

Capstone proj-IBM Toronto_Neighborhoods

July 8, 2020

0.1 Opening a high end steakhouse restaurant in

1 Moore park, Central Toronto

Toronto is the capital of the province of Ontario and it is a major Canadian city and it is located right next to Lake Ontario's. It's a very dynamic metropolitan area with a population of 2.7 million as of 2016. Toronto is an urban city with diverse population and it is an important destination for immigrants to Canada as well. More than 50 percent of residents belong to minority population group which are pretty much are spread out across the city. Toronto has over 200 distinct ethnic origins and while the majority of the population in the city speak English as their primary language, over 160 languages are spoken in the city. The city has 10 boroughs and about 103 neighborhoods.

```
In [196]: toronto_df.head(10)
```

```
Out[196]:
```

| | Borough | Neighborhood | Latitude | \ |
|---|------------------|---|-----------|---|
| 0 | North York | Parkwoods | 43.753259 | |
| 1 | North York | Victoria Village | 43.725882 | |
| 2 | Downtown Toronto | Regent Park, Harbourfront | 43.654260 | |
| 3 | North York | Lawrence Manor, Lawrence Heights | 43.718518 | |
| 4 | Downtown Toronto | Queen's Park, Ontario Provincial Government | 43.662301 | |
| 5 | Etobicoke | Islington Avenue, Humber Valley Village | 43.667856 | |
| 6 | Scarborough | Malvern, Rouge | 43.806686 | |
| 7 | North York | Don Mills | 43.745906 | |
| 8 | East York | Parkview Hill, Woodbine Gardens | 43.706397 | |
| 9 | Downtown Toronto | Garden District, Ryerson | 43.657162 | |

| | Longitude |
|---|------------|
| 0 | -79.329656 |
| 1 | -79.315572 |
| 2 | -79.360636 |
| 3 | -79.464763 |
| 4 | -79.389494 |
| 5 | -79.532242 |
| 6 | -79.194353 |
| 7 | -79.352188 |
| 8 | -79.309937 |
| 9 | -79.378937 |

```
In [197]: map_toronto
```

```
Out[197]: <folium.folium.Map at 0x2af5ab7c6a0>
```

1.0.1 Business problem and Methodology

Finding the right location is a very important factor to open a restaurant. Since in this project I am targeting Toronto City I need to analyze a borough that suits best for to open a high-end steak-house restaurant. This type of restaurant can be close to city center or suburb but in observation it is best open this type of restaurant close to shopping center, mall or city center. Since it would be a high-end restaurant it must be located in high income class neighborhood. Moore park in Central Toronto seems to be more appropriate neighborhood for this type restaurant.

```
In [198]: Central_Toronto
```

```
Out[198]:
```

| | Borough | Neighborhood \ |
|---|-----------------|---|
| 0 | Central Toronto | Lawrence Park |
| 1 | Central Toronto | Roselawn |
| 2 | Central Toronto | Davisville North |
| 3 | Central Toronto | Forest Hill North & West, Forest Hill Road Park |
| 4 | Central Toronto | North Toronto West, Lawrence Park |
| 5 | Central Toronto | The Annex, North Midtown, Yorkville |
| 6 | Central Toronto | Davisville |
| 7 | Central Toronto | Moore Park, Summerhill East |
| 8 | Central Toronto | Summerhill West, Rathnelly, South Hill, Forest... |

| | Latitude | Longitude |
|---|-----------|------------|
| 0 | 43.728020 | -79.388790 |
| 1 | 43.711695 | -79.416936 |
| 2 | 43.712751 | -79.390197 |
| 3 | 43.696948 | -79.411307 |
| 4 | 43.715383 | -79.405678 |
| 5 | 43.672710 | -79.405678 |
| 6 | 43.704324 | -79.388790 |
| 7 | 43.689574 | -79.383160 |
| 8 | 43.686412 | -79.400049 |

```
In [199]: toronto_df1.head()
```

```
Out[199]:
```

| | index | Borough | Neighborhood | Population | Average_Income \ |
|---|-------|------------------|---------------|------------|------------------|
| 0 | 16 | Central Toronto | Lawrence Park | 6653 | 214110.0 |
| 1 | 21 | Downtown Toronto | Rosedale | 7672 | 213941.0 |
| 2 | 7 | East York | Leaside | 13876 | 82670.0 |
| 3 | 5 | East Toronto | The Beaches | 20416 | 67536.0 |
| 4 | 20 | Central Toronto | Davisville | 23727 | 55735.0 |

| | Latitude | Longitude |
|---|-----------|------------|
| 0 | 43.728020 | -79.388790 |
| 1 | 43.679563 | -79.377529 |

```

2  43.709060 -79.363452
3  43.676357 -79.293031
4  43.704324 -79.388790

```

```
In [200]: map_Central_Toronto
```

```
Out[200]: <folium.folium.Map at 0x2af5b134278>
```

1.0.2 Exploring and cleaning data sets city of Toronto

On this project I am using three data sets, the Boroughs data, Latitude and Longitude, and population data. After combining the three data sets on unique value column now we can analyze the combined data and to find where could make more sense to open a high-end steak house.

1.0.3 Result and conclusion section

Based on analysis below we can see the frequency of restaurants in Moore Park in Toronto is about 0.25 and it is a neighborhood with high income household. Additionally, on the fifth cluster we can see that the restaurants are first most common venues in the area. Thus, there would be a decent competition in this neighborhood but quality and customer happiness will win always.

```
In [201]: num_top_venues = 5
```

```

for Neighborhood in Central_Toronto_grouped['Neighborhood']:
    print("----"+Neighborhood+"----")
    temp = Central_Toronto_grouped[Central_Toronto_grouped['Neighborhood'] == Neighborhood]
    temp.columns = ['venue', 'freq']
    temp = temp.iloc[1:]
    temp['freq'] = temp['freq'].astype(float)
    temp = temp.round({'freq': 2})
    print(temp.sort_values('freq', ascending=False).reset_index(drop=True).head(num_top_venues))
    print('\n')

```

```
----Davisville----
```

| | venue | freq |
|---|--------------------|------|
| 0 | Dessert Shop | 0.10 |
| 1 | Coffee Shop | 0.07 |
| 2 | Italian Restaurant | 0.07 |
| 3 | Sandwich Place | 0.07 |
| 4 | Café | 0.07 |

```
----Davisville North----
```

| | venue | freq |
|---|----------------------|------|
| 0 | Hotel | 0.12 |
| 1 | Gym / Fitness Center | 0.12 |
| 2 | Department Store | 0.12 |
| 3 | Food & Drink Shop | 0.12 |
| 4 | Pizza Place | 0.12 |

----Forest Hill North & West, Forest Hill Road Park----

| | venue | freq |
|---|---------------------|------|
| 0 | Jewelry Store | 0.25 |
| 1 | Trail | 0.25 |
| 2 | Mexican Restaurant | 0.25 |
| 3 | Sushi Restaurant | 0.25 |
| 4 | American Restaurant | 0.00 |

----Lawrence Park----

| | venue | freq |
|---|---------------------|------|
| 0 | Bus Line | 0.33 |
| 1 | Park | 0.33 |
| 2 | Swim School | 0.33 |
| 3 | American Restaurant | 0.00 |
| 4 | Salon / Barbershop | 0.00 |

----Moore Park, Summerhill East----

| | venue | freq |
|---|---------------------|------|
| 0 | Restaurant | 0.25 |
| 1 | Trail | 0.25 |
| 2 | Tennis Court | 0.25 |
| 3 | Park | 0.25 |
| 4 | American Restaurant | 0.00 |

----North Toronto West, Lawrence Park----

| | venue | freq |
|---|--------------------|------|
| 0 | Clothing Store | 0.11 |
| 1 | Coffee Shop | 0.11 |
| 2 | Yoga Studio | 0.05 |
| 3 | Cosmetics Shop | 0.05 |
| 4 | Mexican Restaurant | 0.05 |

----Roselawn----

| | venue | freq |
|---|---------------------|------|
| 0 | Home Service | 0.33 |
| 1 | Garden | 0.33 |
| 2 | Pool | 0.33 |
| 3 | American Restaurant | 0.00 |
| 4 | Indian Restaurant | 0.00 |

----Summerhill West, Rathnelly, South Hill, Forest Hill SE, Deer Park----

```

          venue freq
0          Pub 0.12
1    Coffee Shop 0.12
2 American Restaurant 0.06
3    Pizza Place 0.06
4    Bagel Shop 0.06

```

----The Annex, North Midtown, Yorkville----

```

          venue freq
0 Sandwich Place 0.14
1          Café 0.14
2    Coffee Shop 0.09
3 History Museum 0.05
4    Donut Shop 0.05

```

On the k-means cluster based anlysis we figured out that the fifth cluster of the Central which Moor Park has restuarant as most common venue and thus it the very favorable location to open our steak house house restuarant.

In [203]: *# Fifth cluster*

```
Central_Toronto_merged.loc[Central_Toronto_merged['Cluster Labels'] == 4, Central_Toronto_merged
```

```

Out[203]:
Neighborhood 1st Most Common Venue 2nd Most Common Venue \
7 Moore Park, Summerhill East Restaurant Trail

3rd Most Common Venue 4th Most Common Venue 5th Most Common Venue \
7 Park Tennis Court Fast Food Restaurant

6th Most Common Venue 7th Most Common Venue 8th Most Common Venue \
7 Dessert Shop Diner Donut Shop

9th Most Common Venue 10th Most Common Venue
7 Farmers Market Fried Chicken Joint

```

In [184]: *# Loading all the necessary labrararies*

```

import numpy as np

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 if you haven't comp
from geopy.geocoders import Nominatim # convert an address into latitude and longitu

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

#!conda install -c conda-forge folium=0.5.0 --yes # uncomment this line if you haven
import folium # map rendering library

print('Libraries imported.')

```

Libraries imported.

