

IBM DataScience - Capstone - wk3 Part 3 - Sven De Smit

August 25, 2019

1 IBM DataScience - Capstone - wk3 Part 3 - Sven De Smit

2 Segmenting and Clustering Neighborhoods in Toronto

```
In [1]: #!conda install -c conda-forge folium=0.5.0 --yes
```

```
In [2]: import pandas as pd
```

```
import requests
```

```
import folium
```

```
import matplotlib
```

```
import numpy as np
```

```
import matplotlib.pyplot as plt
```

```
%matplotlib inline
```

2.1 Read neighborhood dataset from the file created by the previous exercise

```
In [3]: df_postal_loc = pd.read_csv('toronto_postal_codes_with_location.csv')
print(df_postal_loc.shape)
df_postal_loc.head()
```

(103, 5)

```
Out[3]:
```

	PostalCode	Borough	Neighbourhood	Latitude	\
0	M1B	Scarborough	Rouge, Malvern	43.806686	
1	M1C	Scarborough	Highland Creek, Rouge Hill, Port Union	43.784535	
2	M1E	Scarborough	Guildwood, Morningside, West Hill	43.763573	
3	M1G	Scarborough	Woburn	43.770992	
4	M1H	Scarborough	Cedarbrae	43.773136	

	Longitude
0	-79.194353
1	-79.160497
2	-79.188711
3	-79.216917
4	-79.239476

2.2 Show neighborhoods on a map

2.2.1 Find geo location of the city to make sure that the map is centered correctly

```
In [4]: from geopy.geocoders import Nominatim

        address = 'Canada, Toronto'

        geolocator = Nominatim(user_agent="toronto_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 43.653963, -79.387207.

2.2.2 Show the neighborhoods on a city map

```
In [35]: from IPython.core.display import HTML
        map_toronto = folium.Map(location=[latitude +0.05, longitude], zoom_start=11)

        # add markers to map
        for lat, lng, borough, neighborhood in zip(df_postal_loc['Latitude'], df_postal_loc['Longitude'], df_postal_loc['Borough'], df_postal_loc['Neighborhood']):
            label = '{} {}'.format(neighborhood, borough)
            label = folium.Popup(label, parse_html=True)
            folium.CircleMarker(
                [lat, lng],
                radius=7,
                popup=label,
                color='red',
                fill=True,
                fill_color='orange',
                fill_opacity=0.7,
                parse_html=False).add_to(map_toronto)

        map_toronto
```

Out[35]: <folium.folium.Map at 0x20c46fe8748>

2.3 Create venue dataset with Foursquare API's

2.3.1 Read Foursquare credentials from CSV file

```
In [6]: df_credentials = pd.read_csv('credentials.txt', header=None)
        CLIENT_ID = df_credentials[1][0]
        CLIENT_SECRET = df_credentials[1][1]
```

2.3.2 Get nearby venues for all rows in the neighborhood dataset

```
In [7]: VERSION = '20180605' # Foursquare API version
def getNearbyVenues(names, latitudes, longitudes, radius=500):

    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
        #print(requests.get(url).json())
        results = requests.get(url).json()["response"]["groups"][0]["items"]
        #print(results)
        #print(' ')
        # 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 [8]: LIMIT = 100
toronto_venues = getNearbyVenues(names=df_postal_loc['Neighbourhood'],
                                  latitudes=df_postal_loc['Latitude'],
```

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

```
In [9]: print('There are {} venues in this dataset.'.format(toronto_venues.shape[0]))
        toronto_venues.head()
```

There are 2244 venues in this dataset.

```
Out[9]:
```

	Neighborhood	Neighborhood Latitude \
0	Rouge, Malvern	43.806686
1	Highland Creek, Rouge Hill, Port Union	43.784535
2	Guildwood, Morningside, West Hill	43.763573
3	Guildwood, Morningside, West Hill	43.763573
4	Guildwood, Morningside, West Hill	43.763573

	Neighborhood Longitude	Venue	Venue Latitude \
0	-79.194353	Wendy's	43.807448
1	-79.160497	Royal Canadian Legion	43.782533
2	-79.188711	Swiss Chalet Rotisserie & Grill	43.767697
3	-79.188711	G & G Electronics	43.765309
4	-79.188711	Big Bite Burrito	43.766299

	Venue Longitude	Venue Category
0	-79.199056	Fast Food Restaurant
1	-79.163085	Bar
2	-79.189914	Pizza Place
3	-79.191537	Electronics Store
4	-79.190720	Mexican Restaurant

2.3.3 Get an idea how much venues we have per neighborhood

```
In [10]: res = toronto_venues.groupby('Neighborhood')['Venue'].count()
        #print(type(res))
        #print(res.index)
        for n,vc in zip(res.index,res):
            print(vc,'\t',n)
```

