

Exploring_Neighborhoods_in_the_city_of_Toronto

April 7, 2024

0.0.1 Explore and cluster the neighborhoods in Toronto.

Generate maps to visualize your neighborhoods and how they cluster together

```
[1]: import pandas as pd
import numpy as np
import requests
from bs4 import BeautifulSoup
import os
from sklearn.cluster import KMeans
import folium
from geopy.geocoders import Nominatim
import matplotlib.cm as cm
import matplotlib.colors as colors

print('Libraries imported.')
```

Libraries imported.

```
[2]: List_url='https://en.wikipedia.org/wiki/List_of_postal_codes_of_Canada:_M'
source = requests.get(List_url).text
```

```
[3]: soup = BeautifulSoup(source, 'xml')
table=soup.find('table')
column_names=['Postalcode', 'Borough', 'Neighborhood']
df = pd.DataFrame(columns=column_names)
```

```
[4]: for tr_cell in table.find_all('tr'):
    row_data=[]
    for td_cell in tr_cell.find_all('td'):
        row_data.append(td_cell.text.strip())
    if len(row_data)==3:
        df.loc[len(df)] = row_data
    df.head()
```

```
[5]: df=df[df['Borough']!='Not assigned']
```

```
[7]: df.head()
```

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

```
[8]: temp_df=df.groupby('Postalcode')['Neighborhood'].apply(lambda x: "%s" % ', '.join(x))
temp_df=temp_df.reset_index(drop=False)
temp_df.rename(columns={'Neighborhood':'Neighborhood_joined'},inplace=True)
```

```
[9]: df_merge = pd.merge(df, temp_df, on='Postalcode')
```

```
[10]: df_merge.drop(['Neighborhood'],axis=1,inplace=True)
```

```
[11]: df_merge.drop_duplicates(inplace=True)
```

```
[12]: df_merge.rename(columns={'Neighborhood_joined':'Neighborhood'},inplace=True)
```

```
[13]: df_merge.head()
```

```
[13]:   Postalcode      Borough      Neighborhood
0      M3A      North York      Parkwoods
1      M4A      North York      Victoria Village
2      M5A  Downtown Toronto      Regent Park, Harbourfront
3      M6A      North York      Lawrence Manor, Lawrence Heights
4      M7A  Downtown Toronto  Queen's Park, Ontario Provincial Government
```

```
[14]: df_merge.shape
```

```
[14]: (103, 3)
```

```
[15]: def get_geocode(postal_code):
      # initialize your variable to None
      lat_lng_coords = None
      while(lat_lng_coords is None):
          g = geocoder.google('{}, Toronto, Ontario'.format(postal_code))
          lat_lng_coords = g.latlng
          latitude = lat_lng_coords[0]
          longitude = lat_lng_coords[1]
      return latitude,longitude
```

```
[16]: geo_df=pd.read_csv('http://cocl.us/Geospatial_data')
```

```
[17]: geo_df.head()
```

```
[17]:   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
```

```
[18]: geo_df.rename(columns={'Postal Code':'Postalcode'},inplace=True)
geo_merged = pd.merge(geo_df, df_merge, on='Postalcode')
```

```
[19]: geo_merged.head()
```

```
[19]:   Postalcode  Latitude  Longitude  Borough \
0         M1B  43.806686 -79.194353  Scarborough
1         M1C  43.784535 -79.160497  Scarborough
2         M1E  43.763573 -79.188711  Scarborough
3         M1G  43.770992 -79.216917  Scarborough
4         M1H  43.773136 -79.239476  Scarborough
```

```

           Neighborhood
0           Malvern, Rouge
1  Rouge Hill, Port Union, Highland Creek
2    Guildwood, Morningside, West Hill
3                Woburn
4           Cedarbrae
```

```
[20]: geo_data=geo_merged[['Postalcode','Borough','Neighborhood','Latitude','Longitude']]
geo_data.head()
```

```
[20]:   Postalcode  Borough  Neighborhood  Latitude \
0         M1B  Scarborough  Malvern, Rouge  43.806686
1         M1C  Scarborough  Rouge Hill, Port Union, Highland Creek  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
```

```
[21]: toronto_data=geo_data[geo_data['Borough'].str.contains("Toronto")]
toronto_data.head()
```

```
[21]: Postalcode      Borough      Neighborhood  Latitude \
37      M4E      East Toronto      The Beaches  43.676357
41      M4K      East Toronto      The Danforth West, Riverdale  43.679557
42      M4L      East Toronto      India Bazaar, The Beaches West  43.668999
43      M4M      East Toronto      Studio District  43.659526
44      M4N      Central Toronto      Lawrence Park  43.728020

      Longitude
37 -79.293031
41 -79.352188
42 -79.315572
43 -79.340923
44 -79.388790
```

```
[22]: CLIENT_ID = 'ODW53TD40J43SX3I05J1TO2DTHKWKJRGKWRXPEHXTAHJR33M' # your
      ↪Foursquare ID
CLIENT_SECRET = 'PCTM14K3UAVAHUEQXXYJLGKJVY4NCYVG4MCMAT44ZZ1GXDO' # your
      ↪Foursquare Secret
VERSION = '20180604'
```

```
[23]: def getNearbyVenues(names, latitudes, longitudes):
      radius=500
      LIMIT=100
      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={} &ll={}, {} &radius={} &limit={}'.format(
              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)

