

tter-place-in-scarborough-toronto

July 23, 2023

1 Capstone Project – The Battle of Neighborhoods | Finding a Better Place in Scarborough, Toronto

1.0.1 1. Installing and Importing Python Libraries and Dependencies

```
[1]: !pip install geocoder  
     !pip install folium
```

Collecting geocoder

Downloading <https://files.pythonhosted.org/packages/4f/6b/13166c909ad2f2d76b929a4227c952630ebaf0d729f6317eb09cbceccbab/geocoder-1.38.1-py2.py3-none-any.whl> (98kB)

| 102kB 17.1MB/s ta 0:00:01

Requirement already satisfied: six in

/opt/conda/envs/Python36/lib/python3.6/site-packages (from geocoder) (1.12.0)

Requirement already satisfied: requests in

/opt/conda/envs/Python36/lib/python3.6/site-packages (from geocoder) (2.21.0)

Requirement already satisfied: future in

/opt/conda/envs/Python36/lib/python3.6/site-packages (from geocoder) (0.17.1)

Collecting ratelim (from geocoder)

Downloading <https://files.pythonhosted.org/packages/f2/98/7e6d147fd16a10a5f821db6e25f192265d6ecca3d82957a4fdd592cad49c/ratelim-0.1.6-py2.py3-none-any.whl>

Requirement already satisfied: click in

/opt/conda/envs/Python36/lib/python3.6/site-packages (from geocoder) (7.0)

Requirement already satisfied: certifi>=2017.4.17 in

/opt/conda/envs/Python36/lib/python3.6/site-packages (from requests->geocoder) (2019.9.11)

Requirement already satisfied: idna<2.9,>=2.5 in

/opt/conda/envs/Python36/lib/python3.6/site-packages (from requests->geocoder) (2.8)

Requirement already satisfied: urllib3<1.25,>=1.21.1 in

/opt/conda/envs/Python36/lib/python3.6/site-packages (from requests->geocoder) (1.24.1)

Requirement already satisfied: chardet<3.1.0,>=3.0.2 in

/opt/conda/envs/Python36/lib/python3.6/site-packages (from requests->geocoder) (3.0.4)

Requirement already satisfied: decorator in

/opt/conda/envs/Python36/lib/python3.6/site-packages (from ratelim->geocoder)

(4.3.2)

Installing collected packages: ratelim, geocoder

Successfully installed geocoder-1.38.1 ratelim-0.1.6

Collecting folium

Downloading <https://files.pythonhosted.org/packages/72/ff/004bfe344150a064e558cb2aedeaa02ecbf75e60e148a55a9198f0c41765/folium-0.10.0-py2.py3-none-any.whl> (91kB)

| 92kB 12.7MB/s eta 0:00:01

Requirement already satisfied: jinja2>=2.9 in

/opt/conda/envs/Python36/lib/python3.6/site-packages (from folium) (2.10)

Requirement already satisfied: numpy in

/opt/conda/envs/Python36/lib/python3.6/site-packages (from folium) (1.15.4)

Requirement already satisfied: requests in

/opt/conda/envs/Python36/lib/python3.6/site-packages (from folium) (2.21.0)

Collecting branca>=0.3.0 (from folium)

Downloading <https://files.pythonhosted.org/packages/63/36/1c93318e9653f4e414a2e0c3b98fc898b4970e939afeedeee6075dd3b703/branca-0.3.1-py3-none-any.whl>

Requirement already satisfied: MarkupSafe>=0.23 in

/opt/conda/envs/Python36/lib/python3.6/site-packages (from jinja2>=2.9->folium) (1.1.0)

Requirement already satisfied: certifi>=2017.4.17 in

/opt/conda/envs/Python36/lib/python3.6/site-packages (from requests->folium) (2019.9.11)

Requirement already satisfied: urllib3<1.25,>=1.21.1 in

/opt/conda/envs/Python36/lib/python3.6/site-packages (from requests->folium) (1.24.1)

Requirement already satisfied: idna<2.9,>=2.5 in

/opt/conda/envs/Python36/lib/python3.6/site-packages (from requests->folium) (2.8)

Requirement already satisfied: chardet<3.1.0,>=3.0.2 in

/opt/conda/envs/Python36/lib/python3.6/site-packages (from requests->folium) (3.0.4)

Requirement already satisfied: six in

/opt/conda/envs/Python36/lib/python3.6/site-packages (from branca>=0.3.0->folium) (1.12.0)

Installing collected packages: branca, folium

Successfully installed branca-0.3.1 folium-0.10.0

Importing Libraries

```
[2]: import pandas as pd
import requests
import numpy as np
import geocoder
import folium
import requests
import matplotlib.cm as cm
import matplotlib.colors as colors
```

```

import json
import xml
import matplotlib.pyplot as plt
%matplotlib inline
import warnings
warnings.filterwarnings("ignore")

from pandas.io.json import json_normalize
from sklearn.cluster import KMeans
from geopy.geocoders import Nominatim
from bs4 import BeautifulSoup

pd.set_option('display.max_columns', None)
pd.set_option('display.max_rows', None)

print("All Required Libraries Imported!")

```

All Required Libraries Imported!

1.0.2 2. Data Extraction and Cleaning

Using BeautifulSoup Scraping List of Postal Codes of Given Wikipedia Page. Link: https://en.wikipedia.org/wiki/List_of_postal_codes_of_Canada:_M

```

[3]: url = "https://en.wikipedia.org/wiki/List_of_postal_codes_of_Canada:_M"
      extracting_data = requests.get(url).text
      wiki_data = BeautifulSoup(extracting_data, 'lxml')

```

Converting content of PostalCode HTML table as dataframe

```

[4]: column_names = ['Postalcode', 'Borough', 'Neighborhood']
      toronto = pd.DataFrame(columns = column_names)

      content = wiki_data.find('div', class_='mw-parser-output')
      table = content.table.tbody
      postcode = 0
      borough = 0
      neighborhood = 0

      for tr in table.find_all('tr'):
          i = 0
          for td in tr.find_all('td'):
              if i == 0:
                  postcode = td.text
                  i = i + 1
              elif i == 1:
                  borough = td.text
                  i = i + 1

```

```

elif i == 2:
    neighborhood = td.text.strip('\n').replace(']', '')
    toronto = toronto.append({'Postalcode': postcode, 'Borough':
↳borough, 'Neighborhood': neighborhood}, ignore_index=True)

```

```

[5]: # clean dataframe
toronto = toronto[toronto.Borough!='Not assigned']
toronto = toronto[toronto.Borough!= 0]
toronto.reset_index(drop = True, inplace = True)
i = 0
for i in range(0, toronto.shape[0]):
    if toronto.iloc[i][2] == 'Not assigned':
        toronto.iloc[i][2] = toronto.iloc[i][1]
    i = i+1

```

```

[6]: df = toronto.groupby(['Postalcode', 'Borough'])['Neighborhood'].apply(', '.join).
↳reset_index()
df.head()

```

```

[6]:   Postalcode   Borough   Neighborhood
0      M1B  Scarborough   Rouge, Malvern
1      M1C  Scarborough Highland Creek, Rouge Hill, Port Union
2      M1E  Scarborough   Guildwood, Morningside, West Hill
3      M1G  Scarborough   Woburn
4      M1H  Scarborough   Cedarbrae

```

```

[7]: df.describe()

```

```

[7]:   Postalcode   Borough   Neighborhood
count      103      103      103
unique      103       11      103
top      M2M  North York  York Mills West
freq        1       24        1

```

```

[8]: df = df.dropna()
empty = 'Not assigned'
df = df[(df.Postalcode != empty) & (df.Borough != empty) & (df.Neighborhood !=
↳empty)]

```

```

[9]: df.head()

```

```

[9]:   Postalcode   Borough   Neighborhood
0      M1B  Scarborough   Rouge, Malvern
1      M1C  Scarborough Highland Creek, Rouge Hill, Port Union
2      M1E  Scarborough   Guildwood, Morningside, West Hill
3      M1G  Scarborough   Woburn
4      M1H  Scarborough   Cedarbrae

```

```
[10]: def neighborhood_list(grouped):
        return ', '.join(sorted(grouped['Neighborhood'].tolist()))

grp = df.groupby(['Postalcode', 'Borough'])
df_2 = grp.apply(neighborhood_list).reset_index(name='Neighborhood')
```

```
[11]: df_2.describe()
```

```
[11]:
```

	Postalcode	Borough	Neighborhood
count	103	103	103
unique	103	11	103
top	M2M	North York	York Mills West
freq	1	24	1

```
[12]: print(df_2.shape)
df_2.head()
```

```
(103, 3)
```

```
[12]:
```

