

Geospatial Voronoi Analysis with Python

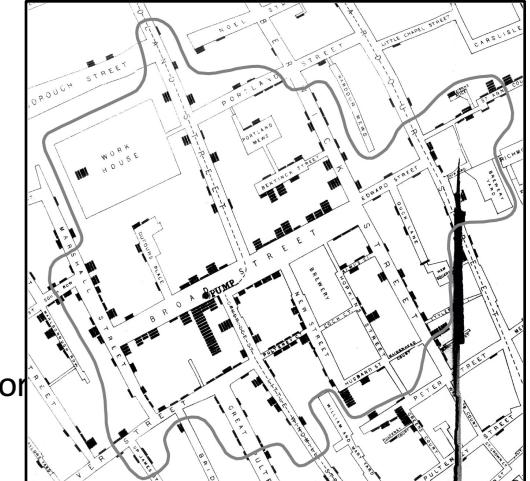
Dan Finkel PhD
Service Management Group (SMG)
Boston Python Meetup
December 20 2016

1854 Broad Street Cholera Outbreak

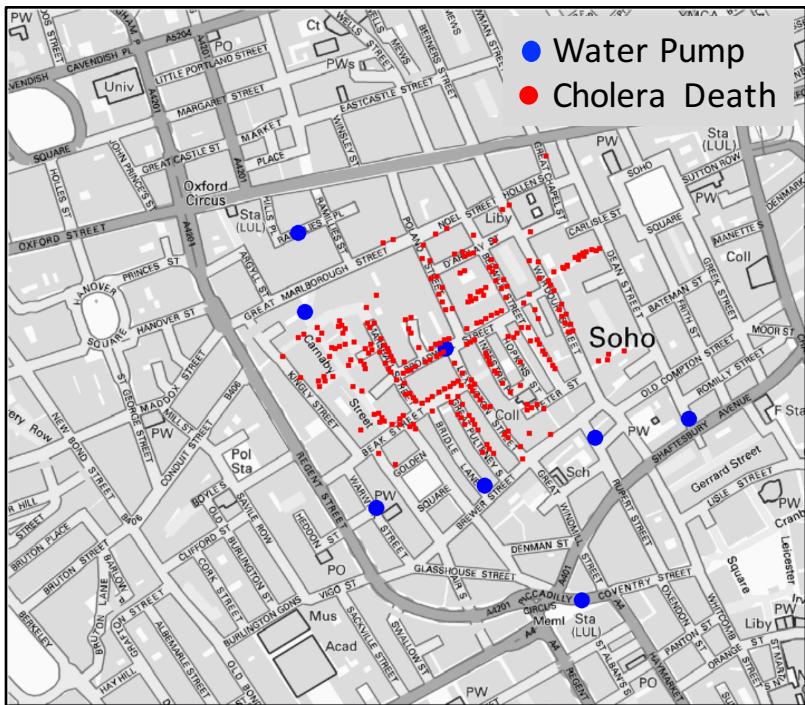
- A major cholera outbreak killed over 600 London Soho residents and caused 75% of residents to flee the area
- A local physician, John Snow, connected the outbreak to filth polluting the water
 - Conclusions were based on graph and pattern analysis
 - Regarded as the founding event of epidemiology
- In the absence of hard biological evidence Snow relied on graph and pattern analysis
- Postscript
 - Snow's evidence caused Broad St well to be disabled
 - After the disease subsided Snow's theory was rejected and air pollution was blamed
 - "To accept Snow's theory of oral-foecal transmission of disease was too unpleasant for most of the public to contemplate"