```
100      Adelaide, King, Richmond
5        Agincourt
2      Agincourt North, L'Amoreaux East, Milliken, Steeles East
9      Albion Gardens, Beaumont Heights, Humbergate, Jamestown, Mount Olive, Silverstone, S
9      Alderwood, Long Branch
22      Bathurst Manor, Downsview North, Wilson Heights
4      Bayview Village
22      Bedford Park, Lawrence Manor East
57      Berczy Park
5      Birch Cliff, Cliffside West
9      Bloordale Gardens, Eringate, Markland Wood, Old Burnhamthorpe
```

21 Brockton, Exhibition Place, Parkdale Village
 19 Business Reply Mail Processing Centre 969 Eastern
 2 CFB Toronto, Downsview East
 14 CN Tower, Bathurst Quay, Island airport, Harbourfront West, King and Spadina, Rail
 45 Cabbagetown, St. James Town
 5 Caledonia-Fairbanks
 11 Canada Post Gateway Processing Centre
 7 Cedarbrae
 84 Central Bay Street
 100 Chinatown, Grange Park, Kensington Market
 15 Christie
 84 Church and Wellesley
 9 Clairlea, Golden Mile, Oakridge
 10 Clarks Corners, Sullivan, Tam O'Shanter
 3 Cliffcrest, Cliffside, Scarborough Village West
 1 Cloverdale, Islington, Martin Grove, Princess Gardens, West Deane Park
 100 Commerce Court, Victoria Hotel
 35 Davisville
 9 Davisville North
 15 Deer Park, Forest Hill SE, Rathnelly, South Hill, Summerhill West
 6 Del Ray, Keelesdale, Mount Dennis, Silverthorn
 100 Design Exchange, Toronto Dominion Centre
 5 Don Mills North
 6 Dorset Park, Scarborough Town Centre, Wexford Heights
 16 Dovercourt Village, Dufferin
 3 Downsview Central
 4 Downsview Northwest
 5 Downsview West
 4 Downsview, North Park, Upwood Park
 6 East Birchmount Park, Ionview, Kennedy Park
 5 East Toronto
 1 Emery, Humberlea
 59 Fairview, Henry Farm, Oriole
 100 First Canadian Place, Underground city
 20 Flemingdon Park, Don Mills South
 4 Forest Hill North, Forest Hill West
 4 Glencairn
 7 Guildwood, Morningside, West Hill
 35 Harbord, University of Toronto
 100 Harbourfront East, Toronto Islands, Union Station
 49 Harbourfront, Regent Park
 23 High Park, The Junction South
 1 Highland Creek, Rouge Hill, Port Union
 6 Hillcrest Village
 14 Humber Bay Shores, Mimico South, New Toronto
 4 Humber Bay, King's Mill Park, Kingsway Park South East, Mimico NE, Old Mill South, T
 2 Humber Summit
 4 Humewood-Cedarvale

2	Kingsview Village, Martin Grove Gardens, Richview Gardens, St. Phillips
15	Kingsway Park South West, Mimico NW, The Queensway West, Royal York South West, So
12	L'Amoreaux West
14	Lawrence Heights, Lawrence Manor
4	Lawrence Park
34	Leaside
65	Little Portugal, Trinity
6	Maryvale, Wexford
4	Moore Park, Summerhill East
1	Newtonbrook, Willowdale
20	North Toronto West
3	Northwest
6	Northwood Park, York University
15	Parkdale, Roncesvalles
3	Parkwoods
40	Queen's Park
5	Rosedale
1	Roselawn
1	Rouge, Malvern
36	Runnymede, Swansea
100	Ryerson, Garden District
3	Scarborough Village
1	Silver Hills, York Mills
100	St. James Town
96	Stn A PO Boxes 25 The Esplanade
39	Studio District
24	The Annex, North Midtown, Yorkville
5	The Beaches
18	The Beaches West, India Bazaar
42	The Danforth West, Riverdale
5	The Junction North, Runnymede
2	The Kingsway, Montgomery Road, Old Mill North
16	Thorncliffe Park
4	Victoria Village
8	Westmount
2	Weston
35	Willowdale South
5	Willowdale West
3	Woburn
11	Woodbine Gardens, Parkview Hill
9	Woodbine Heights
3	York Mills West

2.3.4 Get an overview of the venue categories

```
In [11]: print('There are {} unique categories.'.format(len(toronto_venues['Venue Category']).unique()))
```

There are 280 unique categories.

```
Out[11]: array(['Fast Food Restaurant', 'Bar', 'Pizza Place', 'Electronics Store',
'Mexican Restaurant', 'Rental Car Location', 'Medical Center',
'Intersection', 'Breakfast Spot', 'Coffee Shop',
'Korean Restaurant', 'Hakka Restaurant', 'Caribbean Restaurant',
'Thai Restaurant', 'Athletics & Sports', 'Bank', 'Bakery',
'Fried Chicken Joint', 'Playground', 'Spa', 'Convenience Store',
'Department Store', 'Discount Store', 'Chinese Restaurant',
'Bus Station', 'Bus Line', 'Metro Station', 'Soccer Field',
'Motel', 'Movie Theater', 'American Restaurant', 'Café',
'General Entertainment', 'Farm', 'Skating Rink', 'College Stadium',
'Indian Restaurant', 'Vietnamese Restaurant', 'Pet Store',
'Latin American Restaurant', 'Sandwich Place',
'Middle Eastern Restaurant', 'Shopping Mall', 'Auto Garage',
'Lounge', 'Italian Restaurant', 'Noodle House', 'Pharmacy', 'Park',
'Grocery Store', 'Thrift / Vintage Store', 'Nail Salon',
'Golf Course', 'Pool', 'Mediterranean Restaurant', 'Dog Run',
'Toy / Game Store', 'Burger Joint', 'Tea Room', 'Candy Store',
'Juice Bar', 'Salon / Barbershop', 'Smoothie Shop',
'Clothing Store', 'Theater', 'Japanese Restaurant', 'Food Court',
'Restaurant', 'Cosmetics Shop', 'Liquor Store', 'Video Game Store',
'Asian Restaurant', 'Wings Joint', 'Sporting Goods Shop',
'Burrito Place', 'Shoe Store', 'Deli / Bodega', 'Gift Shop',
'Boutique', 'Supplement Shop', "Women's Store", 'Luggage Store',
'Baseball Field', 'Cafeteria', 'Ramen Restaurant', 'Steakhouse',
'Indonesian Restaurant', 'Arts & Crafts Store', 'Plaza',
'Ice Cream Shop', 'Sushi Restaurant', 'Bubble Tea Shop', 'Hotel',
'Butcher', 'Food & Drink Shop', 'Gym / Fitness Center',
'Basketball Court', 'Gym', 'Bike Shop', 'Beer Store',
'Dim Sum Restaurant', 'Supermarket', 'Bridal Shop', 'Diner',
'Frozen Yogurt Shop', 'Video Store', 'Massage Studio',
'Falafel Restaurant', 'Airport', 'Home Service', 'Food Truck',
'Hockey Arena', 'Portuguese Restaurant',
'Financial or Legal Service', 'Gastropub', 'Curling Ice',
'Bus Stop', 'Trail', 'Health Food Store', 'Pub', 'Neighborhood',
'Sports Bar', 'Fish & Chips Shop', 'Dessert Shop', 'Brewery',
'Furniture / Home Store', 'Bagel Shop', 'Warehouse Store',
'Yoga Studio', 'Housing Development', 'Greek Restaurant',
'Fruit & Vegetable Store', 'Bookstore', 'Fish Market',
'Cheese Shop', 'Comfort Food Restaurant', 'Seafood Restaurant',
'Stationery Store', 'Coworking Space', 'Music Store', 'Gay Bar',
'Swim School', 'Dance Studio', 'Gourmet Shop', 'Farmers Market',
'Costume Shop', 'Light Rail Station', 'Building', 'Jewelry Store',
'Taiwanese Restaurant', 'Market', 'Dive Bar', 'Outdoor Sculpture',
'Snack Place', 'Theme Restaurant', 'Creperie', 'Hobby Shop',
'Ethiopian Restaurant', "Men's Store", 'Smoke Shop', 'Sake Bar',
```