In [185]: # Uploading Boroughs with Postal Code

```

import numpy as np # library to handle data in a vectorized manner

import pandas as pd # library for data analysis

import requests
from bs4 import BeautifulSoup

res = requests.get("https://en.wikipedia.org/wiki/List_of_postal_codes_of_Canada:_M")

soup = BeautifulSoup(res.content, 'lxml')
table = soup.find_all('table')[0]
df = pd.read_html(str(table))

# Creating a dataframe object from listoftuples
df1 = pd.DataFrame(df[0])

df1.columns = df1.iloc[0]
df1 = df1[1:]
print("Shape of data", df1.shape, "\n")
df1.head(10)

```

Shape of data (180, 3)

```

Out[185]: 0  Postal Code      Borough      Neighborhood
          1      M1A      Not assigned      Not assigned

```

| | | | |
|----|-----|------------------|---|
| 2 | M2A | Not assigned | Not assigned |
| 3 | M3A | North York | Parkwoods |
| 4 | M4A | North York | Victoria Village |
| 5 | M5A | Downtown Toronto | Regent Park, Harbourfront |
| 6 | M6A | North York | Lawrence Manor, Lawrence Heights |
| 7 | M7A | Downtown Toronto | Queen's Park, Ontario Provincial Government |
| 8 | M8A | Not assigned | Not assigned |
| 9 | M9A | Etobicoke | Islington Avenue, Humber Valley Village |
| 10 | M1B | Scarborough | Malvern, Rouge |

In [186]: *#Getting only assigned Boroughs from the data*

```
df1 = df1[df1.Borough != 'Not assigned']

print("shape of the data after eliminating not assigned Boroughs", df1.shape, "\n")

df1.head(10)
```

shape of the data after eliminating not assigned Boroughs (103, 3)

```
Out[186]: 0 Postal Code      Borough      Neighborhood
3      M3A      North York      Parkwoods
4      M4A      North York      Victoria Village
5      M5A  Downtown Toronto      Regent Park, Harbourfront
6      M6A      North York      Lawrence Manor, Lawrence Heights
7      M7A  Downtown Toronto  Queen's Park, Ontario Provincial Government
9      M9A      Etobicoke      Islington Avenue, Humber Valley Village
10     M1B      Scarborough      Malvern, Rouge
12     M3B      North York      Don Mills
13     M4B      East York      Parkview Hill, Woodbine Gardens
14     M5B  Downtown Toronto      Garden District, Ryerson
```

In [187]: *#If a cell has a borough but a Not assigned neighborhood, then the neighborhood will*

```
df1['Neighborhood'].loc[(df1['Neighborhood'] == 'Not assigned')] = df1['Borough']
df1.head()
```

```
Out[187]: 0 Postal Code      Borough      Neighborhood
3      M3A      North York      Parkwoods
4      M4A      North York      Victoria Village
5      M5A  Downtown Toronto      Regent Park, Harbourfront
6      M6A      North York      Lawrence Manor, Lawrence Heights
7      M7A  Downtown Toronto  Queen's Park, Ontario Provincial Government
```

In [188]: *#Getting the Latitude and Longitude for each postal code*

```
import io
```

```
import requests
url_1 = "https://cocl.us/Geospatial_data"
ll_df = requests.get(url_1).text
ll_df = pd.read_csv(io.StringIO(ll_df))
ll_df.head()
```

```
Out[188]:
```

| | Postal Code | Latitude | Longitude |
|---|-------------|-----------|------------|
| 0 | M1B | 43.806686 | -79.194353 |
| 1 | M1C | 43.784535 | -79.160497 |
| 2 | M1E | 43.763573 | -79.188711 |
| 3 | M1G | 43.770992 | -79.216917 |
| 4 | M1H | 43.773136 | -79.239476 |

```
In [195]: # Combining Boroughs with Latitude and Longitude data frames
```

```
toronto_df = pd.merge(df1,ll_df, on = "Postal Code")

toronto_df = toronto_df[['Borough','Neighborhood', 'Latitude','Longitude']]
toronto_df.head(10)
```

```
Out[195]:
```

| | Borough | Neighborhood | Latitude | Longitude |
|---|------------------|---|-----------|------------|
| 0 | North York | Parkwoods | 43.753259 | -79.329656 |
| 1 | North York | Victoria Village | 43.725882 | -79.315572 |
| 2 | Downtown Toronto | Regent Park, Harbourfront | 43.654260 | -79.360636 |
| 3 | North York | Lawrence Manor, Lawrence Heights | 43.718518 | -79.464763 |
| 4 | Downtown Toronto | Queen's Park, Ontario Provincial Government | 43.662301 | -79.389494 |

```
In [194]: # To open a high end steak house we to add Average income column and pick one of the
```

```
toronto_pop = pd.read_csv(r"C:\Users\T\Downloads\toronto_pop.csv",encoding= 'unicode')

toronto_pop =toronto_pop.rename(columns = {'Name':'Neighborhood','Average Income':'Average_Income'})

toronto_df1 = pd.merge(toronto_df,toronto_pop, on = "Neighborhood")
toronto_df1 = toronto_df1[['Borough','Neighborhood','Population','Average_Income','Latitude','Longitude']]
toronto_df1 = toronto_df1.sort_values(by = ['Average_Income'], ascending=False, na_position='last')

toronto_df1.head()
```

```
Out[194]:
```

| | index | Borough | Neighborhood | Population | Average_Income | Latitude | Longitude |
|---|-------|-----------------|---------------|------------|----------------|-----------|------------|
| 0 | 16 | Central Toronto | Lawrence Park | 6653 | 214110.0 | 43.662301 | -79.389494 |

| | | | | | |
|---|----|------------------|-------------|-------|----------|
| 1 | 21 | Downtown Toronto | Rosedale | 7672 | 213941.0 |
| 2 | 7 | East York | Leaside | 13876 | 82670.0 |
| 3 | 5 | East Toronto | The Beaches | 20416 | 67536.0 |
| 4 | 20 | Central Toronto | Davisville | 23727 | 55735.0 |

| | Latitude | Longitude |
|---|-----------|------------|
| 0 | 43.728020 | -79.388790 |
| 1 | 43.679563 | -79.377529 |
| 2 | 43.709060 | -79.363452 |
| 3 | 43.676357 | -79.293031 |
| 4 | 43.704324 | -79.388790 |

```
In [191]: print('The dataframe has {} boroughs and {} neighborhoods.'.format(
            len(toronto_df['Borough'].unique()),
            toronto_df.shape[0]
        )
    )
```

The dataframe has 10 boroughs and 103 neighborhoods.

```
In [192]: address = 'Toronto, Canada'
```

```
geolocator = Nominatim(user_agent="ny_explorer")
location = geolocator.geocode(address)
latitude = location.latitude
longitude = location.longitude
print('The geograpical coordinate of Toronto City are {}, {}'.format(latitude, longitude))
```

The geograpical coordinate of Toronto City are 43.6534817, -79.3839347.

```
In [193]: # Creating map of Toronto using folium library
```

```
map_toronto = folium.Map(location=[latitude, longitude], zoom_start=10)

# add markers to map
for lat, lng, borough, neighborhood in zip(toronto_df['Latitude'], toronto_df['Longitude'],
                                            toronto_df['Borough'], toronto_df['Neighborhood']):
    label = '{} {}'.format(neighborhood, borough)
    popup = folium.Popup(label, parse_html=True)
    folium.CircleMarker(
        [lat, lng],
        radius=5,
        popup=popup,
        color='blue',
        fill=True,
        fill_color='#3186cc',
        fill_opacity=0.7,
        parse_html=False).add_to(map_toronto)
```

```
map_toronto
```

```
Out[193]: <folium.folium.Map at 0x2af5ab7c6a0>
```

```
In [168]: # Exploring Central Toronto borough
```

```
b1 = 'Central Toronto' #'Central Toronto'
```

```
Central_Toronto = toronto_df[toronto_df['Borough'] == b1].reset_index(drop=True)
print(Central_Toronto.shape)
Central_Toronto
```

```
(9, 4)
```

```
Out[168]:
```

| | Borough | Neighborhood \ |
|---|-----------------|---|
| 0 | Central Toronto | Lawrence Park |
| 1 | Central Toronto | Roselawn |
| 2 | Central Toronto | Davisville North |
| 3 | Central Toronto | Forest Hill North & West, Forest Hill Road Park |
| 4 | Central Toronto | North Toronto West, Lawrence Park |
| 5 | Central Toronto | The Annex, North Midtown, Yorkville |
| 6 | Central Toronto | Davisville |
| 7 | Central Toronto | Moore Park, Summerhill East |
| 8 | Central Toronto | Summerhill West, Rathnelly, South Hill, Forest... |

