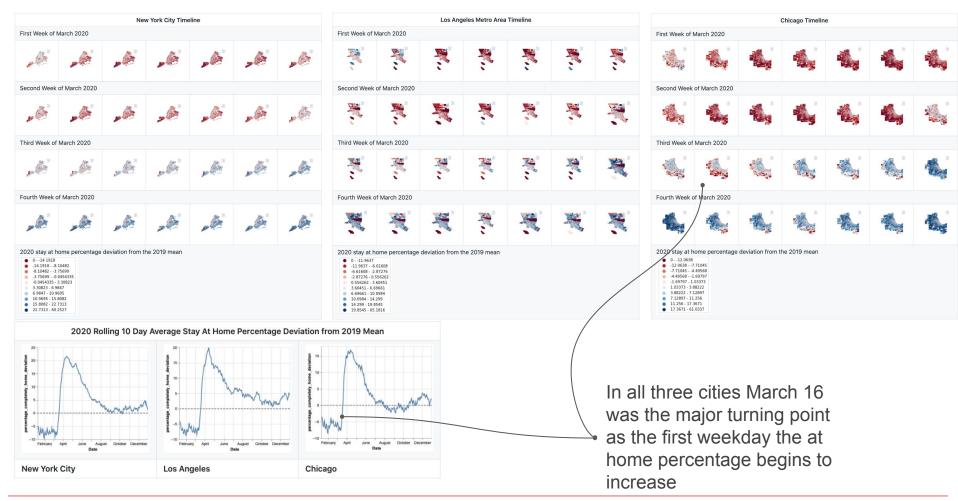
Visualizing the Mobility Gradient

Over Time and in Relation to Poverty

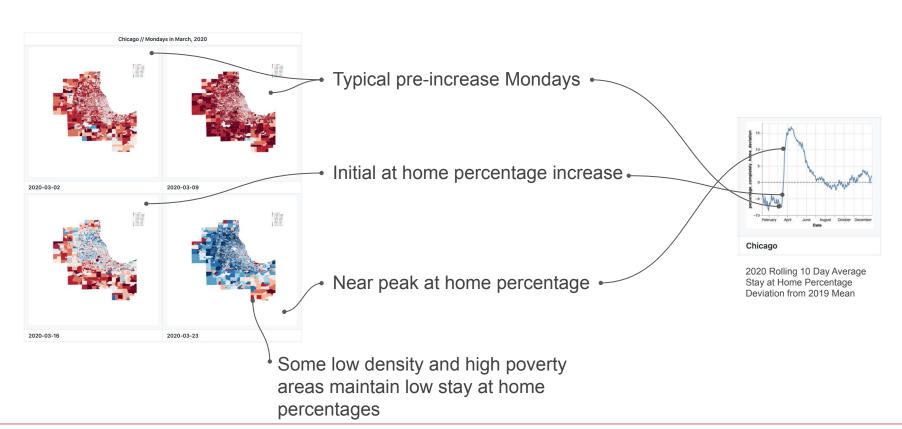
Justin Snider (js10853) Anchit Srivastava (as14022) Diksha Chouhan (dc4454)

