

# Finding the optimal location for opening a Restaurant Supply Store in New York

PRANAV NARAYANAN

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## 1. Introduction

The City of New York is famous for its excellent cuisine. Its food culture includes an array of international cuisines influenced by the city's immigrant history.

Due to this influence, many restaurants are operational in the city catering to a wide variety of audience. This also means that there is a huge demand for the supply of items needed to run these restaurants.

### 1.1 Problem

In this analysis, we will discuss the optimal place to start a restaurant supply chain shop which can cater to maximum restaurants and reduce the operational costs incurred in running the shop.

### 1.2 Target Audience

The target audience for this analysis would be any person or company who is interested in opening a restaurant supply store in New York.

After the analysis they would obtain the optimum position to open the store which will minimise the cost of transport as well as generate sufficient income due to its prime location.

## 2. Data Acquisition and Cleaning

### 2.1 Data Sources

Data related to the city of New York such as the Borough and Neighbourhood information can be obtained from [here](#).

## 2.2 Data Cleaning

The data obtained from above link contains lot of information out of which we are particularly interested in details such as neighbourhood, boroughs, the latitude and longitude for each neighbourhood.

## 3. Exploratory Data Analysis

### 3.1 Obtaining the list of venues related to food for each neighbourhood using Foursquare API

We use the Foursquare API to retrieve the list of venues upto a maximum of 100 venues for each neighbourhood. Since we would like to restrict the venues to only food related ones, we pass a parameter called as “CategoryId” within the request to Foursquare API.

Once the relevant venues are retrieved from Foursquare API, the dataset looks something like this -

|   | Neighborhood | Neighborhood Latitude | Neighborhood Longitude | Venue                                   | Venue Latitude | Venue Longitude | Venue Category       |
|---|--------------|-----------------------|------------------------|---|----------------|-----------------|----------------------|
| 0 | Wakefield    | 40.894705             | -73.847201             | Lollipops Gelato                        | 40.894123      | -73.845892      | Dessert Shop         |
| 1 | Wakefield    | 40.894705             | -73.847201             | Margaritaville Restaurant and Lounge    | 40.898564      | -73.837016      | Caribbean Restaurant |
| 2 | Wakefield    | 40.894705             | -73.847201             | Cooler Runnings Jamaican Restaurant Inc | 40.898083      | -73.850259      | Caribbean Restaurant |
| 3 | Wakefield    | 40.894705             | -73.847201             | Essa Deli                               | 40.892181      | -73.854427      | Deli / Bodega        |
| 4 | Wakefield    | 40.894705             | -73.847201             | Big Daddy's Caribbean Taste Restaurant  | 40.899767      | -73.857135      | Caribbean Restaurant |

### 3.2 Data Wrangling

Once the data has been obtained from Foursquare API, we apply multiple data wrangling techniques such as one-hot encoding and grouping to obtain insights regarding the data.

Below is the result of one-hot encoding-

|   | Neighborhood | Afghan Restaurant | African Restaurant | American Restaurant | Arcade | Arepa Restaurant | Argentinian Restaurant | Art Gallery | Asian Restaurant | Australian Restaurant | Austrian Restaurant | BBQ Joint | Bagel Shop | Bakery | Bangladeshi Restaurant | Bar | Bath House | Beach | Beach Bar | Beer Bar |
|---|--------------|-------------------|--------------------|---------------------|--------|------------------|------------------------|-------------|------------------|-----------------------|---------------------|-----------|------------|--------|------------------------|-----|------------|-------|-----------|----------|
| 0 | Wakefield    | 0                 | 0                  | 0                   | 0      | 0                | 0                      | 0           | 0                | 0                     | 0                   | 0         | 0          | 0      | 0                      | 0   | 0          | 0     | 0         | 0        |
| 1 | Wakefield    | 0                 | 0                  | 0                   | 0      | 0                | 0                      | 0           | 0                | 0                     | 0                   | 0         | 0          | 0      | 0                      | 0   | 0          | 0     | 0         | 0        |
| 2 | Wakefield    | 0                 | 0                  | 0                   | 0      | 0                | 0                      | 0           | 0                | 0                     | 0                   | 0         | 0          | 0      | 0                      | 0   | 0          | 0     | 0         | 0        |
| 3 | Wakefield    | 0                 | 0                  | 0                   | 0      | 0                | 0                      | 0           | 0                | 0                     | 0                   | 0         | 0          | 0      | 0                      | 0   | 0          | 0     | 0         | 0        |
| 4 | Wakefield    | 0                 | 0                  | 0                   | 0      | 0                | 0                      | 0           | 0                | 0                     | 0                   | 0         | 0          | 0      | 0                      | 0   | 0          | 0     | 0         | 0        |

Once the encoding is done, we group the Neighbourhoods and obtain the mean for every venue within get neighbourhood.