	Postalcode	Borough	Neighborhood
0	M1B	Scarborough	Rouge, Malvern
1	M1C	Scarborough	Highland Creek, Rouge Hill, Port Union
2	M1E	Scarborough	Guildwood, Morningside, West Hill
3	M1G	Scarborough	Woburn
4	M1H	Scarborough	Cedarbrae

```
[13]: def get_latilong(postal_code):
        lati_long_coords = None
        while(lati_long_coords is None):
            g = geocoder.arcgis('{} , Toronto, Ontario'.format(postal_code))
            lati_long_coords = g.latlng
        return lati_long_coords

get_latilong('M4G')
```

```
[13]: [43.709495000000006, -79.36398897099997]
```

```
[14]: # Retrieving Postal Code Co-ordinates
postal_codes = df_2['Postalcode']
coords = [ get_latilong(postal_code) for postal_code in postal_codes.tolist() ]
```

```
[15]: # Adding Columns Latitude & Longitude
df_coords = pd.DataFrame(coords, columns=['Latitude', 'Longitude'])
df_2['Latitude'] = df_coords['Latitude']
df_2['Longitude'] = df_coords['Longitude']
```

```
[16]: df_2[df_2.Postalcode == 'M5G']
```

```
[16]:   Postalcode      Borough      Neighborhood  Latitude  Longitude
      57          M5G  Downtown Toronto  Central Bay Street  43.656091  -79.38493
```

```
[17]: df_2.head(10)
```

```
[17]:   Postalcode      Borough      Neighborhood \
0      M1B  Scarborough      Rouge, Malvern
1      M1C  Scarborough  Highland Creek, Rouge Hill, Port Union
2      M1E  Scarborough      Guildwood, Morningside, West Hill
3      M1G  Scarborough      Woburn
4      M1H  Scarborough      Cedarbrae
5      M1J  Scarborough  Scarborough Village
6      M1K  Scarborough  East Birchmount Park, Ionview, Kennedy Park
7      M1L  Scarborough      Clairlea, Golden Mile, Oakridge
8      M1M  Scarborough  Cliffcrest, Cliffside, Scarborough Village West
9      M1N  Scarborough      Birch Cliff, Cliffside West

      Latitude  Longitude
0  43.811525  -79.195517
1  43.785665  -79.158725
2  43.765815  -79.175193
3  43.768369  -79.217590
4  43.769688  -79.239440
5  43.743125  -79.231750
6  43.726276  -79.263625
7  43.713054  -79.285055
8  43.724235  -79.227925
9  43.696770  -79.259967
```

```
[18]: address = 'Scarborough,Toronto'

geolocator = Nominatim()
location = geolocator.geocode(address)
latitude_x = location.latitude
longitude_y = location.longitude
print('The Geographical Co-ordinate of Seattle,Washington are {}, {}'.format(latitude_x, longitude_y))
```

The Geographical Co-ordinate of Seattle,Washington are 43.773077, -79.257774.

1.0.3 3. Map of Scarborough

```
[19]: map_Scarborough = folium.Map(location=[latitude_x, longitude_y], zoom_start=10)

for lat, lng, nei in zip(df_2['Latitude'], df_2['Longitude'],
                        df_2['Neighborhood']):
```

```

label = '{}'.format(nei)
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_Scarborough)

```

map_Scarborough

[19]: <folium.folium.Map at 0x7f24255bdac8>

```

[20]: address = 'Scarborough,Toronto'

geolocator = Nominatim()
location = geolocator.geocode(address)
latitude_n1 = location.latitude
longitude_n1 = location.longitude
print('The Geographical Co-ordinate of Neighborhood_1 are {}, {}'.format(
    latitude_x, longitude_y))

```

The Geographical Co-ordinate of Neighborhood_1 are 43.773077, -79.257774.

```

[21]: # @hiddel_cell
CLIENT_ID = 'DPBY4JUY3DU20ALPSUV40NY2K1G0JJKJ1NIHBB32XEMOVYY' # my Foursquare ID
CLIENT_SECRET = '1MV443TYEP4HU0OWDUW5NQ5W10L2Y4G05NWG11WIR3NUGC5B' # my Foursquare Secret
VERSION = '20180604'
LIMIT = 30
print('Your credentails:')
print('CLIENT_ID: '+CLIENT_ID)
print('CLIENT_SECRET: '+CLIENT_SECRET)

```

Your credentails:

CLIENT_ID: DPBY4JUY3DU20ALPSUV40NY2K1G0JJKJ1NIHBB32XEMOVYY

CLIENT_SECRET: 1MV443TYEP4HU0OWDUW5NQ5W10L2Y4G05NWG11WIR3NUGC5B

```

[22]: radius = 700
LIMIT = 100
url = 'https://api.foursquare.com/v2/venues/explore?
    &client_id={}&client_secret={}&v={}&ll={},{}&radius={}&limit={}'.format(
        CLIENT_ID,

```

```

CLIENT_SECRET,
VERSION,
latitude_n1,
longitude_n1,
radius,
LIMIT)
results = requests.get(url).json()

```

```

[23]: venues=results['response']['groups'][0]['items']
nearby_venues = json_normalize(venues)
nearby_venues.columns

```

```

[23]: Index(['reasons.count', 'reasons.items', 'referralId', 'venue.categories',
          'venue.events.count', 'venue.events.summary', 'venue.id',
          'venue.location.address', 'venue.location.cc', 'venue.location.city',
          'venue.location.country', 'venue.location.crossStreet',
          'venue.location.distance', 'venue.location.formattedAddress',
          'venue.location.labeledLatLngs', 'venue.location.lat',
          'venue.location.lng', 'venue.location.neighborhood',
          'venue.location.postalCode', 'venue.location.state', 'venue.name',
          'venue.photos.count', 'venue.photos.groups', 'venue.venuePage.id'],
          dtype='object')

```

```

[24]: 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']

```

1.0.4 4. Nearby Venues/Locations

```

[25]: filtered_columns = ['venue.name', 'venue.categories', 'venue.location.lat',
                          ↪ 'venue.location.lng']
nearby_venues = nearby_venues.loc[:, filtered_columns]
nearby_venues.head()

```

```

[25]:
          venue.name \
0          Disney Store
1  American Eagle Outfitters
2             SEPHORA
3   Canyon Creek Chophouse

```


4 Coliseum Scarborough Cinemas

```

venue.categories venue.location.lat \
0 [{'id': '4bf58dd8d48988d1f3941735', 'name': 'T... 43.775537
1 [{'id': '4bf58dd8d48988d103951735', 'name': 'C... 43.775908
2 [{'id': '4bf58dd8d48988d10c951735', 'name': 'C... 43.775017
3 [{'id': '4bf58dd8d48988d1cc941735', 'name': 'S... 43.776959
4 [{'id': '4bf58dd8d48988d17f941735', 'name': 'M... 43.775995

venue.location.lng
0 -79.256833
1 -79.258352
2 -79.258109
3 -79.261694
4 -79.255649
```

1.0.5 5. Categories of Nearby Venues/Locations

```
[26]: 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(5)
```

```
[26]:
```

	name	categories	lat	lng
0	Disney Store	Toy / Game Store	43.775537	-79.256833
1	American Eagle Outfitters	Clothing Store	43.775908	-79.258352
2	SEPHORA	Cosmetics Shop	43.775017	-79.258109
3	Canyon Creek Chophouse	Steakhouse	43.776959	-79.261694
4	Coliseum Scarborough Cinemas	Movie Theater	43.775995	-79.255649

```
[27]: # Top 10 Categories
a=pd.Series(nearby_venues.categories)
a.value_counts()[:10]
```

```
[27]: Clothing Store      8
Coffee Shop             5
Restaurant              3
Cosmetics Shop          3
Sporting Goods Shop     2
Pharmacy                2
Wings Joint             2
Tea Room                2
Sandwich Place          2
Electronics Store       1
```

Name: categories, dtype: int64

```
[28]: def getNearbyVenues(names, latitudes, longitudes, radius=700):

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

        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)

        # making GET request
        venue_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 venue_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)
```

```
[29]: # Nearby Venues
Scarborough_venues = getNearbyVenues(names=df_2['Neighborhood'],
                                      latitudes=df_2['Latitude'],
```

```
longitudes=df_2['Longitude']  
)
```

Rouge, Malvern
Highland Creek, Rouge Hill, Port Union
Guildwood, Morningside, West Hill
Woburn
Cedarbrae
Scarborough Village
East Birchmount Park, Ionview, Kennedy Park
Clairlea, Golden Mile, Oakridge
Cliffcrest, Cliffside, Scarborough Village West
Birch Cliff, Cliffside West
Dorset Park, Scarborough Town Centre, Wexford Heights
Maryvale, Wexford
Agincourt
Clarks Corners, Sullivan, Tam O'Shanter
Agincourt North, L'Amoreaux East, Milliken, Steeles East
L'Amoreaux West
Upper Rouge
Hillcrest Village
Fairview, Henry Farm, Oriole
Bayview Village
Silver Hills, York Mills
Newtonbrook, Willowdale
Willowdale South
York Mills West
Willowdale West
Parkwoods
Don Mills North
Flemingdon Park, Don Mills South
Bathurst Manor, Downsview North, Wilson Heights
Northwood Park, York University
CFB Toronto, Downsview East
Downsview West
Downsview Central
Downsview Northwest
Victoria Village
Woodbine Gardens, Parkview Hill
Woodbine Heights
The Beaches
Leaside
Thorncliffe Park
East Toronto
The Danforth West, Riverdale
The Beaches West, India Bazaar
Studio District

