

BATTLE OF THE NEIGHBORHOODS PROJECT

1. Introduction

1.1 Background:

Building a good chain of outlets for any business is a necessity in the current world to survive as a brand. Many brands are unable to sustain in the market even after good quality of products and services for just one most important reason, which is, bad placement of the outlet/branch or not expanding to the correct location at the correct time.

Correct time of expansion depends upon the brand, which usually depends on the quality of products and services as good quality of products and services gain them the necessary funding to expand. After that, it's upon the brand to invest their resources on expansion or modification of existing outlets.

Though modifications of current outlet/branch is a good step, but in most of the cases, in contrast to expansion, its effects on the profits is very less.

Correct placement of the outlet/branch in a given neighbourhood is a very important step which must be done with all the necessary background studies done as one wrong placement can result into huge loss, and thus we decided to deal with this particular problem. Our area of concern for this project will be the state of New York.

1.2 Problem Statement:

Keeping in mind the problem stated in the background study, and for a sample client in our scenario, i.e., a Pizza Place owner. Thus the problem statement can be stated as:

"To find the best locations in New York State for the expansion of a Pizza Place based in Carnegie Hill, Manhattan, NY."

1.3 Interest:

Our current client is very much interested in the project as he will get a narrowed down list of all the places where he can possibly expand, based on location.

Any other business who wish to expand their business might also be interested in this project based on the success/satisfaction of our current client.

2. Data Acquisition and Cleaning

2.1 Data Sources:

The main data source for our project is the neighbourhood JSON data found [here](#). This dataset contains all the neighbourhoods in New York State along with their Latitude and Longitude values.

We also used the [foursquare](#) API to retrieve all the nearby venues in the form of JSON data, which consisted of all the venues with their Latitude and Longitude values along with their Venue category.

2.2 Data Cleaning:

Data downloaded is in below format:

```
{'type': 'FeatureCollection',
  'totalFeatures': 306,
  'features': [{'type': 'Feature',
    'id': Place ID,
    'geometry': {'type': 'Point',
      'coordinates': [Latitude, Longitude]},
    'geometry_name': 'geom',
    'properties': {'name': 'Place Name',
      'stacked': 1,
      'annoline1': 'Place name annotation name 1',
      'annoline2': 'Place name annotation name 2',
      'annoline3': 'Place name annotation name 3',
      'annoangle': 0.0,
      'borough': 'Borough Name',
      'bbox': [Top left x,
        Top left y,
        Bottom right x,
        Bottom right y]}},...
...],
  'crs': {'type': 'name', 'properties': {'name': 'urn:ogc:def:crs:EPSG::4326'}},
  'bbox': [Entire State Top Left x,
    Entire State Top Left y,
    Entire State Bottom Right x,
    Entire State Bottom Right y]
}
```

From the above data, we need the data under “features” tag only. So we access it by calling `file['features']`, and ignore rest of the data. We can use “totalFeatures” tag as well so as to confirm we have read all the features data.

The Foursquare data obtained is in the following format:

```
{
  "response": {
    "groups": {
      "items": {
        "venue": {
          "name": "Venue Name",
          "location": {
            "lat": "latitude",
            "lng": "Longitude"
          },
          "categories": {
            {
              "name": "Category Name",
              (other tags)
            }
            {
              "name": "Category Name",
              (other tags)
            }
            ...
          }
        }
      }
    }
  }
}
```

2.3 Feature Selection:

After cleaning the data, we select features by accessing each feature using `file["features"][feature number - 1]`. We need not perform any specific operations on the neighbourhood JSON file since it is well made.

Of all the features available to us, we will be needing "coordinates", "name" and "borough" tags for our use and we may discard all the other tags since they serve no purpose to our project as per the requirements.