```

```

[24]: toronto_venues = getNearbyVenues(names=toronto_data['Neighborhood'],
                                       latitudes=toronto_data['Latitude'],
                                       longitudes=toronto_data['Longitude']
                                       )

```

The Beaches

The Danforth West, Riverdale

India Bazaar, The Beaches West

Studio District

Lawrence Park

Davisville North

North Toronto West, Lawrence Park

Davisville

Moore Park, Summerhill East

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

Rosedale

St. James Town, Cabbagetown

Church and Wellesley

Regent Park, Harbourfront

Garden District, Ryerson

St. James Town

Berczy Park

Central Bay Street

Richmond, Adelaide, King

Harbourfront East, Union Station, Toronto Islands

Toronto Dominion Centre, Design Exchange

Commerce Court, Victoria Hotel

Roselawn

Forest Hill North & West, Forest Hill Road Park

The Annex, North Midtown, Yorkville

University of Toronto, Harbord

Kensington Market, Chinatown, Grange Park
 CN Tower, King and Spadina, Railway Lands, Harbourfront West, Bathurst Quay,
 South Niagara, Island airport
 Stn A PO Boxes
 First Canadian Place, Underground city
 Christie
 Dufferin, Dovercourt Village
 Little Portugal, Trinity
 Brockton, Parkdale Village, Exhibition Place
 High Park, The Junction South
 Parkdale, Roncesvalles
 Runnymede, Swansea
 Queen's Park, Ontario Provincial Government
 Business reply mail Processing Centre, South Central Letter Processing Plant
 Toronto

```
[25]: toronto_venues.head()
```

```
[25]:
```

	Neighborhood	Neighborhood Latitude	\
0	The Beaches	43.676357	
1	The Beaches	43.676357	
2	The Beaches	43.676357	
3	The Beaches	43.676357	
4	The Danforth West, Riverdale	43.679557	

	Neighborhood Longitude	Venue	Venue Latitude	\
0	-79.293031	Glen Manor Ravine	43.676821	
1	-79.293031	The Big Carrot Natural Food Market	43.678879	
2	-79.293031	Grover Pub and Grub	43.679181	
3	-79.293031	Upper Beaches	43.680563	
4	-79.352188	MenEssentials	43.677820	

	Venue Longitude	Venue Category
0	-79.293942	Trail
1	-79.297734	Health Food Store
2	-79.297215	Pub
3	-79.292869	Neighborhood
4	-79.351265	Cosmetics Shop

```
[26]: toronto_venues.groupby('Neighborhood').count()
```

```
[26]:
```

	Neighborhood	Neighborhood Latitude	\
	Berczy Park	57	
	Brockton, Parkdale Village, Exhibition Place	22	
	Business reply mail Processing Centre, South Ce...	16	
	CN Tower, King and Spadina, Railway Lands, Harb...	16	