```
'Afghan Restaurant', 'Persian Restaurant',
'Health & Beauty Service', 'Strip Club', 'Sculpture Garden',
'Polish Restaurant', 'Historic Site', 'Chocolate Shop',
'Performing Arts Venue', 'French Restaurant', 'Event Space',
'Art Gallery', 'Antique Shop', 'Comic Shop', 'Taco Place',
'Music Venue', 'Beer Bar', 'Vegetarian / Vegan Restaurant',
'College Rec Center', 'Tanning Salon',
'Modern European Restaurant', 'Miscellaneous Shop', 'Hookah Bar',
'Wine Bar', 'Lake', 'Lingerie Store', 'Other Great Outdoors',
'Poutine Place', 'Office', 'BBQ Joint', 'Church', 'Speakeasy',
'Poke Place', 'New American Restaurant', 'Hostel', 'Cocktail Bar',
'Jazz Club', 'Camera Store', 'Tailor Shop', 'Fountain',
'German Restaurant', 'Concert Hall', 'Museum', 'Bistro',
'Basketball Stadium', 'Beach', 'Eastern European Restaurant',
'Irish Pub', 'Nightclub', 'Art Museum', 'Salad Place',
'Donut Shop', 'Opera House', 'Monument / Landmark',
'General Travel', 'Colombian Restaurant', 'Record Shop',
'Brazilian Restaurant', 'Gluten-free Restaurant', 'Roof Deck',
'Aquarium', 'Train Station', 'History Museum', 'Scenic Lookout',
'Baseball Stadium', 'Hotel Bar', 'Soup Place', 'Garden',
'Jewish Restaurant', 'College Gym', 'College Arts Building',
'Flower Shop', 'Organic Grocery', 'Dumpling Restaurant',
'Gaming Cafe', 'Martial Arts Dojo', 'Doner Restaurant',
'Filipino Restaurant', 'Hotpot Restaurant', 'Hospital',
'Airport Lounge', 'Harbor / Marina', 'Airport Food Court',
'Airport Terminal', 'Airport Gate', 'Plane', 'Airport Service',
'Boat or Ferry', 'Molecular Gastronomy Restaurant', 'Optical Shop',
'Accessories Store', 'Fraternity House', 'Field', 'Baby Store',
'Cuban Restaurant', 'Mac & Cheese Joint', 'Malay Restaurant',
'Cupcake Shop', 'Tapas Restaurant',
'Southern / Soul Food Restaurant', 'Climbing Gym', 'Stadium',
'Construction & Landscaping', 'Flea Market',
'Cajun / Creole Restaurant', 'Food', 'Indie Movie Theater',
'College Auditorium', 'College Cafeteria', 'Skate Park',
'Garden Center', 'Auto Workshop', 'Recording Studio', 'River',
'Locksmith', 'Hardware Store', 'Shopping Plaza',
'Empanada Restaurant', 'Drugstore'], dtype=object)
```

2.3.5 Create a one hot encoding matrix, with venue categories as columns and venues from the venues in the venue dataframe as rows

In [12]: *# one hot encoding*

```
toronto_onehot = pd.get_dummies(toronto_venues[['Venue Category']], prefix="", prefix_sep=""
print(toronto_onehot.shape)
toronto_onehot.head()
```

(2244, 280)


```

Out[12]:
Accessories Store  Afghan Restaurant  Airport  Airport Food Court  \
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

Airport Gate  Airport Lounge  Airport Service  Airport Terminal  \
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

American Restaurant  Antique Shop  ...  Train Station  \
0                0            0  ...            0
1                0            0  ...            0
2                0            0  ...            0
3                0            0  ...            0
4                0            0  ...            0

Vegetarian / Vegan Restaurant  Video Game Store  Video Store  \
0                0                0            0
1                0                0            0
2                0                0            0
3                0                0            0
4                0                0            0

Vietnamese Restaurant  Warehouse Store  Wine Bar  Wings Joint  \
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

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

[5 rows x 280 columns]

```

2.3.6 Remove 'Neighborhood' column in the onehot dataset

For some reason there seems to be a 'Neighborhood' column in this dataset, where we would only expect venue categories

```

In [13]: #toronto_onehot['Neighborhood']

In [14]: toronto_onehot.drop(columns=['Neighborhood'],inplace=True)