|   | Neighborhood  | Afghan Restaurant | African Restaurant | American Restaurant | Arcade | Arepa Restaurant | Argentinian Restaurant | Art Gallery | Asian Restaurant | Australian Restaurant | Austrian Restaurant | BBQ Joint | Bagel Shop | Bakery   | Bangladeshi Restaurant | Bar | Bath House | Beach | Beach Bar |
|---|---------------|-------------------|--------------------|---------------------|--------|------------------|------------------------|-------------|------------------|-----------------------|---------------------|-----------|------------|----------|------------------------|-----|------------|-------|-----------|
| 0 | Allerton      | 0.0               | 0.0                | 0.020000            | 0.0    | 0.0              | 0.0                    | 0.0         | 0.000000         | 0.0                   | 0.0                 | 0.000000  | 0.000000   | 0.060000 | 0.0                    | 0.0 | 0.0        | 0.0   | 0         |
| 1 | Annadale      | 0.0               | 0.0                | 0.107143            | 0.0    | 0.0              | 0.0                    | 0.0         | 0.000000         | 0.0                   | 0.0                 | 0.000000  | 0.071429   | 0.071429 | 0.0                    | 0.0 | 0.0        | 0.0   | 0         |
| 2 | Arden Heights | 0.0               | 0.0                | 0.096774            | 0.0    | 0.0              | 0.0                    | 0.0         | 0.000000         | 0.0                   | 0.0                 | 0.032258  | 0.032258   | 0.032258 | 0.0                    | 0.0 | 0.0        | 0.0   | 0         |
| 3 | Arlington     | 0.0               | 0.0                | 0.055556            | 0.0    | 0.0              | 0.0                    | 0.0         | 0.027778         | 0.0                   | 0.0                 | 0.000000  | 0.027778   | 0.027778 | 0.0                    | 0.0 | 0.0        | 0.0   | 0         |
| 4 | Arrochar      | 0.0               | 0.0                | 0.000000            | 0.0    | 0.0              | 0.0                    | 0.0         | 0.000000         | 0.0                   | 0.0                 | 0.000000  | 0.066667   | 0.000000 | 0.0                    | 0.0 | 0.0        | 0.0   | 0         |

### 3.3 Classification Model

Now that we have data in the required format, we need to apply the K-Means grouping algorithm to group the data into similar clusters. This helps us in identifying similarities between the Neighbourhoods.

The main parameter if K-Means algorithm is the cluster size. To find the best cluster size, we use silhouette\_score.

Below is the result of silhouette\_score for different cluster sizes.

```
silhouette_score results : {2: 0.17743357998254314, 3: 0.14066439991387775, 4: 0.12914863106062802, 5: 0.1328577503914053, 6: 0.11736892508853244, 7: 0.12027984810897324, 8: 0.08326902182481445, 9: 0.07694941825109619}
```

The highlighted value shows that Cluster size of 2 is best suited for K-Means algorithm.

### 3.4 After Classification

After the classification model was successfully run on the dataset, we obtain the following result by further wrangling the data. We find out the most common venues within the neighbourhoods and then apply the Cluster Labels(0 and 1) to each of the entries.

|   | 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 | Bronx   | Wakefield    | 40.894705 | -73.847201 | 0              | Caribbean Restaurant  | Deli / Bodega         | Chinese Restaurant    | Fast Food Restaurant  | Donut Shop            | Food                  | Asian Restaurant      | Southern / Soul Food Restaurant | Steakhouse            | Juice Bar              |
| 1 | Bronx   | Co-op City   | 40.874294 | -73.829939 | 0              | Bakery                | Chinese Restaurant    | Deli / Bodega         | Fast Food Restaurant  | Donut Shop            | Caribbean Restaurant  | Fried Chicken Joint   | Mexican Restaurant              | Seafood Restaurant    | American Restaurant    |
| 2 | Bronx   | Eastchester  | 40.887556 | -73.827806 | 0              | Pizza Place           | Deli / Bodega         | Caribbean Restaurant  | Fast Food Restaurant  | Bakery                | Donut Shop            | Chinese Restaurant    | Burger Joint                    | Café                  | Asian Restaurant       |
| 3 | Bronx   | Fieldston    | 40.895437 | -73.905643 | 0              | Pizza Place           | Mexican Restaurant    | Deli / Bodega         | Fast Food Restaurant  | Bagel Shop            | Donut Shop            | Sushi Restaurant      | Bakery                          | Sandwich Place        | Diner                  |
| 4 | Bronx   | Riverdale    | 40.890834 | -73.912585 | 0              | Pizza Place           | Diner                 | Bagel Shop            | Deli / Bodega         | Fast Food Restaurant  | Café                  | Mexican Restaurant    | Food Truck                      | Bakery                | Sandwich Place         |

## 4. Analysis

After the data has been clustered, we analyse each cluster to find which cluster contains more relevant data by finding its shape.

The cluster 0 has 233 rows while Cluster 1 has 73 rows.

Hence we use Cluster 0 as our dataset for further evaluation.

### 4.1 Finding Centroid of Cluster 0

The next step is to find the centroid of Cluster 0. This is in fact the optimal location for opening our restaurant supply store.

We calculate the centroid of geographical coordinates by using the averaging methodology. A detailed explanation is given [here](#).

