

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:

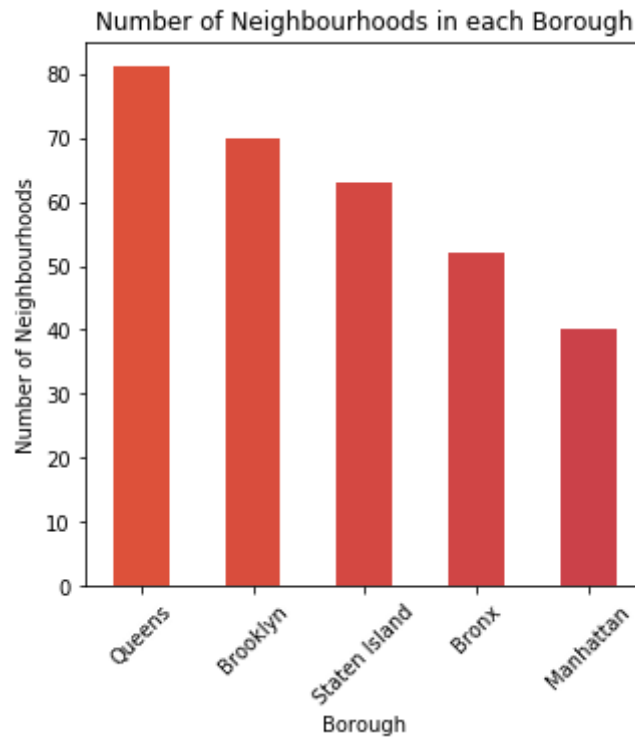


Fig. 1: Number of neighbourhoods in each Borough

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:

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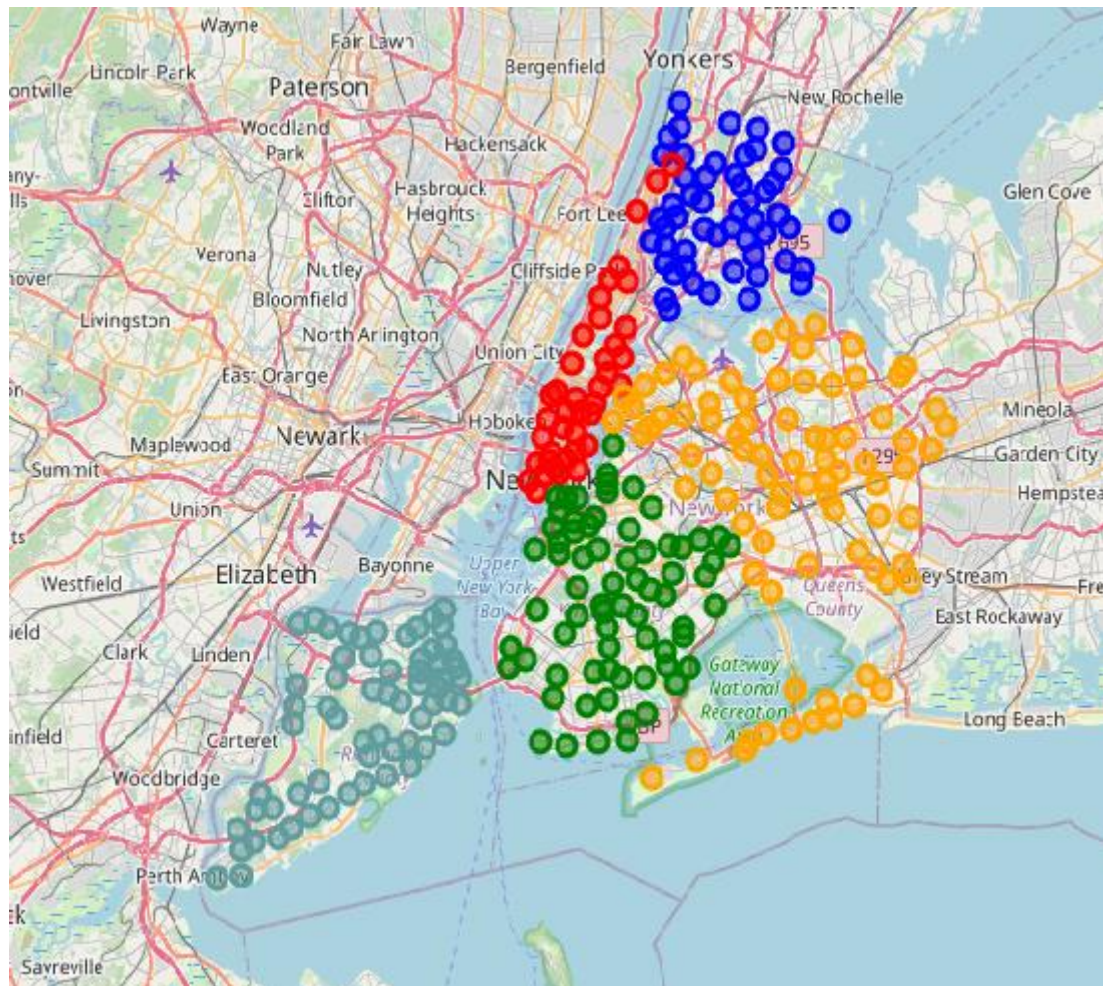