In [15]: #toronto_onehot['Neighborhood']

In [16]: ### Add Neighborhood column to the onehot dataset as a first column

In [17]: # add neighborhood column back to dataframe
toronto_onehot['Neighborhood'] = toronto_venues['Neighborhood']
fixed_columns = [toronto_onehot.columns[-1]] + list(toronto_onehot.columns[:-1])
toronto_onehot = toronto_onehot[fixed_columns]
toronto_onehot.head()

```

```

Out[17]:
      Neighborhood  Accessories Store \
0      Rouge, Malvern                0
1 Highland Creek, Rouge Hill, Port Union    0
2      Guildwood, Morningside, West Hill    0
3      Guildwood, Morningside, West Hill    0
4      Guildwood, Morningside, West Hill    0

      Afghan Restaurant  Airport  Airport Food Court  Airport Gate \
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

      Airport Lounge  Airport Service  Airport Terminal  American 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

      ...      Train Station  Vegetarian / Vegan Restaurant \
0      ...                0                0
1      ...                0                0
2      ...                0                0
3      ...                0                0
4      ...                0                0

      Video Game Store  Video Store  Vietnamese Restaurant  Warehouse 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

```

	Wine Bar	Wings Joint	Women's Store	Yoga Studio
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 rows x 280 columns]

In [18]: *### For each Neighborhood, calculate the average % for each venue category (% of total)*

```
In [19]: toronto_grouped = toronto_onehot.groupby('Neighborhood').mean()
# Neighborhood from index => column
toronto_grouped.reset_index(inplace=True)
toronto_grouped.head()
```

```
Out[19]:
```

	Neighborhood	Accessories Store \
0	Adelaide, King, Richmond	0.0
1	Agincourt	0.0
2	Agincourt North, L'Amoreaux East, Milliken, St...	0.0
3	Albion Gardens, Beaumont Heights, Humbergate, ...	0.0
4	Alderwood, Long Branch	0.0

	Afghan Restaurant	Airport	Airport Food Court	Airport Gate \
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

	Airport Lounge	Airport Service	Airport Terminal	American Restaurant \
0	0.0	0.0	0.0	0.03
1	0.0	0.0	0.0	0.00
2	0.0	0.0	0.0	0.00
3	0.0	0.0	0.0	0.00
4	0.0	0.0	0.0	0.00

	...	Train Station	Vegetarian / Vegan Restaurant \
0	...	0.0	0.01
1	...	0.0	0.00
2	...	0.0	0.00
3	...	0.0	0.00
4	...	0.0	0.00

	Video Game Store	Video Store	Vietnamese Restaurant	Warehouse Store \
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

	Wine Bar	Wings Joint	Women's Store	Yoga Studio
0	0.01	0.0	0.01	0.0
1	0.00	0.0	0.00	0.0
2	0.00	0.0	0.00	0.0
3	0.00	0.0	0.00	0.0
4	0.00	0.0	0.00	0.0

[5 rows x 280 columns]

2.3.7 For each neighborhood, show the top 5 venue categories

```
In [20]: num_top_venues = 5
```

```
for hood in toronto_grouped['Neighborhood']:
    print("----"+hood+"----")
    temp = toronto_grouped[toronto_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	0.08
1	Café	0.05
2	Bar	0.04
3	Steakhouse	0.04
4	Thai Restaurant	0.04

----Agincourt----

	venue	freq
0	Breakfast Spot	0.2
1	Chinese Restaurant	0.2
2	Lounge	0.2
3	Skating Rink	0.2
4	Sandwich Place	0.2

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

	venue	freq
0	Playground	0.5
1	Park	0.5

2	Accessories Store	0.0
3	Metro Station	0.0
4	Monument / Landmark	0.0

----Albion Gardens, Beaumont Heights, Humbergate, Jamestown, Mount Olive, Silverstone, South S

	venue	freq
0	Grocery Store	0.22
1	Pizza Place	0.11
2	Coffee Shop	0.11
3	Sandwich Place	0.11
4	Fast Food Restaurant	0.11

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

	venue	freq
0	Pizza Place	0.22
1	Skating Rink	0.11
2	Sandwich Place	0.11
3	Pub	0.11
4	Coffee Shop	0.11

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

	venue	freq
0	Coffee Shop	0.09
1	Gift Shop	0.05
2	Shopping Mall	0.05
3	Fast Food Restaurant	0.05
4	Sandwich Place	0.05

----Bayview Village----

	venue	freq
0	Japanese Restaurant	0.25
1	Chinese Restaurant	0.25
2	Bank	0.25
3	Café	0.25
4	Modern European Restaurant	0.00

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

	venue	freq
0	Coffee Shop	0.09
1	Italian Restaurant	0.09
2	Japanese Restaurant	0.05
3	Greek Restaurant	0.05
4	Juice Bar	0.05

----Berczy Park----

	venue	freq
0	Coffee Shop	0.09
1	Cocktail Bar	0.05
2	Beer Bar	0.04
3	Seafood Restaurant	0.04
4	Farmers Market	0.04

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

	venue	freq
0	Skating Rink	0.2
1	College Stadium	0.2
2	Farm	0.2
3	General Entertainment	0.2
4	Café	0.2

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

	venue	freq
0	Pizza Place	0.11
1	Shopping Plaza	0.11
2	Café	0.11
3	Coffee Shop	0.11
4	Liquor Store	0.11

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

	venue	freq
0	Coffee Shop	0.10
1	Café	0.10
2	Breakfast Spot	0.10
3	Intersection	0.05
4	Caribbean Restaurant	0.05

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

	venue	freq
0	Light Rail Station	0.11
1	Yoga Studio	0.05
2	Auto Workshop	0.05
3	Pizza Place	0.05
4	Comic Shop	0.05