https://github.com/chouhandiksha/bigdataproject

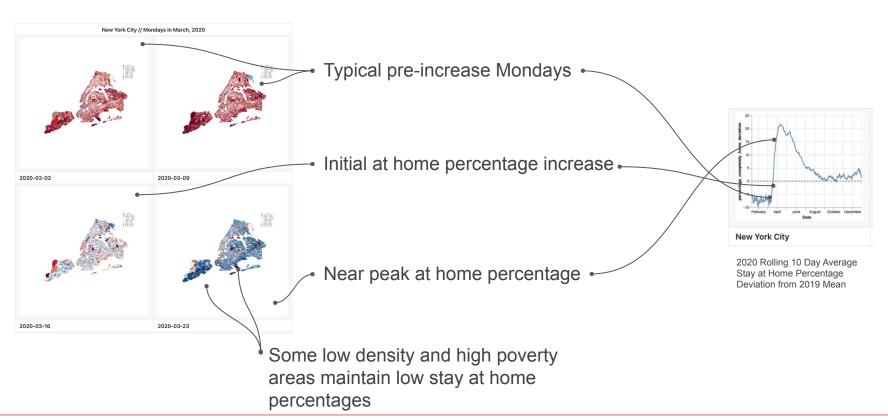
Urban Mobility Gradient



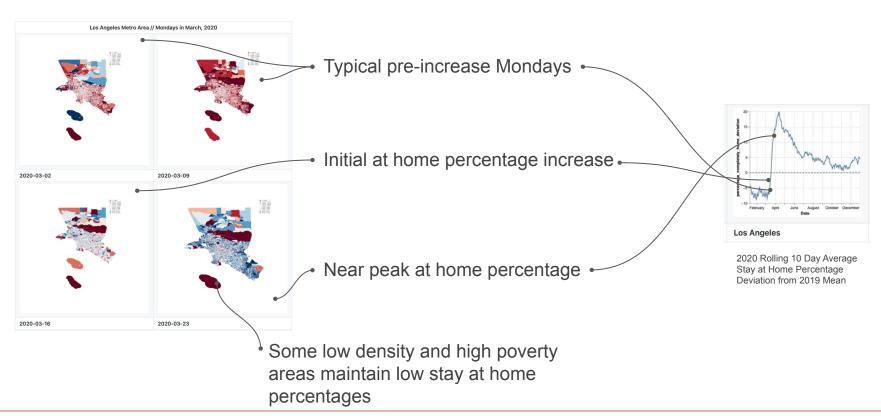
Urban Mobility Gradient // Transformation in Daily Fully at Home Behavior // Similarities



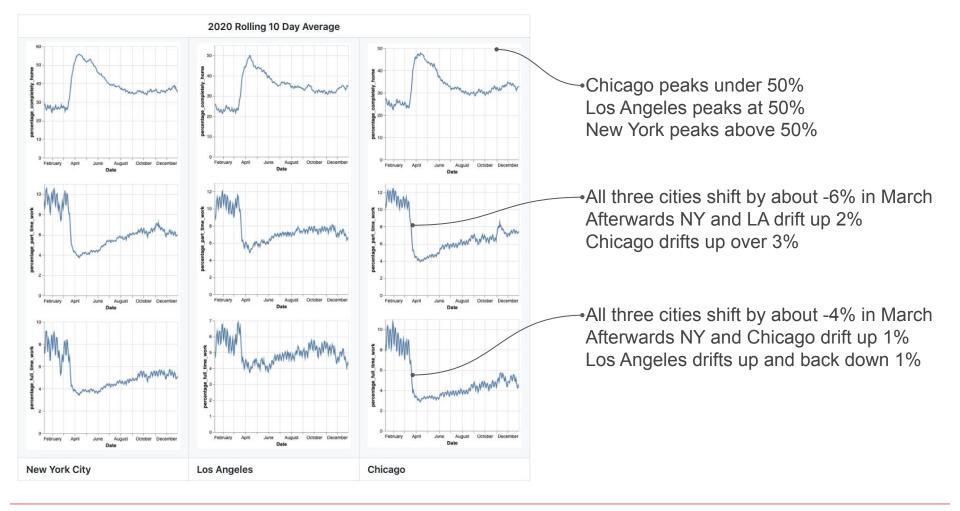
Urban Mobility Gradient // Transformation in Daily Fully at Home Behavior // Similarities



