

Setting up a new restaurant in New York City

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Introduction

- The Business Problem that I wish to help solve is basically in which neighborhood of New York city should a cook/business man build his new restaurant, given the fact that Manhattan is already a place filled with many many restaurants of several cultures.
- “you can’t walk a New City block without passing a restaurant”. It even states that “80 percent of restaurants fail within five years” (<https://vinepair.com/booze-news/new-york-restaurants-eat-at-every-on/>)

The Data

- Foursquare API will be the chosen API to collect the data related to the venues for each geographical point.
- The Source data for the NY City locations should have the following format (after interpreting the JSON file):

	Borough	Neighborhood	Latitude	Longitude
0	Bronx	Wakefield	40.894705	-73.847201
1	Bronx	Co-op City	40.874294	-73.829939
2	Bronx	Eastchester	40.887556	-73.827806
3	Bronx	Fieldston	40.895437	-73.905643
4	Bronx	Riverdale	40.890834	-73.912585

The Data



- The data for the venues to be analyzed should have the following format:

	Neighborhood	Neighborhood Latitude	Neighborhood Longitude	Venue	Venue Latitude	Venue Longitude	Venue Category
0	Marble Hill	40.876551	-73.91066	Arturo's	40.874412	-73.910271	Pizza Place
1	Marble Hill	40.876551	-73.91066	Bikram Yoga	40.876844	-73.906204	Yoga Studio
2	Marble Hill	40.876551	-73.91066	Tibbett Diner	40.880404	-73.908937	Diner
3	Marble Hill	40.876551	-73.91066	Starbucks	40.877531	-73.905582	Coffee Shop
4	Marble Hill	40.876551	-73.91066	Dunkin'	40.877136	-73.906666	Donut Shop

Methodology

- First and foremost, the venues of each Manhattan location described in the Data section of this report, were retrieved from Foursquare's API, with the output being a dataframe with following structure:

	Neighborhood	Neighborhood Latitude	Neighborhood Longitude	Venue	Venue Latitude	Venue Longitude	Venue Category
0	Marble Hill	40.876551	-73.91066	Arturo's	40.874412	-73.910271	Pizza Place
1	Marble Hill	40.876551	-73.91066	Bikram Yoga	40.876844	-73.906204	Yoga Studio
2	Marble Hill	40.876551	-73.91066	Tibbett Diner	40.880404	-73.908937	Diner
3	Marble Hill	40.876551	-73.91066	Starbucks	40.877531	-73.905582	Coffee Shop
4	Marble Hill	40.876551	-73.91066	Astral Fitness & Wellness Center	40.876705	-73.906372	Gym

Methodology

- Then the data would be structured in the form that will be essential for this case study and will be used for Machine Learning (ML) algorithms:

	Neighborhood	1st Most Common Venue	2nd Most Common Venue	3rd Most Common Venue	4th Most Common Venue	5th Most Common Venue	6th Most Common Venue	7th Most Common Venue	8th Most Common Venue	9th Most Common Venue	10th Most Common Venue
0	Battery Park City	Park	Hotel	Coffee Shop	Gym	Memorial Site	Playground	Plaza	Shopping Mall	Gourmet Shop	BBQ Joint
1	Carnegie Hill	Coffee Shop	Café	Pizza Place	Yoga Studio	Bookstore	Bakery	French Restaurant	Japanese Restaurant	Italian Restaurant	Bar
2	Central Harlem	African Restaurant	Seafood Restaurant	Bar	American Restaurant	Chinese Restaurant	French Restaurant	Cosmetics Shop	Market	Caribbean Restaurant	Library
3	Chelsea	Coffee Shop	Art Gallery	American Restaurant	Italian Restaurant	Bakery	Seafood Restaurant	French Restaurant	Market	Hotel	Pizza Place
4	Chinatown	Chinese Restaurant	Cocktail Bar	Bakery	Salon / Barbershop	Vietnamese Restaurant	Spa	Bubble Tea Shop	Hotpot Restaurant	Ice Cream Shop	Dessert Shop

