Capstone Project - The Battle of Neighborhoods (Week 1)

July 9, 2019

1 Battle of Neighborhoods (Week 1)

1.1 1. Introduction Section

1.1.1 Description of the Problem

A restaurant is a business which prepares and serves food and drink to customers. It is no surprise to learn that owning a restaurant comes with its fair share of responsibilities that require you to be on your feet. To be successful in the long run, a restaurant needs an accessible location, a theme or style that appeals to a broad range of customers, and a solid menu. Restaurants must stand out from the crowd and finalizing a space within the budget.

So it is evident that to survive in such competitive market it is very important to startegically plan. Often there's an issue of time constraint that keeps a restaurant owner preoccupied with intricacies. Whether it is sourcing the right ingredients, obtaining too many licences, finalizing a space within your budget to interviewing the right staff — restaurateurs have to consider everything beginning from a pin to piano before taking the right foot forward.

First move is very important, thereby choice of location is very important. So, HSB India (P) Ltd., plan to choose the correct location to start its first restaurant in Toronto, Canada.

1.1.2 Discussion of the Background - Prospects of a Indian Restaurant in Toronto, Canada.

Toronto, Canada's largest city is consistently rated among the world's top most liveable cities. Toronto boasts a multicultural population, diverse neighbourhoods, and world class cultural attractions, entertainment, dining and shopping.

Toronto is an international centre of business, finance, arts, and culture, and is recognized as one of the most multicultural and cosmopolitan cities in the world. Its varied cultural institutions, which include numerous museums and galleries, festivals and public events, entertainment districts, national historic sites, and sports activities, attract over 25 million tourists each year.

People have travelled through and inhabited the Toronto area and the diverse population of Toronto reflects its current and historical role as an important destination for immigrants to Canada. According to the 2011 National Household Survey (NHS), 1,264,395 non-Whites, or 20.2% of Canada's visible minority population, live in the city of Toronto; of this, approximately 70 percent are of Asian ancestry.

These indicates that the market is highly competitive. As it is highly developed city so cost of doing business is also one of the highest. Thus, any new business venture or expansion needs to be analysed carefully. The insights derived from analysis will give good understanding of the business environment which help in strategically targeting the market. This will help in reduction of risk and the Return on Investment will be reasonable.

Target Audience

- 1. Business personnel who wants to invest or open a ethnic restaurant. This analysis will be a comprehensive guide to start of expand restaurants targeting the ethnica diversity in Toronto.
- 2. Any Freelancer who wants to start a new restaurant in Toronto.
- 3. New graduates, to find reasonable lunch/breakfast place close to office.
- 4. Budding Data Scientists, who wants to implement some of the most used Exploratory Data Analysis techniques to obtain necessary data, analyze it and, finally be able to tell a story out of it.

Success Criteria: The success criteria of the project will be a good recommendation of Borough/Neighborhood choice to HSB India (P) Ltd., based on lack of such restaurants in that location and nearest suppliers of ingredients.

1.2 2. Data Section

Neighborhood has a total of 11 boroughs and 103 neighborhoods. In order to segement the neighborhoods and explore them, we will essentially need a dataset that contains the 11 boroughs and the neighborhoods that exist in each borough as well as the latitude and logitude coordinates of each neighborhood.

Postal codes beginning with M are located within the City of Toronto in the province of Ontario. I will use the BeautifulSoup website scraping library to extract the content on the Wikipedia page (https://en.wikipedia.org/wiki/List_of_postal_codes_of_Canada:_M) as XML and transfor data that is in the table of postal codes into a pandas dataframe.

I will read the csv file that has the geographical coordinates of each postal code from http://cocl.us/Geospatial_data into a pandas dataframe. Perform a merge the Latitude and Longitude from the Geospatial dataframe with the Toronto neighbourhoods using inner join.