Lawrence Park
Davisville North
North Toronto West
Davisville
Moore Park, Summerhill East
Deer Park, Forest Hill SE, Rathnelly, South Hill, Summerhill West
Rosedale
Cabbagetown, St. James Town
Church and Wellesley
Harbourfront, Regent Park
Ryerson, Garden District
St. James Town
Berczy Park
Central Bay Street
Adelaide, King, Richmond
Harbourfront East, Toronto Islands, Union Station
Design Exchange, Toronto Dominion Centre
Commerce Court, Victoria Hotel
Bedford Park, Lawrence Manor East
Roselawn
Forest Hill North, Forest Hill West
The Annex, North Midtown, Yorkville
Harbord, University of Toronto
Chinatown, Grange Park, Kensington Market
CN Tower, Bathurst Quay, Island airport, Harbourfront West, King and Spadina,
Railway Lands, South Niagara
Stn A PO Boxes 25 The Esplanade
First Canadian Place, Underground city
Lawrence Heights, Lawrence Manor
Glencairn
Humewood-Cedarvale
Caledonia-Fairbanks
Christie
Dovercourt Village, Dufferin
Little Portugal, Trinity
Brockton, Exhibition Place, Parkdale Village
Downsview, North Park, Upwood Park
Del Ray, Keelesdale, Mount Dennis, Silverthorn
The Junction North, Runnymede
High Park, The Junction South
Parkdale, Roncesvalles
Runnymede, Swansea
Queen's Park
Canada Post Gateway Processing Centre
Business Reply Mail Processing Centre 969 Eastern
Humber Bay Shores, Mimico South, New Toronto
Alderwood, Long Branch
The Kingsway, Montgomery Road, Old Mill North

Humber Bay, King's Mill Park, Kingsway Park South East, Mimico NE, Old Mill South, The Queensway East, Royal York South East, Sunnylea Kingsway Park South West, Mimico NW, The Queensway West, Royal York South West, South of Bloor
 Islington Avenue
 Cloverdale, Islington, Martin Grove, Princess Gardens, West Deane Park
 Bloordale Gardens, Eringate, Markland Wood, Old Burnhamthorpe
 Humber Summit
 Emery, Humberlea
 Weston
 Westmount
 Kingsview Village, Martin Grove Gardens, Richview Gardens, St. Phillips
 Albion Gardens, Beaumond Heights, Humbergate, Jamestown, Mount Olive, Silverstone, South Steeles, Thistletown
 Northwest

```
[30]: print('There are {} Uniques Categories.'.format(len(Scarborough_venues['Venue_
↳Category'].unique()))
Scarborough_venues.groupby('Neighborhood').count().head()
```

There are 299 Uniques Categories.

```
[30]:
```

Neighborhood	Latitude
Adelaide, King, Richmond	100
Agincourt	29
Agincourt North, L'Amoreaux East, Milliken, Ste...	4
Albion Gardens, Beaumond Heights, Humbergate, J...	12
Alderwood, Long Branch	9

Neighborhood	Longitude
Adelaide, King, Richmond	100
Agincourt	29
Agincourt North, L'Amoreaux East, Milliken, Ste...	4
Albion Gardens, Beaumond Heights, Humbergate, J...	12
Alderwood, Long Branch	9

Neighborhood	Venue	Venue Latitude
Adelaide, King, Richmond	100	100
Agincourt	29	29
Agincourt North, L'Amoreaux East, Milliken, Ste...	4	4
Albion Gardens, Beaumond Heights, Humbergate, J...	12	12
Alderwood, Long Branch	9	9

Neighborhood	Venue Longitude
Adelaide, King, Richmond	100
Agincourt	29
Agincourt North, L'Amoreaux East, Milliken, Ste...	4
Albion Gardens, Beaumond Heights, Humbergate, J...	12
Alderwood, Long Branch	9

Neighborhood	
Adelaide, King, Richmond	100
Agincourt	29
Agincourt North, L'Amoreaux East, Milliken, Ste...	4
Albion Gardens, Beaumont Heights, Humbergate, J...	12
Alderwood, Long Branch	9

	Venue Category
Neighborhood	
Adelaide, King, Richmond	100
Agincourt	29
Agincourt North, L'Amoreaux East, Milliken, Ste...	4
Albion Gardens, Beaumont Heights, Humbergate, J...	12
Alderwood, Long Branch	9

1.0.6 One Hot Encoding of Features

```
[31]: # one hot encoding
Scarborough_onehot = pd.get_dummies(Scarborough_venues[['Venue Category']],
    ↪ prefix="", prefix_sep="")

# add neighborhood column back to dataframe
Scarborough_onehot['Neighborhood'] = Scarborough_venues['Neighborhood']

# move neighborhood column to the first column
fixed_columns = [Scarborough_onehot.columns[-1]] + list(Scarborough_onehot.
    ↪ columns[:-1])
Scarborough_onehot = Scarborough_onehot[fixed_columns]
Scarborough_grouped = Scarborough_onehot.groupby('Neighborhood').mean().
    ↪ reset_index()
Scarborough_onehot.head(5)
```

```
[31]: Zoo Exhibit  Accessories Store  Afghan Restaurant  African Restaurant  \
0           0           0           0           0
1           0           0           0           0
2           1           0           0           0
3           0           0           0           0
4           0           0           0           0

      Airport  American Restaurant  Antique Shop  Arcade  Art Gallery  \
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
```

	Art Museum	Arts & Crafts Store	Asian Restaurant	Athletics & Sports	\
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	

	Auto Dealership	BBQ Joint	Baby Store	Bagel Shop	Bakery	Bank	Bar	\
0	0	0	0	0	0	0	0	
1	0	0	0	0	0	0	0	
2	0	0	0	0	0	0	0	
3	0	0	0	0	0	0	0	
4	0	0	0	0	0	0	0	1

	Baseball Field	Basketball Stadium	Beach	Beer Bar	Beer Store	\
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	

	Belgian Restaurant	Big Box Store	Bike Rental / Bike Share	Bike Shop	\
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	

	Bistro	Boat or Ferry	Bookstore	Boutique	Brazilian 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	

	Breakfast Spot	Brewery	Bridge	Bubble Tea Shop	Buffet	Building	\
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	

	Burger Joint	Burrito Place	Bus Line	Bus Station	Bus Stop	\
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
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	Business Service	Butcher	Café	Cajun / Creole Restaurant	Camera Store	\
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	

	Candy Store	Cantonese Restaurant	Caribbean Restaurant	Castle	Cemetery	\
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	

	Cheese Shop	Chinese Restaurant	Chiropractor	Chocolate Shop	Church	\
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	

	Climbing Gym	Clothing Store	Cocktail Bar	Coffee Shop	\
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	

	College Arts Building	College Gym	College Rec Center	College Stadium	\
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	

	College Theater	Colombian Restaurant	Comedy Club	\
0	0	0	0	
1	0	0	0	
2	0	0	0	
3	0	0	0	
4	0	0	0	

	Comfort Food Restaurant	Comic Shop	Concert Hall	\
0	0	0	0	
1	0	0	0	

2	0	0	0
3	0	0	0
4	0	0	0

	Construction & Landscaping	Convenience Store	Cosmetics Shop	Creperie	\
0	0	0	0	0	
1	0	0	0	0	
2	0	0	0	0	
3	1	0	0	0	
4	0	0	0	0	

	Cuban Restaurant	Cupcake Shop	Curling Ice	Dance Studio	Deli / Bodega	\
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	

	Dentist's Office	Department Store	Design Studio	Dessert Shop	\
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	

	Dim Sum Restaurant	Diner	Disc Golf	Discount Store	Doctor's Office	\
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	

	Dog Run	Doner Restaurant	Donut Shop	Dumpling Restaurant	\
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	

	Eastern European Restaurant	Electronics Store	Elementary School	\
0	0	0	0	
1	0	0	0	
2	0	0	0	
3	0	0	0	
4	0	0	0	

	Ethiopian Restaurant	Event Space	Falafel Restaurant	Farm	\
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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

	Farmers Market	Fast Food Restaurant	Field	Filipino Restaurant	\
0	0	1	0	0	
1	0	0	0	0	
2	0	0	0	0	
3	0	0	0	0	
4	0	0	0	0	

	Financial or Legal Service	Fish & Chips Shop	Fish Market	Flea Market	\
0	0	0	0	0	
1	1	0	0	0	
2	0	0	0	0	
3	0	0	0	0	
4	0	0	0	0	

	Flower Shop	Food	Food & Drink Shop	Food Court	Food Truck	Fountain	\
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	

	French Restaurant	Fried Chicken Joint	Frozen Yogurt Shop	\
0	0	0	0	
1	0	0	0	
2	0	0	0	
3	0	0	0	
4	0	0	0	