John Snow Memorial

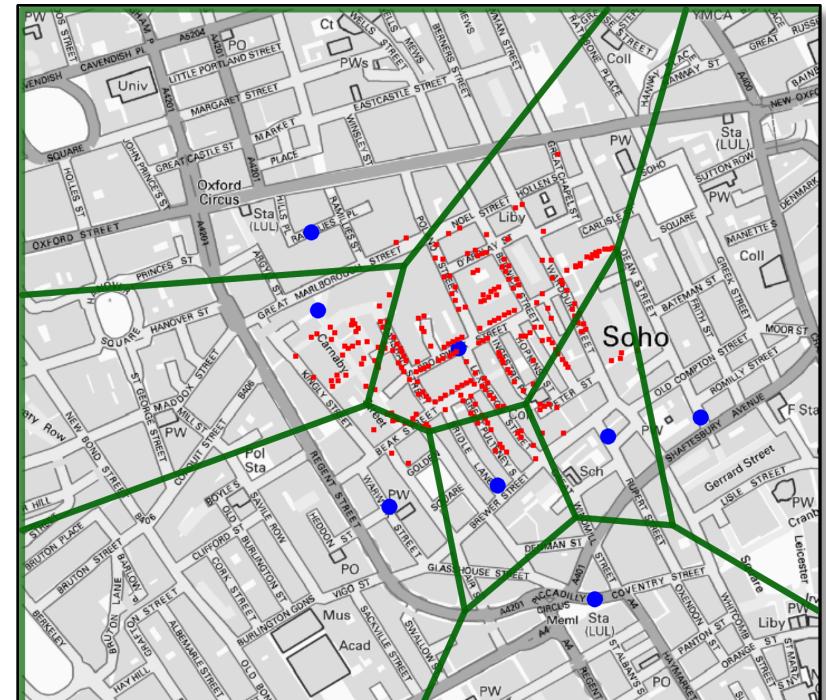
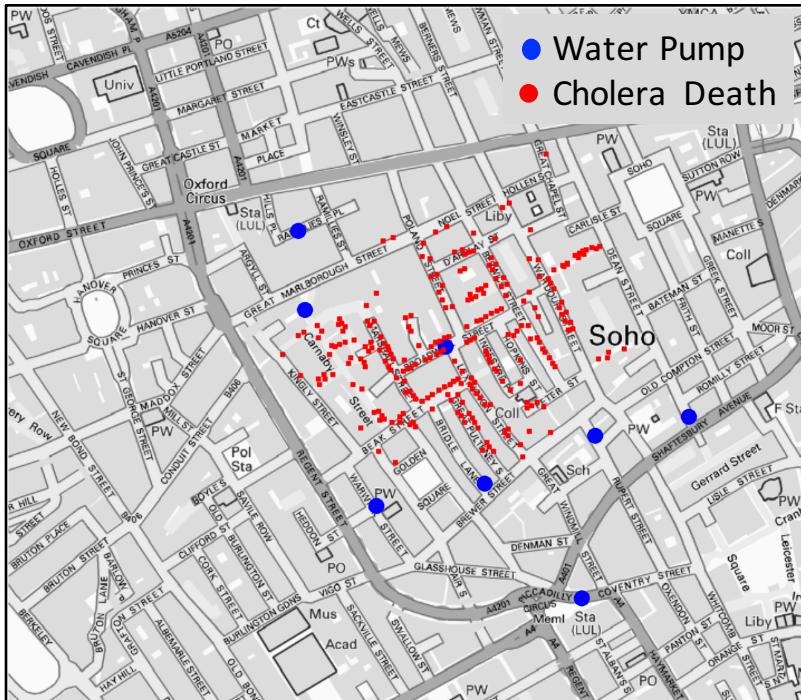


Visualizing the Cholera Outbreak by Water Pump Region



Data obtained from <http://blog.rtwilson.com/john-snows-famous-cholera-analysis-data-in-modern-gis-formats/>

Visualizing the Cholera Outbreak by Water Pump Region



Planar partitions by pump proximity helped Snow argue outbreak was due to polluted water

