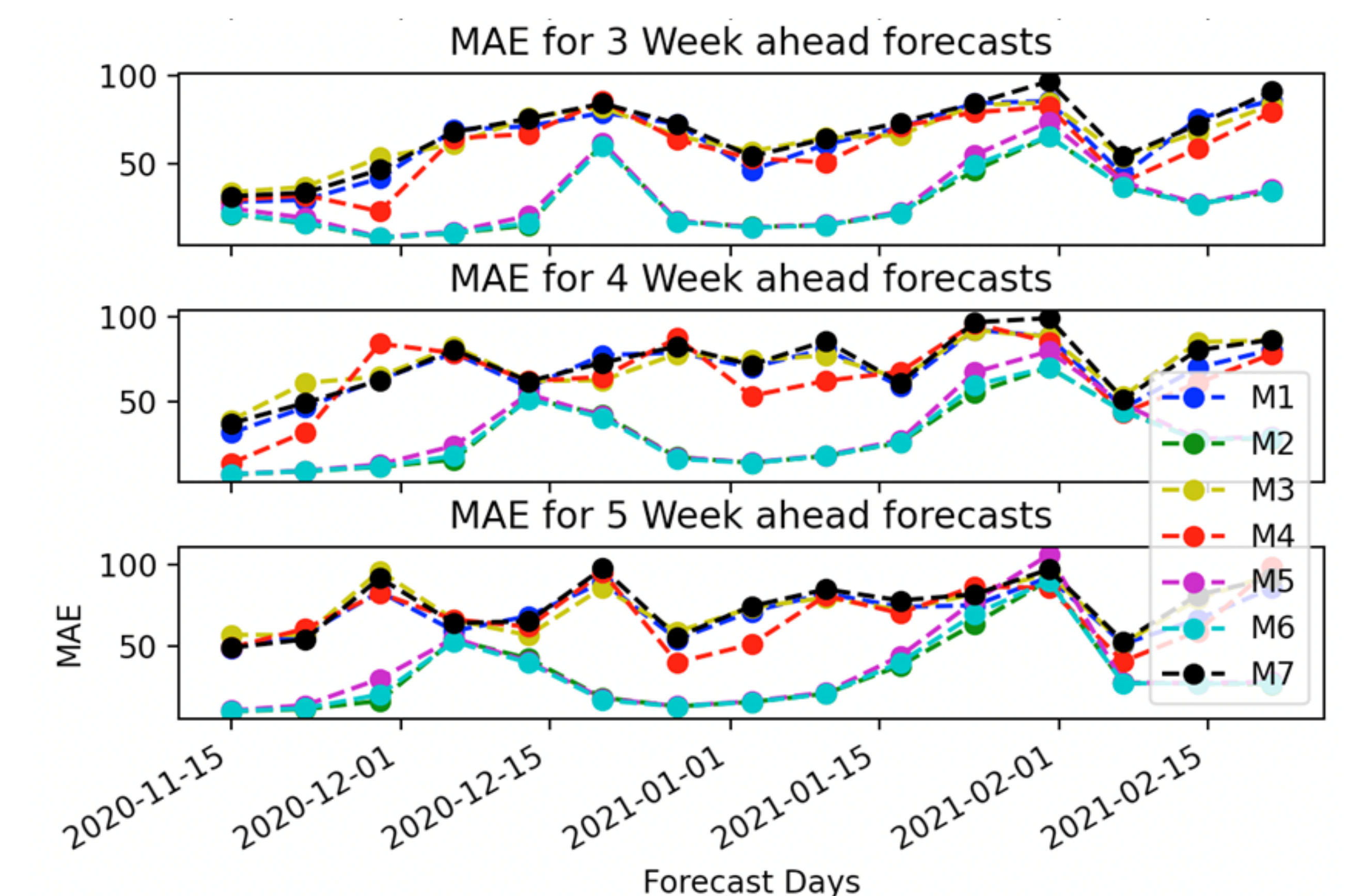


Figure 4: MAE of all the inter-regional mobility models