Capstone Final Project Where to open my new Bar

### **Problem**

York for our example of city). The first reflex that comes is to find where there is a high and low density of Bars to look for a place where he can benifit from his new Bar. For exemple, the majority will choose to open a new Bar where there is no high density of Bars. In this report, we are going to use you will use the Foursquare API to explore neighborhoods in Toronto.

#### Data sources

Using data in our previous lab, we're import it from <a href="https://cocl.us/new\_york\_dataset">https://cocl.us/new\_york\_dataset</a>. The data we're going to explore is composed 5 boroughs and 306 neighborhoods.

Our first data set is composed of 4 columns:

- Borough
- Neighborhood
- Latitude
- Longitude

# Data sources

|   | 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     | ronx Riverdale |           | -73.912585 |  |
| 5 | Bronx     | Kingsbridge    | 40.881687 | -73.902818 |  |
| 6 | Manhattan | Marble Hill    | 40.876551 | -73.910660 |  |
| 7 | Bronx     | Woodlawn       | 40.898273 | -73.867315 |  |
| 8 | Bronx     | Norwood        | 40.877224 | -73.879391 |  |
| 9 | Bronx     | Williamsbridge | 40.881039 | -73.857446 |  |

The data is composed 5 boroughs and 306 neighborhoods.

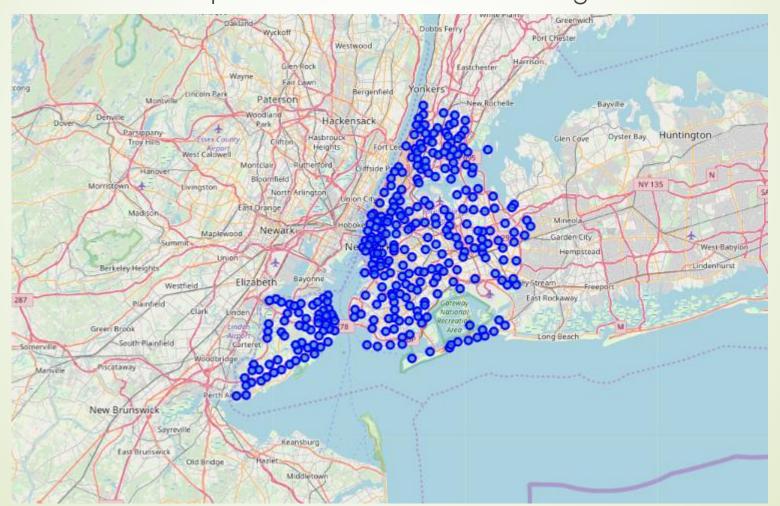
## **Using foursquare API**

After getting the data of every Borough and Neighborhood on New York with Their Latitude and Longitude, we'll use Foursquare API to get venues in every Neighborhod. The result of calling the Api is shown here:

|   | Neighborhood | Neighborhood<br>Latitude | Neighborhood<br>Longitude | Venue                                      | Venue<br>Latitude | Venue<br>Longitude | Venue<br>Category       |  |
|---|--------------|--------------------------|---------------------------|--|-------------------|--------------------|-------------------------|--|
| ( | Wakefield    | 40.894705                | -73.847201                | Lollipops Gelato                           | 40.894123         | -73.845892         | Dessert Shop            |  |
| • | Wakefield    | 40.894705                | -73.847201                | Rite Aid                                   | 40.896649         | -73.844846         | Pharmacy                |  |
| : | 2 Wakefield  | 40.894705                | -73.847201                | Carvel Ice Cream                           | 40.890487         | -73.848568         | Ice Cream<br>Shop       |  |
| ; | Wakefield    | 40.894705                | -73.847201                | Cooler Runnings Jamaican<br>Restaurant Inc | 40.898276         | -73.850381         | Caribbean<br>Restaurant |  |
| 4 | Wakefield    | 40.894705                | -73.847201                | Dunkin'                                    | 40.890459         | -73.849089         | Donut Shop              |  |
| į | Wakefield    | 40.894705                | -73.847201                | SUBWAY                                     | 40.890656         | -73.849192         | Sandwich<br>Place       |  |
| ( | Wakefield    | 40.894705                | -73.847201                | Pitman Deli                                | 40.894149         | -73.845748         | Food                    |  |
| 7 | Wakefield    | 40.894705                | -73.847201                | Baychester Avenue Food<br>Truck            | 40.892293         | -73.843230         | Food Truck              |  |
| 8 | Wakefield    | 40.894705                | -73.847201                | Koss Quick Wash                            | 40.891147         | -73.850230         | Laundromat              |  |
| 9 | Co-op City   | 40.874294                | -73.829939                | Capri II Pizza                             | 40.876374         | -73.829940         | Pizza Place             |  |