From the Foursquare data retrieved, the tags of our concern are as follows:

1. Name

2. Location

3. Categories, which are under the "venue" tag. We may discard all other tags as they serve no purpose to our project.

3. Exploratory Data Analysis

3.1 Initial JSON Data Analysis:

Our initial JSON data consisting of the details of all the neighbourhoods in each Borough of the state of New York. We mine that data and obtain the results that the state of New York consists of 5 Boroughs, namely, “Bronx”, “Manhattan”, “Brooklyn”, “Queens” and “Staten Island”. These boroughs consist of a number of neighbourhoods whose numbers can be identified from the following graph:

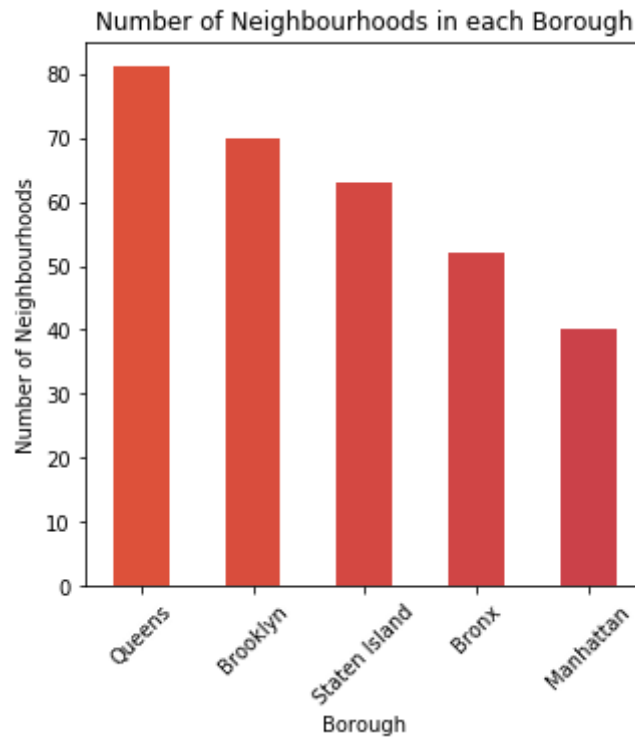


Fig. 1: Number of neighbourhoods in each Borough

From the graph, we can clearly identify that the number of neighbourhoods in each borough are as follows in descending order:

1. Queens: 81
2. Brooklyn: 70
3. Staten Island: 63
4. Bronx: 52
5. Manhattan: 40

Below map picturizes all the neighbourhoods in each separate borough:

P.T.O.