We obtain the centroids as **Latitude : 40.69394045210022, Longitude : -73.93700185136717.**

## 4.2 Reverse lookup of Geographic Coordinates

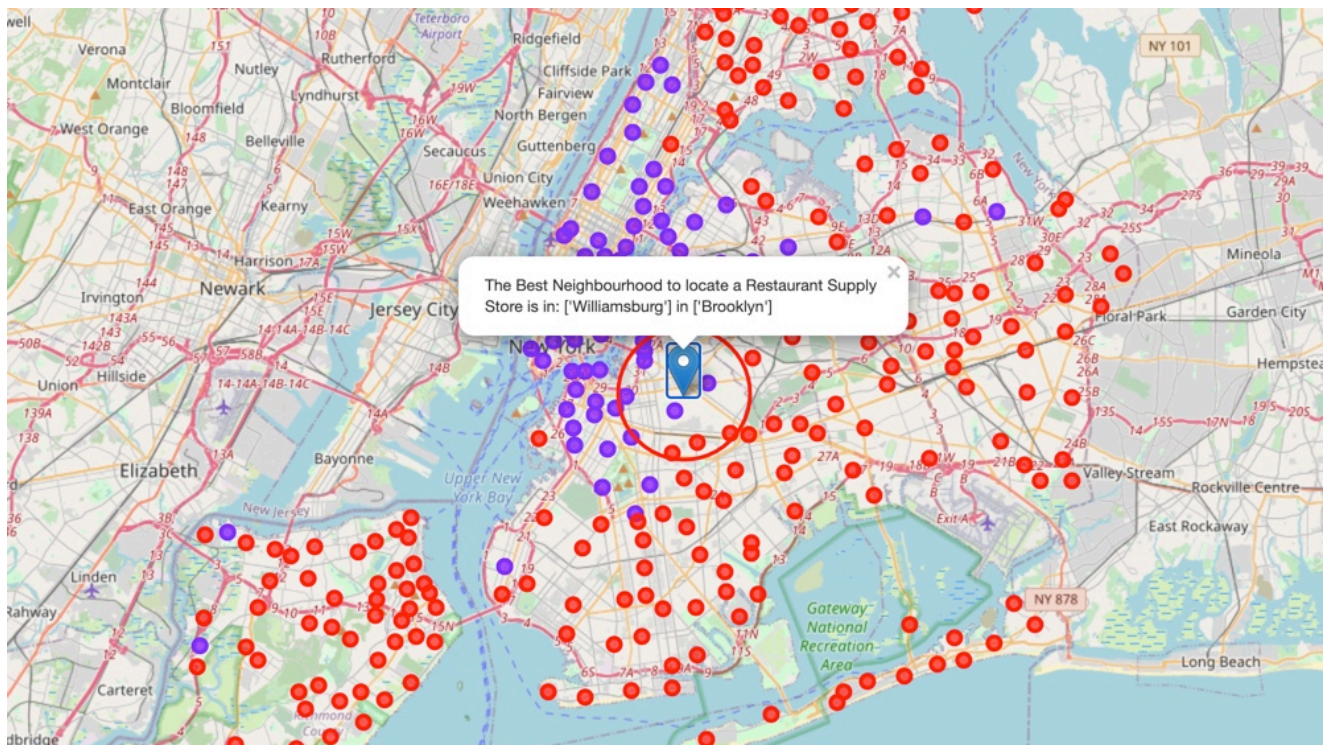
We use OpenCageGeocode to reverse lookup the coordinates and obtain the correct details such as Neighbourhood or Borough. This is used to locate the data on the map using folium.

The reverse lookup yields the following data -

```
'components': {'ISO_3166-1_alpha-2': 'US',
                'ISO_3166-1_alpha-3': 'USA',
                '_category': 'building',
                '_type': 'building',
                'building': 'Building 7',
                'city': 'New York',
                'city_district': 'Kings County',
                'continent': 'North America',
                'country': 'United States',
                'country_code': 'us',
                'house_number': '93',
                'postcode': '11206',
                'residential': 'BEDFORD STUYVESANT/
EXPANDED STUYVEVSANT '
                'HEIGHTS HISTORIC
DISTRICT',
                'road': 'Lewis Avenue',
                'state': 'New York',
                'state_code': 'NY',
                'suburb': 'Brooklyn'},
'confidence': 10,
'formatted': 'Building 7, 93 Lewis Avenue, New York, NY
11206, United States '
            'of America',
'geometry': {'lat': 40.6941276, 'lng': -73.9369924}}
```

This gives us the precise location within the city of New York to open our restaurant supply store.

A map with the cluster and the location marked is shown below -



## 5. Conclusion

I feel confident with the recommendation I have given my client as it is backed up with demonstrated data analysis. While nothing can ever be 100% certain they will certainly be better informed than they were prior to this comprehensive analysis.

Much more inference can be obtained with more work. A potential side business for my client might be assisting new restaurant owners where they might locate a new restaurant, who their competition is and who their clientele might be.