Central Bay Street	66
Christie	16
Church and Wellesley	72
Commerce Court, Victoria Hotel	100
Davisville	35
Davisville North	9
Dufferin, Dovercourt Village	16
First Canadian Place, Underground city	100
Forest Hill North & West, Forest Hill Road Park	4
Garden District, Ryerson	100
Harbourfront East, Union Station, Toronto Islands	100
High Park, The Junction South	25
India Bazaar, The Beaches West	22
Kensington Market, Chinatown, Grange Park	66
Lawrence Park	4
Little Portugal, Trinity	44
Moore Park, Summerhill East	3
North Toronto West, Lawrence Park	20
Parkdale, Roncesvalles	14
Queen's Park, Ontario Provincial Government	35
Regent Park, Harbourfront	44
Richmond, Adelaide, King	97
Rosedale	4
Roselawn	1
Runnymede, Swansea	37
St. James Town	86
St. James Town, Cabbagetown	45
Stn A PO Boxes	98
Studio District	42
Summerhill West, Rathnelly, South Hill, Forest ...	16
The Annex, North Midtown, Yorkville	20
The Beaches	4
The Danforth West, Riverdale	42
Toronto Dominion Centre, Design Exchange	100
University of Toronto, Harbord	36

Neighborhood Longitude \

Neighborhood	
Berczy Park	57
Brockton, Parkdale Village, Exhibition Place	22
Business reply mail Processing Centre, South Ce...	16
CN Tower, King and Spadina, Railway Lands, Harb...	16
Central Bay Street	66
Christie	16
Church and Wellesley	72
Commerce Court, Victoria Hotel	100
Davisville	35

Davisville North	9
Dufferin, Dovercourt Village	16
First Canadian Place, Underground city	100
Forest Hill North & West, Forest Hill Road Park	4
Garden District, Ryerson	100
Harbourfront East, Union Station, Toronto Islands	100
High Park, The Junction South	25
India Bazaar, The Beaches West	22
Kensington Market, Chinatown, Grange Park	66
Lawrence Park	4
Little Portugal, Trinity	44
Moore Park, Summerhill East	3
North Toronto West, Lawrence Park	20
Parkdale, Roncesvalles	14
Queen's Park, Ontario Provincial Government	35
Regent Park, Harbourfront	44
Richmond, Adelaide, King	97
Rosedale	4
Roselawn	1
Runnymede, Swansea	37
St. James Town	86
St. James Town, Cabbagetown	45
Stn A PO Boxes	98
Studio District	42
Summerhill West, Rathnelly, South Hill, Forest ...	16
The Annex, North Midtown, Yorkville	20
The Beaches	4
The Danforth West, Riverdale	42
Toronto Dominion Centre, Design Exchange	100
University of Toronto, Harbord	36

	Venue	Venue Latitude \
Neighborhood		
Berczy Park	57	57
Brockton, Parkdale Village, Exhibition Place	22	22
Business reply mail Processing Centre, South Ce...	16	16
CN Tower, King and Spadina, Railway Lands, Harb...	16	16
Central Bay Street	66	66
Christie	16	16
Church and Wellesley	72	72
Commerce Court, Victoria Hotel	100	100
Davisville	35	35
Davisville North	9	9
Dufferin, Dovercourt Village	16	16
First Canadian Place, Underground city	100	100
Forest Hill North & West, Forest Hill Road Park	4	4
Garden District, Ryerson	100	100

Harbourfront East, Union Station, Toronto Islands	100	100
High Park, The Junction South	25	25
India Bazaar, The Beaches West	22	22
Kensington Market, Chinatown, Grange Park	66	66
Lawrence Park	4	4
Little Portugal, Trinity	44	44
Moore Park, Summerhill East	3	3
North Toronto West, Lawrence Park	20	20
Parkdale, Roncesvalles	14	14
Queen's Park, Ontario Provincial Government	35	35
Regent Park, Harbourfront	44	44
Richmond, Adelaide, King	97	97
Rosedale	4	4
Roselawn	1	1
Runnymede, Swansea	37	37
St. James Town	86	86
St. James Town, Cabbagetown	45	45
Stn A PO Boxes	98	98
Studio District	42	42
Summerhill West, Rathnelly, South Hill, Forest ...	16	16
The Annex, North Midtown, Yorkville	20	20
The Beaches	4	4
The Danforth West, Riverdale	42	42
Toronto Dominion Centre, Design Exchange	100	100
University of Toronto, Harbord	36	36