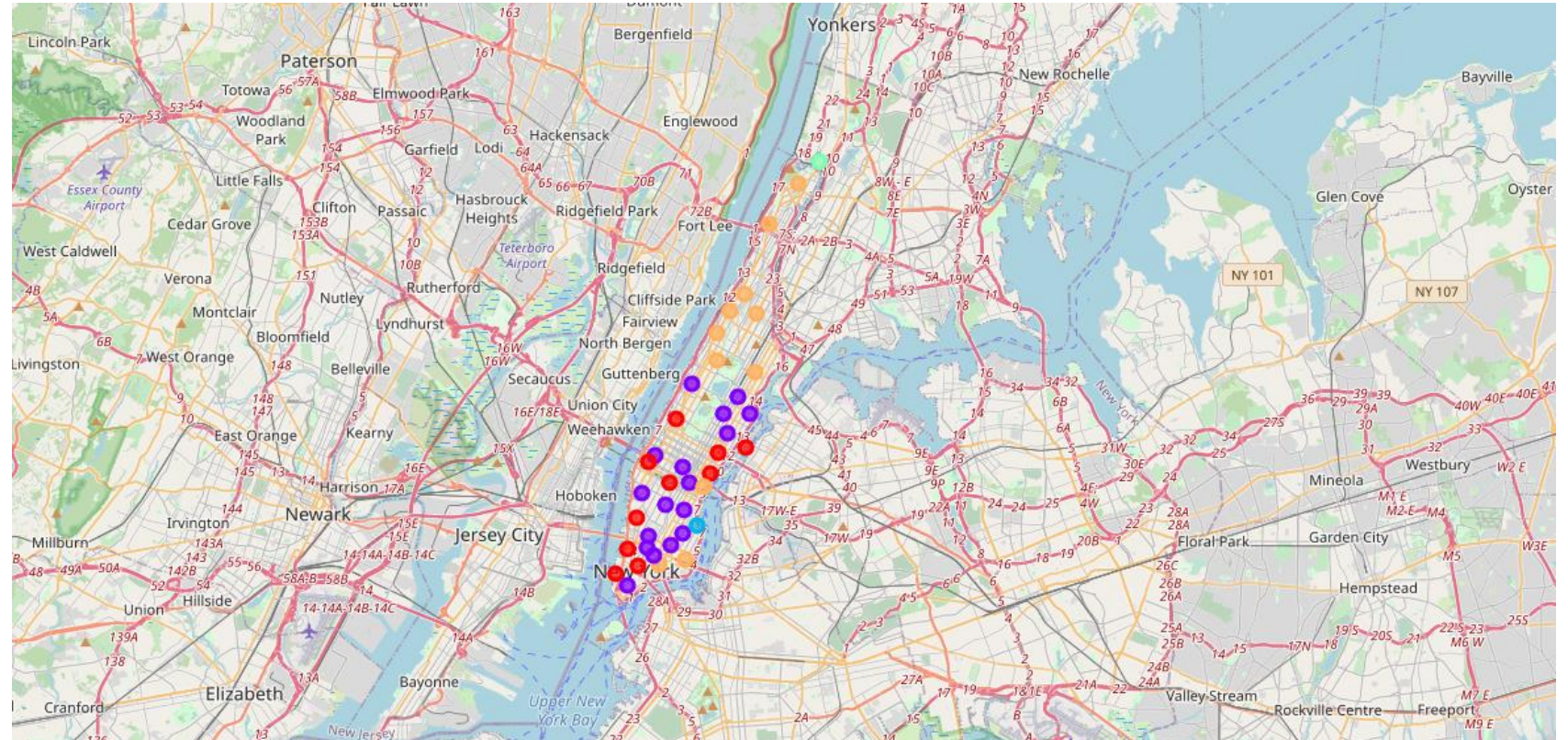
Methodology

- After using clustering to allow us to label our data:

	Borough	Neighborhood	Latitude	Longitude	Cluster Labels	1st Most Common Venue	2nd Most Common Venue	3rd Most Common Venue	4th Most Common Venue	5th Most Common Venue	6th Most Common Venue	7th Most Common Venue	8th Most Common Venue	9th Most Common Venue	10th Most Common Venue
0	Manhattan	Marble Hill	40.876551	-73.910660	3	Gym	Coffee Shop	Yoga Studio	Diner	Seafood Restaurant	Sandwich Place	Supplement Shop	Tennis Stadium	Donut Shop	Shopping Mall
1	Manhattan	Chinatown	40.715618	-73.994279	4	Chinese Restaurant	Cocktail Bar	Bakery	Salon / Barbershop	Vietnamese Restaurant	Spa	Bubble Tea Shop	Hotpot Restaurant	Ice Cream Shop	Dessert Shop
2	Manhattan	Washington Heights	40.851903	-73.936900	4	Café	Bakery	Mobile Phone Shop	Chinese Restaurant	Seafood Restaurant	Bank	Tapas Restaurant	Mexican Restaurant	Coffee Shop	Italian Restaurant
3	Manhattan	Inwood	40.867684	-73.921210	4	Mexican Restaurant	Lounge	Restaurant	Café	Frozen Yogurt Shop	Bakery	Spanish Restaurant	Chinese Restaurant	Caribbean Restaurant	American Restaurant
4	Manhattan	Hamilton Heights	40.823604	-73.949688	4	Pizza Place	Coffee Shop	Café	Deli / Bodega	Mexican Restaurant	Bakery	Park	Cocktail Bar	Sandwich Place	Chinese Restaurant

Methodology

K = 5 Clusters



Methodology

– Using a new test entry to obtain the candidate cluster

- 8 – Park
- 11 – Clothing Store
- 11 – Plaza
- 11 – Coffee Shop
- 12 – American Restaurant
- 1 – Spa
- 20 – Dog Run
- 7 – Liquor Store
- 0 – Donut Shop
- 20 - Thai Restaurant

Results

ML Techniques Accuracy Scores		
Algorithm	Jaccard	F1-score
KNN	0.625	0.643
Decision Tree	0.375	0.383
SVM	0.625	0.625
Logistic Regression	0.625	0.611

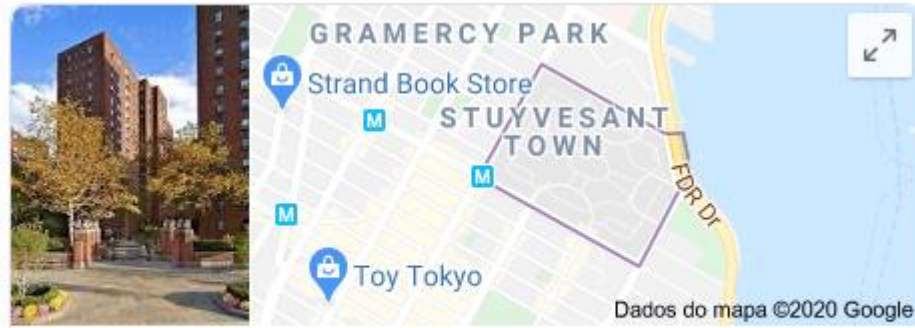
KNN was selected to check the label for our candidate cluster, thus cluster 1 was obtained!

Results

	Neighborhood	1st Most Common Venue	2nd Most Common Venue	3rd Most Common Venue	4th Most Common Venue	5th Most Common Venue	6th Most Common Venue	7th Most Common Venue	8th Most Common Venue	9th Most Common Venue	10th Most Common Venue
11	Roosevelt Island	Park	Hotel	School	Gym	Coffee Shop	Greek Restaurant	Dry Cleaner	Sandwich Place	Liquor Store	Noodle House
13	Lincoln Square	Café	Gym / Fitness Center	Plaza	Theater	Concert Hall	Performing Arts Venue	Wine Shop	American Restaurant	Italian Restaurant	Coffee Shop
21	Tribeca	Park	Italian Restaurant	Café	American Restaurant	Wine Bar	Spa	Coffee Shop	Skate Park	Hotel	Greek Restaurant
24	West Village	Italian Restaurant	New American Restaurant	American Restaurant	Park	Cocktail Bar	French Restaurant	Jazz Club	Coffee Shop	Wine Bar	Theater
28	Battery Park City	Park	Hotel	Coffee Shop	Gym	Memorial Site	Playground	Plaza	Shopping Mall	Gourmet Shop	BBQ Joint
32	Civic Center	Coffee Shop	Cocktail Bar	Hotel	Gym / Fitness Center	Spa	Yoga Studio	Café	French Restaurant	Italian Restaurant	Bakery
33	Midtown South	Korean Restaurant	Hotel	Japanese Restaurant	Burger Joint	Cosmetics Shop	Gym / Fitness Center	Clothing Store	Coffee Shop	Bakery	Pizza Place
34	Sutton Place	Italian Restaurant	Gym / Fitness Center	Furniture / Home Store	Park	Coffee Shop	Gym	Bakery	Thai Restaurant	Beer Bar	Spa
35	Turtle Bay	Coffee Shop	Sushi Restaurant	Italian Restaurant	Wine Bar	Park	Seafood Restaurant	Café	Japanese Restaurant	Deli / Bodega	French Restaurant
39	Hudson Yards	Gym / Fitness Center	Hotel	American Restaurant	Café	Italian Restaurant	Burger Joint	Dog Run	Gym	Park	Coffee Shop

