

# VASA: mapping spatiotemporal structure of mobility patterns during the COVID-19 pandemic

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# I. The problem

# Introduction

- First and second order spatial effects violate assumptions of classic inferential statistic (independence and iid)
- Exploratory tools from spatial statistics and econometrics that help us assess phenomena while taking into consideration these spatial effects
  - Moran's I, Geary's C, Ripley's K, etc.
  - ~~Space-time data-mining and visual analytics~~

# Problem

- Most common visual display in the geostack (PySAL, GeoDa, ArcGIS, QGIS, spdep, etc.) utilize choropleth maps, where statistically significant clusters (hotspots/coldspots) are denoted in blue/red.
- While this graphical display is ideally suited for purely spatial data, spatiotemporal and multivariate representations are problematic. How do we plot study areas with large number of spatial units?

## II. The software



# VASA - Visual Analytics for Spatial Associations

- FOSS written on Python
- Ideally suited for exploratory spatial analysis on spatio-temporal areal data (e.g. Census units). Works with vector data.
- Offers several utility functions for spatial aggregation and visualizations



- Check out the [Github repo](#)
- Navigate example [notebooks](#)
- Install via pip (currently on [test](#), alpha version)

```
pip install -i https://test.pypi.org/simple/VASA
```



# III. The Case Study





# Assessing COVID-19 induced mobility in USA in 2020

- 2 data sources: *SafeGraph* (Open) & *Cuebiq* (Proprietary)
- 2 metrics: %-sheltered at home & distance travelled
- Approximately 3,200 counties over the period of 2020 (daily data)



# General Workflow

0. Import data and clean
1. Create a VASA Object
2. Aggregate by spatial units
3. Downsample to Week
4. Calculate Local Moran's  $I$  for each week and county
5. Map *recency* | *consistency* | *week number* to visual variables
6. Plot on a graphic display

# Create a VASA Object

```
v = VASA(  
    df=data,  
    gdf=counties_map,  
    group_summary=lambda g: g[:2], # state level  
    df_group_col="fips",  
    gdf_group_col="GEOID",  
    date_col="ds_m4",  
    temp_res="week",  
    date_format="%Y-%m-%d"  
) .group()
```

# Impute missing values

Caution! VASA is very finicky about missing values!

```
v.pct_full_missing()
```

```
array([0.03217503])
```

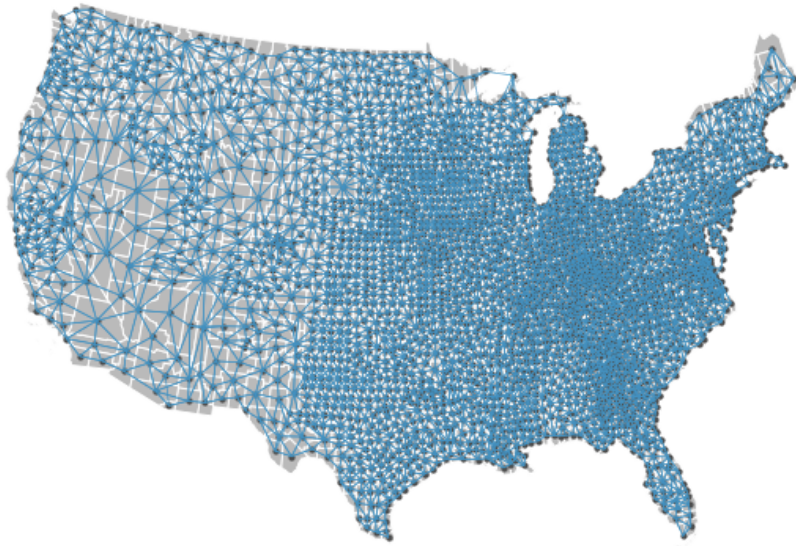
```
v.fill_missing()
```

```
<VASA.vasa.VASA at 0x7f844422e250>
```

# Build Spatial Weights Matrix

Since VASA utilizes Moran's  $I$ , we need to construct spatial weights matrix (SWM). There are different methods to create SWM (Rook's, Queen's, kNN, etc.). Please refer to [PySAL docs](#) to learn more.

```
v.show_weights_connection(k=3, type="union")
```



# Recency and Consistency Map

Better suited to study the aggregate (summary-level) details.

```
stacked.plot_both(a=2500, b=600)
```

