Introduction: Business Problem

In this project we will try to find an **optimal location for a Bar**. Specifically, this report will be targeted to stakeholders interested in **opening a Bar in Toronto**, **Canada**.

Canada is a country which allows migrants from other countries, as the number of immigrants are getting permanent residence in the Canada, most of them try to open a new business for the survival. As, Toronto is the city with high population, high facilities immigrants try to find places to open a new business in Toronto but struggles which business can be open at which location of the city. To overcome this, we will analyse the venues at different Borough by taking the example of Bar business.

Since there are lots of Bar in Toronto, we will try to detect locations that are not already crowded with Bar. We are also particularly interested in areas with no Bar is in vicinity. We would also prefer locations as close to city centre as possible, assuming that first two conditions are met.

We will use our data science powers to generate a few most promising neighbourhoods based on these criteria. Advantages of each area will then be clearly expressed so that best possible final location can be chosen by stakeholders.

Data

Based on definition of our problem, factors that will influence our decision are:

- number of existing Bar in the neighbourhood
- number of and distance to Bars in the neighbourhood, if any
- distance of neighbourhood from city centre

We decided to use regularly spaced grid of locations, cantered around city center, to define our neighbourhoods.

Following data sources will be needed to extract/generate the required information:

- centers of candidate areas will be generated algorithmically and approximate addresses of centers of those areas will be obtained using **Google Maps API reverse geocoding**
- number of restaurants and their type and location in every neighbourhood will be obtained using Foursquare API
- coordinate of Toronto center will be obtained using **Google Maps API geocoding** of well-known Toronto location

we will us the Wikipedia page for extracting the data

https://en.wikipedia.org/wiki/List_of_postal_codes_of_Canada:_M

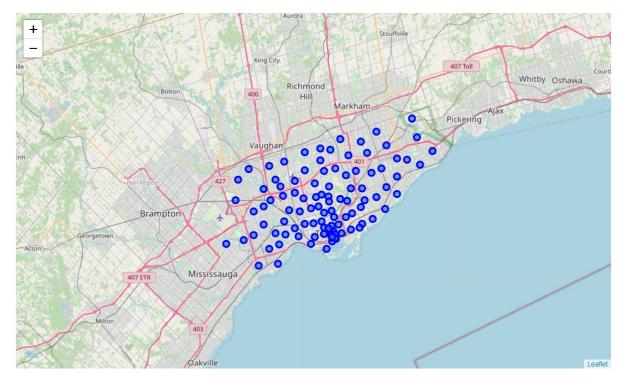
The following shows the dataframe we obtained after wrangling -

| Neighbourhood | Bourough | Postalcode | |
|--|------------------|------------|----|
| Parkwoods | North York | МЗА | 0 |
| Victoria Village | North York | M4A | 1 |
| Regent Park, Harbourfront | Downtown Toronto | M5A | 2 |
| Lawrence Manor, Lawrence Heights | North York | M6A | 3 |
| Queen's Park, Ontario Provincial Government | Downtown Toronto | M7A | 4 |
| Islington Avenue | Etobicoke | M9A | 5 |
| Malvern, Rouge | Scarborough | M1B | 6 |
| Don Mills | North York | МЗВ | 7 |
| Parkview Hill, Woodbine Gardens | East York | M4B | 8 |
| Garden District, Ryerson | Downtown Toronto | M5B | 9 |
| Glencairn | North York | M6B | 10 |
| West Deane Park, Princess Gardens, Martin Grov | Etobicoke | M9B | 11 |

We obtained the longitude and latitude of all the venues so as to identify the exact location.

| | Neighbourhood | Neighbourhood Latitude | Neighbourhood Longitude | Venue | Venue Latitude | Venue Longitude | Venue Category |
|---|---|---------------------------|----------------------------|---------------------------------|-------------------|--------------------|-------------------------|
| 0 | Malvern, Rouge | 43.806686 | -79.194353 | Wendy's | 43.807448 | -79.199056 | Fast Food Restaurant |
| 1 | Rouge Hill, Port Union, Highland Creek | 43.784535 | -79.160497 | RIGHT WAY TO GOLF | 43.785177 | -79.161108 | Golf Course |
| 2 | Rouge Hill, Port Union, Highland Creek | 43.784535 | -79.160497 | Royal Canadian Legion | 43.782533 | -79.163085 | Bar |
| 3 | Rouge Hill, Port Union, Highland Creek | 43.784535 | -79.160497 | Affordable Toronto Movers | 43.787919 | -79.162977 | Moving Target |
| 4 | Guildwood, Morningside, West Hill | 43.763573 | -79.188711 | RBC Royal Bank | 43.766790 | -79.191151 | Bank |
| 5 | Guildwood, Morningside, West Hill | 43.763573 | -79.188711 | G & G Electronics | 43.765309 | -79.191537 | Electronics Store |
| 6 | Guildwood, Morningside, West Hill | 43.763573 | -79.188711 | Big Bite Burrito | 43.766299 | -79.190720 | Mexican Restaurant |
| 7 | Guildwood, Morningside, West Hill | 43.763573 | -79.188711 | Enterprise Rent-A-Car | 43.764076 | -79.193406 | Rental Car Location |
| 8 | Guildwood, Morningside, West Hill | 43.763573 | -79.188711 | Woburn Medical Centre | 43.766631 | -79.192286 | Medical Center |
| 9 | Guildwood, Morningside, West Hill | 43.763573 | -79.188711 | Lawrence Ave E & Kingston Rd | 43.767704 | -79.189490 | Intersection |