	Furniture / Home Store	Gaming Cafe	Garden	Gas Station	Gastropub	\
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	

	Gay Bar	General Entertainment	General Travel	German Restaurant	\
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	

	Gift Shop	Golf Course	Golf Driving Range	Gourmet Shop	Greek 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	

	Grocery Store	Gym	Gym / Fitness Center	Gym Pool	Hakka 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	

	Harbor / Marina	Hardware Store	Hawaiian Restaurant	\
0	0	0	0	
1	0	0	0	
2	0	0	0	
3	0	0	0	
4	0	0	0	

	Health & Beauty Service	Health Food Store	Historic Site	History Museum	\
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	

	Hobby Shop	Hockey Arena	Home Service	Hong Kong Restaurant	Hookah Bar	\
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	

	Hostel	Hot Dog Joint	Hotel	Hotel Bar	Hotpot Restaurant	Ice Cream Shop	\
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	

	Indian Restaurant	Indie Movie Theater	Indonesian Restaurant	\
0	0	0	0	
1	0	0	0	
2	0	0	0	

3	0	0	0
4	0	0	0

	Intersection	Irish Pub	Italian Restaurant	Japanese Restaurant	\
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	

	Jazz Club	Jewelry Store	Jewish Restaurant	Juice Bar	Karaoke Bar	\
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	

	Kids Store	Kitchen Supply Store	Korean Restaurant	\
0	0	0	0	
1	0	0	0	
2	0	0	0	
3	0	0	0	
4	0	0	0	

	Latin American Restaurant	Laundry Service	Leather Goods Store	Library	\
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	

	Light Rail Station	Lingerie Store	Liquor Store	Lounge	\
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	

	Mac & Cheese Joint	Malay Restaurant	Market	Martial Arts Dojo	\
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	

	Massage Studio	Mediterranean Restaurant	Men's Store	Metro Station	\
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

	Mexican Restaurant	Middle Eastern Restaurant	Miscellaneous Shop	\
0	0	0	0	
1	0	0	0	
2	0	0	0	
3	0	0	0	
4	0	0	0	

	Mobile Phone Shop	Modern European Restaurant	\
0	0	0	
1	0	0	
2	0	0	
3	0	0	
4	0	0	

	Molecular Gastronomy Restaurant	Monument / Landmark	Movie Theater	\
0	0	0	0	
1	0	0	0	
2	0	0	0	
3	0	0	0	
4	0	0	0	

	Moving Target	Museum	Music Store	Music Venue	\
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	

	Neighborhood	New American Restaurant	Nightclub	\
0	Rouge, Malvern	0	0	
1	Rouge, Malvern	0	0	
2	Rouge, Malvern	0	0	
3	Rouge, Malvern	0	0	
4	Highland Creek, Rouge Hill, Port Union	0	0	

	Noodle House	Office	Opera House	Optical Shop	Organic Grocery	\
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	

	Other Great Outdoors	Other Nightlife	Paper / Office Supplies Store	Park \
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

	Pastry Shop	Performing Arts Venue	Peruvian Restaurant	Pet 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

	Pharmacy	Pier	Pilates Studio	Pizza Place	Platform	Playground	Plaza \
0	0	0	0	0	0	0	0
1	0	0	0	0	0	0	0
2	0	0	0	0	0	0	0
3	0	0	0	0	0	0	0
4	0	0	0	0	0	0	0

	Poke Place	Pool	Pool Hall	Portuguese Restaurant	Poutine Place \
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

	Print Shop	Pub	Ramen Restaurant	Record Shop	Rental Car Location \
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

	Residential Building (Apartment / Condo)	Restaurant	Road \
0	0	0	0
1	0	0	0
2	0	0	0
3	0	0	0
4	0	0	0

	Rock Climbing Spot	Rock Club	Sake Bar	Salad Place	Salon / Barbershop \
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
---	---	---	---	---	---

	Sandwich Place	Scenic Lookout	Science Museum	Seafood Restaurant	\
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	

	Shanghai Restaurant	Shoe Store	Shopping Mall	Skate Park	Skating Rink	\
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	

	Smoke Shop	Smoothie Shop	Snack Place	Soccer Field	Soccer Stadium	\
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	

	Social Club	Soup Place	Southern / Soul Food Restaurant	Souvlaki Shop	\
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

	Spa	Spanish Restaurant	Speakeasy	Sporting Goods Shop	Sports Bar	\
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	

	Sports Club	Stationery Store	Steakhouse	Storage Facility	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	

	Sushi Restaurant	Taco Place	Tailor Shop	Taiwanese Restaurant	\
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

	Tanning Salon	Tapas Restaurant	Tea Room	Tech Startup	Tennis Court	\
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	

	Thai Restaurant	Theater	Theme Park	Theme Park Ride / Attraction	\
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	

	Theme Restaurant	Thrift / Vintage Store	Toy / Game Store	Track	Trail	\
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	

	Train Station	University	Vegetarian / Vegan Restaurant	Video Game 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	

	Video Store	Vietnamese Restaurant	Warehouse Store	Wine Bar	Wine 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	0	0	0	

	Wings Joint	Women's Store	Yoga Studio
0	0	0	0
1	0	0	0
2	0	0	0
3	0	0	0
4	0	0	0


```
[32]: num_top_venues = 5
for hood in Scarborough_grouped['Neighborhood']:
    print("---- "+hood+" ----")
    temp =Scarborough_grouped[Scarborough_grouped['Neighborhood'] == hood].T.
    ↪reset_index()
    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')
```

---- Adelaide, King, Richmond ----

	venue	freq
0	Coffee Shop	0.09
1	Café	0.07
2	Hotel	0.06
3	Burger Joint	0.03
4	Bar	0.03

---- Agincourt ----

	venue	freq
0	Shopping Mall	0.07
1	Chinese Restaurant	0.07
2	Hong Kong Restaurant	0.03
3	Filipino Restaurant	0.03
4	Shanghai Restaurant	0.03

---- Agincourt North, L'Amoreaux East, Milliken, Steeles East ----

	venue	freq
0	Pharmacy	0.50
1	Sandwich Place	0.25
2	Sushi Restaurant	0.25
3	Zoo Exhibit	0.00
4	Movie Theater	0.00

---- Albion Gardens, Beaumont Heights, Humbertgate, Jamestown, Mount Olive, Silverstone, South Steeles, Thistletown ----

	venue	freq
0	Grocery Store	0.17
1	Hardware Store	0.08
2	Fried Chicken Joint	0.08
3	Sandwich Place	0.08

4 Discount Store 0.08

---- Alderwood, Long Branch ----

	venue	freq
0	Convenience Store	0.11
1	Coffee Shop	0.11
2	Sandwich Place	0.11
3	Dance Studio	0.11
4	Pub	0.11

---- Bathurst Manor, Downsview North, Wilson Heights ----

	venue	freq
0	Pizza Place	0.18
1	Coffee Shop	0.09
2	Mediterranean Restaurant	0.09
3	Deli / Bodega	0.09
4	Middle Eastern Restaurant	0.09

---- Bayview Village ----

	venue	freq
0	Asian Restaurant	0.25
1	Flower Shop	0.25
2	Park	0.25
3	Trail	0.25
4	Zoo Exhibit	0.00

---- Bedford Park, Lawrence Manor East ----

	venue	freq
0	Sushi Restaurant	0.07
1	Coffee Shop	0.07
2	Italian Restaurant	0.07
3	Café	0.04
4	Juice Bar	0.04

---- Berczy Park ----

	venue	freq
0	Coffee Shop	0.09
1	Restaurant	0.05
2	Café	0.05
3	Hotel	0.05
4	Beer Bar	0.04

---- Birch Cliff, Cliffside West ----

	venue	freq
0	Park	0.22
1	Baseball Field	0.11
2	Skating Rink	0.11
3	General Entertainment	0.11
4	Gym	0.11

---- Bloordale Gardens, Eringate, Markland Wood, Old Burnhamthorpe ----

	venue	freq
0	Baseball Field	0.09
1	Fish & Chips Shop	0.09
2	Sandwich Place	0.09
3	Shopping Mall	0.09
4	Grocery Store	0.09

---- Brockton, Exhibition Place, Parkdale Village ----

	venue	freq
0	Coffee Shop	0.08
1	Café	0.07
2	Bar	0.06
3	Restaurant	0.04
4	Furniture / Home Store	0.04

---- Business Reply Mail Processing Centre 969 Eastern ----

	venue	freq
0	Coffee Shop	0.10
1	Café	0.06
2	Hotel	0.04
3	Bar	0.04
4	Sushi Restaurant	0.03

---- CFB Toronto, Downsview East ----

	venue	freq
0	Soccer Field	0.2
1	Food Court	0.2
2	Airport	0.2
3	Park	0.2
4	Coffee Shop	0.2