Fig. 2: Map of all Neighbourhoods in each of the 5 Boroughs

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	Chinese Restaurant	Bus Station	Fast Food Restaurant	Martial Arts Dojo	Spanish Restaurant	Pharmacy	Food
1	Baychester	Electronics Store	Donut Shop	Bank	Baseball Field	Bus Station	Mattress Store	Mexican Restaurant	Pet Store	Spanish Restaurant	Fast Food Restaurant
2	Bedford Park	Diner	Chinese Restaurant	Deli / Bodega	Pizza Place	Mexican Restaurant	Spanish Restaurant	Sandwich Place	Supermarket	Fried Chicken Joint	Bus Station
3	Belmont	Italian Restaurant	Pizza Place	Deli / Bodega	Bakery	Donut Shop	Bank	Dessert Shop	Grocery Store	Bar	Mexican Restaurant
4	Bronxdale	Mexican Restaurant	Spanish Restaurant	Pizza Place	Performing Arts Venue	Breakfast Spot	Italian Restaurant	Gym	Paper / Office Supplies Store	Chinese Restaurant	Bank

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	Memorial Site	Gym	Wine Shop	Shopping Mall	Gourmet Shop	Plaza	Boat or Ferry
1	Carnegie Hill	Coffee Shop	Pizza Place	Café	Yoga Studio	Gym	Wine Shop	Bar	Bookstore	Japanese Restaurant	Grocery Store
2	Central Harlem	African Restaurant	Chinese Restaurant	French Restaurant	American Restaurant	Bar	Cosmetics Shop	Fried Chicken Joint	Seafood Restaurant	Food Truck	Market
3	Chelsea	Art Gallery	Coffee Shop	Café	Bakery	Ice Cream Shop	American Restaurant	Italian Restaurant	Theater	Seafood Restaurant	Hotel
4	Chinatown	Chinese Restaurant	Optical Shop	Bakery	Cocktail Bar	Bubble Tea Shop	Salon / Barbershop	Spa	Ice Cream Shop	American Restaurant	Coffee Shop

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	Chinese Restaurant	Sushi Restaurant	Pizza Place	Italian Restaurant	Bubble Tea Shop	Gas Station	Fast Food Restaurant	Ice Cream Shop	Sandwich Place
1	Bay Ridge	Spa	Italian Restaurant	Pizza Place	Greek Restaurant	American Restaurant	Bar	Thai Restaurant	Ice Cream Shop	Playground	Pharmacy
2	Bedford Stuyvesant	Bus Stop	Café	Pizza Place	Coffee Shop	Bar	Discount Store	Cocktail Bar	Thrift / Vintage Store	Gourmet Shop	Basketball Court
3	Bensonhurst	Chinese Restaurant	Ice Cream Shop	Sushi Restaurant	Italian Restaurant	Bakery	Donut Shop	Pizza Place	Noodle House	Cosmetics Shop	Butcher
4	Bergen Beach	Harbor / Marina	Donut Shop	Baseball Field	Playground	Athletics & Sports	Filipino Restaurant	Falafel Restaurant	Farm	Farmers Market	Fast Food Restaurant