## **Exploratory Data Analysis**

Let's first create a map where we can render the neighborhoods in New York:



## **Exploratory Data Analysis**

We'll group rows by neighborhood and by taking the mean of the frequency of occurrence of each category where we'll find the name 'Bar'

|   | Neighborhood  | Bar | Beer<br>Bar | Cocktail<br>Bar | Dive<br>Bar | Gay<br>Bar | Hookah<br>Bar | Hotel<br>Bar | Juice<br>Bar | Karaoke<br>Bar | Sake<br>Bar | Salon /<br>Barbershop |       | Whisky<br>Bar | Win∈<br>Bar |
|---|---------------|-----|-------------|-----------------|-------------|------------|---------------|--------------|--------------|----------------|-------------|-----------------------|-------|---------------|-------------|
| 0 | Allerton      | 0.0 | 0.0         | 0.0             | 0.0         | 0.0        | 0.0           | 0.0          | 0.0          | 0.0            | 0.0         | 0.0                   | 0.000 | 0.0           | 0.0         |
| 1 | Annadale      | 0.0 | 0.0         | 0.0             | 0.0         | 0.0        | 0.0           | 0.0          | 0.0          | 0.0            | 0.0         | 0.0                   | 0.125 | 0.0           | 0.0         |
| 2 | Arden Heights | 0.0 | 0.0         | 0.0             | 0.0         | 0.0        | 0.0           | 0.0          | 0.0          | 0.0            | 0.0         | 0.0                   | 0.000 | 0.0           | 0.0         |
| 3 | Arlington     | 0.0 | 0.0         | 0.0             | 0.0         | 0.0        | 0.0           | 0.0          | 0.0          | 0.0            | 0.0         | 0.0                   | 0.000 | 0.0           | 0.0         |
| 4 | Arrochar      | 0.0 | 0.0         | 0.0             | 0.0         | 0.0        | 0.0           | 0.0          | 0.0          | 0.0            | 0.0         | 0.0                   | 0.000 | 0.0           | 0.0         |

## **Exploratory Data Analysis**

Let's create the new dataframe and display the top 10 venues that contain Bars.

|   | Neighborhood    | 1st Most<br>Common<br>Venue | 2nd<br>Most<br>Common<br>Venue | 3rd Most<br>Common<br>Venue | 4th Most<br>Common<br>Venue | 5th Most<br>Common<br>Venue | 6th Most<br>Common<br>Venue | 7th Most<br>Common<br>Venue | 8th Most<br>Common<br>Venue | 9th Most<br>Common<br>Venue | 10th<br>Most<br>Commo<br>Venue |
|---|-----------------|-----------------------------|--------------------------------|-----------------------------|-----------------------------|-----------------------------|-----------------------------|-----------------------------|-----------------------------|-----------------------------|--------------------------------|
| 0 | Allerton        | Wine Bar                    | Whisky<br>Bar                  | Sports<br>Bar               | Salon /<br>Barbershop       | Sake Bar                    | Karaoke<br>Bar              | Juice Bar                   | Hotel Bar                   | Hookah<br>Bar               | Gay Bar                        |
| 1 | Annadale        | Sports<br>Bar               | Wine Bar                       | Whisky<br>Bar               | Salon /<br>Barbershop       | Sake Bar                    | Karaoke<br>Bar              | Juice Bar                   | Hotel Bar                   | Hookah<br>Bar               | Gay Bar                        |
| 2 | Arden Heights   | Wine Bar                    | Whisky<br>Bar                  | Sports<br>Bar               | Salon /<br>Barbershop       | Sake Bar                    | Karaoke<br>Bar              | Juice Bar                   | Hotel Bar                   | Hookah<br>Bar               | Gay Bar                        |
| 3 | Arlington       | Wine Bar                    | Whisky<br>Bar                  | Sports<br>Bar               | Salon /<br>Barbershop       | Sake Bar                    | Karaoke<br>Bar              | Juice Bar                   | Hotel Bar                   | Hookah<br>Bar               | Gay Bar                        |
| 4 | Arrochar        | Wine Bar                    | Whisky<br>Bar                  | Sports<br>Bar               | Salon /<br>Barbershop       | Sake Bar                    | Karaoke<br>Bar              | Juice Bar                   | Hotel Bar                   | Hookah<br>Bar               | Gay Bar                        |
| 5 | Arverne         | Wine Bar                    | Whisky<br>Bar                  | Sports<br>Bar               | Salon /<br>Barbershop       | Sake Bar                    | Karaoke<br>Bar              | Juice Bar                   | Hotel Bar                   | Hookah<br>Bar               | Gay Bar                        |
| 6 | Astoria         | Wine Bar                    | Whisky<br>Bar                  | Sports<br>Bar               | Salon /<br>Barbershop       | Sake Bar                    | Karaoke<br>Bar              | Juice Bar                   | Hotel Bar                   | Hookah<br>Bar               | Gay Bar                        |
| 7 | Astoria Heights | Wine Bar                    | Whisky<br>Bar                  | Sports<br>Bar               | Salon /<br>Barbershop       | Sake Bar                    | Karaoke<br>Bar              | Juice Bar                   | Hotel Bar                   | Hookah<br>Bar               | Gay Bar                        |
| 8 | Auburndale      | Wine Bar                    | Whisky<br>Bar                  | Sports<br>Bar               | Salon /<br>Barbershop       | Sake Bar                    | Karaoke<br>Bar              | Juice Bar                   | Hotel Bar                   | Hookah<br>Bar               | Gay Bar                        |
| 9 | Bath Beach      | Wine Bar                    | Whisky<br>Bar                  | Sports<br>Bar               | Salon /<br>Barbershop       | Sake Bar                    | Karaoke<br>Bar              | Juice Bar                   | Hotel Bar                   | Hookah<br>Bar               | Gay Bar                        |