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

	venue	freq
0	Airport	0.5
1	Park	0.5
2	Accessories Store	0.0
3	Metro Station	0.0
4	Monument / Landmark	0.0

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

	venue	freq
0	Airport Terminal	0.14
1	Airport Lounge	0.14
2	Airport Service	0.14
3	Plane	0.07
4	Coffee Shop	0.07

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

	venue	freq
0	Coffee Shop	0.07
1	Park	0.07
2	Convenience Store	0.04
3	Pub	0.04
4	Bakery	0.04

----Caledonia-Fairbanks----

	venue	freq
0	Park	0.4
1	Women's Store	0.2
2	Market	0.2
3	Fast Food Restaurant	0.2
4	Accessories Store	0.0

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

	venue	freq
0	Hotel	0.18
1	Coffee Shop	0.18
2	Mediterranean Restaurant	0.09
3	Burrito Place	0.09
4	Gym / Fitness Center	0.09

----Cedarbrae----

	venue	freq
0	Bank	0.14
1	Thai Restaurant	0.14

2	Hakka Restaurant	0.14
3	Caribbean Restaurant	0.14
4	Athletics & Sports	0.14

----Central Bay Street----

	venue	freq
0	Coffee Shop	0.14
1	Ice Cream Shop	0.05
2	Italian Restaurant	0.05
3	Sandwich Place	0.04
4	Burger Joint	0.04

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

	venue	freq
0	Café	0.07
1	Vegetarian / Vegan Restaurant	0.06
2	Chinese Restaurant	0.05
3	Mexican Restaurant	0.04
4	Dumpling Restaurant	0.04

----Christie----

	venue	freq
0	Grocery Store	0.20
1	Café	0.20
2	Park	0.13
3	Baby Store	0.07
4	Coffee Shop	0.07

----Church and Wellesley----

	venue	freq
0	Coffee Shop	0.07
1	Japanese Restaurant	0.06
2	Sushi Restaurant	0.05
3	Gay Bar	0.04
4	Restaurant	0.04

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

	venue	freq
0	Bus Line	0.22
1	Bakery	0.22
2	Soccer Field	0.11
3	Intersection	0.11
4	Bus Station	0.11

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

	venue	freq
0	Pizza Place	0.2
1	Pharmacy	0.1
2	Noodle House	0.1
3	Fast Food Restaurant	0.1
4	Fried Chicken Joint	0.1

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

	venue	freq
0	Movie Theater	0.33
1	Motel	0.33
2	American Restaurant	0.33
3	Mexican Restaurant	0.00
4	Monument / Landmark	0.00

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

	venue	freq
0	Bank	1.0
1	Accessories Store	0.0
2	Mexican Restaurant	0.0
3	Monument / Landmark	0.0
4	Molecular Gastronomy Restaurant	0.0

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

	venue	freq
0	Coffee Shop	0.09
1	Hotel	0.06
2	Café	0.06
3	Restaurant	0.04
4	American Restaurant	0.04

----Davisville----

	venue	freq
0	Pizza Place	0.09
1	Dessert Shop	0.09
2	Sandwich Place	0.09
3	Restaurant	0.06
4	Café	0.06

----Davisville North----

	venue	freq
0	Dog Run	0.11
1	Gym	0.11
2	Food & Drink Shop	0.11
3	Sandwich Place	0.11
4	Clothing Store	0.11

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

	venue	freq
0	Coffee Shop	0.13
1	Pub	0.13
2	Restaurant	0.07
3	Liquor Store	0.07
4	Sushi Restaurant	0.07

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

	venue	freq
0	Fast Food Restaurant	0.33
1	Restaurant	0.17
2	Sandwich Place	0.17
3	Fried Chicken Joint	0.17
4	Coffee Shop	0.17

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

	venue	freq
0	Coffee Shop	0.12
1	Café	0.08
2	Hotel	0.06
3	Restaurant	0.05
4	Bakery	0.03

----Don Mills North----

	venue	freq
0	Japanese Restaurant	0.2
1	Caribbean Restaurant	0.2
2	Gym / Fitness Center	0.2
3	Café	0.2
4	Basketball Court	0.2

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

	venue	freq
0	Indian Restaurant	0.33
1	Pet Store	0.17

2	Vietnamese Restaurant	0.17
3	Latin American Restaurant	0.17
4	Chinese Restaurant	0.17

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

	venue	freq
0	Pharmacy	0.12
1	Bakery	0.12
2	Supermarket	0.12
3	Middle Eastern Restaurant	0.06
4	Bank	0.06

----Downsview Central----

	venue	freq
0	Food Truck	0.33
1	Home Service	0.33
2	Baseball Field	0.33
3	Mexican Restaurant	0.00
4	Monument / Landmark	0.00

----Downsview Northwest----

	venue	freq
0	Liquor Store	0.25
1	Gym / Fitness Center	0.25
2	Grocery Store	0.25
3	Athletics & Sports	0.25
4	Accessories Store	0.00

----Downsview West----

	venue	freq
0	Grocery Store	0.4
1	Bank	0.2
2	Park	0.2
3	Shopping Mall	0.2
4	Accessories Store	0.0

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

	venue	freq
0	Construction & Landscaping	0.25
1	Bakery	0.25
2	Park	0.25
3	Basketball Court	0.25
4	Accessories Store	0.00

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

	venue	freq
0	Discount Store	0.33
1	Department Store	0.17
2	Coffee Shop	0.17
3	Chinese Restaurant	0.17
4	Bus Station	0.17

----East Toronto----

	venue	freq
0	Park	0.4
1	Pizza Place	0.2
2	Coffee Shop	0.2
3	Convenience Store	0.2
4	Martial Arts Dojo	0.0