Venue Longitude \

Neighborhood	
Berczy Park	57
Brockton, Parkdale Village, Exhibition Place	22
Business reply mail Processing Centre, South Ce...	16
CN Tower, King and Spadina, Railway Lands, Harb...	16
Central Bay Street	66
Christie	16
Church and Wellesley	72
Commerce Court, Victoria Hotel	100
Davisville	35
Davisville North	9
Dufferin, Dovercourt Village	16
First Canadian Place, Underground city	100
Forest Hill North & West, Forest Hill Road Park	4
Garden District, Ryerson	100
Harbourfront East, Union Station, Toronto Islands	100
High Park, The Junction South	25
India Bazaar, The Beaches West	22
Kensington Market, Chinatown, Grange Park	66
Lawrence Park	4

Little Portugal, Trinity	44
Moore Park, Summerhill East	3
North Toronto West, Lawrence Park	20
Parkdale, Roncesvalles	14
Queen's Park, Ontario Provincial Government	35
Regent Park, Harbourfront	44
Richmond, Adelaide, King	97
Rosedale	4
Roselawn	1
Runnymede, Swansea	37
St. James Town	86
St. James Town, Cabbagetown	45
Stn A PO Boxes	98
Studio District	42
Summerhill West, Rathnelly, South Hill, Forest ...	16
The Annex, North Midtown, Yorkville	20
The Beaches	4
The Danforth West, Riverdale	42
Toronto Dominion Centre, Design Exchange	100
University of Toronto, Harbord	36

Venue Category

Neighborhood	
Berczy Park	57
Brockton, Parkdale Village, Exhibition Place	22
Business reply mail Processing Centre, South Ce...	16
CN Tower, King and Spadina, Railway Lands, Harb...	16
Central Bay Street	66
Christie	16
Church and Wellesley	72
Commerce Court, Victoria Hotel	100
Davisville	35
Davisville North	9
Dufferin, Dovercourt Village	16
First Canadian Place, Underground city	100
Forest Hill North & West, Forest Hill Road Park	4
Garden District, Ryerson	100
Harbourfront East, Union Station, Toronto Islands	100
High Park, The Junction South	25
India Bazaar, The Beaches West	22
Kensington Market, Chinatown, Grange Park	66
Lawrence Park	4
Little Portugal, Trinity	44
Moore Park, Summerhill East	3
North Toronto West, Lawrence Park	20
Parkdale, Roncesvalles	14
Queen's Park, Ontario Provincial Government	35

Regent Park, Harbourfront	44
Richmond, Adelaide, King	97
Rosedale	4
Roselawn	1
Runnymede, Swansea	37
St. James Town	86
St. James Town, Cabbagetown	45
Stn A PO Boxes	98
Studio District	42
Summerhill West, Rathnelly, South Hill, Forest ...	16
The Annex, North Midtown, Yorkville	20
The Beaches	4
The Danforth West, Riverdale	42
Toronto Dominion Centre, Design Exchange	100
University of Toronto, Harbord	36

```
[27]: # one hot encoding
toronto_onehot = pd.get_dummies(toronto_venues[['Venue Category']], prefix="",
    ↪prefix_sep="")
toronto_onehot.drop(['Neighborhood'],axis=1,inplace=True)
toronto_onehot.insert(loc=0, column='Neighborhood',
    ↪value=toronto_venues['Neighborhood'] )
toronto_onehot.shape
```

[27]: (1634, 234)

```
[28]: toronto_grouped = toronto_onehot.groupby('Neighborhood').mean().reset_index()
toronto_grouped.head()
```

```
[28]:
```