### Modeling

#### K-means Algorithm

- Let's use k-means to cluster the neighborhood into 5 clusters. It means that the value of k for this exemple is 5.
- k-means is one of the simplest and popular unsupervised machine learning algorithms. In other words, the K-means algorithm identifies k number of centroids, and then allocates every data point to the nearest cluster, while keeping the centroids as small as possible.

## Modeling

# K-means Algorithm

Here is the result of clustering with 5 labels after fiting our k-means algorithm:

|   | Borough   | Neighborhood | Latitude  | Longitude  | Cluster<br>Labels | 1st Most<br>Common<br>Venue | 2nd<br>Most<br>Common<br>Venue | 3rd Most<br>Common<br>Venue | 4th Most<br>Common<br>Venue | 5th Most<br>Common<br>Venue |
|---|-----------|--------------|-----------|------------|-------------------|-----------------------------|--------------------------------|-----------------------------|-----------------------------|-----------------------------|
| 0 | Bronx     | Wakefield    | 40.894705 | -73.847201 | 0.0               | Wine Bar                    | Whisky<br>Bar                  | Sports<br>Bar               | Salon /<br>Barbershop       | Sake Bar                    |
| 1 | Bronx     | Co-op City   | 40.874294 | -73.829939 | 0.0               | Wine Bar                    | Whisky<br>Bar                  | Sports<br>Bar               | Salon /<br>Barbershop       | Sake Bar                    |
| 2 | Bronx     | Eastchester  | 40.887556 | -73.827806 | 0.0               | Wine Bar                    | /ine Bar                       |                             | Salon /<br>Barbershop       | Sake Bar                    |
| 3 | Bronx     | Fieldston    | 40.895437 | -73.905643 | 0.0               | Wine Bar                    | Whisky<br>Bar                  | Sports<br>Bar               | Salon /<br>Barbershop       | Sake Bar                    |
| 4 | Bronx     | Riverdale    | 40.890834 | -73.912585 | 0.0               | Wine Bar                    | Whisky<br>Bar                  | Sports<br>Bar               | Salon /<br>Barbershop       | Sake Bar                    |
| 5 | Bronx     | Kingsbridge  | 40.881687 | -73.902818 | 0.0               | Beer Bar                    | Wine Bar                       | r Whisky<br>Bar Sports Bar  |                             | Salon /<br>Barbershop       |
| 6 | Manhattan | Marble Hill  | 40.876551 | -73.910660 | 0.0               | Wine Bar                    | Whisky<br>Bar                  | Sports<br>Bar               | Salon /<br>Barbershop       | Sake Bar                    |
| 7 | Bronx     | Woodlawn     | 40.898273 | -73.867315 | 3.0               | Bar                         | Wine Bar                       | Whisky<br>Bar               | Sports Bar                  | Salon /<br>Barbershop       |
| 8 | Bronx     | Norwood      | 40.877224 | -73.879391 | 0.0               | Wine Bar                    | Whisky<br>Bar                  | Sports<br>Bar               | Salon /<br>Barbershop       | Sake Bar                    |

#### Result

Let's show our clusters on a map using Folium to get a visual version of the clusters



As you can see on the map, we have 5 clusters of every type of bars. This clusters can be labeled depending on the density of bars in the cluster and help the person who want to open a bar to find the best neighborhood for that