----Emery, Humberlea----

	venue	freq
0	Baseball Field	1.0
1	Accessories Store	0.0
2	Mexican Restaurant	0.0
3	Monument / Landmark	0.0
4	Molecular Gastronomy Restaurant	0.0

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

	venue	freq
0	Clothing Store	0.14
1	Coffee Shop	0.08
2	Fast Food Restaurant	0.08
3	Japanese Restaurant	0.03
4	Shoe Store	0.03

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

	venue	freq
0	Coffee Shop	0.09
1	Café	0.07
2	Steakhouse	0.04
3	Hotel	0.04
4	Restaurant	0.04

----Flemingdon Park, Don Mills South----

	venue	freq
0	Coffee Shop	0.10
1	Beer Store	0.10
2	Gym	0.10
3	Asian Restaurant	0.05
4	Chinese Restaurant	0.05

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

	venue	freq
0	Park	0.25
1	Sushi Restaurant	0.25
2	Jewelry Store	0.25
3	Trail	0.25
4	Accessories Store	0.00

----Glencairn----

	venue	freq
0	Italian Restaurant	0.25
1	Pub	0.25
2	Bakery	0.25
3	Japanese Restaurant	0.25
4	Motel	0.00

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

	venue	freq
0	Pizza Place	0.14
1	Rental Car Location	0.14
2	Medical Center	0.14
3	Mexican Restaurant	0.14
4	Electronics Store	0.14

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

	venue	freq
0	Café	0.11
1	Bakery	0.06
2	Japanese Restaurant	0.06
3	Bookstore	0.06
4	Restaurant	0.06

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

	venue	freq
0	Coffee Shop	0.11
1	Hotel	0.05

2	Aquarium	0.05
3	Café	0.04
4	Italian Restaurant	0.04

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

	venue	freq
0	Coffee Shop	0.16
1	Café	0.06
2	Bakery	0.06
3	Park	0.06
4	Pub	0.06

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

	venue	freq
0	Café	0.09
1	Mexican Restaurant	0.09
2	Music Venue	0.04
3	Grocery Store	0.04
4	Cajun / Creole Restaurant	0.04

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

	venue	freq
0	Bar	1.0
1	Accessories Store	0.0
2	Mexican Restaurant	0.0
3	Monument / Landmark	0.0
4	Molecular Gastronomy Restaurant	0.0

----Hillcrest Village----

	venue	freq
0	Fast Food Restaurant	0.17
1	Mediterranean Restaurant	0.17
2	Athletics & Sports	0.17
3	Pool	0.17
4	Dog Run	0.17

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

	venue	freq
0	Pet Store	0.07
1	Coffee Shop	0.07
2	Fast Food Restaurant	0.07
3	Seafood Restaurant	0.07
4	Liquor Store	0.07

----Humber Bay, King's Mill Park, Kingsway Park South East, Mimico NE, Old Mill South, The Queen's Park

	venue	freq
0	Construction & Landscaping	0.25
1	Locksmith	0.25
2	Baseball Field	0.25
3	Pool	0.25
4	Middle Eastern Restaurant	0.00

----Humber Summit----

	venue	freq
0	Pizza Place	0.5
1	Empanada Restaurant	0.5
2	Monument / Landmark	0.0
3	Molecular Gastronomy Restaurant	0.0
4	Modern European Restaurant	0.0

----Humewood-Cedarvale----

	venue	freq
0	Dog Run	0.25
1	Trail	0.25
2	Field	0.25
3	Hockey Arena	0.25
4	Motel	0.00

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

	venue	freq
0	Bus Line	0.5
1	Park	0.5
2	Accessories Store	0.0
3	Monument / Landmark	0.0
4	Molecular Gastronomy Restaurant	0.0

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

	venue	freq
0	Hardware Store	0.07
1	Grocery Store	0.07
2	Flower Shop	0.07
3	Fast Food Restaurant	0.07
4	Sandwich Place	0.07

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

	venue	freq
0	Chinese Restaurant	0.17
1	Fast Food Restaurant	0.17
2	Breakfast Spot	0.08
3	Thrift / Vintage Store	0.08
4	Pizza Place	0.08

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

	venue	freq
0	Clothing Store	0.21
1	Accessories Store	0.07
2	Shoe Store	0.07
3	Furniture / Home Store	0.07
4	Coffee Shop	0.07

----Lawrence Park----

	venue	freq
0	Bus Line	0.25
1	Park	0.25
2	Swim School	0.25
3	Dim Sum Restaurant	0.25
4	Mexican Restaurant	0.00

----Leaside----

	venue	freq
0	Coffee Shop	0.12
1	Sporting Goods Shop	0.09
2	Sushi Restaurant	0.06
3	Furniture / Home Store	0.06
4	Burger Joint	0.06

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

	venue	freq
0	Bar	0.12
1	Coffee Shop	0.06
2	Asian Restaurant	0.05
3	New American Restaurant	0.03
4	French Restaurant	0.03

----Maryvale, Wexford----

	venue	freq
0	Breakfast Spot	0.17
1	Shopping Mall	0.17

2	Auto Garage	0.17
3	Sandwich Place	0.17
4	Middle Eastern Restaurant	0.17

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

	venue	freq
0	Park	0.25
1	Playground	0.25
2	Gym	0.25
3	Restaurant	0.25
4	Accessories Store	0.00

----Newtonbrook, Willowdale----

	venue	freq
0	Park	1.0
1	Accessories Store	0.0
2	Metro Station	0.0
3	Monument / Landmark	0.0
4	Molecular Gastronomy Restaurant	0.0

----North Toronto West----

	venue	freq
0	Sporting Goods Shop	0.10
1	Clothing Store	0.10
2	Coffee Shop	0.10
3	Yoga Studio	0.05
4	Gift Shop	0.05