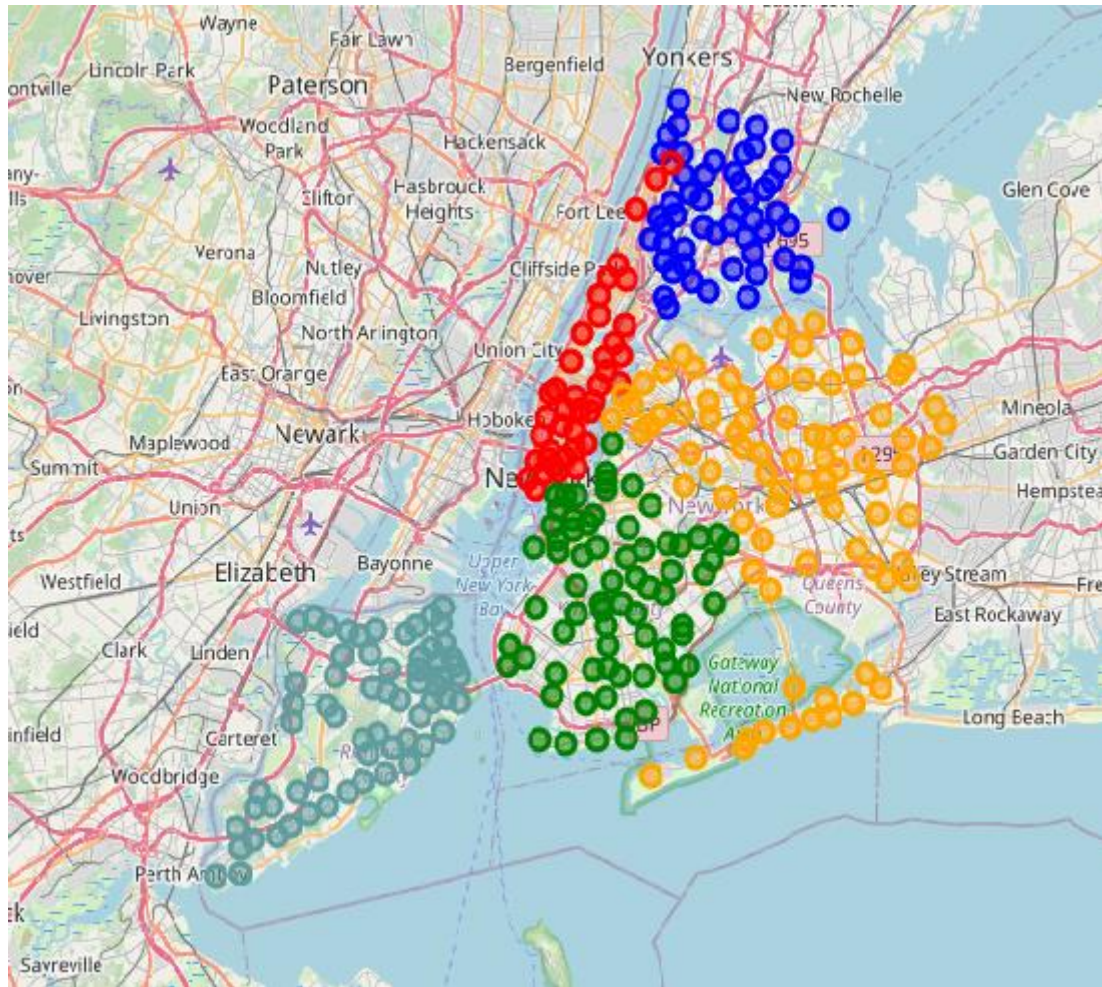


Fig. 2: Map of all Neighbourhoods in each of the 5 Boroughs

3.2 Venues in each neighbourhood:

Since we are using the foursquare API to retrieve details of all the venues in 500m range of the neighbourhood, we cover almost all major venues and in the neighbourhood. Since there are a lot of neighbourhoods in the, it is impossible for us to plot them on a single map. So, we classify them on the basis of "Venue Category" and find the top 10 most common venues. Below tables suggest top 10 most common venues in each neighbourhood (we display only 5 of the neighbourhoods in each borough so as to not cover a lot of space):

	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	Allerton	Pizza Place	Supermarket	Deli / Bodega	Donut Shop	Breakfast Spot	Pharmacy	Spa	Fast Food Restaurant	Electronics Store	Smoke Shop
1	Baychester	Donut Shop	Pet Store	Bank	Men's Store	Mattress Store	Spanish Restaurant	Fast Food Restaurant	Electronics Store	Supermarket	Fried Chicken Joint
2	Bedford Park	Diner	Pizza Place	Chinese Restaurant	Deli / Bodega	Pharmacy	Mexican Restaurant	Bus Station	Spanish Restaurant	Sandwich Place	Supermarket
3	Belmont	Italian Restaurant	Pizza Place	Deli / Bodega	Bakery	Bank	Donut Shop	Mexican Restaurant	Grocery Store	Dessert Shop	Sandwich Place
4	Bronxdale	Breakfast Spot	Bank	Chinese Restaurant	Performing Arts Venue	Park	Paper / Office Supplies Store	Eastern European Restaurant	Spanish Restaurant	Mexican Restaurant	Deli / Bodega

Table 1: Top 10 most common venues in neighbourhoods of Bronx

	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	Coffee Shop	Hotel	Gym	Memorial Site	Boat or Ferry	Plaza	Playground	Shopping Mall	Mexican Restaurant
1	Carnegie Hill	Coffee Shop	Pizza Place	Café	Yoga Studio	Wine Shop	Bookstore	Gym	Bar	Japanese Restaurant	Shipping Store
2	Central Harlem	African Restaurant	Chinese Restaurant	Seafood Restaurant	Cosmetics Shop	French Restaurant	Bar	American Restaurant	Food Truck	Beer Bar	Gym / Fitness Center
3	Chelsea	Art Gallery	Coffee Shop	Café	Bakery	Ice Cream Shop	American Restaurant	Italian Restaurant	Seafood Restaurant	Market	Park
4	Chinatown	Chinese Restaurant	Bakery	Bubble Tea Shop	Cocktail Bar	Coffee Shop	Ice Cream Shop	Salon / Barbershop	American Restaurant	Optical Shop	Vietnamese Restaurant