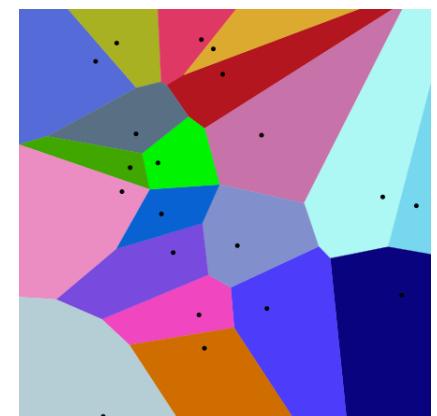
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Voronoi Diagram Overview



Georgy Voronoy
1868 - 1908

- A Voronoi diagram is a planar partition based on distance to a set of points (“seeds”) within the plane
 - Each seed defines a partition as those points in the plane closer to that seed than any other
- $$R_k = \{x \in X \mid d(x, P_k) \leq d(x, P_j) \text{ for all } j \neq k\}$$
- Algorithms / implementations
 - Typically $O(n \log n)$
 - Can scale to multiple dimensions
 - Python implementation in `scipy`
 - Utilizes `Qhull` library
 - Computes via its dual (Delaunay triangulation)



Voronoi Diagram Example (Wikipedia)

Constructing Voronoi Polygons in Python

Simple Functionality

1. Import the library

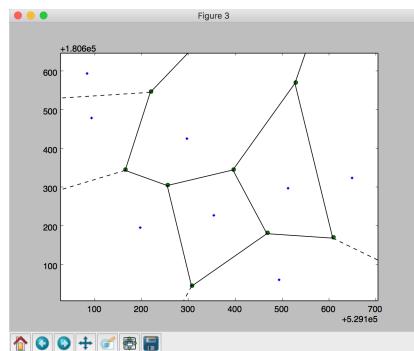
```
> from scipy.spatial import Voronoi, voronoi_plot_2d
```

2. Instantiate

```
> vor = Voronoi(pumps_xy)
```

3. Plot / Visualize

```
> voronoi_plot_2d(vor)
```



Constructing Voronoi Polygons in Python

Simple Functionality

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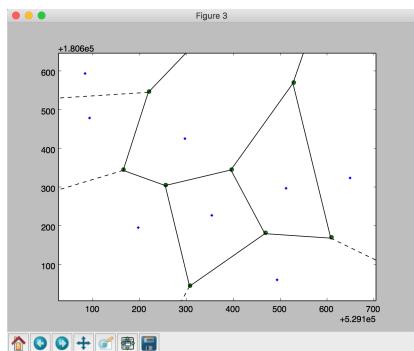
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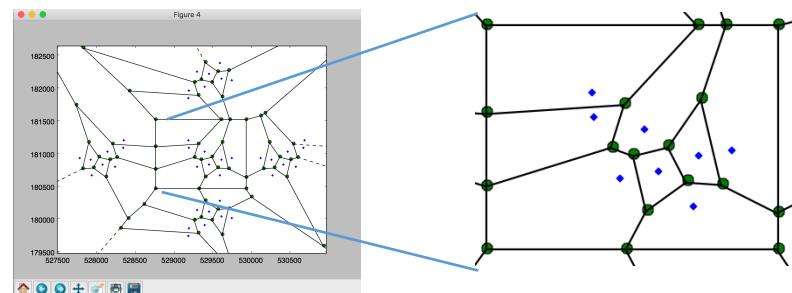
3. Plot / Visualize

```
> voronoi_plot_2d(v)
```



Additional Modifications

Mirror pumps and build Voronoi off expanded set
(to create hard boundaries)



Convert Voronoi regions into polygons

```
def build_vor_polys(vor, soho_poly, pumps_xy):
    polys = []

    # Build all lines in the voronoi region set
    lines = [shapely.geometry.LineString(vor.vertices[line])
             for line in vor.ridge_vertices
             if -1 not in line]

    # polygonize the lines and check if
    # they are original or copies
    for poly in shapely.ops.polygonize(lines):
        if is_in(poly, pumps_xy):
            poly = poly.intersection(soho_poly)
            x, y = poly.exterior.xy
            polys.append(poly)

    return polys
```