| | Latitude | Longitude |
|---|-----------|------------|
| 0 | 43.728020 | -79.388790 |
| 1 | 43.711695 | -79.416936 |
| 2 | 43.712751 | -79.390197 |
| 3 | 43.696948 | -79.411307 |
| 4 | 43.715383 | -79.405678 |
| 5 | 43.672710 | -79.405678 |
| 6 | 43.704324 | -79.388790 |
| 7 | 43.689574 | -79.383160 |
| 8 | 43.686412 | -79.400049 |

```
In [129]: address = b1+', Toronto, Canada'
```

```
geolocator = Nominatim(user_agent="Toronto_explorer")
location = geolocator.geocode(address)
latitude = location.latitude
longitude = location.longitude
print('The geograpical coordinate of {} are {}, {}'.format(b1,latitude, longitude))
```

```
The geograpical coordinate of Central Toronto are 43.6534817, -79.3839347.
```

```
In [130]: # create map of Central_Toronto using latitude and longitude values
map_Central_Toronto = folium.Map(location=[latitude, longitude], zoom_start=11)

# add markers to map
for lat, lng, label in zip(Central_Toronto['Latitude'], Central_Toronto['Longitude']):
    label = folium.Popup(label, parse_html=True)
    folium.CircleMarker(
        [lat, lng],
        radius=5,
        popup=label,
        color='blue',
        fill=True,
        fill_color='#3186cc',
        fill_opacity=0.7,
        parse_html=False).add_to(map_Central_Toronto)

map_Central_Toronto
```

```
Out[130]: <folium.folium.Map at 0x2af5b134278>
```

```
In [131]: # Searching for steakhouse in the neighborhood within 1000 meter radios
```

```
search_query = 'Steakhouse'
radius = 1000
LIMIT = 30
print(search_query)
```

```
Steakhouse
```

```
In [132]: # Credentials for Foursquare website
```

```
CLIENT_ID = 'ABRUHMRTPLBF4BTPZOHADD5SDSUGBUOSSS2SYSPSH2SYLYZU' # your Foursquare ID
CLIENT_SECRET = 'BVLKZZIDZUNVRVKUT5UZ3DJ13BV300PXOCAR01CREFTMI5CI' # your Foursquare
VERSION = '20180605' # Foursquare API version

print('Your credentials:')
print('CLIENT_ID: ' + CLIENT_ID)
print('CLIENT_SECRET: ' + CLIENT_SECRET)
```

```
Your credentials:
```

```
CLIENT_ID: ABRUHMRTPLBF4BTPZOHADD5SDSUGBUOSSS2SYSPSH2SYLYZU
CLIENT_SECRET: BVLKZZIDZUNVRVKUT5UZ3DJ13BV300PXOCAR01CREFTMI5CI
```

```
In [133]: url = 'https://api.foursquare.com/v2/venues/explore?&client_id={}&client_secret={}&v=
            CLIENT_ID,
            CLIENT_SECRET,
            VERSION,
```

```

        lat,
        lng,
        radius,
        LIMIT)

url

Out[133]: 'https://api.foursquare.com/v2/venues/explore?&client_id=ABRUHMRTPLBF4BTPZOHADD5SDSU

In [171]: results = requests.get(url).json()
          #results

In [135]: # function that extracts the category of the venue
          def get_category_type(row):
              try:
                  categories_list = row['categories']
              except:
                  categories_list = row['venue.categories']

              if len(categories_list) == 0:
                  return None
              else:
                  return categories_list[0]['name']

In [136]: # Searching for near by venues

venues = results['response']['groups'][0]['items']

nearby_venues = json_normalize(venues) # flatten JSON

# filter columns
filtered_columns = ['venue.name', 'venue.categories', 'venue.location.lat', 'venue.l
nearby_venues = nearby_venues.loc[:, filtered_columns]

# filter the category for each row
nearby_venues['venue.categories'] = nearby_venues.apply(get_category_type, axis=1)

# clean columns
nearby_venues.columns = [col.split(".")[1] for col in nearby_venues.columns]

nearby_venues.head()

Out[136]:
```

| | name | categories | lat | lng |
|---|----------------------|-------------------|-----------|------------|
| 0 | The Market By Longos | Supermarket | 43.686711 | -79.399536 |
| 1 | LCBO | Liquor Store | 43.686991 | -79.399238 |
| 2 | The Bagel House | Bagel Shop | 43.687374 | -79.393696 |
| 3 | Scaramouche | French Restaurant | 43.681293 | -79.399492 |
| 4 | DAVIDsTEA | Tea Room | 43.688421 | -79.394385 |

```

In [180]: def getNearbyVenues(names, latitudes, longitudes, radius=1000):

    venues_list=[]
    for name, lat, lng in zip(names, latitudes, longitudes):
        print(name)

        # create the API request URL
        url = 'https://api.foursquare.com/v2/venues/explore?&client_id={}&client_secret={}&v={}&lat={}&lng={}&radius={}&limit={}'
        CLIENT_ID,
        CLIENT_SECRET,
        VERSION,
        lat,
        lng,
        radius,
        LIMIT)

        # make the GET request
        results = requests.get(url).json()["response"]["groups"][0]["items"]

        # return only relevant information for each nearby venue
        venues_list.append([
            name,
            lat,
            lng,
            v['venue']['name'],
            v['venue']['location']['lat'],
            v['venue']['location']['lng'],
            v['venue']['categories'][0]['name']) for v in results])

    nearby_venues = pd.DataFrame([item for venue_list in venues_list for item in venue_list])
    nearby_venues.columns = ['Neighborhood',
                            'Neighborhood Latitude',
                            'Neighborhood Longitude',
                            'Venue',
                            'Venue Latitude',
                            'Venue Longitude',
                            'Venue Category']

    return(nearby_venues)

```

```

In [181]: # Listing all the neighborhoods for Central Toronto

```

```

Central_Toronto_venues = getNearbyVenues(names=Central_Toronto['Neighborhood'],
                                          latitudes=Central_Toronto['Latitude'],
                                          longitudes=Central_Toronto['Longitude']
                                          )

```

Lawrence Park
Roselawn

Davisville North
 Forest Hill North & West, Forest Hill Road Park
 North Toronto West, Lawrence Park
 The Annex, North Midtown, Yorkville
 Davisville
 Moore Park, Summerhill East
 Summerhill West, Rathnelly, South Hill, Forest Hill SE, Deer Park

In [182]: # list venues for each neighborhood for Central Toronto

```
print(Central_Toronto_venues.shape)
Central_Toronto_venues.head()
```

(238, 7)

```
Out[182]:
```

| | Neighborhood | Neighborhood Latitude | Neighborhood Longitude | \ |
|---|---------------|-----------------------|------------------------|---|
| 0 | Lawrence Park | 43.72802 | -79.38879 | |
| 1 | Lawrence Park | 43.72802 | -79.38879 | |
| 2 | Lawrence Park | 43.72802 | -79.38879 | |
| 3 | Lawrence Park | 43.72802 | -79.38879 | |
| 4 | Lawrence Park | 43.72802 | -79.38879 | |

| | Venue | Venue Latitude | Venue Longitude | Venue Category |
|---|----------------------|----------------|-----------------|----------------------|
| 0 | Lawrence Park Ravine | 43.726963 | -79.394382 | Park |
| 1 | Granite Club | 43.733043 | -79.381986 | Gym / Fitness Center |
| 2 | Tim Hortons | 43.727324 | -79.379563 | Coffee Shop |
| 3 | Glendon Bookstore | 43.727024 | -79.378976 | Bookstore |
| 4 | Glendon Forest | 43.727226 | -79.378413 | Trail |

In [183]: # Grouping the venues for each neighborhood

```
Central_Toronto_venues.groupby('Neighborhood').count()
```

```
Out[183]:
```

| | Neighborhood | Neighborhood Latitude | \ |
|--|--|-----------------------|----|
| | Davisville | | 30 |
| | Davisville North | | 30 |
| | Forest Hill North & West, Forest Hill Road Park | | 30 |
| | Lawrence Park | | 8 |
| | Moore Park, Summerhill East | | 30 |
| | North Toronto West, Lawrence Park | | 30 |
| | Roselawn | | 20 |
| | Summerhill West, Rathnelly, South Hill, Forest ... | | 30 |
| | The Annex, North Midtown, Yorkville | | 30 |

| | Neighborhood Longitude | \ |
|--|------------------------|---|
|--|------------------------|---|

| | |
|--|----|
| Neighborhood | |
| Davisville | 30 |
| Davisville North | 30 |
| Forest Hill North & West, Forest Hill Road Park | 30 |
| Lawrence Park | 8 |
| Moore Park, Summerhill East | 30 |
| North Toronto West, Lawrence Park | 30 |
| Roselawn | 20 |
| Summerhill West, Rathnelly, South Hill, Forest ... | 30 |
| The Annex, North Midtown, Yorkville | 30 |

| | Venue | Venue Latitude \ |
|--|-------|------------------|
| Neighborhood | | |
| Davisville | 30 | 30 |
| Davisville North | 30 | 30 |
| Forest Hill North & West, Forest Hill Road Park | 30 | 30 |
| Lawrence Park | 8 | 8 |
| Moore Park, Summerhill East | 30 | 30 |
| North Toronto West, Lawrence Park | 30 | 30 |
| Roselawn | 20 | 20 |
| Summerhill West, Rathnelly, South Hill, Forest ... | 30 | 30 |
| The Annex, North Midtown, Yorkville | 30 | 30 |

| | Venue Longitude \ |
|--|-------------------|
| Neighborhood | |
| Davisville | 30 |
| Davisville North | 30 |
| Forest Hill North & West, Forest Hill Road Park | 30 |
| Lawrence Park | 8 |
| Moore Park, Summerhill East | 30 |
| North Toronto West, Lawrence Park | 30 |
| Roselawn | 20 |
| Summerhill West, Rathnelly, South Hill, Forest ... | 30 |
| The Annex, North Midtown, Yorkville | 30 |

| | Venue Category |
|--|----------------|
| Neighborhood | |
| Davisville | 30 |
| Davisville North | 30 |
| Forest Hill North & West, Forest Hill Road Park | 30 |
| Lawrence Park | 8 |
| Moore Park, Summerhill East | 30 |
| North Toronto West, Lawrence Park | 30 |
| Roselawn | 20 |
| Summerhill West, Rathnelly, South Hill, Forest ... | 30 |
| The Annex, North Midtown, Yorkville | 30 |

```
In [149]: print('There are {} uniques categories.'.format(len(Central_Toronto_venues['Venue Ca
```