---- CN Tower, Bathurst Quay, Island airport, Harbourfront West, King and Spadina, Railway Lands, South Niagara ----

	venue	freq
--	-------	------

0	Coffee Shop	0.10
1	Restaurant	0.05
2	Café	0.05
3	Italian Restaurant	0.05
4	Gym	0.04

---- Cabbagetown, St. James Town ----

	venue	freq
0	Restaurant	0.06
1	Coffee Shop	0.06
2	Pizza Place	0.05
3	Japanese Restaurant	0.03
4	Bakery	0.03

---- Caledonia-Fairbanks ----

	venue	freq
0	Park	0.17
1	Sporting Goods Shop	0.08
2	Bus Stop	0.08
3	Café	0.08
4	Market	0.08

---- Canada Post Gateway Processing Centre ----

	venue	freq
0	Coffee Shop	0.10
1	Café	0.06
2	Hotel	0.04
3	Bar	0.04
4	Sushi Restaurant	0.03

---- Cedarbrae ----

	venue	freq
0	Caribbean Restaurant	0.12
1	Flower Shop	0.12
2	Indian Restaurant	0.12
3	Bank	0.12
4	Bakery	0.12

---- Central Bay Street ----

	venue	freq
0	Coffee Shop	0.06
1	Clothing Store	0.04
2	Cosmetics Shop	0.04

3	Ice Cream Shop	0.03
4	Italian Restaurant	0.03

---- Chinatown, Grange Park, Kensington Market ----

	venue	freq
0	Café	0.06
1	Chinese Restaurant	0.05
2	Bar	0.05
3	Vietnamese Restaurant	0.04
4	Vegetarian / Vegan Restaurant	0.04

---- Christie ----

	venue	freq
0	Korean Restaurant	0.18
1	Grocery Store	0.10
2	Diner	0.04
3	Indian Restaurant	0.04
4	Café	0.04

---- Church and Wellesley ----

	venue	freq
0	Coffee Shop	0.09
1	Japanese Restaurant	0.04
2	Café	0.03
3	Sushi Restaurant	0.03
4	Restaurant	0.03

---- Clairlea, Golden Mile, Oakridge ----

	venue	freq
0	Intersection	0.17
1	Bus Line	0.11
2	Coffee Shop	0.11
3	Diner	0.11
4	Bakery	0.11

---- Clarks Corners, Sullivan, Tam O'Shanter ----

	venue	freq
0	Fast Food Restaurant	0.10
1	Pizza Place	0.10
2	Vietnamese Restaurant	0.07
3	Golf Course	0.03
4	Market	0.03

---- Cliffcrest, Cliffside, Scarborough Village West ----

	venue	freq
0	Fast Food Restaurant	0.17
1	Park	0.08
2	Flower Shop	0.08
3	Liquor Store	0.08
4	Pizza Place	0.08

---- Cloverdale, Islington, Martin Grove, Princess Gardens, West Deane Park ----

	venue	freq
0	Pizza Place	0.25
1	Convenience Store	0.12
2	Mexican Restaurant	0.12
3	Bank	0.12
4	Gym	0.12

---- Commerce Court, Victoria Hotel ----

	venue	freq
0	Coffee Shop	0.10
1	Hotel	0.06
2	Café	0.05
3	Restaurant	0.04
4	Gastropub	0.04

---- Davisville ----

	venue	freq
0	Dessert Shop	0.10
1	Pizza Place	0.10
2	Coffee Shop	0.07
3	Sandwich Place	0.07
4	Gym	0.05

---- Davisville North ----

	venue	freq
0	Gym	0.09
1	Yoga Studio	0.04
2	Pharmacy	0.04
3	Breakfast Spot	0.04
4	Brewery	0.04

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

	venue	freq
--	-------	------

0	Pub	0.10
1	Coffee Shop	0.10
2	Skating Rink	0.10
3	Sushi Restaurant	0.10
4	Yoga Studio	0.05

---- Del Ray, Keelesdale, Mount Dennis, Silverthorn ----

	venue	freq
0	Construction & Landscaping	0.2
1	Playground	0.2
2	Storage Facility	0.2
3	Coffee Shop	0.2
4	Fast Food Restaurant	0.2

---- Design Exchange, Toronto Dominion Centre ----

	venue	freq
0	Coffee Shop	0.13
1	Café	0.08
2	Hotel	0.07
3	Restaurant	0.04
4	Gastropub	0.03

---- Don Mills North ----

	venue	freq
0	Coffee Shop	0.11
1	Burger Joint	0.05
2	Italian Restaurant	0.05
3	Soccer Field	0.05
4	Smoothie Shop	0.05

---- Dorset Park, Scarborough Town Centre, Wexford Heights ----

	venue	freq
0	Bakery	0.33
1	Park	0.33
2	Wine Shop	0.17
3	Coffee Shop	0.17
4	Zoo Exhibit	0.00

---- Dovercourt Village, Dufferin ----

	venue	freq
0	Bar	0.10
1	Coffee Shop	0.10
2	Bakery	0.08

3	Café	0.06
4	Cocktail Bar	0.04

---- Downsview Central ----

	venue	freq
0	Home Service	0.29
1	Moving Target	0.14
2	Mobile Phone Shop	0.14
3	Park	0.14
4	Baseball Field	0.14

---- Downsview Northwest ----

	venue	freq
0	Grocery Store	0.07
1	Pizza Place	0.07
2	Fast Food Restaurant	0.07
3	Vietnamese Restaurant	0.07
4	Shopping Mall	0.07

---- Downsview West ----

	venue	freq
0	Coffee Shop	0.10
1	Grocery Store	0.05
2	Discount Store	0.05
3	Sandwich Place	0.05
4	Big Box Store	0.05

---- Downsview, North Park, Upwood Park ----

	venue	freq
0	Convenience Store	0.2
1	Construction & Landscaping	0.2
2	Home Service	0.2
3	Bakery	0.2
4	Park	0.2

---- East Birchmount Park, Ionview, Kennedy Park ----

	venue	freq
0	Convenience Store	0.25
1	Discount Store	0.25
2	Coffee Shop	0.25
3	Department Store	0.25
4	Music Venue	0.00

---- East Toronto ----

	venue	freq
0	Sandwich Place	0.14
1	Italian Restaurant	0.14
2	Coffee Shop	0.14
3	Park	0.14
4	Athletics & Sports	0.14

---- Emery, Humberlea ----

	venue	freq
0	Coffee Shop	0.29
1	Park	0.14
2	Discount Store	0.14
3	Café	0.14
4	Nightclub	0.14

---- Fairview, Henry Farm, Oriole ----

	venue	freq
0	Clothing Store	0.16
1	Fast Food Restaurant	0.10
2	Coffee Shop	0.07
3	Women's Store	0.04
4	Bakery	0.03

---- First Canadian Place, Underground city ----

	venue	freq
0	Coffee Shop	0.14
1	Hotel	0.08
2	Café	0.07
3	Gastropub	0.03
4	American Restaurant	0.03

---- Flemington Park, Don Mills South ----

	venue	freq
0	Japanese Restaurant	0.08
1	Science Museum	0.08
2	Asian Restaurant	0.08
3	Beer Store	0.08
4	Gym	0.08

---- Forest Hill North, Forest Hill West ----

	venue	freq
--	-------	------

0	Café	0.2
1	Coffee Shop	0.1
2	Bookstore	0.1
3	Burger Joint	0.1
4	Ice Cream Shop	0.1

---- Glencairn ----

	venue	freq
0	Fast Food Restaurant	0.18
1	Grocery Store	0.18
2	Metro Station	0.09
3	Japanese Restaurant	0.09
4	Mediterranean Restaurant	0.09

---- Guildwood, Morningside, West Hill ----

	venue	freq
0	Park	0.29
1	Pool	0.14
2	Indian Restaurant	0.14
3	Fried Chicken Joint	0.14
4	Gym / Fitness Center	0.14

---- Harbord, University of Toronto ----

	venue	freq
0	Café	0.07
1	Restaurant	0.06
2	Coffee Shop	0.05
3	Pizza Place	0.04
4	Bar	0.04

---- Harbourfront East, Toronto Islands, Union Station ----

	venue	freq
0	Café	0.29
1	Harbor / Marina	0.14
2	Park	0.14
3	Pier	0.14
4	Beach	0.14

---- Harbourfront, Regent Park ----

	venue	freq
0	Coffee Shop	0.12
1	Theater	0.06
2	Bakery	0.06

3	Café	0.04
4	Restaurant	0.04

---- High Park, The Junction South ----

	venue	freq
0	Convenience Store	0.09
1	Park	0.07
2	Café	0.07
3	Bar	0.07
4	Thai Restaurant	0.04

---- Highland Creek, Rouge Hill, Port Union ----

	venue	freq
0	Bar	1.0
1	Museum	0.0
2	Opera House	0.0
3	Office	0.0
4	Noodle House	0.0

---- Hillcrest Village ----

	venue	freq
0	Park	0.29
1	Fast Food Restaurant	0.14
2	Bakery	0.14
3	Residential Building (Apartment / Condo)	0.14
4	Pharmacy	0.14