Table 2: Top 10 most common venues in neighbourhoods of Manhattan

	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	Bath Beach	Pharmacy	Pizza Place	Chinese Restaurant	Gas Station	Sushi Restaurant	Italian Restaurant	Asian Restaurant	Bubble Tea Shop	Fast Food Restaurant	Burger Joint
1	Bay Ridge	Spa	Pizza Place	Italian Restaurant	Greek Restaurant	Gym / Fitness Center	Bar	American Restaurant	Middle Eastern Restaurant	Hookah Bar	Ice Cream Shop
2	Bedford Stuyvesant	Pizza Place	Coffee Shop	Café	Bar	Tiki Bar	Deli / Bodega	Discount Store	Park	Cocktail Bar	New American Restaurant
3	Bensonhurst	Sushi Restaurant	Dessert Shop	Ice Cream Shop	Flower Shop	Italian Restaurant	Donut Shop	Bakery	Noodle House	Liquor Store	Cha Chaan Teng
4	Bergen Beach	Harbor / Marina	Donut Shop	Baseball Field	Playground	Athletics & Sports	Fish & Chips Shop	Farm	Farmers Market	Fast Food Restaurant	Field

Table 3: Top 10 most common venues in neighbourhoods of Brooklyn

	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	Arverne	Surf Spot	Sandwich Place	Metro Station	Board Shop	Bus Stop	Thai Restaurant	Bed & Breakfast	Donut Shop	Coffee Shop	Beach
1	Astoria	Bar	Middle Eastern Restaurant	Mediterranean Restaurant	Greek Restaurant	Hookah Bar	Indian Restaurant	Deli / Bodega	Bakery	Seafood Restaurant	Café
2	Astoria Heights	Playground	Italian Restaurant	Plaza	Bus Station	Bowling Alley	Supermarket	Laundromat	Bakery	Cocktail Bar	Burger Joint
3	Auburndale	Bar	Discount Store	Fast Food Restaurant	Furniture / Home Store	Supermarket	Noodle House	Sushi Restaurant	Korean Restaurant	Miscellaneous Shop	Italian Restaurant
4	Bay Terrace	Clothing Store	Cosmetics Shop	Lingerie Store	Mobile Phone Shop	Shoe Store	Donut Shop	Women's Store	American Restaurant	Kids Store	Movie Theater

Table 4: Top 10 most common venues in neighbourhoods of Queens

	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	Annadale	Park	Sports Bar	Pizza Place	Train Station	Restaurant	Diner	Pharmacy	Food	Eastern European Restaurant	Fish & Chips Shop
1	Arden Heights	Coffee Shop	Pizza Place	Pharmacy	Deli / Bodega	Yoga Studio	Falafel Restaurant	Food & Drink Shop	Food	Flower Shop	Fish & Chips Shop
2	Arlington	Bus Stop	Deli / Bodega	Boat or Ferry	Grocery Store	Frame Store	Food Truck	Food & Drink Shop	Food	Flower Shop	Fish & Chips Shop
3	Arrochar	Bus Stop	Deli / Bodega	Italian Restaurant	Polish Restaurant	Bagel Shop	Middle Eastern Restaurant	Supermarket	Mediterranean Restaurant	Sandwich Place	Outdoors & Recreation
4	Bay Terrace	Italian Restaurant	Supermarket	Donut Shop	Shipping Store	Home Service	Playground	Train Station	Sushi Restaurant	Salon / Barbershop	Flower Shop

Table 5: Top 10 most common venues in neighbourhoods of Staten Island

4. Predictive Procedure

4.1 Narrowing down based on top common venues in a Neighbourhood:

First we collect all the neighbourhoods where our venue of concern, i.e., “Pizza Place” doesn't come under top 10 most common place so as to make sure our client doesn't place their new outlet in a region where their product is already very common. New table of suggested places is as follows (only 5 neighbourhoods are displayed so as to save space):

	Borough	Neighborhood	Latitude	Longitude
0	Bronx	Baychester	40.866858	-73.835798
1	Bronx	Bronxdale	40.852723	-73.861726
2	Bronx	City Island	40.847247	-73.786488
3	Bronx	Clason Point	40.806551	-73.854144
4	Bronx	Country Club	40.844246	-73.824099

Table 6: Neighbourhoods where Pizza Place is not in top 10 most common venues

Following graph visualizes all the narrowed down areas:

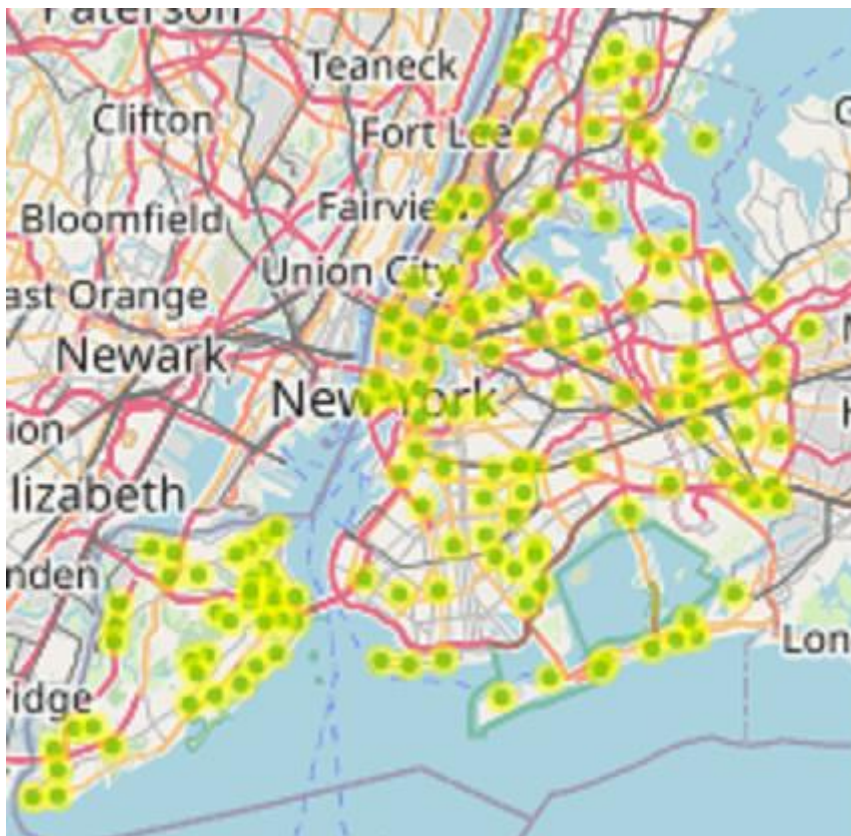


Fig. 3: Map of all suggested Neighbourhoods

In order to find where all the pizza places are present with respect to our identified neighbourhoods, we plot a map with all the Pizza Places, along with all our identified neighbourhoods as given below:

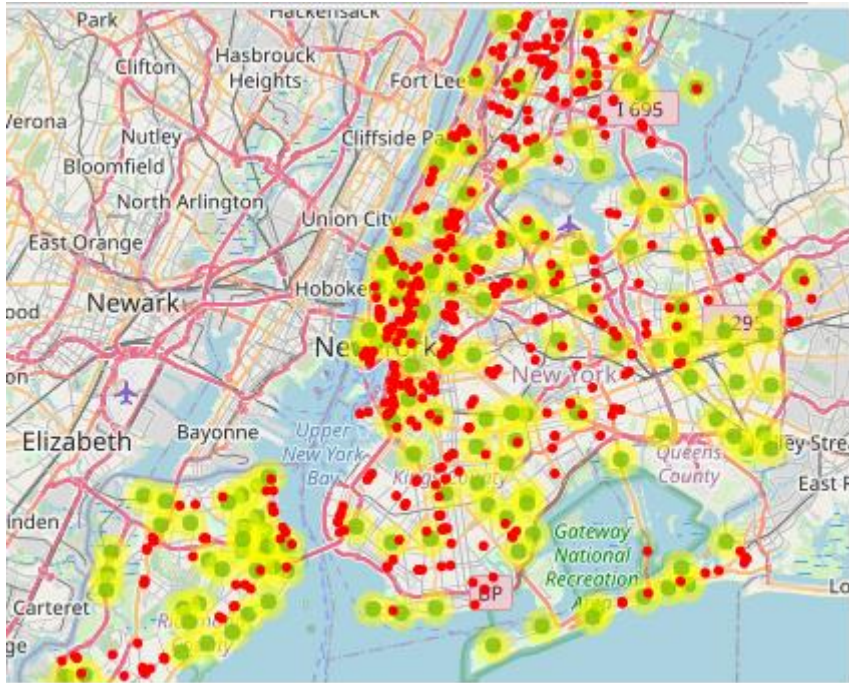


Fig. 4: Map of all suggested Neighbourhoods with all Pizza Places

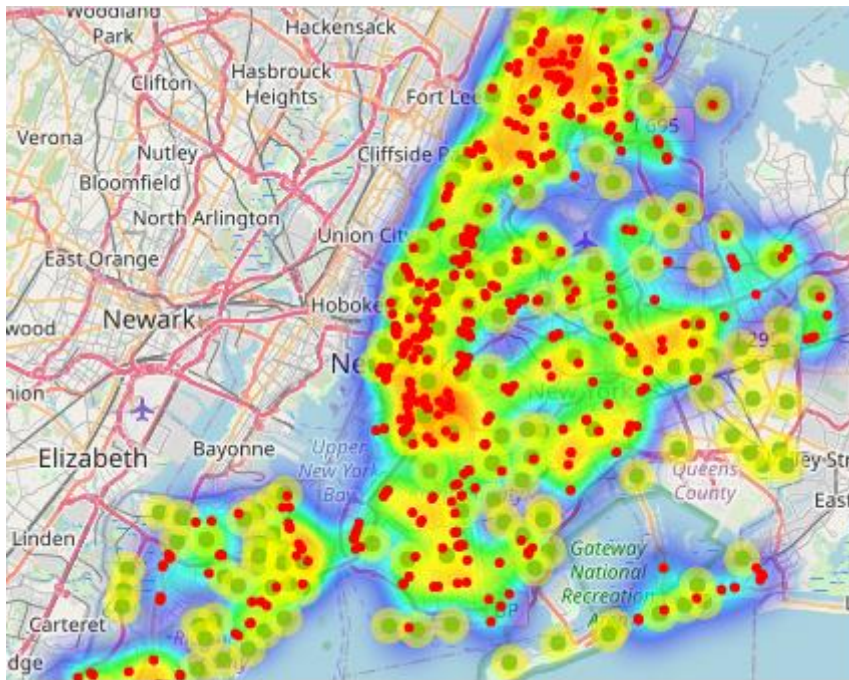


Fig. 4: HeatMap of all suggested Neighbourhoods with all Pizza Places

4.2 Narrowing down based number of competitors in 1Km radius:

Now we will narrow down the list of suggested places further by finding all the competitors of our client within 1Km radius using Foursquare API again. Based on the number of competitors, we divide the neighbourhoods into 3 categories, namely: "Best", "Moderate" and "Worst". This is done based on the number of competitors in 1Km radius.

1. Best: 0 or 1 competitors
2. Moderate: 2-4 competitors
3. Worst: 5 or more competitors

Following table suggests all the neighbourhoods categorized into the stated categories:

	Borough	Neighborhood	Latitude	Longitude
0	Bronx	City Island	40.847247	-73.786488
1	Bronx	Clason Point	40.806551	-73.854144
2	Bronx	Soundview	40.821012	-73.865746
3	Bronx	Wakefield	40.894705	-73.847201
4	Manhattan	Chelsea	40.744035	-74.003116

Table 7: Best suggested neighbourhoods

	Borough	Neighborhood	Latitude	Longitude
0	Bronx	Baychester	40.866858	-73.835798
1	Bronx	Country Club	40.844246	-73.824099
2	Bronx	Eastchester	40.887556	-73.827806
3	Bronx	Edenwald	40.884561	-73.848083
4	Bronx	Fieldston	40.895437	-73.905643

Table 8: Moderate suggested neighbourhoods

	Borough	Neighborhood	Latitude	Longitude
0	Bronx	Bronxdale	40.852723	-73.861726
1	Bronx	Mount Hope	40.848842	-73.908299
2	Bronx	Williamsbridge	40.881039	-73.857446
3	Manhattan	East Harlem	40.792249	-73.944182
4	Manhattan	Stuyvesant Town	40.731000	-73.974052

Table 9: Worst suggested neighbourhoods

Then we plot all the regions on the map as given below:

P.T.O.