There are 62 uniques categories.

```
In [150]: # one hot encoding
```

```
Central_Toronto_onehot = pd.get_dummies(Central_Toronto_venues[['Venue Category']],
```

```
# add neighborhood column back to dataframe
```

```
Central_Toronto_onehot['Neighborhood'] = Central_Toronto_venues['Neighborhood']
```

```
# move neighborhood column to the first column
```

```
Central_Toronto_onehot = Central_Toronto_onehot.set_index('Neighborhood').reset_index
```

```
Central_Toronto_onehot.head(20)
```

```
Out[150]:
```

| | Neighborhood | American Restaurant \ |
|----|---|-----------------------|
| 0 | Lawrence Park | 0 |
| 1 | Lawrence Park | 0 |
| 2 | Lawrence Park | 0 |
| 3 | Roselawn | 0 |
| 4 | Roselawn | 0 |
| 5 | Roselawn | 0 |
| 6 | Davisville North | 0 |
| 7 | Davisville North | 0 |
| 8 | Davisville North | 0 |
| 9 | Davisville North | 0 |
| 10 | Davisville North | 0 |
| 11 | Davisville North | 0 |
| 12 | Davisville North | 0 |
| 13 | Davisville North | 0 |
| 14 | Forest Hill North & West, Forest Hill Road Park | 0 |
| 15 | Forest Hill North & West, Forest Hill Road Park | 0 |
| 16 | Forest Hill North & West, Forest Hill Road Park | 0 |
| 17 | Forest Hill North & West, Forest Hill Road Park | 0 |
| 18 | North Toronto West, Lawrence Park | 0 |
| 19 | North Toronto West, Lawrence Park | 0 |

| | BBQ Joint | Bagel Shop | Bank | Breakfast Spot | Brewery | Burger Joint \ |
|---|-----------|------------|------|----------------|---------|----------------|
| 0 | 0 | 0 | 0 | 0 | 0 | 0 |
| 1 | 0 | 0 | 0 | 0 | 0 | 0 |
| 2 | 0 | 0 | 0 | 0 | 0 | 0 |
| 3 | 0 | 0 | 0 | 0 | 0 | 0 |
| 4 | 0 | 0 | 0 | 0 | 0 | 0 |
| 5 | 0 | 0 | 0 | 0 | 0 | 0 |
| 6 | 0 | 0 | 0 | 0 | 0 | 0 |
| 7 | 0 | 0 | 0 | 0 | 0 | 0 |
| 8 | 0 | 0 | 0 | 1 | 0 | 0 |
| 9 | 0 | 0 | 0 | 0 | 0 | 0 |

| | | | | | | |
|----|---|---|---|---|---|---|
| 10 | 0 | 0 | 0 | 0 | 0 | 0 |
| 11 | 0 | 0 | 0 | 0 | 0 | 0 |
| 12 | 0 | 0 | 0 | 0 | 0 | 0 |
| 13 | 0 | 0 | 0 | 0 | 0 | 0 |
| 14 | 0 | 0 | 0 | 0 | 0 | 0 |
| 15 | 0 | 0 | 0 | 0 | 0 | 0 |
| 16 | 0 | 0 | 0 | 0 | 0 | 0 |
| 17 | 0 | 0 | 0 | 0 | 0 | 0 |
| 18 | 0 | 0 | 0 | 0 | 0 | 0 |
| 19 | 0 | 0 | 0 | 0 | 0 | 0 |

| | Bus Line | Café | Cheese Shop | Chinese Restaurant | Clothing Store | \ |
|----|----------|------|-------------|--------------------|----------------|---|
| 0 | 0 | 0 | 0 | 0 | 0 | |
| 1 | 0 | 0 | 0 | 0 | 0 | |
| 2 | 1 | 0 | 0 | 0 | 0 | |
| 3 | 0 | 0 | 0 | 0 | 0 | |
| 4 | 0 | 0 | 0 | 0 | 0 | |
| 5 | 0 | 0 | 0 | 0 | 0 | |
| 6 | 0 | 0 | 0 | 0 | 0 | |
| 7 | 0 | 0 | 0 | 0 | 0 | |
| 8 | 0 | 0 | 0 | 0 | 0 | |
| 9 | 0 | 0 | 0 | 0 | 0 | |
| 10 | 0 | 0 | 0 | 0 | 0 | |
| 11 | 0 | 0 | 0 | 0 | 0 | |
| 12 | 0 | 0 | 0 | 0 | 0 | |
| 13 | 0 | 0 | 0 | 0 | 0 | |
| 14 | 0 | 0 | 0 | 0 | 0 | |
| 15 | 0 | 0 | 0 | 0 | 0 | |
| 16 | 0 | 0 | 0 | 0 | 0 | |
| 17 | 0 | 0 | 0 | 0 | 0 | |
| 18 | 0 | 0 | 0 | 0 | 0 | |
| 19 | 0 | 0 | 0 | 0 | 0 | |

| | Coffee Shop | Convenience Store | Cosmetics Shop | Department Store | \ |
|----|-------------|-------------------|----------------|------------------|---|
| 0 | 0 | 0 | 0 | 0 | |
| 1 | 0 | 0 | 0 | 0 | |
| 2 | 0 | 0 | 0 | 0 | |
| 3 | 0 | 0 | 0 | 0 | |
| 4 | 0 | 0 | 0 | 0 | |
| 5 | 0 | 0 | 0 | 0 | |
| 6 | 0 | 0 | 0 | 0 | |
| 7 | 0 | 0 | 0 | 0 | |
| 8 | 0 | 0 | 0 | 0 | |
| 9 | 0 | 0 | 0 | 0 | 1 |
| 10 | 0 | 0 | 0 | 0 | 0 |
| 11 | 0 | 0 | 0 | 0 | 0 |
| 12 | 0 | 0 | 0 | 0 | 0 |
| 13 | 0 | 0 | 0 | 0 | 0 |

| | | | | |
|----|---|---|---|---|
| 14 | 0 | 0 | 0 | 0 |
| 15 | 0 | 0 | 0 | 0 |
| 16 | 0 | 0 | 0 | 0 |
| 17 | 0 | 0 | 0 | 0 |
| 18 | 0 | 0 | 0 | 0 |
| 19 | 1 | 0 | 0 | 0 |

| | Dessert Shop | Diner | Donut Shop | Farmers Market | Fast Food Restaurant | \ |
|----|--------------|-------|------------|----------------|----------------------|---|
| 0 | 0 | 0 | 0 | 0 | 0 | |
| 1 | 0 | 0 | 0 | 0 | 0 | |
| 2 | 0 | 0 | 0 | 0 | 0 | |
| 3 | 0 | 0 | 0 | 0 | 0 | |
| 4 | 0 | 0 | 0 | 0 | 0 | |
| 5 | 0 | 0 | 0 | 0 | 0 | |
| 6 | 0 | 0 | 0 | 0 | 0 | |
| 7 | 0 | 0 | 0 | 0 | 0 | |
| 8 | 0 | 0 | 0 | 0 | 0 | |
| 9 | 0 | 0 | 0 | 0 | 0 | |
| 10 | 0 | 0 | 0 | 0 | 0 | |
| 11 | 0 | 0 | 0 | 0 | 0 | |
| 12 | 0 | 0 | 0 | 0 | 0 | |
| 13 | 0 | 0 | 0 | 0 | 0 | |
| 14 | 0 | 0 | 0 | 0 | 0 | |
| 15 | 0 | 0 | 0 | 0 | 0 | |
| 16 | 0 | 0 | 0 | 0 | 0 | |
| 17 | 0 | 0 | 0 | 0 | 0 | |
| 18 | 0 | 0 | 0 | 0 | 0 | |
| 19 | 0 | 0 | 0 | 0 | 0 | |

| | Food & Drink Shop | Fried Chicken Joint | Garden | Gas Station | Gourmet Shop | \ |
|----|-------------------|---------------------|--------|-------------|--------------|---|
| 0 | 0 | 0 | 0 | 0 | 0 | |
| 1 | 0 | 0 | 0 | 0 | 0 | |
| 2 | 0 | 0 | 0 | 0 | 0 | |
| 3 | 0 | 0 | 0 | 0 | 0 | |
| 4 | 0 | 0 | 1 | 0 | 0 | |
| 5 | 0 | 0 | 0 | 0 | 0 | |
| 6 | 0 | 0 | 0 | 0 | 0 | |
| 7 | 1 | 0 | 0 | 0 | 0 | |
| 8 | 0 | 0 | 0 | 0 | 0 | |
| 9 | 0 | 0 | 0 | 0 | 0 | |
| 10 | 0 | 0 | 0 | 0 | 0 | |
| 11 | 0 | 0 | 0 | 0 | 0 | |
| 12 | 0 | 0 | 0 | 0 | 0 | |
| 13 | 0 | 0 | 0 | 0 | 0 | |
| 14 | 0 | 0 | 0 | 0 | 0 | |
| 15 | 0 | 0 | 0 | 0 | 0 | |
| 16 | 0 | 0 | 0 | 0 | 0 | |
| 17 | 0 | 0 | 0 | 0 | 0 | |

| | | | | | |
|----|---|---|---|---|---|
| 18 | 0 | 0 | 0 | 0 | 0 |
| 19 | 0 | 0 | 0 | 0 | 0 |

| | Greek Restaurant | Gym | Gym / Fitness Center | History Museum | Home Service \ |
|----|------------------|-----|----------------------|----------------|----------------|
| 0 | 0 | 0 | 0 | 0 | 0 |
| 1 | 0 | 0 | 0 | 0 | 0 |
| 2 | 0 | 0 | 0 | 0 | 0 |
| 3 | 0 | 0 | 0 | 0 | 1 |
| 4 | 0 | 0 | 0 | 0 | 0 |
| 5 | 0 | 0 | 0 | 0 | 0 |
| 6 | 0 | 0 | 0 | 0 | 0 |
| 7 | 0 | 0 | 0 | 0 | 0 |
| 8 | 0 | 0 | 0 | 0 | 0 |
| 9 | 0 | 0 | 0 | 0 | 0 |
| 10 | 0 | 0 | 0 | 0 | 0 |
| 11 | 0 | 0 | 0 | 0 | 0 |
| 12 | 0 | 0 | 1 | 0 | 0 |
| 13 | 0 | 0 | 0 | 0 | 0 |
| 14 | 0 | 0 | 0 | 0 | 0 |
| 15 | 0 | 0 | 0 | 0 | 0 |
| 16 | 0 | 0 | 0 | 0 | 0 |
| 17 | 0 | 0 | 0 | 0 | 0 |
| 18 | 0 | 0 | 0 | 0 | 0 |
| 19 | 0 | 0 | 0 | 0 | 0 |

| | Hotel | Indian Restaurant | Italian Restaurant | Jewelry Store \ |
|----|-------|-------------------|--------------------|-----------------|
| 0 | 0 | 0 | 0 | 0 |
| 1 | 0 | 0 | 0 | 0 |
| 2 | 0 | 0 | 0 | 0 |
| 3 | 0 | 0 | 0 | 0 |
| 4 | 0 | 0 | 0 | 0 |
| 5 | 0 | 0 | 0 | 0 |
| 6 | 0 | 0 | 0 | 0 |
| 7 | 0 | 0 | 0 | 0 |
| 8 | 0 | 0 | 0 | 0 |
| 9 | 0 | 0 | 0 | 0 |
| 10 | 1 | 0 | 0 | 0 |
| 11 | 0 | 0 | 0 | 0 |
| 12 | 0 | 0 | 0 | 0 |
| 13 | 0 | 0 | 0 | 0 |
| 14 | 0 | 0 | 0 | 0 |
| 15 | 0 | 0 | 0 | 0 |
| 16 | 0 | 0 | 0 | 0 |
| 17 | 0 | 0 | 0 | 1 |
| 18 | 0 | 0 | 0 | 0 |
| 19 | 0 | 0 | 0 | 0 |