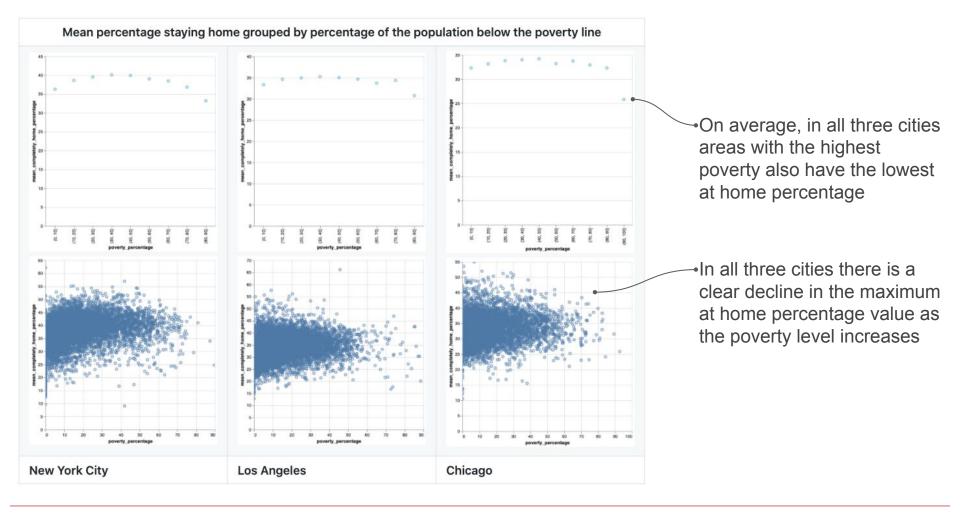
Urban Mobility Gradient // Transformation in Daily Fully at Home Behavior // Similarities

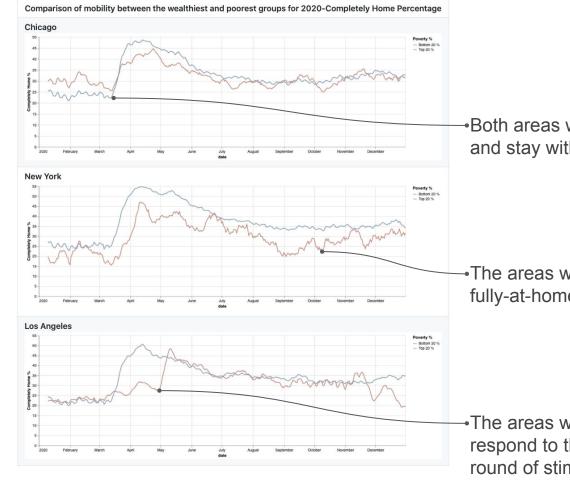


Urban Mobility Gradient // Transformation in Daily Fully at Home Behavior // Similarities