----Northwest----

	venue	freq
0	Drugstore	0.33
1	Rental Car Location	0.33
2	Bar	0.33
3	Accessories Store	0.00
4	Movie Theater	0.00

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

	venue	freq
0	Massage Studio	0.17
1	Coffee Shop	0.17
2	Bar	0.17
3	Metro Station	0.17
4	Falafel Restaurant	0.17

```

----Parkdale, Roncesvalles----
      venue  freq
0      Breakfast Spot  0.13
1      Gift Shop  0.13
2  Eastern European Restaurant  0.07
3      Bank  0.07
4      Movie Theater  0.07

```

```

----Parkwoods----
      venue  freq
0  Fast Food Restaurant  0.33
1      Park  0.33
2    Food & Drink Shop  0.33
3    Accessories Store  0.00
4    Mexican Restaurant  0.00

```

```

----Queen's Park----
      venue  freq
0  Coffee Shop  0.22
1      Park  0.05
2      Diner  0.05
3      Gym  0.05
4  Yoga Studio  0.02

```

```

----Rosedale----
      venue  freq
0      Park  0.4
1      Trail  0.2
2    Playground  0.2
3      Building  0.2
4  Accessories Store  0.0

```

```

----Roselawn----
      venue  freq
0      Garden  1.0
1  Accessories Store  0.0
2  Mexican Restaurant  0.0
3      Motel  0.0
4  Monument / Landmark  0.0

```

```

----Rouge, Malvern----

```

	venue	freq
0	Fast Food Restaurant	1.0
1	Accessories Store	0.0
2	Metro Station	0.0
3	Monument / Landmark	0.0
4	Molecular Gastronomy Restaurant	0.0

----Runnymede, Swansea----

	venue	freq
0	Coffee Shop	0.08
1	Café	0.08
2	Pizza Place	0.06
3	Diner	0.06
4	Sushi Restaurant	0.06

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

	venue	freq
0	Coffee Shop	0.10
1	Clothing Store	0.06
2	Cosmetics Shop	0.04
3	Café	0.03
4	Middle Eastern Restaurant	0.03

----Scarborough Village----

	venue	freq
0	Convenience Store	0.33
1	Spa	0.33
2	Playground	0.33
3	Accessories Store	0.00
4	Metro Station	0.00

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

	venue	freq
0	Cafeteria	1.0
1	Accessories Store	0.0
2	Mexican Restaurant	0.0
3	Monument / Landmark	0.0
4	Molecular Gastronomy Restaurant	0.0

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

	venue	freq
0	Hotel	0.05
1	Restaurant	0.05

2	Coffee Shop	0.05
3	Café	0.05
4	Italian Restaurant	0.05

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

	venue	freq
0	Coffee Shop	0.11
1	Café	0.04
2	Restaurant	0.04
3	Italian Restaurant	0.03
4	Cocktail Bar	0.03

----Studio District----

	venue	freq
0	Café	0.10
1	Coffee Shop	0.08
2	American Restaurant	0.05
3	Bakery	0.05
4	Italian Restaurant	0.05

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

	venue	freq
0	Coffee Shop	0.12
1	Café	0.12
2	Sandwich Place	0.12
3	Pizza Place	0.08
4	Indian Restaurant	0.04

----The Beaches----

	venue	freq
0	Health Food Store	0.2
1	Coffee Shop	0.2
2	Pub	0.2
3	Trail	0.2
4	Monument / Landmark	0.0

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

	venue	freq
0	Pizza Place	0.11
1	Pet Store	0.06
2	Pub	0.06
3	Fish & Chips Shop	0.06
4	Brewery	0.06

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

	venue	freq
0	Greek Restaurant	0.21
1	Coffee Shop	0.10
2	Italian Restaurant	0.07
3	Furniture / Home Store	0.05
4	Ice Cream Shop	0.05

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

	venue	freq
0	Convenience Store	0.2
1	Bus Line	0.2
2	Pizza Place	0.2
3	Caribbean Restaurant	0.2
4	Brewery	0.2

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

	venue	freq
0	River	0.5
1	Park	0.5
2	Accessories Store	0.0
3	Mexican Restaurant	0.0
4	Monument / Landmark	0.0

----Thorncliffe Park----

	venue	freq
0	Burger Joint	0.12
1	Indian Restaurant	0.12
2	Grocery Store	0.06
3	Sandwich Place	0.06
4	Supermarket	0.06

----Victoria Village----

	venue	freq
0	Financial or Legal Service	0.25
1	Hockey Arena	0.25
2	Coffee Shop	0.25
3	Portuguese Restaurant	0.25
4	Massage Studio	0.00

----Westmount----

	venue	freq
0	Pizza Place	0.25
1	Coffee Shop	0.12
2	Middle Eastern Restaurant	0.12
3	Sandwich Place	0.12
4	Intersection	0.12

----Weston----

	venue	freq
0	Convenience Store	0.5
1	Park	0.5
2	Accessories Store	0.0
3	Metro Station	0.0
4	Monument / Landmark	0.0

----Willowdale South----

	venue	freq
0	Ramen Restaurant	0.09
1	Coffee Shop	0.09
2	Café	0.06
3	Sushi Restaurant	0.06
4	Pizza Place	0.06

----Willowdale West----

	venue	freq
0	Pizza Place	0.2
1	Coffee Shop	0.2
2	Discount Store	0.2
3	Butcher	0.2
4	Pharmacy	0.2

----Woburn----

	venue	freq
0	Coffee Shop	0.67
1	Korean Restaurant	0.33
2	Museum	0.00
3	Motel	0.00
4	Monument / Landmark	0.00

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

	venue	freq
0	Pizza Place	0.18
1	Fast Food Restaurant	0.18

2	Athletics & Sports	0.09
3	Pharmacy	0.09
4	Bank	0.09

----Woodbine Heights----

	venue	freq
0	Skating Rink	0.22
1	Pharmacy	0.11
2	Bus Stop	0.11
3	Cosmetics Shop	0.11
4	Park	0.11