| Light Rail Station | Liquor Store | Mexican Restaurant \ |
|--------------------|--------------|----------------------|
|--------------------|--------------|----------------------|

| | | | |
|----|---|---|---|
| 0 | 0 | 0 | 0 |
| 1 | 0 | 0 | 0 |
| 2 | 0 | 0 | 0 |
| 3 | 0 | 0 | 0 |
| 4 | 0 | 0 | 0 |
| 5 | 0 | 0 | 0 |
| 6 | 0 | 0 | 0 |
| 7 | 0 | 0 | 0 |
| 8 | 0 | 0 | 0 |
| 9 | 0 | 0 | 0 |
| 10 | 0 | 0 | 0 |
| 11 | 0 | 0 | 0 |
| 12 | 0 | 0 | 0 |
| 13 | 0 | 0 | 0 |
| 14 | 0 | 0 | 0 |
| 15 | 0 | 0 | 0 |
| 16 | 0 | 0 | 1 |
| 17 | 0 | 0 | 0 |
| 18 | 0 | 0 | 0 |
| 19 | 0 | 0 | 0 |

| | Middle Eastern Restaurant | Park | Pharmacy | Pizza Place | Pool | Pub | \ |
|----|---------------------------|------|----------|-------------|------|-----|---|
| 0 | 0 | 1 | 0 | 0 | 0 | 0 | |
| 1 | 0 | 0 | 0 | 0 | 0 | 0 | |
| 2 | 0 | 0 | 0 | 0 | 0 | 0 | |
| 3 | 0 | 0 | 0 | 0 | 0 | 0 | |
| 4 | 0 | 0 | 0 | 0 | 0 | 0 | |
| 5 | 0 | 0 | 0 | 0 | 1 | 0 | |
| 6 | 0 | 1 | 0 | 0 | 0 | 0 | |
| 7 | 0 | 0 | 0 | 0 | 0 | 0 | |
| 8 | 0 | 0 | 0 | 0 | 0 | 0 | |
| 9 | 0 | 0 | 0 | 0 | 0 | 0 | |
| 10 | 0 | 0 | 0 | 0 | 0 | 0 | |
| 11 | 0 | 0 | 0 | 0 | 0 | 0 | |
| 12 | 0 | 0 | 0 | 0 | 0 | 0 | |
| 13 | 0 | 0 | 0 | 1 | 0 | 0 | |
| 14 | 0 | 0 | 0 | 0 | 0 | 0 | |
| 15 | 0 | 0 | 0 | 0 | 0 | 0 | |
| 16 | 0 | 0 | 0 | 0 | 0 | 0 | |
| 17 | 0 | 0 | 0 | 0 | 0 | 0 | |
| 18 | 0 | 0 | 0 | 0 | 0 | 0 | |
| 19 | 0 | 0 | 0 | 0 | 0 | 0 | |

| | Rental Car Location | Restaurant | Salon / Barbershop | Sandwich Place | \ |
|---|---------------------|------------|--------------------|----------------|---|
| 0 | 0 | 0 | 0 | 0 | |
| 1 | 0 | 0 | 0 | 0 | |
| 2 | 0 | 0 | 0 | 0 | |
| 3 | 0 | 0 | 0 | 0 | |

| | | | | |
|----|---|---|---|---|
| 4 | 0 | 0 | 0 | 0 |
| 5 | 0 | 0 | 0 | 0 |
| 6 | 0 | 0 | 0 | 0 |
| 7 | 0 | 0 | 0 | 0 |
| 8 | 0 | 0 | 0 | 0 |
| 9 | 0 | 0 | 0 | 0 |
| 10 | 0 | 0 | 0 | 0 |
| 11 | 0 | 0 | 0 | 1 |
| 12 | 0 | 0 | 0 | 0 |
| 13 | 0 | 0 | 0 | 0 |
| 14 | 0 | 0 | 0 | 0 |
| 15 | 0 | 0 | 0 | 0 |
| 16 | 0 | 0 | 0 | 0 |
| 17 | 0 | 0 | 0 | 0 |
| 18 | 0 | 0 | 0 | 0 |
| 19 | 0 | 0 | 0 | 0 |

| | Seafood Restaurant | Spa | Sporting Goods Shop | Sports Bar | Supermarket | \ |
|----|--------------------|-----|---------------------|------------|-------------|---|
| 0 | 0 | 0 | 0 | 0 | 0 | |
| 1 | 0 | 0 | 0 | 0 | 0 | |
| 2 | 0 | 0 | 0 | 0 | 0 | |
| 3 | 0 | 0 | 0 | 0 | 0 | |
| 4 | 0 | 0 | 0 | 0 | 0 | |
| 5 | 0 | 0 | 0 | 0 | 0 | |
| 6 | 0 | 0 | 0 | 0 | 0 | |
| 7 | 0 | 0 | 0 | 0 | 0 | |
| 8 | 0 | 0 | 0 | 0 | 0 | |
| 9 | 0 | 0 | 0 | 0 | 0 | |
| 10 | 0 | 0 | 0 | 0 | 0 | |
| 11 | 0 | 0 | 0 | 0 | 0 | |
| 12 | 0 | 0 | 0 | 0 | 0 | |
| 13 | 0 | 0 | 0 | 0 | 0 | |
| 14 | 0 | 0 | 0 | 0 | 0 | |
| 15 | 0 | 0 | 0 | 0 | 0 | |
| 16 | 0 | 0 | 0 | 0 | 0 | |
| 17 | 0 | 0 | 0 | 0 | 0 | |
| 18 | 0 | 0 | 0 | 0 | 0 | |
| 19 | 0 | 0 | 0 | 0 | 0 | |

| | Sushi Restaurant | Swim School | Tennis Court | Thai Restaurant | \ |
|---|------------------|-------------|--------------|-----------------|---|
| 0 | 0 | 0 | 0 | 0 | |
| 1 | 0 | 1 | 0 | 0 | |
| 2 | 0 | 0 | 0 | 0 | |
| 3 | 0 | 0 | 0 | 0 | |
| 4 | 0 | 0 | 0 | 0 | |
| 5 | 0 | 0 | 0 | 0 | |
| 6 | 0 | 0 | 0 | 0 | |
| 7 | 0 | 0 | 0 | 0 | |

| | | | | |
|----|---|---|---|---|
| 8 | 0 | 0 | 0 | 0 |
| 9 | 0 | 0 | 0 | 0 |
| 10 | 0 | 0 | 0 | 0 |
| 11 | 0 | 0 | 0 | 0 |
| 12 | 0 | 0 | 0 | 0 |
| 13 | 0 | 0 | 0 | 0 |
| 14 | 0 | 0 | 0 | 0 |
| 15 | 1 | 0 | 0 | 0 |
| 16 | 0 | 0 | 0 | 0 |
| 17 | 0 | 0 | 0 | 0 |
| 18 | 0 | 0 | 0 | 0 |
| 19 | 0 | 0 | 0 | 0 |

| | Toy / Game Store | Trail | Vegetarian / Vegan Restaurant \ |
|----|------------------|-------|---------------------------------|
| 0 | 0 | 0 | 0 |
| 1 | 0 | 0 | 0 |
| 2 | 0 | 0 | 0 |
| 3 | 0 | 0 | 0 |
| 4 | 0 | 0 | 0 |
| 5 | 0 | 0 | 0 |
| 6 | 0 | 0 | 0 |
| 7 | 0 | 0 | 0 |
| 8 | 0 | 0 | 0 |
| 9 | 0 | 0 | 0 |
| 10 | 0 | 0 | 0 |
| 11 | 0 | 0 | 0 |
| 12 | 0 | 0 | 0 |
| 13 | 0 | 0 | 0 |
| 14 | 0 | 1 | 0 |
| 15 | 0 | 0 | 0 |
| 16 | 0 | 0 | 0 |
| 17 | 0 | 0 | 0 |
| 18 | 0 | 0 | 0 |
| 19 | 0 | 0 | 0 |

| | Vietnamese Restaurant | Yoga Studio |
|----|-----------------------|-------------|
| 0 | 0 | 0 |
| 1 | 0 | 0 |
| 2 | 0 | 0 |
| 3 | 0 | 0 |
| 4 | 0 | 0 |
| 5 | 0 | 0 |
| 6 | 0 | 0 |
| 7 | 0 | 0 |
| 8 | 0 | 0 |
| 9 | 0 | 0 |
| 10 | 0 | 0 |
| 11 | 0 | 0 |

| | | |
|----|---|---|
| 12 | 0 | 0 |
| 13 | 0 | 0 |
| 14 | 0 | 0 |
| 15 | 0 | 0 |
| 16 | 0 | 0 |
| 17 | 0 | 0 |
| 18 | 0 | 1 |
| 19 | 0 | 0 |

In [151]: Central_Toronto_onehot.shape

Out[151]: (109, 63)

In [152]: Central_Toronto_grouped = Central_Toronto_onehot.groupby('Neighborhood').mean().reset_index()
Central_Toronto_grouped