Cluster 1

Discussion



Stuyvesant Town

- Cluster 3, which only has the neighborhood Stuyvesant Town, can be discarded since by just googling location we notice that this is a large private residential development, which are usually not a great ideal place to build your restaurant.
- Cluster 4 can also be discarded, but for a different reason. We only have one entry for this particular cluster which, if the model's chosen label were to be this one, would not be a result giving us great confidence in it.
- In cluster 5 restaurants are predominant, thus it isn't a good choice!

Discussion

- Cluster 2 has the following statistics:

	Neighborhood	1st Most Common Venue	2nd Most Common Venue	3rd Most Common Venue	4th Most Common Venue	5th Most Common Venue	6th Most Common Venue	7th Most Common Venue	8th Most Common Venue	9th Most Common Venue	10th Most Common Venue
count	17	17	17	17	17	17	17	17	17	17	17
unique	17	6	12	14	17	14	13	12	15	15	14
top	Little Italy	Italian Restaurant	Coffee Shop	Pizza Place	Mexican Restaurant	Mediterranean Restaurant	Bakery	French Restaurant	Mediterranean Restaurant	Italian Restaurant	Japanese Restaurant
freq	1	7	3	3	1	2	3	4	2	2	2

- Which features a great predominance of restaurants with a total number of 20 most common restaurants (not counting the Pizza Places). In contrast with cluster 1, which has the following statistics:

	Neighborhood	1st Most Common Venue	2nd Most Common Venue	3rd Most Common Venue	4th Most Common Venue	5th Most Common Venue	6th Most Common Venue	7th Most Common Venue	8th Most Common Venue	9th Most Common Venue	10th Most Common Venue
count	10	10	10	10	10	10	10	10	10	10	10
unique	10	6	6	9	8	9	10	9	9	9	9
top	West Village	Park	Hotel	American Restaurant	Park	Coffee Shop	French Restaurant	Café	Coffee Shop	Italian Restaurant	Coffee Shop
freq	1	3	4	2	2	2	1	2	2	2	2

Discussion

- Looking back at Cluster 1's dataframe:

	Neighborhood	1st Most Common Venue	2nd Most Common Venue	3rd Most Common Venue	4th Most Common Venue	5th Most Common Venue	6th Most Common Venue	7th Most Common Venue	8th Most Common Venue	9th Most Common Venue	10th Most Common Venue
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21	Tribeca	Park	Italian Restaurant	Café	American Restaurant	Wine Bar	Spa	Coffee Shop	Skate Park	Hotel	Greek Restaurant
24	West Village	Italian Restaurant	New American Restaurant	American Restaurant	Park	Cocktail Bar	French Restaurant	Jazz Club	Coffee Shop	Wine Bar	Theater
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32	Civic Center	Coffee Shop	Cocktail Bar	Hotel	Gym / Fitness Center	Spa	Yoga Studio	Café	French Restaurant	Italian Restaurant	Bakery
33	Midtown South	Korean Restaurant	Hotel	Japanese Restaurant	Burger Joint	Cosmetics Shop	Gym / Fitness Center	Clothing Store	Coffee Shop	Bakery	Pizza Place
34	Sutton Place	Italian Restaurant	Gym / Fitness Center	Furniture / Home Store	Park	Coffee Shop	Gym	Bakery	Thai Restaurant	Beer Bar	Spa
35	Turtle Bay	Coffee Shop	Sushi Restaurant	Italian Restaurant	Wine Bar	Park	Seafood Restaurant	Café	Japanese Restaurant	Deli / Bodega	French Restaurant
39	Hudson Yards	Gym / Fitness Center	Hotel	American Restaurant	Café	Italian Restaurant	Burger Joint	Dog Run	Gym	Park	Coffee Shop