---- Humber Bay Shores, Mimico South, New Toronto ----

	venue	freq
0	Park	0.14
1	Grocery Store	0.10
2	Convenience Store	0.10
3	Skating Rink	0.05
4	Breakfast Spot	0.05

---- Humber Bay, King's Mill Park, Kingsway Park South East, Mimico NE, Old Mill South, The Queensway East, Royal York South East, Sunnylea ----

	venue	freq
0	Coffee Shop	0.19
1	Sandwich Place	0.06
2	Italian Restaurant	0.06
3	Liquor Store	0.06
4	Bank	0.06

---- Humber Summit ----

	venue	freq
0	Electronics Store	1.0
1	Zoo Exhibit	0.0
2	Opera House	0.0
3	Noodle House	0.0
4	Nightclub	0.0

---- Humewood-Cedarvale ----

	venue	freq
0	Convenience Store	0.14
1	Trail	0.14
2	Hockey Arena	0.14
3	Grocery Store	0.14
4	Deli / Bodega	0.14

---- Islington Avenue ----

	venue	freq
0	Pharmacy	0.25
1	Convenience Store	0.08
2	Shopping Mall	0.08
3	Grocery Store	0.08
4	Bank	0.08

---- Kingsview Village, Martin Grove Gardens, Richview Gardens, St. Phillips

	venue	freq
0	Pharmacy	0.2
1	Mobile Phone Shop	0.2
2	Bus Line	0.2
3	American Restaurant	0.2
4	Coffee Shop	0.2

---- Kingsway Park South West, Mimico NW, The Queensway West, Royal York South West, South of Bloor ----

	venue	freq
0	Burrito Place	0.12
1	Gym	0.08
2	Bank	0.04
3	Grocery Store	0.04
4	Thai Restaurant	0.04

---- L'Amoreaux West ----

	venue	freq
0	Pharmacy	0.13
1	Fast Food Restaurant	0.13
2	Chinese Restaurant	0.13
3	Other Great Outdoors	0.07
4	Grocery Store	0.07

---- Lawrence Heights, Lawrence Manor ----

	venue	freq
0	Clothing Store	0.17
1	Dessert Shop	0.05
2	Toy / Game Store	0.03
3	Restaurant	0.03
4	Greek Restaurant	0.03

---- Lawrence Park ----

	venue	freq
0	Bus Line	0.14
1	Park	0.14
2	Café	0.14
3	Coffee Shop	0.14
4	Restaurant	0.14

---- Leaside ----

	venue	freq
0	Coffee Shop	0.10
1	Electronics Store	0.06
2	Sporting Goods Shop	0.06
3	Furniture / Home Store	0.06
4	Brewery	0.04

---- Little Portugal, Trinity ----

	venue	freq
0	Bar	0.08
1	Coffee Shop	0.05
2	Restaurant	0.05
3	Café	0.04
4	Italian Restaurant	0.04

---- Maryvale, Wexford ----

	venue	freq
--	-------	------

0	Hookah Bar	0.2
1	Pizza Place	0.2
2	Middle Eastern Restaurant	0.2
3	Gas Station	0.2
4	Burger Joint	0.2

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

	venue	freq
0	Park	0.4
1	Grocery Store	0.2
2	Gym	0.2
3	Thai Restaurant	0.2
4	Zoo Exhibit	0.0

---- Newtonbrook, Willowdale ----

	venue	freq
0	Korean Restaurant	0.12
1	Middle Eastern Restaurant	0.12
2	Café	0.09
3	Coffee Shop	0.09
4	Fried Chicken Joint	0.03

---- North Toronto West ----

	venue	freq
0	Coffee Shop	0.09
1	Sporting Goods Shop	0.09
2	Café	0.09
3	Clothing Store	0.09
4	Restaurant	0.06

---- Northwest ----

	venue	freq
0	Sandwich Place	0.17
1	Coffee Shop	0.17
2	Storage Facility	0.17
3	Burger Joint	0.17
4	Gym	0.17

---- Northwood Park, York University ----

	venue	freq
0	Japanese Restaurant	0.12
1	Bar	0.12
2	Bank	0.12

3	Pizza Place	0.12
4	Massage Studio	0.12

---- Parkdale, Roncesvalles ----

	venue	freq
0	Coffee Shop	0.09
1	Bakery	0.06
2	Eastern European Restaurant	0.04
3	Sushi Restaurant	0.04
4	Thai Restaurant	0.04

---- Parkwoods ----

	venue	freq
0	Food & Drink Shop	0.33
1	Park	0.33
2	Pet Store	0.33
3	Middle Eastern Restaurant	0.00
4	Miscellaneous Shop	0.00

---- Queen's Park ----

	venue	freq
0	Coffee Shop	0.20
1	Café	0.08
2	Sandwich Place	0.05
3	Italian Restaurant	0.04
4	Japanese Restaurant	0.04

---- Rosedale ----

	venue	freq
0	Park	0.4
1	Playground	0.2
2	Grocery Store	0.2
3	Candy Store	0.2
4	Zoo Exhibit	0.0

---- Roselawn ----

	venue	freq
0	Playground	0.4
1	Garden	0.2
2	Business Service	0.2
3	Pet Store	0.2
4	Zoo Exhibit	0.0

---- Rouge, Malvern ----

	venue	freq
0	Zoo Exhibit	0.25
1	Financial or Legal Service	0.25
2	Fast Food Restaurant	0.25
3	Construction & Landscaping	0.25
4	Hotpot Restaurant	0.00

---- Runnymede, Swansea ----

	venue	freq
0	Café	0.10
1	Coffee Shop	0.08
2	Bakery	0.05
3	Pizza Place	0.05
4	Sushi Restaurant	0.03

---- Ryerson, Garden District ----

	venue	freq
0	Coffee Shop	0.07
1	Clothing Store	0.05
2	Cosmetics Shop	0.04
3	Middle Eastern Restaurant	0.04
4	Restaurant	0.03

---- Scarborough Village ----

	venue	freq
0	Fast Food Restaurant	0.33
1	Sandwich Place	0.22
2	Train Station	0.11
3	Big Box Store	0.11
4	Indian Restaurant	0.11

---- Silver Hills, York Mills ----

	venue	freq
0	Park	1.0
1	Zoo Exhibit	0.0
2	Moving Target	0.0
3	Noodle House	0.0
4	Nightclub	0.0

---- St. James Town ----

	venue	freq
--	-------	------

0	Coffee Shop	0.07
1	Hotel	0.06
2	Café	0.06
3	Italian Restaurant	0.04
4	Bakery	0.04

---- Stn A PO Boxes 25 The Esplanade ----

	venue	freq
0	Coffee Shop	0.10
1	Café	0.06
2	Hotel	0.04
3	Bar	0.04
4	Sushi Restaurant	0.03

---- Studio District ----

	venue	freq
0	Coffee Shop	0.07
1	Café	0.05
2	Italian Restaurant	0.05
3	Diner	0.05
4	Bakery	0.05

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

	venue	freq
0	Italian Restaurant	0.08
1	Café	0.08
2	Coffee Shop	0.08
3	Sandwich Place	0.08
4	History Museum	0.05

---- The Beaches ----

	venue	freq
0	Pub	0.06
1	Pizza Place	0.06
2	Japanese Restaurant	0.06
3	Bar	0.06
4	Café	0.03

---- The Beaches West, India Bazaar ----

	venue	freq
0	Park	0.08
1	Café	0.05
2	Bakery	0.05

3	Italian Restaurant	0.05
4	Brewery	0.05

---- The Danforth West, Riverdale ----

	venue	freq
0	Greek Restaurant	0.13
1	Ice Cream Shop	0.07
2	Café	0.07
3	Yoga Studio	0.03
4	Bubble Tea Shop	0.03

---- The Junction North, Runnymede ----

	venue	freq
0	Brewery	0.38
1	Park	0.25
2	Athletics & Sports	0.12
3	Coffee Shop	0.12
4	Furniture / Home Store	0.12

---- The Kingsway, Montgomery Road, Old Mill North ----

	venue	freq
0	Sushi Restaurant	0.07
1	Breakfast Spot	0.07
2	French Restaurant	0.07
3	Burger Joint	0.07
4	Toy / Game Store	0.04

---- Thorncliffe Park ----

	venue	freq
0	Burger Joint	0.11
1	Indian Restaurant	0.11
2	Yoga Studio	0.06
3	Discount Store	0.06
4	Bridge	0.06

---- Victoria Village ----

	venue	freq
0	Middle Eastern Restaurant	0.14
1	Portuguese Restaurant	0.14
2	Pizza Place	0.14
3	French Restaurant	0.14
4	Park	0.14

---- Westmount ----

	venue	freq
0	Pizza Place	0.18
1	Golf Driving Range	0.09
2	Discount Store	0.09
3	Supermarket	0.09
4	Chinese Restaurant	0.09

---- Weston ----

	venue	freq
0	Train Station	0.15
1	Coffee Shop	0.15
2	Sandwich Place	0.08
3	Middle Eastern Restaurant	0.08
4	Pizza Place	0.08

