

### Everybody Wants Venue Clusters Visualized



#### **Individuals**

- Tourists
- Newcomers
- Locals



#### **Venue Operators**

- Restaurants
- Bars
- Coffee Places



**Investors** 



Media



City Planners



**Guide Editors** 

### Data

• Content: Restaurant location data

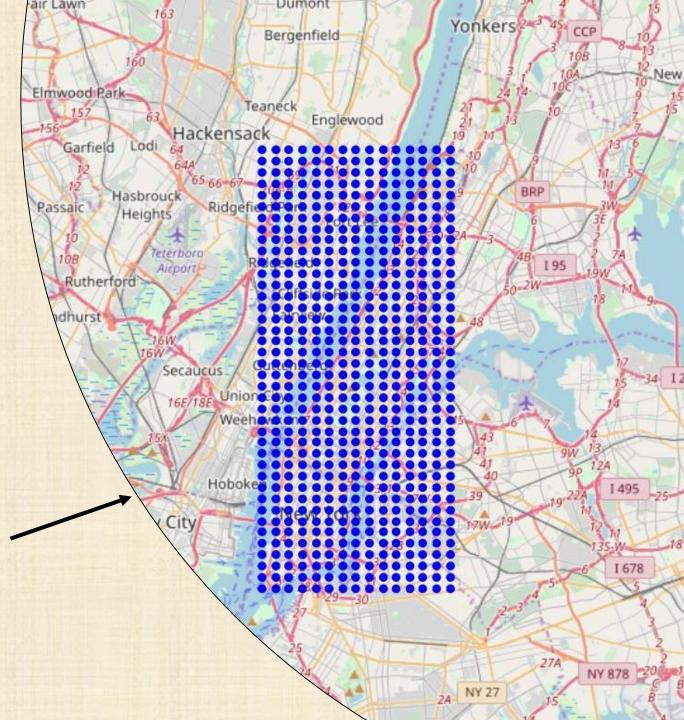
with ratings

Source: Foursquare API

Problem: Limitation to 50 venues

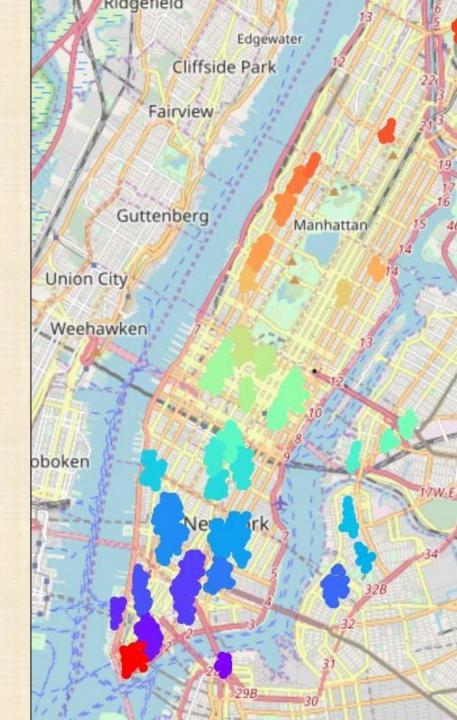
per search

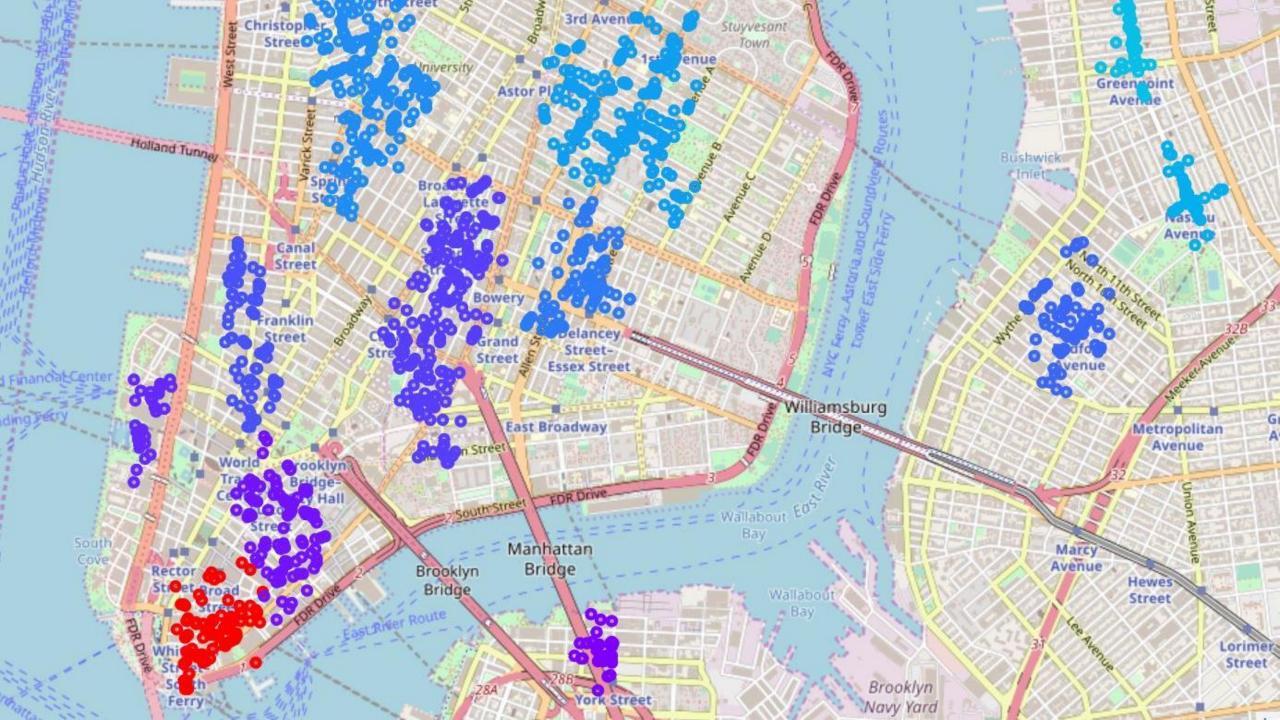
• Solution: Grid with location points



# Clustering

- Type: Density-Based Clustering
- Algorithm: DBSCAN from Scikit-learn
- Reasons: Allows arbitrary shapes
  - Identification of Outliers
  - No specification of no. of clusters
- Parameters Radius R
  - Number within R





# Rating Data

Problem: Limitation to 500 queries per day

 Solution: Sample of 12 clusters for lower Manhattan and subset thereof

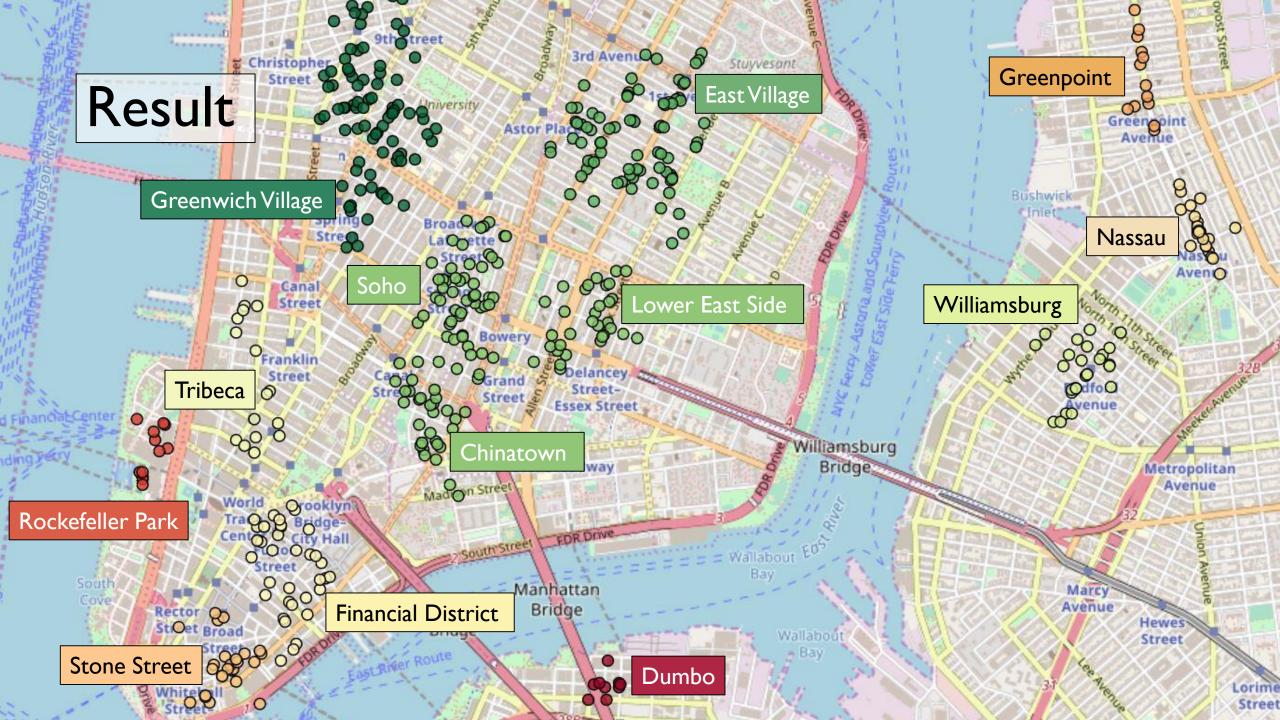
Procedure: I. Collect rating data

2. Group by clusters and take mean

3. Create cluster rating column in restaurant data frame

4. Visualize wiht RdYlGn color map

	Latitude	Longitude	Rating
Clus_Db			
0	40.703855	-74.011015	7.6
1	40.702890	-73.987510	7.0
2	40.708698	-74.007128	7.8
3	40.713981	-74.015500	7.2
4	40.719137	-73.996982	8.3
5	40.717426	-74.009285	7.9
6	40.717707	-73.958365	8.0
7	40.720156	-73.988669	8.3
8	40.730004	-74.001853	8.7
9	40.728620	-73.985708	8.4
10	40.724280	-73.951003	7.7
11	40.731437	-73.954612	7.5



### Discussion

Value:

- For tourists, individuals, businesses,
  investors, guide editors, city planners etc.
- Easily replicable for other cities and venues

- Caveats:
- More crowding to certain cluster
- Adverse effect on restaurants in bad clusters
- Exceptions not highlighted
- Cluster parameters matter
- Example: Chinatown and SoHo clustered together, but the restaurants in SoHo have better ratings (cf. right)
- Food Query possibly too general (fast food)
- Hypothesis: Size of cluster → better average rating