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	Metro Station	Sandwich Place	Playground	Wine Shop	Bed & Breakfast	Pizza Place	Thai Restaurant	Bus Stop	Beach
1	Astoria	Middle Eastern Restaurant	Bar	Indian Restaurant	Hookah Bar	Mediterranean Restaurant	Greek Restaurant	Bakery	Deli / Bodega	Café	Seafood Restaurant
2	Astoria Heights	Burger Joint	Italian Restaurant	Bakery	Supermarket	Bus Station	Pizza Place	Bowling Alley	Hostel	Playground	Plaza
3	Auburndale	Mobile Phone Shop	Italian Restaurant	Supermarket	Korean Restaurant	Bar	Fast Food Restaurant	Furniture / Home Store	Toy / Game Store	Noodle House	Athletics & Sports
4	Bay Terrace	Clothing Store	Women's Store	Mobile Phone Shop	American Restaurant	Lingerie Store	Cosmetics Shop	Donut Shop	Kids Store	Shoe Store	Gift Shop

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	Pizza Place	Train Station	Sports Bar	Bakery	Liquor Store	Restaurant	Diner	Dance Studio	Pub	American Restaurant
1	Arden Heights	Pharmacy	Bus Stop	Lawyer	Coffee Shop	Pizza Place	Falafel Restaurant	Food & Drink Shop	Food	Flower Shop	Fish & Chips Shop
2	Arlington	Intersection	Grocery Store	Coffee Shop	Bus Stop	Boat or Ferry	Deli / Bodega	American Restaurant	Hotel	Falafel Restaurant	Food Truck
3	Arrochar	Bus Stop	Italian Restaurant	Deli / Bodega	Supermarket	Mediterranean Restaurant	Middle Eastern Restaurant	Food Truck	Outdoors & Recreation	Pizza Place	Polish Restaurant
4	Bay Terrace	Supermarket	Insurance Office	Train Station	Sushi Restaurant	Donut Shop	Salon / Barbershop	Shipping Store	Diner	Farmers Market	Food

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	City Island	40.847247	-73.786488
2	Bronx	Clason Point	40.806551	-73.854144
3	Bronx	Co-op City	40.874294	-73.829939
4	Bronx	Concourse Village	40.824780	-73.915847

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

Following graph visualizes all the narrowed down areas:

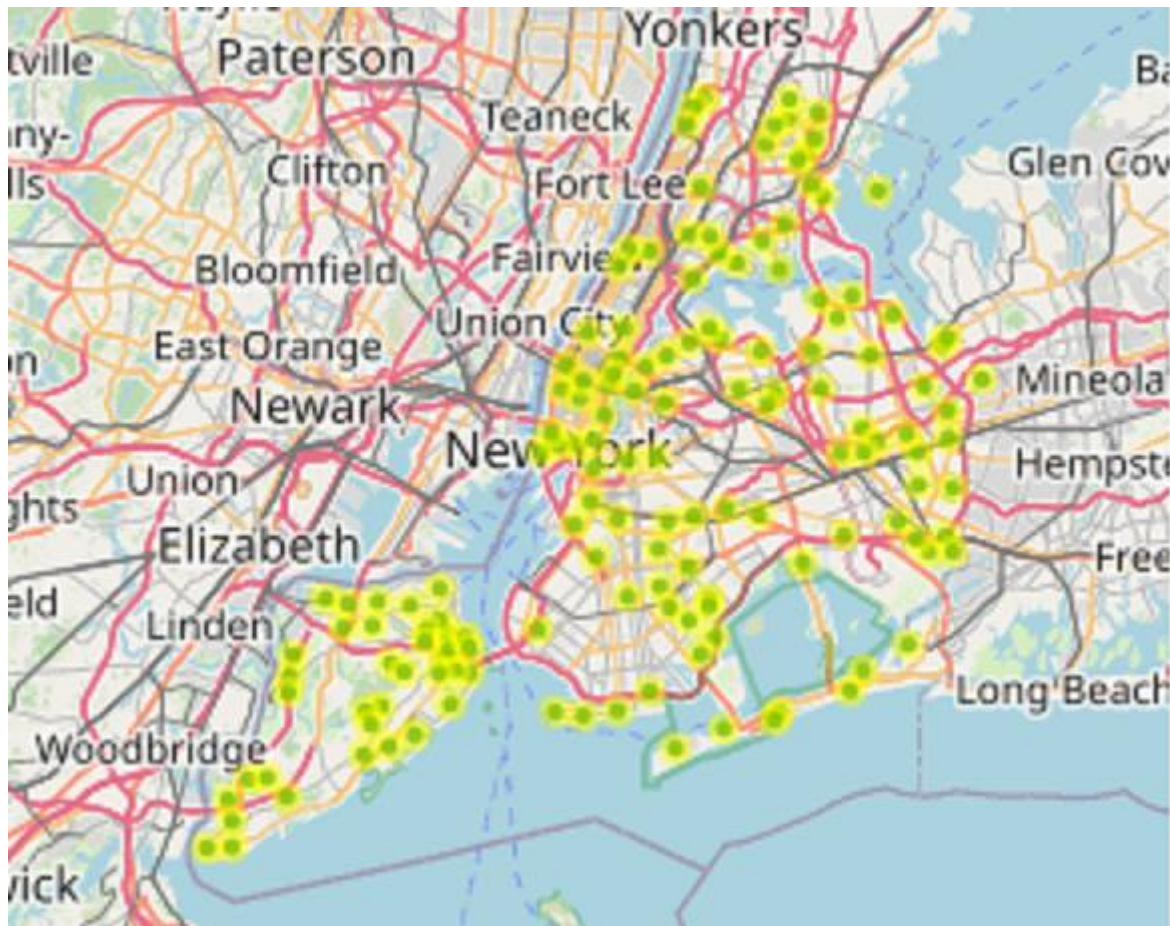


Fig. 3: Map of all suggested Neighbourhoods

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:

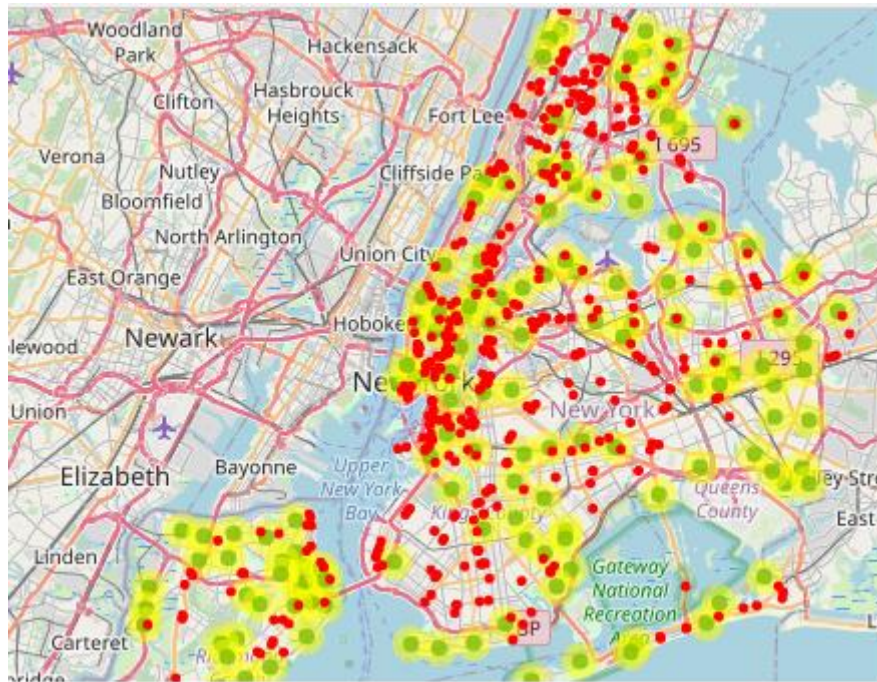


Fig. 4: Map of all suggested Neighbourhoods with all Pizza Places



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	Hunts Point	40.809730	-73.883315
3	Bronx	Soundview	40.821012	-73.865746
4	Bronx	Wakefield	40.894705	-73.847201

Table 7: Best suggested neighbourhoods

	Borough	Neighborhood	Latitude	Longitude
0	Bronx	Baychester	40.866858	-73.835798
1	Bronx	Co-op City	40.874294	-73.829939
2	Bronx	Concourse Village	40.824780	-73.915847
3	Bronx	Country Club	40.844246	-73.824099
4	Bronx	Eastchester	40.887556	-73.827806

Table 8: Moderate suggested neighbourhoods

	Borough	Neighborhood	Latitude	Longitude
0	Bronx	Fieldston	40.895437	-73.905643
1	Bronx	Mount Hope	40.848842	-73.908299
2	Bronx	Olinville	40.871371	-73.863324
3	Bronx	Unionport	40.829774	-73.850535
4	Bronx	Williamsbridge	40.881039	-73.857446

Table 9: Worst suggested neighbourhoods

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

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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:

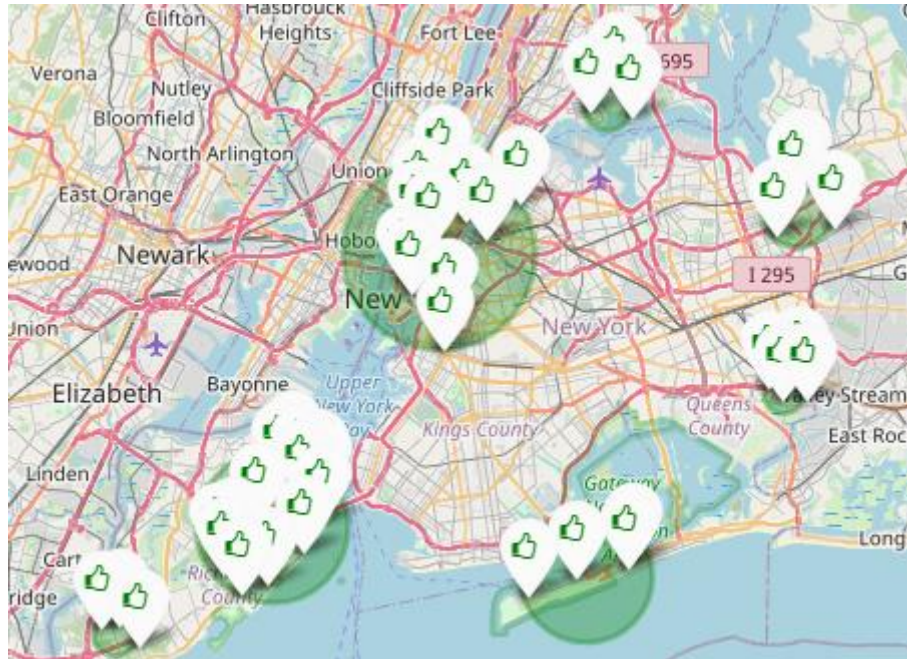


Fig. 7: Clusters of best suggested neighbourhoods along with their locations

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	Staten Island	Arrochar	40.596313	-74.067124
1	Staten Island	Egbertville	40.579119	-74.127272
2	Staten Island	Fox Hills	40.617311	-74.081740
3	Staten Island	Grymes Hill	40.624185	-74.087248
4	Staten Island	Lighthouse Hill	40.576506	-74.137927
5	Staten Island	New Dorp Beach	40.564255	-74.104327
6	Staten Island	Oakwood	40.558462	-74.121566
7	Staten Island	Park Hill	40.609190	-74.080157
8	Staten Island	Richmond Town	40.569606	-74.134057
9	Staten Island	Rosebank	40.615305	-74.069805
10	Staten Island	Shore Acres	40.609719	-74.066678
11	Staten Island	Silver Lake	40.619193	-74.096290
12	Staten Island	South Beach	40.580247	-74.079553
13	Staten Island	Todt Hill	40.597069	-74.111329

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
Arrochar	Staten Island
Egbertville	Staten Island
Fox Hills	Staten Island
Grymes Hill	Staten Island
Lighthouse Hill	Staten Island
New Dorp Beach	Staten Island
Oakwood	Staten Island
Park Hill	Staten Island
Richmond Town	Staten Island
Rosebank	Staten Island
Shore Acres	Staten Island
Silver Lake	Staten Island
South Beach	Staten Island
Todt Hill	Staten Island

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
Hunts Point	Bronx
Soundview	Bronx
Wakefield	Bronx
Chelsea	Manhattan
Civic Center	Manhattan
Flatiron	Manhattan
Hudson Yards	Manhattan
Lincoln Square	Manhattan
Tribeca	Manhattan
Turtle Bay	Manhattan
Bergen Beach	Brooklyn
Boerum Hill	Brooklyn
Brighton Beach	Brooklyn
East Flatbush	Brooklyn
Sea Gate	Brooklyn
Vinegar Hill	Brooklyn
Bayswater	Queens
Belle Harbor	Queens
Breezy Point	Queens
Brookville	Queens
Douglaston	Queens
Glen Oaks	Queens
Hunters Point	Queens
Laurelton	Queens
Little Neck	Queens
Neponsit	Queens
Oakland Gardens	Queens
Queensboro Hill	Queens
Ravenswood	Queens
Rosedale	Queens
Roxbury	Queens
Somerville	Queens
Springfield Gardens	Queens
Arlington	Staten Island
Bloomfield	Staten Island
Butler Manor	Staten Island
Huguenot	Staten Island
Sandy Ground	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.