Next will slice the dataframe to have only the boroughs that contain the word Toronto. I will use the Foursquare API to explore neighborhoods in Toronto City. Toronto City geographical coordinates data used as input for the Foursquare API, to explore venues information in each neighborhood.

1.2.1 2.1 Preparation for Data

First, let's download all the dependencies that we will need.

```
[2]: import numpy as np # library to handle data in a vectorized manner import pandas as pd # library for data analysis pd.set_option('display.max_columns', None) pd.set_option('display.max_rows', None) import json # library to handle JSON files #!conda install -c conda-forge geopy --yes # uncomment this line to install geopy library from geopy.geocoders import Nominatim # convert an address into latitude and longitude values import requests # library to handle requests
```

```
from pandas.io.json import json_normalize # tranform JSON file into a pandas dataframe

# Matplotlib and associated plotting modules
import matplotlib.cm as cm
import matplotlib.colors as colors

# import k-means from clustering stage
from sklearn.cluster import KMeans

# import Beautiful Soup for webscraping
from bs4 import BeautifulSoup

#!conda install -c conda-forge folium=0.5.0 --yes # uncomment this line to install Folium library
import folium # map rendering library

print('Libraries imported.')
```

Libraries imported.

Download and Scrape the Data I used the BeautifulSoup website scraping library to extract the content on the Wikipedia page as XML.

Next, obtain the data that is in the table of postal codes and transform the data into a pandas dataframe.

```
[4]: table_LoP = LoP_Canada.find('table')
LoP_values = table_LoP.find_all('td')

postcode = []
borough = []
neighbourhood = []

for i in range(0, len(LoP_values), 3):
    postcode.append(LoP_values[i].text.strip())
    borough.append(LoP_values[i+1].text.strip())
    neighbourhood.append(LoP_values[i+2].text.strip())

df_LoP = pd.DataFrame(data=[postcode, borough, neighbourhood]).transpose()

# dataframe consist of three columns: PostalCode, Borough, and Neighborhood
df_LoP.columns = ['PostalCode', 'Borough', 'Neighbourhood']
```

Let's take a quick look at the data.

```
[5]: df_LoP.head()
```

```
[5]: PostalCode Borough Neighbourhood 0 M1A Not assigned Not assigned
```

```
1 M2A Not assigned Not assigned
2 M3A North York Parkwoods
3 M4A North York Victoria Village
4 M5A Downtown Toronto Harbourfront
```

Cleanup the Dataset

| | _01 _ | | |
|----|---|---|--|
| | PostalCode | Borough \ | |
| 0 | M1B | Scarborough | |
| 1 | M1C | Scarborough | |
| 2 | M1E | Scarborough | |
| 3 | M1G | Scarborough | |
| 4 | M1H | Scarborough | |
| 5 | M1J | Scarborough | |
| 6 | M1K | Scarborough | |
| 7 | M1L | Scarborough | |
| 8 | M1M | Scarborough | |
| 9 | M1N | Scarborough | |
| 10 | M1P | Scarborough | |
| 11 | M1R | Scarborough | |
| 12 | M1S | Scarborough | |
| 13 | M1T | $\operatorname{Scarborough}$ | |
| 14 | M1V | Scarborough | |
| 15 | M1W | Scarborough | |
| 16 | M1X | Scarborough | |
| 17 | M2H | North York | |
| 18 | M2J | North York | |
| 19 | M2K | North York | |
| 20 | M2L | North York | |
| 21 | M2M | North York | |
| 22 | M2N | North York | |
| 23 | M2P | North York | |
| 24 | M2R | North York | |
| | $ \begin{array}{c} 1 \\ 2 \\ 3 \\ 4 \\ 5 \\ 6 \\ 7 \\ 8 \\ 9 \\ 10 \\ 11 \\ 12 \\ 13 \\ 14 \\ 15 \\ 16 \\ 17 \\ 18 \\ 19 \\ 20 \\ 21 \\ 22 \\ 23 \\ \end{array} $ | 0 M1B 1 M1C 2 M1E 3 M1G 4 M1H 5 M1J 6 M1K 7 M1L 8 M1M 9 M1N 10 M1P 11 M1R 12 M1S 13 M1T 14 M1V 15 M1W 16 M1X 17 M2H 18 M2J 19 M2K 20 M2L 21 M2M 22 M2N 23 M2P | |