---- Willowdale South ----

	venue	freq
0	Coffee Shop	0.07
1	Ramen Restaurant	0.05
2	Pizza Place	0.05
3	Café	0.05
4	Sandwich Place	0.05

---- Willowdale West ----

	venue	freq
0	Coffee Shop	0.15
1	Park	0.08
2	Eastern European Restaurant	0.08
3	Grocery Store	0.08
4	Bakery	0.08

---- Woburn ----

	venue	freq
0	Coffee Shop	0.4
1	Park	0.2
2	Business Service	0.2
3	Fast Food Restaurant	0.2
4	Zoo Exhibit	0.0

---- Woodbine Gardens, Parkview Hill ----

	venue	freq
--	-------	------

0	Fast Food Restaurant	0.10
1	Brewery	0.10
2	Pizza Place	0.10
3	Home Service	0.10
4	Restaurant	0.05

---- Woodbine Heights ----

	venue	freq
0	Bus Line	0.06
1	Pizza Place	0.06
2	Coffee Shop	0.06
3	Grocery Store	0.06
4	Sandwich Place	0.06

---- York Mills West ----

	venue	freq
0	Coffee Shop	0.16
1	Restaurant	0.12
2	Gym	0.12
3	Burrito Place	0.04
4	Bank	0.04

```
[33]: 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]
```

1.0.7 Most Common venues near neighborhood

```
[34]: import numpy as np
      num_top_venues = 10

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

      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))

      neighborhoods_venues_sorted = pd.DataFrame(columns=columns)
```

```

neighborhoods_venues_sorted['Neighborhood'] =
↳ Scarborough_grouped['Neighborhood']

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

neighborhoods_venues_sorted.head()

```

```

[34]:
      Neighborhood 1st Most Common Venue \
0      Adelaide, King, Richmond      Coffee Shop
1      Agincourt      Chinese Restaurant
2  Agincourt North, L'Amoreaux East, Milliken, St...      Pharmacy
3  Albion Gardens, Beaumond Heights, Humbergate, ...      Grocery Store
4      Alderwood, Long Branch      Convenience Store

      2nd Most Common Venue 3rd Most Common Venue 4th Most Common Venue \
0      Café      Hotel      Gastropub
1      Shopping Mall      Pizza Place      Supermarket
2      Sandwich Place      Sushi Restaurant      Doner Restaurant
3      Park      Sandwich Place      Discount Store
4      Pub      Sandwich Place      Coffee Shop

      5th Most Common Venue 6th Most Common Venue      7th Most Common Venue \
0      Burger Joint      Asian Restaurant      Bar
1      Sushi Restaurant      Breakfast Spot      Print Shop
2      Donut Shop      Dumpling Restaurant      Eastern European Restaurant
3      Japanese Restaurant      Fried Chicken Joint      Beer Store
4      Gas Station      Dance Studio      Gym

      8th Most Common Venue 9th Most Common Venue 10th Most Common Venue
0      Restaurant      American Restaurant      Steakhouse
1      Mediterranean Restaurant      Coffee Shop      Pool
2      Electronics Store      Elementary School      Ethiopian Restaurant
3      Hardware Store      Pizza Place      Fast Food Restaurant
4      Pharmacy      Pizza Place      Falafel Restaurant

```

1.0.8 K-Means Clustering Approach

```

[35]: # Using K-Means to cluster neighborhood into 3 clusters
Scarborough_grouped_clustering = Scarborough_grouped.drop('Neighborhood', 1)
kmeans = KMeans(n_clusters=3, random_state=0).
↳ fit(Scarborough_grouped_clustering)
kmeans.labels_

```

```
[36]: neighborhoods_venues_sorted.insert(0, 'Cluster Labels', kmeans.labels_)

Scarborough_merged = df_2.iloc[:16,:]

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

Scarborough_merged.head()# check the last columns!
```

	Longitude	Cluster Labels	1st Most Common Venue
0	-79.195517	0	Zoo Exhibit
1	-79.158725	0	Bar
2	-79.175193	2	Park
3	-79.217590	0	Coffee Shop
4	-79.239440	0	Flower Shop

	4th Most Common Venue	5th Most Common Venue
0	Construction & Landscaping	Fish & Chips Shop
1	Dumpling Restaurant	Eastern European Restaurant
2	Fried Chicken Joint	Indian Restaurant
3	Park	Yoga Studio
4	Bank	Bakery

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1	Electronics Store	Elementary School	Ethiopian Restaurant
2	Athletics & Sports	Ethiopian Restaurant	Donut Shop
3	Dumpling Restaurant	Eastern European Restaurant	Electronics Store
4	Caribbean Restaurant	Hakka Restaurant	Indian Restaurant

	9th Most Common Venue	10th Most Common Venue
0	Farmers Market	Doner Restaurant
1	Event Space	Yoga Studio
2	Dumpling Restaurant	Eastern European Restaurant
3	Elementary School	Ethiopian Restaurant
4	Eastern European Restaurant	Electronics Store

1.0.9 Map of Clusters

```
[37]: kclusters = 10
```

```
[38]: # create map
map_clusters = folium.Map(location=[latitude_x, longitude_y], zoom_start=11)

# set color scheme for the clusters
x = np.arange(kclusters)
colors_array = cm.rainbow(np.linspace(0, 1, kclusters))
rainbow = [colors.rgb2hex(i) for i in colors_array]
print(rainbow)
# add markers to the map

markers_colors = []
for lat, lon, nei, cluster in zip(Scarborough_merged['Latitude'],
                                   Scarborough_merged['Longitude'],
                                   Scarborough_merged['Neighborhood'],
                                   Scarborough_merged['Cluster Labels']):
    label = folium.Popup(str(nei) + ' 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
```

```
['#8000ff', '#4856fb', '#10a2f0', '#2adddd', '#62fbc4', '#9cfba4', '#d4dd80',
 '#ffa256', '#ff562c', '#ff0000']
```

```
[38]: <folium.folium.Map at 0x7f2424441128>
```

```

[39]: df1=Scarborough_merged.loc[Scarborough_merged['Cluster Labels'] ==
↳0,Scarborough_merged.columns[[2] + list(range(5, Scarborough_merged.
↳shape[1]))]]
df2=Scarborough_merged.loc[Scarborough_merged['Cluster Labels'] ==
↳1,Scarborough_merged.columns[[2] + list(range(5, Scarborough_merged.
↳shape[1]))]]
df3=Scarborough_merged.loc[Scarborough_merged['Cluster Labels'] ==
↳2,Scarborough_merged.columns[[2] + list(range(5, Scarborough_merged.
↳shape[1]))]]

[40]: Scarborough_Avg_HousingPrice=pd.DataFrame({"Neighborhood":df_2["Neighborhood"],
"Average_Housing_Price": [335000.0,286600.
↳0,175000.0,225900.0,219400.0,
573900.0,225000.
↳0,370500.0,370500.0,433500.0,279200.0,
279200.0,225000.
↳0,370500.,255400.0,433500.0,433500.0,
435000.0,289500.
↳0,265000.0,285900.0,239400.0,
589900.0,295000.
↳0,380500.0,378500.0,438500.0,229200.0,
229200.0,365000.
↳0,388500.,285400.0,493500.0,477500.0,378000.0,316600.0,195000.0,225900.
↳0,219400.0,
573900.0,367000.
↳0,370500.0,370500.0,363500.0,279200.0,
279200.0,271000.
↳0,370500.,255400.0,383500.0,433500.0,335000.0,286600.0,185000.0,225900.
↳0,219400.0,
573900.0,329000.
↳0,370500.0,370500.0,533500.0,279200.0,
279200.0,375000.
↳0,370500.,255400.0,493500.0,433500.0,335000.0,286600.0,165000.0,225900.
↳0,219400.0,
573900.0,425000.
↳0,370500.0,370500.0,433500.0,279200.0,
279200.0,195000.
↳0,370500.,255400.0,403500.0,433500.0,335000.0,286600.0,187000.0,225900.
↳0,219400.0,
573900.0,325000.
↳0,370500.0,370500.0,333500.0,279200.0,
279200.0,289000.
↳0,370500.,255400.0,413500.0,433500.0,254800.0

    ])

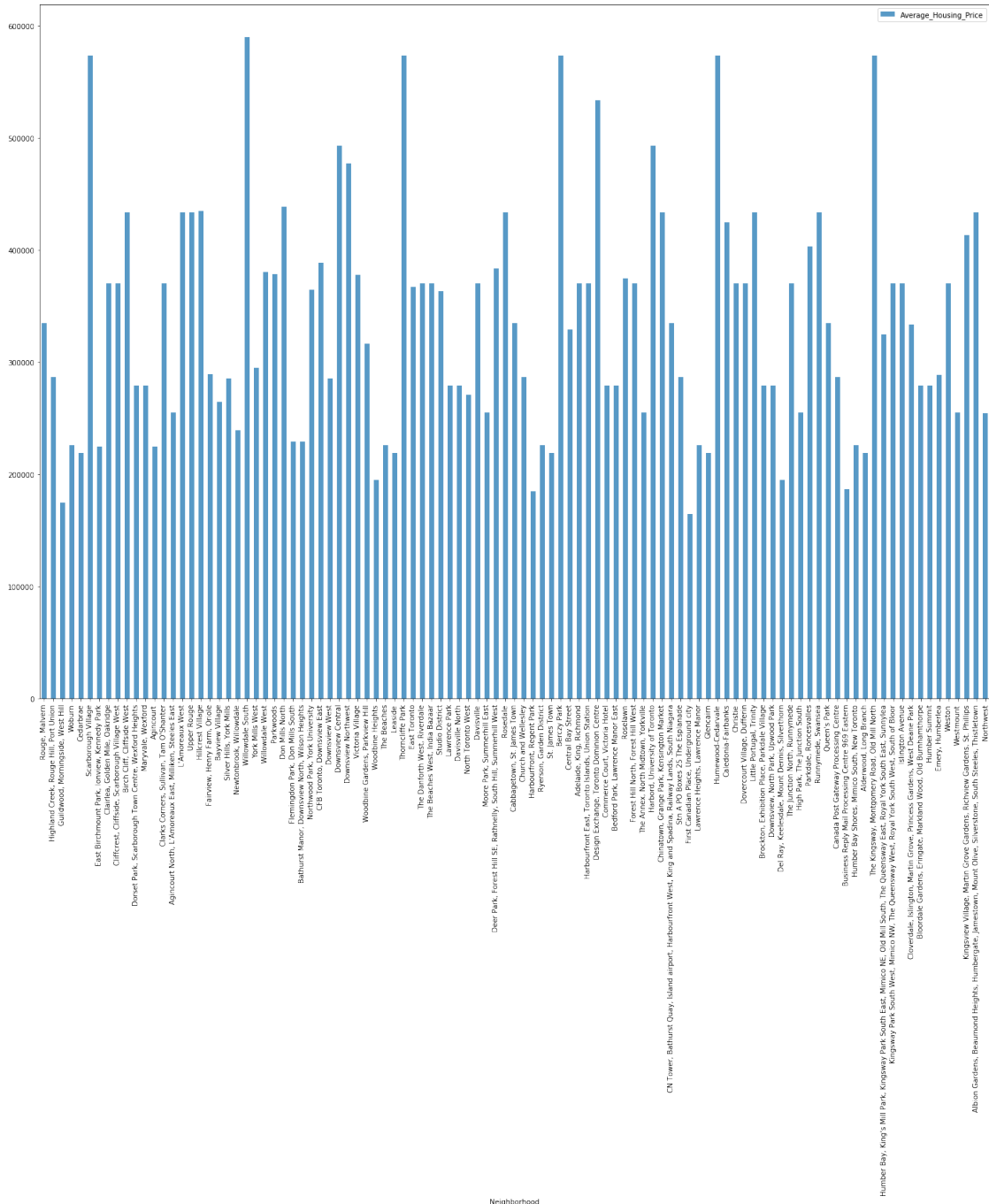
```



```
[41]: Scarborough_Avg_HousingPrice.set_index('Neighborhood', inplace=True, drop=True)
```

```
[42]: Scarborough_Avg_HousingPrice.plot(kind='bar', figsize=(24,18), alpha=0.75)
```

```
[42]: <matplotlib.axes._subplots.AxesSubplot at 0xf24243f3320>
```