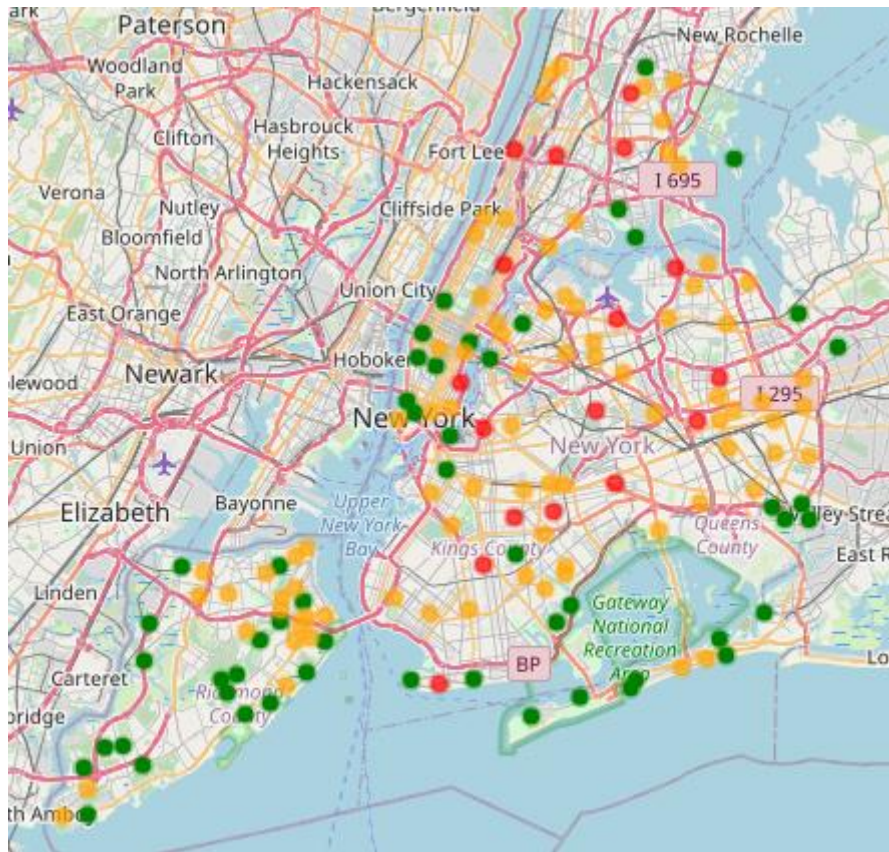


Fig. 5: Best (in green), Moderate (in orange) and Worst (in red) suggested neighbourhoods

4.3 Narrowing down based on clusters of Best Suggested Neighbourhoods:

Now we use DBSCAN clustering algorithm to cluster the Best Suggested Neighbourhoods with at least 3 neighbourhoods per cluster with epsilon value of 0.005. Upon using the algorithm, we find the centres and radii of each cluster and plot them on the map as given below:



Fig. 6: Clusters of best suggested neighbourhoods

We also plot all the neighbourhoods which qualify as “Best” neighbourhoods for our client to open a new branch/outlet as given below:

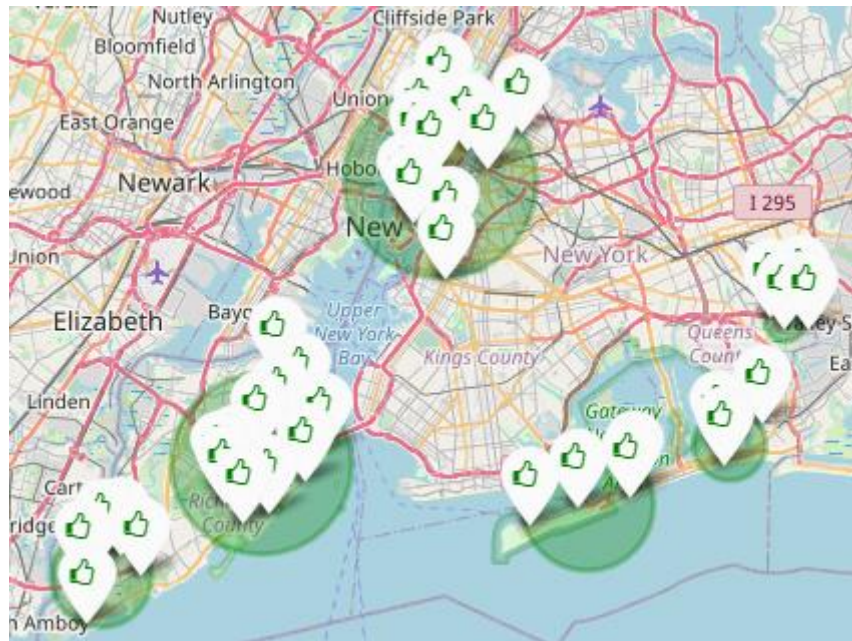


Fig. 7: Clusters of best suggested neighbourhoods along with their locations

The biggest cluster among all these clusters suggests that there are a large number of neighbourhoods in close proximity where there are very few “Pizza Places” compared to other neighbourhoods and thus will be best suitable for our client to open a new branch/outlet, thus increasing their chances of success by a great factor.

Below table includes a list of all the neighbourhoods which come under the biggest cluster of “Best” suggested places:

	Borough	Neighborhood	Latitude	Longitude
0	Manhattan	Chelsea	40.744035	-74.003116
1	Manhattan	Civic Center	40.715229	-74.005415
2	Manhattan	Flatiron	40.739673	-73.990947
3	Manhattan	Hudson Yards	40.756658	-74.000111
4	Manhattan	Lincoln Square	40.773529	-73.985338
5	Manhattan	Tribeca	40.721522	-74.010683
6	Manhattan	Turtle Bay	40.752042	-73.967708
7	Brooklyn	Boerum Hill	40.685683	-73.983748
8	Brooklyn	Vinegar Hill	40.703321	-73.981116
9	Queens	Hunters Point	40.743414	-73.953868
10	Queens	Ravenswood	40.761705	-73.931575

Table 10: Neighbourhoods which come under the biggest cluster of “Best” suggested Neighbourhoods

5. Conclusions

From all the above data processing and analysis, we can come to a final conclusion that following neighbourhoods are the best for our client to carry out further research on the likings and average money spent on eating out, by the residents, to further narrow down their new outlet location:

Neighborhood	Borough
Chelsea	Manhattan
Civic Center	Manhattan
Flatiron	Manhattan
Hudson Yards	Manhattan
Lincoln Square	Manhattan
Tribeca	Manhattan
Turtle Bay	Manhattan
Boerum Hill	Brooklyn
Vinegar Hill	Brooklyn
Hunters Point	Queens
Ravenswood	Queens

Table 11: Best suggested Neighbourhoods

Client may also want to look into following neighbourhoods, although they do not make a huge cluster and thus may not be as beneficial for our client to invest their resources on:

Neighborhood	Borough
City Island	Bronx
Clason Point	Bronx
Soundview	Bronx
Wakefield	Bronx
Bergen Beach	Brooklyn
Brighton Beach	Brooklyn
East Flatbush	Brooklyn
Mill Island	Brooklyn
Sea Gate	Brooklyn
Arverne	Queens
Bayswater	Queens
Belle Harbor	Queens
Breezy Point	Queens
Brookville	Queens
Douglaston	Queens
Glen Oaks	Queens
Laurelton	Queens
Neponsit	Queens
Rosedale	Queens
Roxbury	Queens
Somerville	Queens
Springfield Gardens	Queens
Arlington	Staten Island
Arrochar	Staten Island
Bloomfield	Staten Island
Butler Manor	Staten Island
Charleston	Staten Island
Egbertville	Staten Island
Emerson Hill	Staten Island
Fox Hills	Staten Island
Huguenot	Staten Island
Lighthouse Hill	Staten Island
New Dorp Beach	Staten Island
Oakwood	Staten Island
Randall Manor	Staten Island
Richmond Town	Staten Island
Sandy Ground	Staten Island
South Beach	Staten Island
Todt Hill	Staten Island
Travis	Staten Island
Woodrow	Staten Island

Table 12: Other suggested Neighbourhoods

6. Future Directions:

Based on our current research and methodology, we were only able to suggest our client a possible neighbourhood based on location. Later on for further improvements, we may also want to look into the average salaries and average spending of people living in our target neighbourhoods so as to better analyse and further narrow down the list of suggested neighbourhoods. We may also consider the general likes and dislikes of people as they also play an important factor in deciding whether any business will be successful in the region or not.