| 25 | M3A North York |
|----|----------------------|
| 26 | M3B North York |
| 27 | M3C North York |
| | |
| 28 | M3H North York |
| 29 | M3J North York |
| 30 | M3K North York |
| | |
| 31 | M3L North York |
| 32 | M3M North York |
| 33 | M3N North York |
| 34 | M4A North York |
| | |
| 35 | M4B East York |
| 36 | M4C East York |
| 37 | M4E East Toronto |
| | |
| 38 | M4G East York |
| 39 | M4H East York |
| 40 | M4J East York |
| 41 | M4K East Toronto |
| | |
| 42 | M4L East Toronto |
| 43 | M4M East Toronto |
| 44 | M4N Central Toronto |
| | |
| 45 | |
| 46 | M4R Central Toronto |
| 47 | M4S Central Toronto |
| 48 | M4T Central Toronto |
| | |
| 49 | M4V Central Toronto |
| 50 | M4W Downtown Toronto |
| 51 | M4X Downtown Toronto |
| 52 | M4Y Downtown Toronto |
| | |
| 53 | M5A Downtown Toronto |
| 54 | M5B Downtown Toronto |
| 55 | M5C Downtown Toronto |
| | |
| 56 | M5E Downtown Toronto |
| 57 | M5G Downtown Toronto |
| 58 | M5H Downtown Toronto |
| 59 | M5J Downtown Toronto |
| | |
| 60 | M5K Downtown Toronto |
| 61 | M5L Downtown Toronto |
| 62 | M5M North York |
| 63 | M5N Central Toronto |
| | |
| 64 | M5P Central Toronto |
| 65 | M5R Central Toronto |
| 66 | M5S Downtown Toronto |
| | |
| 67 | M5T Downtown Toronto |
| 68 | M5V Downtown Toronto |
| 69 | M5W Downtown Toronto |
| 70 | M5X Downtown Toronto |
| | |
| 71 | M6A North York |
| | |

```
72
        M6B
                  North York
73
        M6C
                       York
74
                       York
        M6E
        M6G Downtown Toronto
75
76
        M6H
                 West Toronto
77
        M6J
                 West Toronto
78
        M6K
                 West Toronto
79
                  North York
        M6L
80
        M6M
                       York
81
        M6N
                       York
82
        M6P
                 West Toronto
83
        M6R
                 West Toronto
84
        M6S
                 West Toronto
85
        M7A
                 Queen's Park
86
                 Mississauga
        M7R
                 East Toronto
87
        M7Y
88
        M8V
                   Etobicoke
89
        M8W
                    Etobicoke
90
        M8X
                   Etobicoke
91
        M8Y
                   Etobicoke
92
        M8Z
                   Etobicoke
93
        M9A
                   Etobicoke
94
        M9B
                   Etobicoke
95
        M9C
                   Etobicoke
96
        M9L
                  North York
97
        M9M
                   North York
98
                       York
        M9N
99
        M9P
                   Etobicoke
                   Etobicoke
100
        M9R
101
        M9V
                   Etobicoke
102
        M9W
                    Etobicoke
                             Neighbourhood
                            Rouge, Malvern
0
           Highland Creek, Rouge Hill, Port Union
1
2
               Guildwood, Morningside, West Hill
3
                                  Woburn
4
                                Cedarbrae
                         Scarborough Village
5
6
        East Birchmount Park, Ionview, Kennedy Park
7
                Clairlea, Golden Mile, Oakridge
8
     Cliffcrest, Cliffside, Scarborough Village West
9
                   Birch Cliff, Cliffside West
10
    Dorset Park, Scarborough Town Centre, Wexford ...
```