	Neighborhood	Airport	\
0	Berczy Park	0.0000	
1	Brockton, Parkdale Village, Exhibition Place	0.0000	
2	Business reply mail Processing Centre, South C...	0.0000	
3	CN Tower, King and Spadina, Railway Lands, Har...	0.0625	
4	Central Bay Street	0.0000	

	Airport Food Court	Airport Lounge	Airport Service	Airport Terminal	\
0	0.0000	0.000	0.000	0.000	
1	0.0000	0.000	0.000	0.000	
2	0.0000	0.000	0.000	0.000	
3	0.0625	0.125	0.125	0.125	
4	0.0000	0.000	0.000	0.000	

	American Restaurant	Antique Shop	Aquarium	Art Gallery	...	\
0	0.0	0.0	0.0	0.017544	...	
1	0.0	0.0	0.0	0.000000	...	
2	0.0	0.0	0.0	0.000000	...	

3	0.0	0.0	0.0	0.000000 ...
4	0.0	0.0	0.0	0.000000 ...

	Theme Restaurant	Toy / Game Store	Trail	Train Station \
0	0.0	0.0	0.0	0.0
1	0.0	0.0	0.0	0.0
2	0.0	0.0	0.0	0.0
3	0.0	0.0	0.0	0.0
4	0.0	0.0	0.0	0.0

	Vegetarian / Vegan Restaurant	Video Game Store	Vietnamese Restaurant \
0	0.017544	0.0	0.0
1	0.000000	0.0	0.0
2	0.000000	0.0	0.0
3	0.000000	0.0	0.0
4	0.015152	0.0	0.0

	Wine Bar	Women's Store	Yoga Studio
0	0.000000	0.0	0.000000
1	0.000000	0.0	0.000000
2	0.000000	0.0	0.000000
3	0.000000	0.0	0.000000
4	0.015152	0.0	0.015152

[5 rows x 234 columns]

```
[29]: 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]
```

```
[30]: 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'] = toronto_grouped['Neighborhood']
```

```

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

neighborhoods_venues_sorted.head()

```

```

[30]:

```

	Neighborhood	1st Most Common Venue	\
0	Berczy Park	Coffee Shop	
1	Brockton, Parkdale Village, Exhibition Place	Café	
2	Business reply mail Processing Centre, South C...	Light Rail Station	
3	CN Tower, King and Spadina, Railway Lands, Har...	Airport Lounge	
4	Central Bay Street	Coffee Shop	

	2nd Most Common Venue	3rd Most Common Venue	4th Most Common Venue	\
0	Cocktail Bar	Bakery	Café	
1	Breakfast Spot	Coffee Shop	Pet Store	
2	Comic Shop	Auto Workshop	Brewery	
3	Airport Service	Airport Terminal	Airport	
4	Italian Restaurant	Sandwich Place	Café	

	5th Most Common Venue	6th Most Common Venue	7th Most Common Venue	\
0	Beer Bar	Cheese Shop	Farmers Market	
1	Furniture / Home Store	Convenience Store	Stadium	
2	Burrito Place	Spa	Farmers Market	
3	Rental Car Location	Bar	Coffee Shop	
4	Japanese Restaurant	Burger Joint	Bubble Tea Shop	

	8th Most Common Venue	9th Most Common Venue	10th Most Common Venue
0	Pharmacy	Restaurant	Seafood Restaurant
1	Restaurant	Italian Restaurant	Intersection
2	Fast Food Restaurant	Restaurant	Skate Park
3	Boat or Ferry	Plane	Boutique
4	Bar	Department Store	Salad Place

```

[31]: # set number of clusters
kclusters = 5

toronto_grouped_clustering = toronto_grouped.drop('Neighborhood', 1)

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

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