Poverty and Mobility

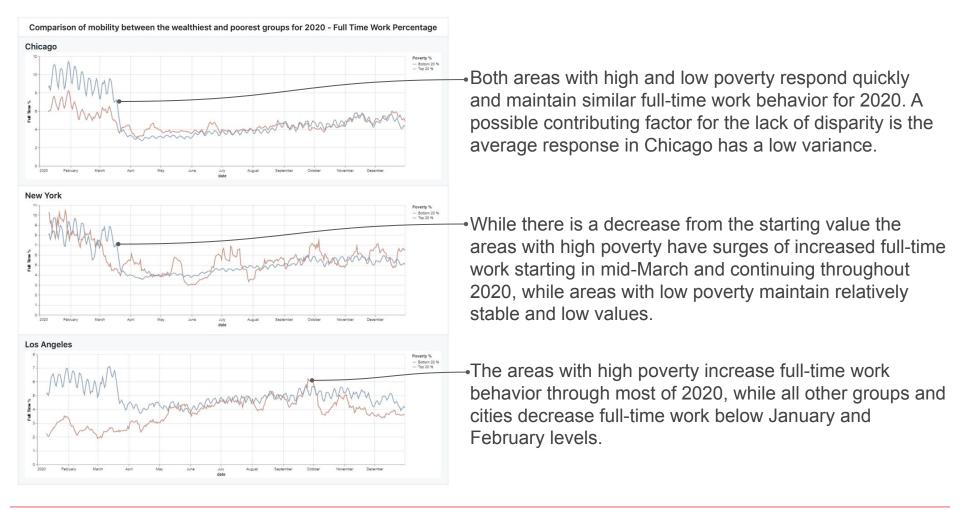


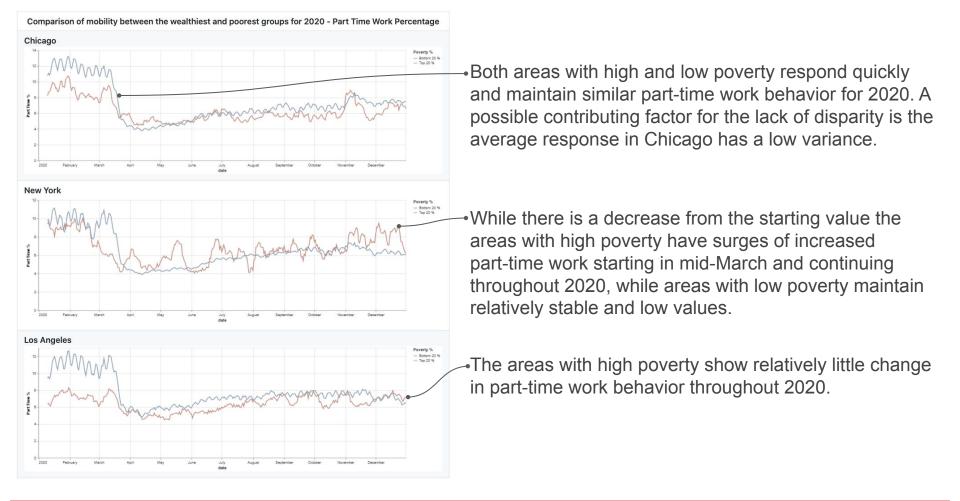


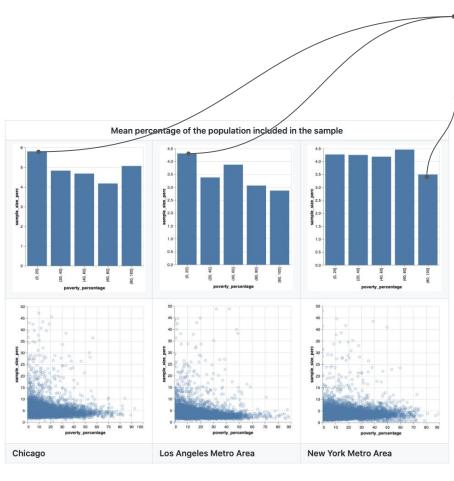
→Both areas with high and low poverty respond quickly and stay within about 5% for 2020

•The areas with high poverty have 5% to 15% lower fully-at-home percentage sustained over most of 2020.

•The areas with high poverty take one and a half months to respond to the crisis. This is a few weeks after the first round of stimulus checks.







The areas with low poverty have the highest percentage of the population included in the SafeGraph Social Distancing data.

The areas with highest poverty have the lowest percentage of the population included in the SafeGraph Social Distancing data.

In general the data includes on average 2.5% to 6% of the population. So we should keep in mind those without phones and not included will not be represented in this dataset.

You can see specific census block groups have higher and lower representation in the data.

Conclusion

Challenges

- Finding clear and useful information in such a large dataset was a substantial challenge.
- We repeatedly hit the Google Drive 1,000,000,000 file read limit while doing analysis. As a solution we created multiple clones of the Google Drive account to spread out usage.
- The SafeGraph Social Distancing Metrics data set is 81 GB and took 6 hours to download. The download sometimes fails due to the long download time.
 Download requires the use of the AWS command line interface.
- The 731 days of data for each the 11,078,297 census block groups in the United States was challenging to process. All together there are approximately 8,098,235,107 rows of data. Using Spark and filtering out just the census blocks for the three cities made analysis much more efficient.
- The American Census Survey data was not intuitive to understand. There data attributes do not have intuitive names and they provide population counts. We had to extract just the useful columns and transform them into percentages to make the data useful.

Key Findings

Mobility

- Chicago show the lowest overall change in mobility.
- New York showed the highest change in mobility.

Poverty

- New York has the most sustained disparity between areas with high and low poverty.
- Areas with high poverty in Los Angeles have the most delayed response to taking a month and a half to increase their stay at home percentage.
- Chicago had the least disparity at least in part because the wealthy show relatively low levels of change in mobility compared to NY and LA.

Sampling Bias

A sampling bias exists in the SafeGraph Social Distancing data. The percentage
of the population included in the SafeGraph dataset is higher in areas with less
poverty in all three cities.