11

12

13

Maryvale, Wexford

Clarks Corners, Sullivan, Tam O'Shanter

Agincourt

| 14 | Agincourt North, L'Amoreaux East, Milliken, St |
|----|---|
| 15 | L'Amoreaux West |
| 16 | Upper Rouge |
| | |
| 17 | Hillcrest Village |
| 18 | Fairview, Henry Farm, Oriole |
| 19 | Bayview Village |
| 20 | Silver Hills, York Mills |
| 21 | Newtonbrook, Willowdale |
| | |
| 22 | Willowdale South |
| 23 | York Mills West |
| 24 | Willowdale West |
| 25 | Parkwoods |
| 26 | Don Mills North |
| | |
| 27 | Flemingdon Park, Don Mills South |
| 28 | Bathurst Manor, Downsview North, Wilson Heights |
| 29 | Northwood Park, York University |
| 30 | CFB Toronto, Downsview East |
| 31 | Downsview West |
| | |
| 32 | Downsview Central |
| 33 | Downsview Northwest |
| 34 | Victoria Village |
| 35 | Woodbine Gardens, Parkview Hill |
| 36 | Woodbine Heights |
| | The Beaches |
| 37 | |
| 38 | $\operatorname{Leaside}$ |
| 39 | Thorncliffe Park |
| 40 | East Toronto |
| 41 | The Danforth West, Riverdale |
| 42 | The Beaches West, India Bazaar |
| | , |
| 43 | Studio District |
| 44 | Lawrence Park |
| 45 | Davisville North |
| 46 | North Toronto West |
| 47 | Davisville |
| 48 | Moore Park, Summerhill East |
| | |
| 49 | Deer Park, Forest Hill SE, Rathnelly, South Hi |
| 50 | Rosedale |
| 51 | Cabbagetown, St. James Town |
| 52 | Church and Wellesley |
| 53 | Harbourfront, Regent Park |
| | · - |
| 54 | Ryerson, Garden District |
| 55 | St. James Town |
| 56 | Berczy Park |
| 57 | Central Bay Street |
| 58 | Adelaide, King, Richmond |
| 59 | · |
| | Harbourfront East, Toronto Islands, Union Station |
| 60 | Design Exchange, Toronto Dominion Centre |
| | |

```
61
                  Commerce Court, Victoria Hotel
62
                Bedford Park, Lawrence Manor East
63
                                  Roselawn
              Forest Hill North, Forest Hill West
64
65
              The Annex, North Midtown, Yorkville
66
                  Harbord, University of Toronto
67
          Chinatown, Grange Park, Kensington Market
68
    CN Tower, Bathurst Quay, Island airport, Harbo...
69
                 Stn A PO Boxes 25 The Esplanade
70
            First Canadian Place, Underground city
71
                Lawrence Heights, Lawrence Manor
72
                                 Glencairn
73
                          Humewood-Cedarvale
74
                          Caledonia-Fairbanks
75
                                  Christie
76
                   Dovercourt Village, Dufferin
77
                      Little Portugal, Trinity
78
       Brockton, Exhibition Place, Parkdale Village
79
               Downsview, North Park, Upwood Park
80
      Del Ray, Keelesdale, Mount Dennis, Silverthorn
81
                   The Junction North, Runnymede
82
                   High Park, The Junction South
83
                       Parkdale, Roncesvalles
84
                          Runnymede, Swansea
85
                               Queen's Park
86
             Canada Post Gateway Processing Centre
    Business Reply Mail Processing Centre 969 Eastern
87
88
       Humber Bay Shores, Mimico South, New Toronto
89
                        Alderwood, Long Branch
90
       The Kingsway, Montgomery Road, Old Mill North
    Humber Bay, King's Mill Park, Kingsway Park So...
91
92
    Kingsway Park South West, Mimico NW, The Queen...
93
                            Islington Avenue
94
    Cloverdale, Islington, Martin Grove, Princess ...
95
    Bloordale Gardens, Eringate, Markland Wood, Ol...
96
                              Humber Summit
97
                            Emery, Humberlea
98
                                   Weston
99
                                 Westmount
    Kingsview Village, Martin Grove Gardens, Richv...
101
    Albion Gardens, Beaumond Heights, Humbergate, ...
102
                                 Northwest
```