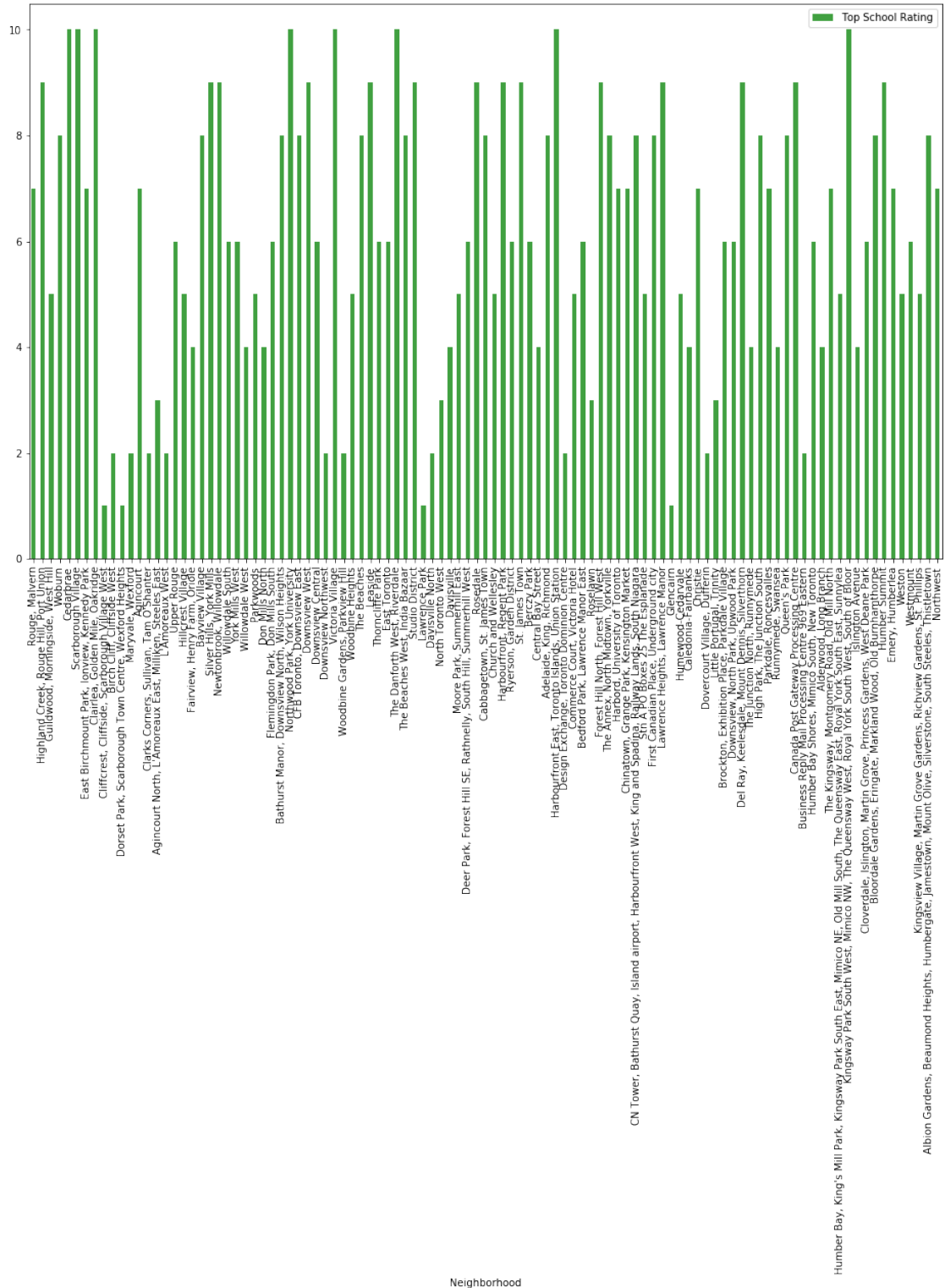
1.0.10 School Rating by Clusters

```
[43]: clusters=pd.DataFrame({"Cluster1":df1["Neighborhood"],
                             "Cluster2":df2["Neighborhood"],
                             "Cluster4":df3["Neighborhood"]})
clusters = clusters.replace(np.nan, '', regex=True)

[44]: new_Scarborough=Scarborough_merged.set_index("Neighborhood",drop=True)
#Source:https://www.greatschools.org
Scarborough_school_ratings=pd.DataFrame({"Neighborhood":df["Neighborhood"],
                                          "Top School Rating":
      ↪ [7,9,5,8,10,10,7,10,1,2,1,2,7,2,3,2,6,
      ↪ 5,4,8,9,9,6,6,4,5,4,6,8,10,8,9,6,2,
      ↪ 10,2,5,8,9,6,6,10,8,9,1,2,3,4,5,6,9,
      ↪ 8,5,9,6,9,6,4,8,10,2,5,6,3,9,8,7,
      ↪ 7,8,5,8,9,1,5,4,7,2,3,6,6,9,4,8,7,
      ↪ 4,8,9,2,6,4,7,5,10,4,6,8,9,7,5,6,5,8,7
                                          ]})

[45]: Scarborough_school_ratings.set_index('Neighborhood',inplace=True,drop=True)

[46]: Scarborough_school_ratings.
      ↪ plot(kind='bar',figsize=(16,10),color='green',alpha=0.75);
```



Conclusion: In this project, using k-means cluster algorithm I separated the neighborhood into 10(Ten) different clusters and for 103 different latitude and longitude from dataset, which have

very-similar neighborhoods around them. Using the charts above results presented to a particular neighborhood based on average house prices and school rating have been made.