- Shapely Geospatial functions
- Compute area
- Intersect, overlap with other polygons
- Join / union with other objects

All code available on <http://github.com/danfinkel>

MLB Stadium Market Research Example



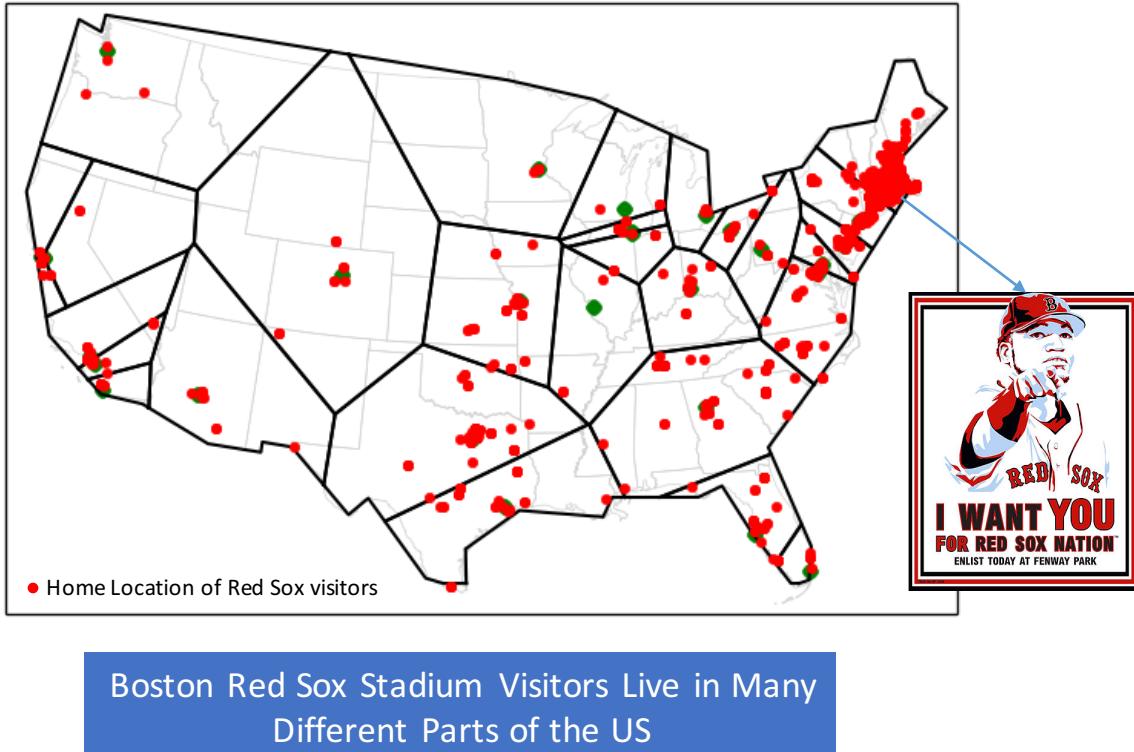
- SMG maintains a large panel of individuals who report behavior and customer experiences
- During 2016 MLB season we monitored stadium visits to identify geospatial patterns and preferences