The map shows all the neighbourhoods locations as per there longitude and latitudes-



Methodology

In this project we will direct our efforts on detecting the 5 most common venues in the Neighbourhoods along with its frequency.

```
----Agincourt----
                               freq
                        venue
0
                                0.2
                       Lounge
              Breakfast Spot
1
                                0.2
   Latin American Restaurant
2
                                0.2
3
                Skating Rink
                                0.2
4
              Clothing Store
                                0.2
  --Alderwood, Long Branch----
            venue
                   freq
      Pizza Place
0
1
         Pharmacy
                    0.1
2
              Gym
     Skating Rink
3
                    0.1
   Sandwich Place
                    0.1
```

Second step in our analysis will be looking at the top 10 most common venues in each Neighbourhood and storing the data in new data frame for further analysis after **clustering. We used KMeans for clustering.

| | Neighbourhood | 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 | Agincourt | Lounge | Latin American Restaurant | Skating Rink | Clothing Store | Breakfast Spot | Dog Run | Dim Sum Restaurant | Diner | Discount Store | Distribution Center |
| 1 | Alderwood, Long Branch | Pizza Place | Gym | Pharmacy | Sandwich Place | Athletics & Sports | Pool | Pub | Skating Rink | Coffee Shop | Curling Ice |
| 2 | Bathurst Manor, Wilson Heights, Downsview North | Coffee Shop | Bank | Pizza Place | Pharmacy | Sushi Restaurant | Middle Eastern Restaurant | Shopping Mall | Deli / Bodega | Restaurant | Fried Chicken Joint |
| 3 | Bayview Village | Café | Bank | Chinese Restaurant | Japanese Restaurant | Yoga Studio | Department Store | Dim Sum Restaurant | Diner | Discount Store | Distribution Center |
| 4 | Bedford Park, Lawrence Manor East | Coffee Shop | Restaurant | Sandwich Place | Italian Restaurant | Sushi Restaurant | Comfort Food Restaurant | Pharmacy | Pizza Place | Café | Pub |

Following map shows clustering with K=5



In third and final step we will focus on most promising areas and within those create clusters of locations that meet some basic requirements established in discussion with stakeholders: we will take into consideration locations with less no of Bars. We will present map of all such locations but also create clusters (using k-means clustering) of those locations. We will observe and find out the locations where we can set-up Bars for better Business.

Below is the example of how we examined each cluster for the different venues.

| Clu | ster 3 | 3 - No Bars | | | | | | | | | | | |
|------------|--------|-------------|----------------|-----------------------------|-----------------------------|-----------------------------|-----------------------------|-----------------------------|-----------------------------|-----------------------------|-----------------------------|-----------------------------|------------------------------|
| In [41]: ▶ | df_c | | _merged.loc[t | oronto_mer | ged['Clust | er_Labels | '] == 2, t | coronto_mer | ged.column | s[[1] + li | st(range(| 5, toronto | merged.sha |
| | 4 | | | | | | | | | | | | > |
| Out[41]: | | Bourough | 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 5 | Scarborough | 2 | Fast Food Restaurant | Deli / Bodega | Event Space | Ethiopian Restaurant | Electronics Store | Eastern European | Drugstore | Donut Shop | Doner Restaurant | Dog Run |

Cluster 2 - Bars are present only in North York



Results

The following are the outcomes of analysis of 5 clusters-

- In cluster 1 Bars are present in Scarborough and Etobicoke Borough.
- In cluster 2 Bars are present in only North York.
- In cluster 3, 4 and 5 Bars are not present.

Discussion

Hence, for people who are interested in opening the Bar business can easily opt for any on cluster 3 or 4 or 5.Ba

In case the stakeholders wants the business to be in cluster 1 or 2, they can choose Borough other than Scarborough and Etobicoke for cluster-1 and borough except North York in cluster 2.

Conclusion

Purpose of this project was to identify Toronto areas close to center with low number of Bars in order to aid stakeholders in narrowing down the search for optimal location for a business. By calculating Bar density(frequency) distribution from Foursquare data we have first identified general boroughs that justify further analysis, and then generated extensive collection of locations which satisfy some basic requirements regarding existing nearby bars. Clustering of those locations was then performed in order to create major zones of interest and addresses of those zone centers were created to be used as starting points for final exploration by stakeholders.

Final decision on optimal Bar location will be made by stakeholders based on specific characteristics of neighbourhoods and locations in every recommended zone, taking into consideration additional factors like attractiveness of each location (proximity to park or water), levels of noise / proximity to major roads, real estate availability, prices, social and economic dynamics of every neighbourhood etc.