```

```
[31]: array([0, 0, 0, 0, 0, 0, 0, 0, 0, 0], dtype=int32)
```

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

toronto_merged = toronto_data

# merge toronto_grouped with toronto_data to add latitude/longitude for each
↳ neighborhood
toronto_merged = toronto_merged.join(neighborhoods_venues_sorted.
↳ set_index('Neighborhood'), on='Neighborhood')

toronto_merged.head()
```

[32]:	Postalcode	Borough	Neighborhood	Latitude	\
37	M4E	East Toronto	The Beaches	43.676357	
41	M4K	East Toronto	The Danforth West, Riverdale	43.679557	
42	M4L	East Toronto	India Bazaar, The Beaches West	43.668999	
43	M4M	East Toronto	Studio District	43.659526	
44	M4N	Central Toronto	Lawrence Park	43.728020	

	Longitude	Cluster Labels	1st Most Common Venue	2nd Most Common Venue	\
37	-79.293031	0	Pub	Health Food Store	
41	-79.352188	0	Greek Restaurant	Italian Restaurant	
42	-79.315572	0	Sandwich Place	Park	
43	-79.340923	0	Café	Coffee Shop	
44	-79.388790	4	Park	Bus Line	

	3rd Most Common Venue	4th Most Common Venue	5th Most Common Venue	\
37	Trail	Yoga Studio	Discount Store	
41	Coffee Shop	Furniture / Home Store	Restaurant	
42	Pizza Place	Sushi Restaurant	Pub	
43	American Restaurant	Bakery	Brewery	
44	Swim School	Yoga Studio	Discount Store	

	6th Most Common Venue	7th Most Common Venue	8th Most Common Venue	\
37	Distribution Center	Dog Run	Doner Restaurant	
41	Ice Cream Shop	Yoga Studio	Spa	
42	Brewery	Fast Food Restaurant	Fish & Chips Shop	
43	Gastropub	Yoga Studio	Food	
44	Falafel Restaurant	Event Space	Ethiopian Restaurant	

	9th Most Common Venue	10th Most Common Venue
37	Donut Shop	Eastern European Restaurant
41	Bookstore	Juice Bar
42	Restaurant	Italian Restaurant
43	Pet Store	Park

44 Electronics Store Eastern European Restaurant

```
[33]: neighborhoods_venues_sorted.head()
```

```
[33]:
```

	Cluster Labels		Neighborhood \
0	0		Berczy Park
1	0	Brockton, Parkdale Village, Exhibition Place	
2	0	Business reply mail Processing Centre, South C...	
3	0	CN Tower, King and Spadina, Railway Lands, Har...	
4	0		Central Bay Street

	1st Most Common Venue	2nd Most Common Venue	3rd Most Common Venue \
0	Coffee Shop	Cocktail Bar	Bakery
1	Café	Breakfast Spot	Coffee Shop
2	Light Rail Station	Comic Shop	Auto Workshop
3	Airport Lounge	Airport Service	Airport Terminal
4	Coffee Shop	Italian Restaurant	Sandwich Place

	4th Most Common Venue	5th Most Common Venue	6th Most Common Venue \
0	Café	Beer Bar	Cheese Shop
1	Pet Store	Furniture / Home Store	Convenience Store
2	Brewery	Burrito Place	Spa
3	Airport	Rental Car Location	Bar
4	Café	Japanese Restaurant	Burger Joint

	7th Most Common Venue	8th Most Common Venue	9th Most Common Venue \
0	Farmers Market	Pharmacy	Restaurant
1	Stadium	Restaurant	Italian Restaurant
2	Farmers Market	Fast Food Restaurant	Restaurant
3	Coffee Shop	Boat or Ferry	Plane
4	Bubble Tea Shop	Bar	Department Store

	10th Most Common Venue
0	Seafood Restaurant
1	Intersection
2	Skate Park
3	Boutique
4	Salad Place

```
[34]: address = 'Toronto, CA'

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

The geographical coordinate of Manhattan are 43.6534817, -79.3839347.

0.0.2 Generating Maps

```
[36]: 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(toronto_merged['Latitude'],
    ↳toronto_merged['Longitude'], toronto_merged['Neighborhood'],
    ↳toronto_merged['Cluster Labels']):
    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
```

```
[36]: <folium.folium.Map at 0x7ffd0c36ca10>
```

```
[ ]:
```