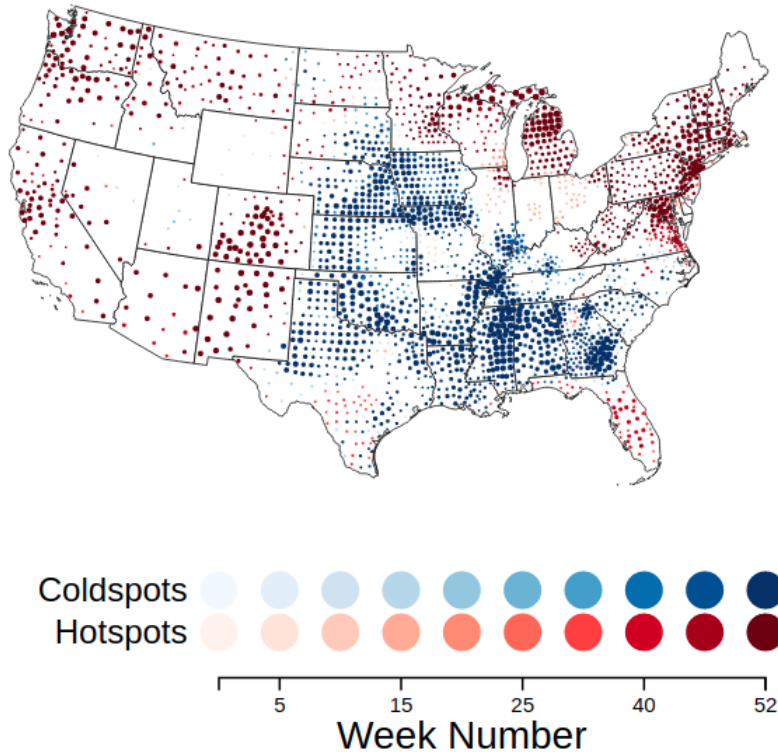
**hotspots**: high values co-located with other high values.

**coldspots**: low values co-located with other low values.

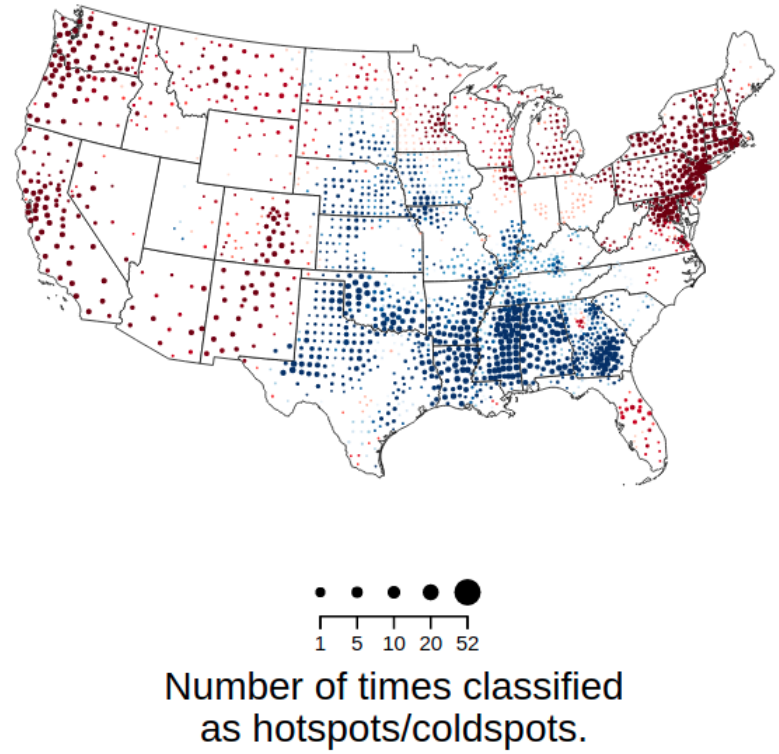
Clustering **consistency** is illustrated by **size**.

Clustering **recency** is depicted by **color intensity**.

**Percentage of Sheltered in Place (Cuebiq)**



**Percentage of Sheltered in Place (Safegraph)**



# Line-path scatter plot

Better suited to study the finer (object-level) details.

```
from VASA.scatter import Scatter  
sc = Scatter(v)  
sc.plot(highlight="06") # this filters out CA
```

**hotspots**: high values co-located with other high values.

**coldspots**: low values co-located with other low values.

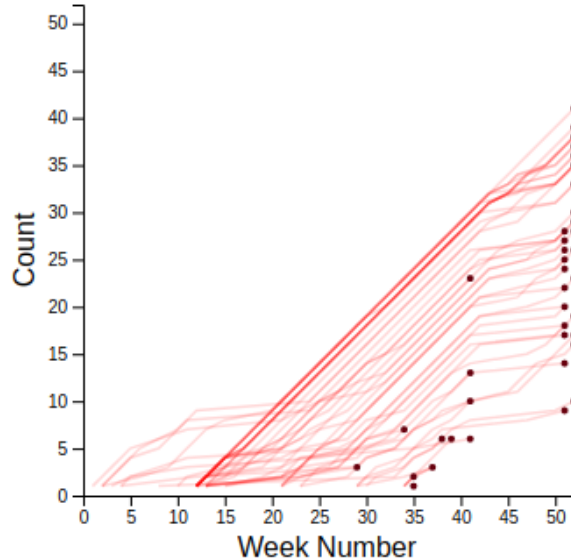
**Week number** is mapped onto X-axis.