- We can already discard neighborhoods Tribeca, West Village, Midtown South, Sutton Place, Turtle Bay and Hudson Yards because they have restaurants at least as their 3rd most common venue. Leaving us with just Roosevelt Island, Lincoln Square, Battery Park City and Civic Center neighborhoods.

	Neighborhood	1st Most Common Venue	2nd Most Common Venue	3rd Most Common Venue	4th Most Common Venue	5th Most Common Venue	6th Most Common Venue	7th Most Common Venue	8th Most Common Venue	9th Most Common Venue	10th Most Common Venue
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33											
34											
35											
39											

Discussion

- Then in order to choose just one neighborhood, we started to analyze from left to right until we found the first neighborhood with a restaurant as its most common venue, which is Roosevelt Island with a Greek Restaurant as its 6th most common venue.
- Again, reading from left to right, we get to the “8th Most Common Venue” column, which has, for both Lincoln Square and Civic Center a restaurant. We eliminate these entries and reach the conclusion that our best choice of a neighborhood is **Battery Park City**, featuring virtually no restaurant as its most common venue, featuring entries such as Park and Plaza which are in the single-entry dataframe that we used as test data.

Discussion

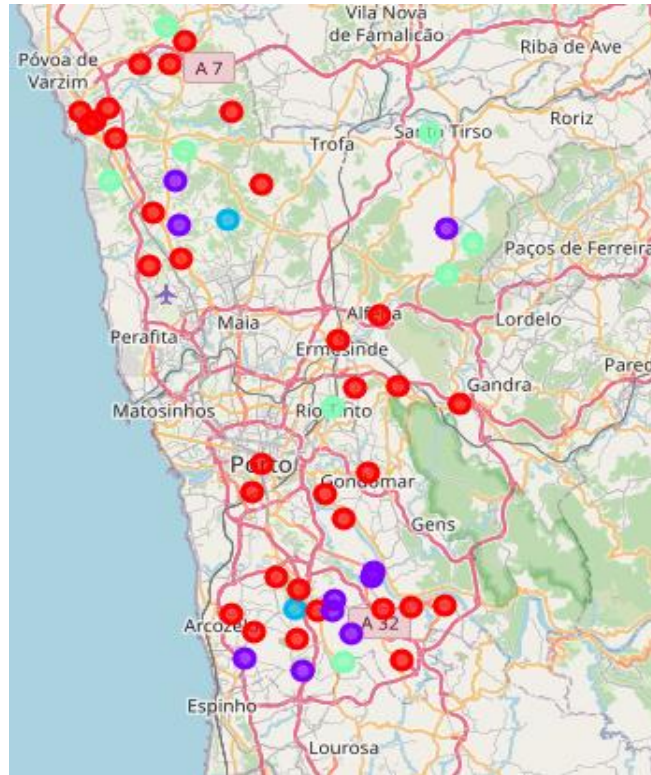


Battery Park City

Our Winner!

Discussion

- Now which type of restaurant do we build? -> Porto as the most different city overall



Portuguese Cuisine is the differentiator!



Conclusion

- Lack of geographical points to some cities (like Porto for example, in which the entire district of Porto had to be considered and not just the city itself)
- An American restaurant is in that same dataframe, but I opted for a more realistic approach, since it is difficult to assume that in New York City, a place already filled up to the top with so many restaurants (as discussed in the Introduction section), would have a neighborhood with no restaurants whatsoever.
- The test data set could be, of course, enhanced in order to feature more entries to check the validity of the model according to the results obtained in this work.
- And, also, many more cities could have been researched upon in order to place the new restaurant, however New York City, as it was shown, proved to be a great challenge.

Thank you