Read the csv file that has the geographical coordinates of each postal code from http://cocl.us/Geospatial_data into a pandas dataframe.

```
[7]: df_geo = pd.read_csv('http://cocl.us/Geospatial_data')
df_geo.columns = ['PostalCode', 'Latitude', 'Longitude']
```

Perform a merge the Latitude and Longitude from the Geospatial dataframe with the Toronto neighbourhoods using inner join. Print the top 5 from the merged dataframe.

```
[8]: df_pos = pd.merge(df_gp_LoP, df_geo, on=['PostalCode'], how='inner')

df_tor_neigh = df_pos[['PostalCode','Borough', 'Neighbourhood', 'Latitude', 'Longitude']].

-copy()

df_tor_neigh.head()
```

| [8]: | Posta | $_{ m llCode}$ | Borough | $egin{array}{cccc} egin{array}{ccccc} egin{array}{ccccccccc} egin{array}{cccccccccc} egin{array}{ccccccccc} egin{array}{cccccccc} egin{array}{ccccccccc} egin{array}{ccccccccc} egin{array}{ccccccccc} egin{array}{cccccccccc} egin{array}{ccccccccc} egin{array}{cccccccccc} egin{array}{cccccccccc} egin{array}{cccccccccccccccccccccccccccccccccccc$ |
|------|-------|----------------|-------------|---|
| | 0 | M1B | Scarborough | Rouge, Malvern 43.806686 |
| | 1 | M1C | Scarborough | Highland Creek, Rouge Hill, Port Union 43.784535 |
| | 2 | M1E | Scarborough | Guildwood, Morningside, West Hill 43.763573 |
| | 3 | M1G | Scarborough | Woburn 43.770992 |
| | 4 | M1H | Scarborough | Cedarbrae 43.773136 |

Longitude

- 0 79 . 194353
- 1 79.160497
- 2 79.188711
- 3 79.216917
- 4 79.239476

Print the number of rows of the dataframe.

- [9]: df_tor_neigh_shape
- **[9]**: (103, 5)

Let's simplify the dataframe and cluster only the neighborhoods in Toronto. So let's slice the original dataframe and just take portion of dataframe where Boroughs contain word Toronto.

| [10]: | ${\bf PostalCode}$ | | Borough | $ m Neighbourhood \ Latitude \ ackslash$ |
|-------|--------------------|-----|-----------------|--|
| | 0 | M4E | East Toronto | The Beaches 43.676357 |
| | 1 | M4K | East Toronto | The Danforth West, Riverdale 43.679557 |
| | 2 | M4L | East Toronto | The Beaches West, India Bazaar 43.668999 |
| | 3 | M4M | East Toronto | Studio District 43.659526 |
| | 4 | M4N | Central Toronto | Lawrence Park 43.728020 |

Longitude

- 0 -79.293031
- 1 79.352188
- 2 79.315572
- 3 79.340923
- 4 79.388790

1.3 Conclusion

1.3.1 Week 1: Description of Problem, Discussion of Background and Data

We get the initial Dataframe with Postalcode, Borough and corresponding Neighbourhoods in each Borough and the coordinates of those major Neighbourhoods.

So as the next step I will use Foursquare data and obtain information on restaurants. With these, we can start with our battle of neighborhoods for opening a restaurant in Toronto.

1.3.2 Thank You!