**Number of times classified as hs/cs** is mapped onto Y-axis.

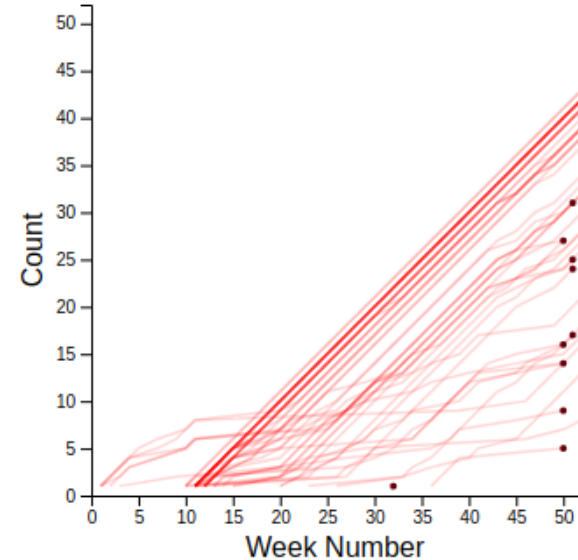


# Social distancing in California

Percentage of Sheltered in Place (Cuebiq)

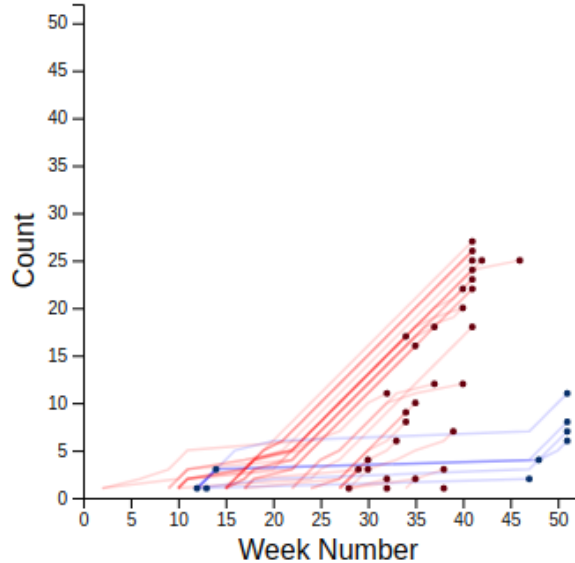


Percentage of Sheltered in Place (Safegraph)

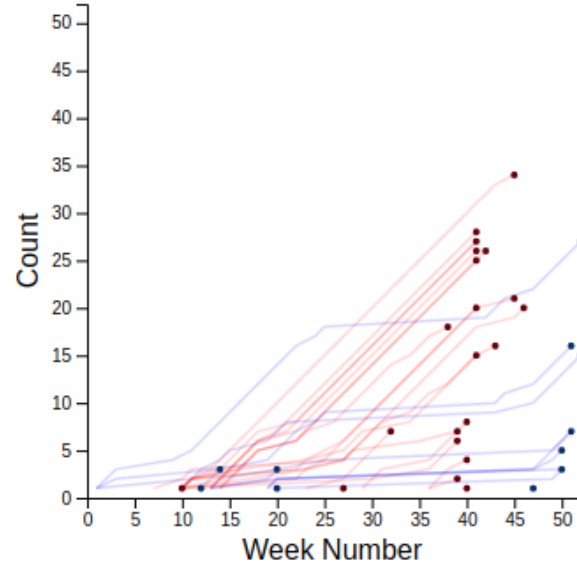


# Social distancing in Florida

Percentage of Sheltered in Place (Cuebiq)



Percentage of Sheltered in Place (Safegraph)



# IV. Conclusion



- Proposed methodology provides a robust framework for assessing spatiotemporal structure of the data (NOT only mobility!).
- VASA allows exploratory spatial data analysis at the object-level and summary-level, which makes comparison of multiple spatiotemporal series easy
- VASA adds novel visual displays to traditional choropleth mapping prevalent in spatial econometrics
- *Planned: Adding more visual displays for micro-scale analysis on areal data (e.g. LA Census Block Groups ~ 25,000)*

# Questions?

# Thank you for Zoom-listening!



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