Out[152]:

| | Neighborhood | American Restaurant | \ |
|---|---|---------------------|---|
| 0 | Davisville | 0.0000 | |
| 1 | Davisville North | 0.0000 | |
| 2 | Forest Hill North & West, Forest Hill Road Park | 0.0000 | |
| 3 | Lawrence Park | 0.0000 | |
| 4 | Moore Park, Summerhill East | 0.0000 | |
| 5 | North Toronto West, Lawrence Park | 0.0000 | |
| 6 | Roselawn | 0.0000 | |
| 7 | Summerhill West, Rathnelly, South Hill, Forest... | 0.0625 | |
| 8 | The Annex, North Midtown, Yorkville | 0.0000 | |

| | BBQ Joint | Bagel Shop | Bank | Breakfast Spot | Brewery | Burger Joint | \ |
|---|-----------|------------|--------|----------------|----------|--------------|---|
| 0 | 0.000000 | 0.0000 | 0.0000 | 0.000 | 0.033333 | 0.000000 | |
| 1 | 0.000000 | 0.0000 | 0.0000 | 0.125 | 0.000000 | 0.000000 | |
| 2 | 0.000000 | 0.0000 | 0.0000 | 0.000 | 0.000000 | 0.000000 | |
| 3 | 0.000000 | 0.0000 | 0.0000 | 0.000 | 0.000000 | 0.000000 | |
| 4 | 0.000000 | 0.0000 | 0.0000 | 0.000 | 0.000000 | 0.000000 | |
| 5 | 0.000000 | 0.0000 | 0.0000 | 0.000 | 0.000000 | 0.000000 | |
| 6 | 0.000000 | 0.0000 | 0.0000 | 0.000 | 0.000000 | 0.000000 | |
| 7 | 0.000000 | 0.0625 | 0.0625 | 0.000 | 0.000000 | 0.000000 | |
| 8 | 0.045455 | 0.0000 | 0.0000 | 0.000 | 0.000000 | 0.045455 | |

| | Bus Line | Café | Cheese Shop | Chinese Restaurant | Clothing Store | \ |
|---|----------|----------|-------------|--------------------|----------------|---|
| 0 | 0.000000 | 0.066667 | 0.000000 | 0.000000 | 0.000000 | |
| 1 | 0.000000 | 0.000000 | 0.000000 | 0.000000 | 0.000000 | |
| 2 | 0.000000 | 0.000000 | 0.000000 | 0.000000 | 0.000000 | |
| 3 | 0.333333 | 0.000000 | 0.000000 | 0.000000 | 0.000000 | |
| 4 | 0.000000 | 0.000000 | 0.000000 | 0.000000 | 0.000000 | |
| 5 | 0.000000 | 0.052632 | 0.000000 | 0.052632 | 0.105263 | |
| 6 | 0.000000 | 0.000000 | 0.000000 | 0.000000 | 0.000000 | |
| 7 | 0.000000 | 0.000000 | 0.000000 | 0.000000 | 0.000000 | |
| 8 | 0.000000 | 0.136364 | 0.045455 | 0.000000 | 0.000000 | |

| | Coffee Shop | Convenience Store | Cosmetics Shop | Department Store | \ |
|---|-------------|-------------------|----------------|------------------|---|
| 0 | 0.066667 | 0.000000 | 0.000000 | 0.000 | |
| 1 | 0.000000 | 0.000000 | 0.000000 | 0.125 | |
| 2 | 0.000000 | 0.000000 | 0.000000 | 0.000 | |
| 3 | 0.000000 | 0.000000 | 0.000000 | 0.000 | |
| 4 | 0.000000 | 0.000000 | 0.000000 | 0.000 | |
| 5 | 0.105263 | 0.000000 | 0.052632 | 0.000 | |
| 6 | 0.000000 | 0.000000 | 0.000000 | 0.000 | |
| 7 | 0.125000 | 0.000000 | 0.000000 | 0.000 | |
| 8 | 0.090909 | 0.045455 | 0.000000 | 0.000 | |

| | Dessert Shop | Diner | Donut Shop | Farmers Market | Fast Food Restaurant | \ |
|---|--------------|----------|------------|----------------|----------------------|---|
| 0 | 0.1 | 0.033333 | 0.000000 | 0.033333 | 0.000000 | |
| 1 | 0.0 | 0.000000 | 0.000000 | 0.000000 | 0.000000 | |
| 2 | 0.0 | 0.000000 | 0.000000 | 0.000000 | 0.000000 | |
| 3 | 0.0 | 0.000000 | 0.000000 | 0.000000 | 0.000000 | |
| 4 | 0.0 | 0.000000 | 0.000000 | 0.000000 | 0.000000 | |
| 5 | 0.0 | 0.052632 | 0.000000 | 0.000000 | 0.052632 | |
| 6 | 0.0 | 0.000000 | 0.000000 | 0.000000 | 0.000000 | |
| 7 | 0.0 | 0.000000 | 0.000000 | 0.000000 | 0.000000 | |
| 8 | 0.0 | 0.000000 | 0.045455 | 0.000000 | 0.000000 | |

| | Food & Drink Shop | Fried Chicken Joint | Garden | Gas Station | \ |
|---|-------------------|---------------------|----------|-------------|---|
| 0 | 0.000 | 0.0000 | 0.000000 | 0.033333 | |
| 1 | 0.125 | 0.0000 | 0.000000 | 0.000000 | |
| 2 | 0.000 | 0.0000 | 0.000000 | 0.000000 | |
| 3 | 0.000 | 0.0000 | 0.000000 | 0.000000 | |
| 4 | 0.000 | 0.0000 | 0.000000 | 0.000000 | |
| 5 | 0.000 | 0.0000 | 0.000000 | 0.000000 | |
| 6 | 0.000 | 0.0000 | 0.333333 | 0.000000 | |
| 7 | 0.000 | 0.0625 | 0.000000 | 0.000000 | |
| 8 | 0.000 | 0.0000 | 0.000000 | 0.000000 | |

| | Gourmet Shop | Greek Restaurant | Gym | Gym / Fitness Center | \ |
|---|--------------|------------------|----------|----------------------|---|
| 0 | 0.033333 | 0.033333 | 0.066667 | 0.000000 | |
| 1 | 0.000000 | 0.000000 | 0.000000 | 0.125000 | |
| 2 | 0.000000 | 0.000000 | 0.000000 | 0.000000 | |
| 3 | 0.000000 | 0.000000 | 0.000000 | 0.000000 | |
| 4 | 0.000000 | 0.000000 | 0.000000 | 0.000000 | |
| 5 | 0.000000 | 0.000000 | 0.000000 | 0.052632 | |
| 6 | 0.000000 | 0.000000 | 0.000000 | 0.000000 | |
| 7 | 0.000000 | 0.000000 | 0.000000 | 0.000000 | |
| 8 | 0.000000 | 0.000000 | 0.000000 | 0.000000 | |

| | History Museum | Home Service | Hotel | Indian Restaurant | Italian Restaurant | \ |
|---|----------------|--------------|-------|-------------------|--------------------|---|
| 0 | 0.000000 | 0.000000 | 0.000 | 0.033333 | 0.066667 | |
| 1 | 0.000000 | 0.000000 | 0.125 | 0.000000 | 0.000000 | |
| 2 | 0.000000 | 0.000000 | 0.000 | 0.000000 | 0.000000 | |

| | | | | | |
|---|----------|----------|-------|----------|----------|
| 3 | 0.000000 | 0.000000 | 0.000 | 0.000000 | 0.000000 |
| 4 | 0.000000 | 0.000000 | 0.000 | 0.000000 | 0.000000 |
| 5 | 0.000000 | 0.000000 | 0.000 | 0.000000 | 0.000000 |
| 6 | 0.000000 | 0.333333 | 0.000 | 0.000000 | 0.000000 |
| 7 | 0.000000 | 0.000000 | 0.000 | 0.000000 | 0.000000 |
| 8 | 0.045455 | 0.000000 | 0.000 | 0.045455 | 0.000000 |

| | Jewelry Store | Light Rail Station | Liquor Store | Mexican Restaurant | \ |
|---|---------------|--------------------|--------------|--------------------|---|
| 0 | 0.00 | 0.0000 | 0.000000 | 0.000000 | |
| 1 | 0.00 | 0.0000 | 0.000000 | 0.000000 | |
| 2 | 0.25 | 0.0000 | 0.000000 | 0.250000 | |
| 3 | 0.00 | 0.0000 | 0.000000 | 0.000000 | |
| 4 | 0.00 | 0.0000 | 0.000000 | 0.000000 | |
| 5 | 0.00 | 0.0000 | 0.000000 | 0.052632 | |
| 6 | 0.00 | 0.0000 | 0.000000 | 0.000000 | |
| 7 | 0.00 | 0.0625 | 0.062500 | 0.000000 | |
| 8 | 0.00 | 0.0000 | 0.045455 | 0.000000 | |

| | Middle Eastern Restaurant | Park | Pharmacy | Pizza Place | Pool | \ |
|---|---------------------------|----------|----------|-------------|----------|---|
| 0 | 0.000000 | 0.033333 | 0.033333 | 0.066667 | 0.000000 | |
| 1 | 0.000000 | 0.125000 | 0.000000 | 0.125000 | 0.000000 | |
| 2 | 0.000000 | 0.000000 | 0.000000 | 0.000000 | 0.000000 | |
| 3 | 0.000000 | 0.333333 | 0.000000 | 0.000000 | 0.000000 | |
| 4 | 0.000000 | 0.250000 | 0.000000 | 0.000000 | 0.000000 | |
| 5 | 0.000000 | 0.052632 | 0.000000 | 0.000000 | 0.000000 | |
| 6 | 0.000000 | 0.000000 | 0.000000 | 0.000000 | 0.333333 | |
| 7 | 0.000000 | 0.000000 | 0.000000 | 0.062500 | 0.000000 | |
| 8 | 0.045455 | 0.045455 | 0.045455 | 0.045455 | 0.000000 | |

| | Pub | Rental Car Location | Restaurant | Salon / Barbershop | \ |
|---|----------|---------------------|------------|--------------------|---|
| 0 | 0.000000 | 0.000000 | 0.033333 | 0.000000 | |
| 1 | 0.000000 | 0.000000 | 0.000000 | 0.000000 | |
| 2 | 0.000000 | 0.000000 | 0.000000 | 0.000000 | |
| 3 | 0.000000 | 0.000000 | 0.000000 | 0.000000 | |
| 4 | 0.000000 | 0.000000 | 0.250000 | 0.000000 | |
| 5 | 0.000000 | 0.052632 | 0.052632 | 0.052632 | |
| 6 | 0.000000 | 0.000000 | 0.000000 | 0.000000 | |
| 7 | 0.125000 | 0.000000 | 0.062500 | 0.000000 | |
| 8 | 0.045455 | 0.000000 | 0.000000 | 0.000000 | |

