The Battle of Neighborhoods

(IBM Applied Data Science Capstone Final Project Report)

Introduction

Study and analysis of data by various methods is critical for any activity in today's world. This project is to help a friend to identify a neighborhood location in Toronto city like his current location in New York city with shopping malls, parks and food courts. My friend can decide his location to relocate based on the outcome of my analysis. This is achieved with the help of Machine Language Algorithms.

Business Problem

In this project, I want to help my friend to find the best neighborhood in Toronto which is like his current place of residence with amenities like shopping malls, parks and food courts. The challenge is to find a suitable neighborhood which is close enough meeting his specific requirements.

Data

A list of neighborhoods in New York and Toronto is downloaded and their respective location. The data is extracted from the below sources

New York neighborhoods:

https://ibm.box.com/shared/static/fbpwbovar7lf8p5sgddm06cgipa2rxpe.json

Toronto neighborhoods:

https://en.wikipedia.org/wiki/List of postal codes of Canada: M

Transform the data into Pandas dataframe, use Geopy Python package to get the latitude and the longitude of all the neighborhoods of Toronto, use Folium Python library to map the neighborhoods, use Foursquare API to get information about venues around the neighborhoods. Look for a group of venues in walking distance of each neighborhood like shopping malls, parks and food courts.

Methodology

For the purpose of doing unsupervised learning to find similarities between neighborhoods, clustering algorithm K-Means is used. K-Means clustering helps to classify the neighborhoods based on the nearby venues. To make the decision, the neighborhoods are sorted in each cluster.

| | Neighborhood | Shopping Mall | Cluster Labels | Neighborhood Latitude | Neighborhood Longitude | Venue | Venue Latitude | Venue Longitude | Venue Category |
|----|-----------------------------------|------------------|-------------------|--------------------------|---------------------------|--------------------------------------|-------------------|--------------------|------------------------|
| 0 | Bath Beach | 0.000000 | 0 | 40.599519 | -73.998752 | Bensonhurst Park | 40.597065 | -73.998340 | Park |
| 66 | Mill Basin | 0.000000 | 0 | 40.615974 | -73.915154 | Key Food | 40.617269 | -73.909776 | Supermarket |
| 66 | Mill Basin | 0.000000 | 0 | 40.615974 | -73.915154 | MTA B3, B100 (Ave U/East 66th St) | 40.617206 | -73.911299 | Bus Station |
| | | | ••• | | | | | | |
| 24 | Commerce Court, Victoria Hotel | 0.010000 | 2 | 43.648198 | -79.379817 | Jump | 43.648147 | -79.378752 | American Restaurant |
| 95 | Stn A PO Boxes | 0.010417 | 2 | 43.646435 | -79.374846 | Joe Fresh | 43.644285 | -79.369771 | Clothing Store |
| 95 | Stn A PO Boxes | 0.010417 | 2 | 43.646435 | -79.374846 | Loblaws | 43.645427 | -79.369789 | Grocery Store |

4392 rows x 9 columns

Map the cluster neighborhoods



Results

The results of the clusters are as below

Cluster 1

| | Shopping Mall | Venue | Venue Latitude | Venue Longitude | Venue Category |
|----|---------------|-----------------------------------|----------------|-----------------|---------------------|
| 0 | 0.0 | Bensonhurst Park | 40.597065 | -73.998340 | Park |
| 66 | 0.0 | Key Food | 40.617269 | -73.909776 | Supermarket |
| 66 | 0.0 | MTA B3, B100 (Ave U/East 66th St) | 40.617206 | -73.911299 | Bus Station |
| | | | | | |
| 33 | 0.0 | Dumbo Archway | 40.703056 | -73.987985 | Monument / Landmark |
| 30 | 0.0 | Kumo Sushi | 40.640900 | -73.965067 | Japanese Restaurant |
| 30 | 0.0 | Cortelyou Gourmet Deli | 40.641665 | -73.962767 | Deli / Bodega |

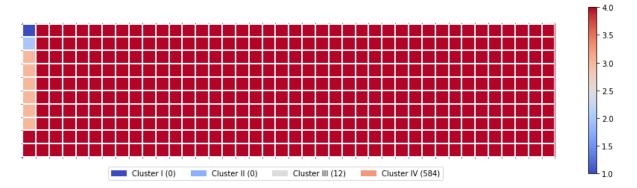
Cluster 2

| | Shopping Mall | Venue | Venue Latitude | Venue Longitude | Venue Category |
|----|---------------|---------------------------------------|----------------|-----------------|-------------------|
| 94 | 0.083333 | MTA Bus - B82, B82 LTD, B83, BM2, BM5 | 40.646296 | -73.879967 | Bus Station |
| 94 | 0.083333 | Dunkin' | 40.646776 | -73.883176 | Donut Shop |
| 94 | 0.083333 | CVS pharmacy | 40.647316 | -73.882990 | Pharmacy |
| | | | | | |
| 94 | 0.083333 | Starrett City Shopping Center | 40.649342 | -73.884329 | Shopping Mall |
| 94 | 0.083333 | Fresh Creek Natural Reserve | 40.643372 | -73.880271 | River |
| 94 | 0.083333 | 50% Half Cards | 40.647151 | -73.883584 | Convenience Store |

Cluster 3

| | Shopping Mall | Venue | Venue Latitude | Venue Longitude | Venue Category |
|-----|---------------|--------------------------|----------------|-----------------|---------------------|
| 102 | 0.010000 | Maple Leaf Square | 43.642925 | -79.380892 | Plaza |
| 102 | 0.010000 | The Ritz-Carlton | 43.645330 | -79.387089 | Hotel |
| 102 | 0.010000 | Boxcar Social Temperance | 43.650557 | -79.381956 | Bar |
| | | | | | |
| 24 | 0.010000 | Jump | 43.648147 | -79.378752 | American Restaurant |
| 95 | 0.010417 | Joe Fresh | 43.644285 | -79.369771 | Clothing Store |
| 95 | 0.010417 | Loblaws | 43.645427 | -79.369789 | Grocery Store |

In the above image it is obvious that cluster algorithm is not segmenting the neighborhoods for location areas. It is possible to see which neighborhoods within Brooklyn, New York are more like the neighborhoods within Toronto. Those neighborhoods that are similar among them belong to the same cluster. Hence, they have the same color in the image above.



Discussion

This work is useful only for those who live in Brooklyn, New York or in the neighborhoods near the center of Toronto, Canada. That is because there is a limited amount of data we can request using de Foursquare API.

Conclusion

The K-Means clustering algorithm is used for finding similiraties between all the neighborhoods listed in the feature matrix. Results show that there are 2 major groups and 2 minor groups. In this work a segmentation between two different countries is done. This segementation involves the neighborhoods in Brooklyn, New York and the neighborhoods near to the center of Toronto. The data is downloaded and the venues around the neighborhoods is selected using the Foursquare API. One Hot Encoding is used for converting the categories of the venues into a feature matrix. Then, all venues are grouped by neighborhoods and at the same time the mean is calculated. Hence, the resulting features used are the frequency of occurrence from each category in a neighborhood.

The description of the clusters is as below

Cluster

- I: Neighborhoods that have around parks, bus station and Restaurant places.
- II: Neighborhoods that have around shopping mall, bus station and convenience store
- III: Neighborhoods that have around Plaza, hotels, bar and stores
- IV: Neighborhood that has nothing as per requirement

My friend can now decide and move from Brooklyn, New York to Toronto, Canada using this system to get a notion or idea about what is the best suitable place for him.