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Behavioral Analysis

Red Sox Nation

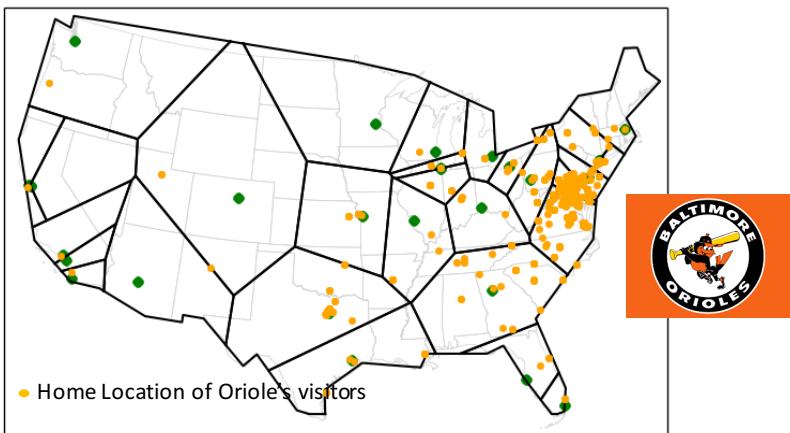
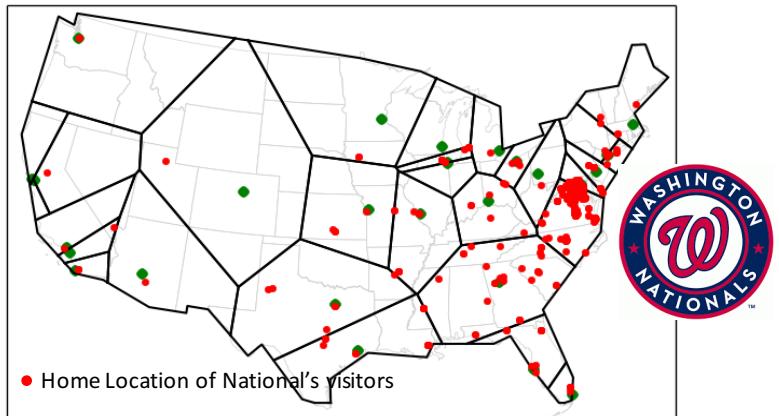


SMG Panelists Visit to Voronoi "Home" Team (2016)

| Stadium | Home Ratio | Away Ratio | Rival Ratio |
|------------------------------|------------|------------|-------------------|
| Nationals Park | 80.56% | 15.10% | 4.34% |
| Citizens Bank Park | 83.74% | 16.26% | 0.00% |
| Citi Field | 63.02% | 18.63% | 18.35% |
| Safeco Field | 79.41% | 20.59% | 0.00% |
| AT&T Park | 20.98% | 33.26% | 45.76% |
| Tropicana Field | 82.35% | 17.65% | 0.00% |
| PETCO Park | 62.79% | 37.21% | 0.00% |
| Great American Ball Park | 88.94% | 11.06% | 0.00% |
| U.S. Cellular Field | 59.51% | 13.59% | 26.90% |
| Comerica Park | 84.26% | 15.74% | 0.00% |
| Chase Field | 85.46% | 14.54% | 0.00% |
| Fenway Park | 62.82% | 37.18% | 0.00% |
| Progressive Field | 82.36% | 17.64% | 0.00% |
| Busch Stadium | 79.67% | 20.33% | 0.00% |
| Minute Maid Park | 88.64% | 11.36% | 0.00% |
| Globe Life Park in Arlington | 94.28% | 5.72% | 0.00% |
| Camden Yards | 56.89% | 20.38% | 22.72% |
| Angel Stadium of Anaheim | 69.47% | 15.09% | 15.44% |
| PNC Park | 83.82% | 16.18% | 0.00% |
| Target Field | 90.75% | 9.25% | 0.00% |
| Turner Field | 90.70% | 9.30% | 0.00% |
| Wrigley Field | 35.34% | 29.33% | 35.34% |
| Kauffman Stadium | 93.11% | 6.89% | 0.00% |
| Yankee Stadium | 64.24% | 16.19% | 19.57% |
| O.co Coliseum | 60.10% | 30.39% | 9.51% |
| Miller Park | 81.01% | 18.99% | 0.00% |
| Coors Field | 82.99% | 17.01% | 0.00% |
| Marlins Park | 0.97% | 99.03% | 0.0 Mapping Error |
| Dodger Stadium | 66.47% | 12.60% | 20.93% |

Behavioral Analysis

Rivals Compete for Guests



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Asymmetry an opportunity for Nationals?

Mapping Error

Thanks!

Dan Finkel

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<http://github.com/danfinkel>