| | Sandwich Place | Seafood Restaurant | Spa | Sporting Goods Shop | \ |
|---|----------------|--------------------|----------|---------------------|---|
| 0 | 0.066667 | 0.033333 | 0.000000 | 0.000000 | |
| 1 | 0.125000 | 0.000000 | 0.000000 | 0.000000 | |
| 2 | 0.000000 | 0.000000 | 0.000000 | 0.000000 | |
| 3 | 0.000000 | 0.000000 | 0.000000 | 0.000000 | |
| 4 | 0.000000 | 0.000000 | 0.000000 | 0.000000 | |
| 5 | 0.000000 | 0.052632 | 0.052632 | 0.052632 | |
| 6 | 0.000000 | 0.000000 | 0.000000 | 0.000000 | |

| | | | | |
|---|----------|----------|----------|----------|
| 7 | 0.000000 | 0.000000 | 0.000000 | 0.000000 |
| 8 | 0.136364 | 0.000000 | 0.000000 | 0.000000 |

| | Sports Bar | Supermarket | Sushi Restaurant | Swim School | Tennis Court \ |
|---|------------|-------------|------------------|-------------|----------------|
| 0 | 0.0000 | 0.0000 | 0.066667 | 0.000000 | 0.00 |
| 1 | 0.0000 | 0.0000 | 0.000000 | 0.000000 | 0.00 |
| 2 | 0.0000 | 0.0000 | 0.250000 | 0.000000 | 0.00 |
| 3 | 0.0000 | 0.0000 | 0.000000 | 0.333333 | 0.00 |
| 4 | 0.0000 | 0.0000 | 0.000000 | 0.000000 | 0.25 |
| 5 | 0.0000 | 0.0000 | 0.000000 | 0.000000 | 0.00 |
| 6 | 0.0000 | 0.0000 | 0.000000 | 0.000000 | 0.00 |
| 7 | 0.0625 | 0.0625 | 0.062500 | 0.000000 | 0.00 |
| 8 | 0.0000 | 0.0000 | 0.000000 | 0.000000 | 0.00 |

| | Thai Restaurant | Toy / Game Store | Trail | Vegetarian / Vegan Restaurant \ |
|---|-----------------|------------------|-------|---------------------------------|
| 0 | 0.033333 | 0.033333 | 0.00 | 0.000000 |
| 1 | 0.000000 | 0.000000 | 0.00 | 0.000000 |
| 2 | 0.000000 | 0.000000 | 0.25 | 0.000000 |
| 3 | 0.000000 | 0.000000 | 0.00 | 0.000000 |
| 4 | 0.000000 | 0.000000 | 0.25 | 0.000000 |
| 5 | 0.000000 | 0.000000 | 0.00 | 0.000000 |
| 6 | 0.000000 | 0.000000 | 0.00 | 0.000000 |
| 7 | 0.000000 | 0.000000 | 0.00 | 0.000000 |
| 8 | 0.000000 | 0.000000 | 0.00 | 0.045455 |

| | Vietnamese Restaurant | Yoga Studio |
|---|-----------------------|-------------|
| 0 | 0.0000 | 0.000000 |
| 1 | 0.0000 | 0.000000 |
| 2 | 0.0000 | 0.000000 |
| 3 | 0.0000 | 0.000000 |
| 4 | 0.0000 | 0.000000 |
| 5 | 0.0000 | 0.052632 |
| 6 | 0.0000 | 0.000000 |
| 7 | 0.0625 | 0.000000 |
| 8 | 0.0000 | 0.000000 |

```
In [153]: # On this analysis we can see the the first venue on Moore Park is restuarant so it is
# a restaurant in this neighborhood
```

```
num_top_venues = 5
```

```
for Neighborhood in Central_Toronto_grouped['Neighborhood']:
    print("----"+Neighborhood+"----")
    temp = Central_Toronto_grouped[Central_Toronto_grouped['Neighborhood'] == Neighborhood]
    temp.columns = ['venue', 'freq']
    temp = temp.iloc[1:]
    temp['freq'] = temp['freq'].astype(float)
    temp = temp.round({'freq': 2})
```

```
print(temp.sort_values('freq', ascending=False).reset_index(drop=True).head(num_t
print('\n')
```

----Davisville----

| | venue | freq |
|---|--------------------|------|
| 0 | Dessert Shop | 0.10 |
| 1 | Coffee Shop | 0.07 |
| 2 | Italian Restaurant | 0.07 |
| 3 | Sandwich Place | 0.07 |
| 4 | Café | 0.07 |

----Davisville North----

| | venue | freq |
|---|----------------------|------|
| 0 | Hotel | 0.12 |
| 1 | Gym / Fitness Center | 0.12 |
| 2 | Department Store | 0.12 |
| 3 | Food & Drink Shop | 0.12 |
| 4 | Pizza Place | 0.12 |

----Forest Hill North & West, Forest Hill Road Park----

| | venue | freq |
|---|---------------------|------|
| 0 | Jewelry Store | 0.25 |
| 1 | Trail | 0.25 |
| 2 | Mexican Restaurant | 0.25 |
| 3 | Sushi Restaurant | 0.25 |
| 4 | American Restaurant | 0.00 |

----Lawrence Park----

| | venue | freq |
|---|---------------------|------|
| 0 | Bus Line | 0.33 |
| 1 | Park | 0.33 |
| 2 | Swim School | 0.33 |
| 3 | American Restaurant | 0.00 |
| 4 | Salon / Barbershop | 0.00 |

----Moore Park, Summerhill East----

| | venue | freq |
|---|---------------------|------|
| 0 | Restaurant | 0.25 |
| 1 | Trail | 0.25 |
| 2 | Tennis Court | 0.25 |
| 3 | Park | 0.25 |
| 4 | American Restaurant | 0.00 |

----North Toronto West, Lawrence Park----

| | venue | freq |
|---|--------------------|------|
| 0 | Clothing Store | 0.11 |
| 1 | Coffee Shop | 0.11 |
| 2 | Yoga Studio | 0.05 |
| 3 | Cosmetics Shop | 0.05 |
| 4 | Mexican Restaurant | 0.05 |

----Roselawn----

| | venue | freq |
|---|---------------------|------|
| 0 | Home Service | 0.33 |
| 1 | Garden | 0.33 |
| 2 | Pool | 0.33 |
| 3 | American Restaurant | 0.00 |
| 4 | Indian Restaurant | 0.00 |

----Summerhill West, Rathnelly, South Hill, Forest Hill SE, Deer Park----

| | venue | freq |
|---|---------------------|------|
| 0 | Pub | 0.12 |
| 1 | Coffee Shop | 0.12 |
| 2 | American Restaurant | 0.06 |
| 3 | Pizza Place | 0.06 |
| 4 | Bagel Shop | 0.06 |

----The Annex, North Midtown, Yorkville----

| | venue | freq |
|---|----------------|------|
| 0 | Sandwich Place | 0.14 |
| 1 | Café | 0.14 |
| 2 | Coffee Shop | 0.09 |
| 3 | History Museum | 0.05 |
| 4 | Donut Shop | 0.05 |

```
In [154]: def return_most_common_venues(row, num_top_venues):  
           row_categories = row.iloc[1:]  
           row_categories_sorted = row_categories.sort_values(ascending=False)  
  
           return row_categories_sorted.index.values[0:num_top_venues]
```

```
In [155]: num_top_venues = 10
```

```
           indicators = ['st', 'nd', 'rd']
```

```

# create columns according to number of top venues
columns = ['Neighborhood']
for ind in np.arange(num_top_venues):
    try:
        columns.append('{} {} Most Common Venue'.format(ind+1, indicators[ind]))
    except:
        columns.append('{}th Most Common Venue'.format(ind+1))

# create a new dataframe
neighborhoods_venues_sorted = pd.DataFrame(columns=columns)
neighborhoods_venues_sorted['Neighborhood'] = Central_Toronto_grouped['Neighborhood']

for ind in np.arange(Central_Toronto_grouped.shape[0]):
    neighborhoods_venues_sorted.iloc[ind, 1:] = return_most_common_venues(Central_Toronto_grouped, ind+1, num_top_venues)

neighborhoods_venues_sorted.head(20)

```

```

Out[155]:

```

| | Neighborhood | 1st Most Common Venue | \ |
|---|---|-----------------------|---|
| 0 | Davisville | Dessert Shop | |
| 1 | Davisville North | Food & Drink Shop | |
| 2 | Forest Hill North & West, Forest Hill Road Park | Sushi Restaurant | |
| 3 | Lawrence Park | Bus Line | |
| 4 | Moore Park, Summerhill East | Restaurant | |
| 5 | North Toronto West, Lawrence Park | Coffee Shop | |
| 6 | Roselawn | Home Service | |
| 7 | Summerhill West, Rathnelly, South Hill, Forest... | Coffee Shop | |
| 8 | The Annex, North Midtown, Yorkville | Sandwich Place | |

| | 2nd Most Common Venue | 3rd Most Common Venue | 4th Most Common Venue | \ |
|---|-----------------------|-----------------------|-----------------------|---|
| 0 | Coffee Shop | Gym | Italian Restaurant | |
| 1 | Park | Gym / Fitness Center | Pizza Place | |
| 2 | Trail | Jewelry Store | Mexican Restaurant | |
| 3 | Park | Swim School | Yoga Studio | |
| 4 | Trail | Park | Tennis Court | |
| 5 | Clothing Store | Yoga Studio | Seafood Restaurant | |
| 6 | Garden | Pool | Department Store | |
| 7 | Pub | Sports Bar | Fried Chicken Joint | |
| 8 | Café | Coffee Shop | Pub | |

| | 5th Most Common Venue | 6th Most Common Venue | 7th Most Common Venue | \ |
|---|-----------------------|-----------------------|-----------------------|---|
| 0 | Sushi Restaurant | Café | Sandwich Place | |
| 1 | Sandwich Place | Department Store | Breakfast Spot | |
| 2 | Food & Drink Shop | Diner | Donut Shop | |
| 3 | Diner | Donut Shop | Farmers Market | |
| 4 | Fast Food Restaurant | Dessert Shop | Diner | |
| 5 | Gym / Fitness Center | Fast Food Restaurant | Diner | |
| 6 | Gym | Greek Restaurant | Gourmet Shop | |
| 7 | Vietnamese Restaurant | Light Rail Station | Liquor Store | |

| | Cheese Shop | Park | Pharmacy |
|---|-----------------------|-----------------------|------------------------|
| 8 | 8th Most Common Venue | 9th Most Common Venue | 10th Most Common Venue |
| 0 | Pizza Place | Restaurant | Gourmet Shop |
| 1 | Hotel | Gas Station | Garden |
| 2 | Farmers Market | Fast Food Restaurant | Yoga Studio |
| 3 | Fast Food Restaurant | Food & Drink Shop | Fried Chicken Joint |
| 4 | Donut Shop | Farmers Market | Fried Chicken Joint |
| 5 | Mexican Restaurant | Park | Cosmetics Shop |
| 6 | Gas Station | Fried Chicken Joint | Food & Drink Shop |
| 7 | Pizza Place | Restaurant | American Restaurant |
| 8 | Pizza Place | Liquor Store | Donut Shop |

```
In [156]: # set number of clusters
kclusters = 5
```

```
Central_Toronto_grouped_clustering = Central_Toronto_grouped.drop('Neighborhood', 1)

# run k-means clustering
kmeans = KMeans(n_clusters=kclusters, random_state=0).fit(Central_Toronto_grouped_clustering)

# check cluster labels generated for each row in the dataframe
kmeans.labels_[0:10]
```

```
Out[156]: array([0, 0, 3, 2, 4, 0, 1, 0, 0])
```

```
In [160]: # add clustering labels
neighborhoods_venues_sorted.insert(0, 'Cluster Labels', kmeans.labels_)
```

```
Central_Toronto_merged = Central_Toronto
```

```
# merge toronto_grouped with toronto_data to add latitude/longitude for each neighborhood
Central_Toronto_merged = Central_Toronto_merged.join(neighborhoods_venues_sorted.set_index('Neighborhood'))
```

```
Central_Toronto_merged.head(20) # check the last columns!
```

```
Out[160]:
```

| | Borough | Neighborhood \ |
|---|-----------------|---|
| 0 | Central Toronto | Lawrence Park |
| 1 | Central Toronto | Roselawn |
| 2 | Central Toronto | Davisville North |
| 3 | Central Toronto | Forest Hill North & West, Forest Hill Road Park |
| 4 | Central Toronto | North Toronto West, Lawrence Park |
| 5 | Central Toronto | The Annex, North Midtown, Yorkville |
| 6 | Central Toronto | Davisville |
| 7 | Central Toronto | Moore Park, Summerhill East |
| 8 | Central Toronto | Summerhill West, Rathnelly, South Hill, Forest... |

| | Latitude | Longitude | Cluster Labels | 1st Most Common Venue \ |
|---|-----------|------------|----------------|-------------------------|
| 0 | 43.728020 | -79.388790 | 2 | Bus Line |

| | | | | |
|---|-----------|------------|---|-------------------|
| 1 | 43.711695 | -79.416936 | 1 | Home Service |
| 2 | 43.712751 | -79.390197 | 0 | Food & Drink Shop |
| 3 | 43.696948 | -79.411307 | 3 | Sushi Restaurant |
| 4 | 43.715383 | -79.405678 | 0 | Coffee Shop |
| 5 | 43.672710 | -79.405678 | 0 | Sandwich Place |
| 6 | 43.704324 | -79.388790 | 0 | Dessert Shop |
| 7 | 43.689574 | -79.383160 | 4 | Restaurant |
| 8 | 43.686412 | -79.400049 | 0 | Coffee Shop |

| | 2nd Most Common Venue | 3rd Most Common Venue | 4th Most Common Venue | \ |
|---|-----------------------|-----------------------|-----------------------|---|
| 0 | Park | Swim School | Yoga Studio | |
| 1 | Garden | Pool | Department Store | |
| 2 | Park | Gym / Fitness Center | Pizza Place | |
| 3 | Trail | Jewelry Store | Mexican Restaurant | |
| 4 | Clothing Store | Yoga Studio | Seafood Restaurant | |
| 5 | Café | Coffee Shop | Pub | |
| 6 | Coffee Shop | Gym | Italian Restaurant | |
| 7 | Trail | Park | Tennis Court | |
| 8 | Pub | Sports Bar | Fried Chicken Joint | |

| | 5th Most Common Venue | 6th Most Common Venue | 7th Most Common Venue | \ |
|---|-----------------------|-----------------------|-----------------------|---|
| 0 | Diner | Donut Shop | Farmers Market | |
| 1 | Gym | Greek Restaurant | Gourmet Shop | |
| 2 | Sandwich Place | Department Store | Breakfast Spot | |
| 3 | Food & Drink Shop | Diner | Donut Shop | |
| 4 | Gym / Fitness Center | Fast Food Restaurant | Diner | |
| 5 | Cheese Shop | Park | Pharmacy | |
| 6 | Sushi Restaurant | Café | Sandwich Place | |
| 7 | Fast Food Restaurant | Dessert Shop | Diner | |
| 8 | Vietnamese Restaurant | Light Rail Station | Liquor Store | |

| | 8th Most Common Venue | 9th Most Common Venue | 10th Most Common Venue |
|---|-----------------------|-----------------------|------------------------|
| 0 | Fast Food Restaurant | Food & Drink Shop | Fried Chicken Joint |
| 1 | Gas Station | Fried Chicken Joint | Food & Drink Shop |
| 2 | Hotel | Gas Station | Garden |
| 3 | Farmers Market | Fast Food Restaurant | Yoga Studio |
| 4 | Mexican Restaurant | Park | Cosmetics Shop |
| 5 | Pizza Place | Liquor Store | Donut Shop |
| 6 | Pizza Place | Restaurant | Gourmet Shop |
| 7 | Donut Shop | Farmers Market | Fried Chicken Joint |
| 8 | Pizza Place | Restaurant | American Restaurant |

```
In [161]: # Creating map with clusters and the fifth cluster seems to have restaurant a the mo
# This location would the most appropriate location for us open this steak house res
```

```
map_clusters = folium.Map(location=[latitude, longitude], zoom_start=11)

# set color scheme for the clusters
```

```

x = np.arange(kclusters)
ys = [i + x + (i*x)**2 for i in range(kclusters)]
colors_array = cm.rainbow(np.linspace(0, 1, len(ys)))
rainbow = [colors.rgb2hex(i) for i in colors_array]

# add markers to the map
markers_colors = []
for lat, lon, poi, cluster in zip(Central_Toronto_merged['Latitude'], Central_Toronto_merged['Longitude'], Central_Toronto_merged['Name'], Central_Toronto_merged['Cluster']):
    label = folium.Popup(str(poi) + ' Cluster ' + str(cluster), parse_html=True)
    folium.CircleMarker(
        [lat, lon],
        radius=5,
        popup=label,
        color=rainbow[cluster-1],
        fill=True,
        fill_color=rainbow[cluster-1],
        fill_opacity=0.7).add_to(map_clusters)

map_clusters

```

Out[161]: <folium.folium.Map at 0x2af5acf6dd8>

In [162]: # First cluster

```
Central_Toronto_merged.loc[Central_Toronto_merged['Cluster Labels'] == 0, Central_Toronto_merged]
```

```

Out[162]:
      Neighborhood 1st Most Common Venue \
2      Davisville North      Food & Drink Shop
4      North Toronto West, Lawrence Park      Coffee Shop
5      The Annex, North Midtown, Yorkville      Sandwich Place
6      Davisville      Dessert Shop
8      Summerhill West, Rathnelly, South Hill, Forest...      Coffee Shop

      2nd Most Common Venue 3rd Most Common Venue 4th Most Common Venue \
2      Park Gym / Fitness Center      Pizza Place
4      Clothing Store      Yoga Studio      Seafood Restaurant
5      Café      Coffee Shop      Pub
6      Coffee Shop      Gym      Italian Restaurant
8      Pub      Sports Bar      Fried Chicken Joint

      5th Most Common Venue 6th Most Common Venue 7th Most Common Venue \
2      Sandwich Place      Department Store      Breakfast Spot
4      Gym / Fitness Center      Fast Food Restaurant      Diner
5      Cheese Shop      Park      Pharmacy
6      Sushi Restaurant      Café      Sandwich Place
8      Vietnamese Restaurant      Light Rail Station      Liquor Store

      8th Most Common Venue 9th Most Common Venue 10th Most Common Venue
2      Hotel      Gas Station      Garden

```


| | | | |
|---|--------------------|--------------|---------------------|
| 4 | Mexican Restaurant | Park | Cosmetics Shop |
| 5 | Pizza Place | Liquor Store | Donut Shop |
| 6 | Pizza Place | Restaurant | Gourmet Shop |
| 8 | Pizza Place | Restaurant | American Restaurant |

In [163]: # *Second cluster*

Central_Toronto_merged.loc[Central_Toronto_merged['Cluster Labels'] == 1, Central_Toronto_merged.columns]

Out[163]:

| Neighborhood | 1st Most Common Venue | 2nd Most Common Venue | \ |
|-----------------------|------------------------|-----------------------|---|
| 1 Roselawn | Home Service | Garden | |
| 3rd Most Common Venue | 4th Most Common Venue | 5th Most Common Venue | \ |
| 1 Pool | Department Store | Gym | |
| 6th Most Common Venue | 7th Most Common Venue | 8th Most Common Venue | \ |
| 1 Greek Restaurant | Gourmet Shop | Gas Station | |
| 9th Most Common Venue | 10th Most Common Venue | | |
| 1 Fried Chicken Joint | Food & Drink Shop | | |

In [164]: # *Fifth cluster*

Central_Toronto_merged.loc[Central_Toronto_merged['Cluster Labels'] == 4, Central_Toronto_merged.columns]

Out[164]:

| Neighborhood | 1st Most Common Venue | 2nd Most Common Venue | \ |
|-------------------------------|------------------------|-----------------------|---|
| 7 Moore Park, Summerhill East | Restaurant | Trail | |
| 3rd Most Common Venue | 4th Most Common Venue | 5th Most Common Venue | \ |
| 7 Park | Tennis Court | Fast Food Restaurant | |
| 6th Most Common Venue | 7th Most Common Venue | 8th Most Common Venue | \ |
| 7 Dessert Shop | Diner | Donut Shop | |
| 9th Most Common Venue | 10th Most Common Venue | | |
| 7 Farmers Market | Fried Chicken Joint | | |