----York Mills West----

	venue	freq
0	Convenience Store	0.33
1	Bank	0.33
2	Park	0.33
3	Accessories Store	0.00
4	Mexican Restaurant	0.00

2.3.8 Create a dataframe containing the 10 most common venue categories for each neighborhood

```
In [21]: 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 [22]: import numpy as np
    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, ind)

print(neighborhoods_venues_sorted.shape)
neighborhoods_venues_sorted.head()

```

(101, 11)

```

Out [22]:

```

	Neighborhood	1st Most Common Venue	\
0	Adelaide, King, Richmond	Coffee Shop	
1	Agincourt	Lounge	
2	Agincourt North, L'Amoreaux East, Milliken, St...	Park	
3	Albion Gardens, Beaumond Heights, Humbergate, ...	Grocery Store	
4	Alderwood, Long Branch	Pizza Place	

	2nd Most Common Venue	3rd Most Common Venue	4th Most Common Venue	\
0	Café	Steakhouse	Bar	
1	Breakfast Spot	Skating Rink	Chinese Restaurant	
2	Playground	Yoga Studio	Eastern European Restaurant	
3	Fast Food Restaurant	Pizza Place	Sandwich Place	
4	Pool	Skating Rink	Gym	

	5th Most Common Venue	6th Most Common Venue	7th Most Common Venue	\
0	Thai Restaurant	Gym	Breakfast Spot	
1	Sandwich Place	Eastern European Restaurant	Doner Restaurant	
2	Dive Bar	Dog Run	Doner Restaurant	
3	Coffee Shop	Beer Store	Pharmacy	
4	Pharmacy	Coffee Shop	Pub	

	8th Most Common Venue	9th Most Common Venue	10th Most Common Venue
0	Hotel	Asian Restaurant	Restaurant
1	Donut Shop	Drugstore	Dumpling Restaurant
2	Donut Shop	Drugstore	Dumpling Restaurant
3	Fried Chicken Joint	Empanada Restaurant	Electronics Store
4	Sandwich Place	Yoga Studio	Diner

2.3.9 Create neighborhood clusters based on the top 10 venue categories

```
In [23]: from sklearn.cluster import KMeans
```

```
kclusters = 5
```

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


	4th Most Common Venue	5th Most Common Venue	6th Most Common Venue	\
0	Dog Run	Doner Restaurant	Donut Shop	
1	Dog Run	Doner Restaurant	Donut Shop	
2	Intersection	Mexican Restaurant	Medical Center	
3	Electronics Store	Dog Run	Doner Restaurant	
4	Bank	Bakery	Athletics & Sports	

	7th Most Common Venue	8th Most Common Venue	9th Most Common Venue	\
0	Drugstore	Dumpling Restaurant	Electronics Store	
1	Drugstore	Dumpling Restaurant	Electronics Store	
2	Pizza Place	Empanada Restaurant	Ethiopian Restaurant	
3	Donut Shop	Drugstore	Dumpling Restaurant	
4	Caribbean Restaurant	Cuban Restaurant	Costume Shop	

	10th Most Common Venue
0	Field
1	Discount Store
2	Eastern European Restaurant
3	Eastern European Restaurant
4	Farmers Market

2.3.11 Remove rows where the cluster label could not be calculated

```
In [25]: print(df_postal_loc.shape)
print(neighborhoods_venues_sorted.shape)
df1 = toronto_merged[toronto_merged.isna().any(axis=1)]
df1
```

```
(103, 5)
(101, 12)
```

```
Out [25]:
```

	PostalCode	Borough	Neighbourhood	Latitude	Longitude	\
16	M1X	Scarborough	Upper Rouge	43.836125	-79.205636	
93	M9A	Etobicoke	Islington Avenue	43.667856	-79.532242	

	Cluster Labels	1st Most Common Venue	2nd Most Common Venue	\
16	NaN	NaN	NaN	
93	NaN	NaN	NaN	

	3rd Most Common Venue	4th Most Common Venue	5th Most Common Venue	\
16	NaN	NaN	NaN	
93	NaN	NaN	NaN	

	6th Most Common Venue	7th Most Common Venue	8th Most Common Venue	\
16	NaN	NaN	NaN	
93	NaN	NaN	NaN	

	9th Most Common Venue	10th Most Common Venue
16	NaN	NaN
93	NaN	NaN

```
In [26]: toronto_merged.dropna(inplace=True)
         print(toronto_merged.shape)
```

```
(101, 16)
```

```
In [27]: df1 = toronto_merged[toronto_merged.isna().any(axis=1)]
         df1
```

```
Out[27]: Empty DataFrame
```

```
Columns: [PostalCode, Borough, Neighbourhood, Latitude, Longitude, Cluster Labels, 1st
Index: []
```

2.3.12 Generate a map of Toronto with the neighborhoods and how they cluster together (colour indicated)

```
In [31]: import matplotlib.cm as cm
         import matplotlib.colors as colors

         # create map
         map_clusters = folium.Map(location=[latitude + 0.05, 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'],
         label = folium.Popup(str(poi) + ' Cluster ' + str(cluster), parse_html=True)
         cluster = int(cluster)
         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[31]: <folium.folium.Map at 0x20c46e68978>
```

2.4 Conclusion

I currently live in the Humber Summit neighborhood and I want to move to the Guildwood, Morningside, West Hill neighborhood.

The map shows that both neighborhoods are in the same cluster and are thus pretty similar.

Below we also see that both are neighborhoods with plenty of restaurants.

```
In [29]: f = (toronto_merged['PostalCode'] == 'M9L') | (toronto_merged['PostalCode'] == 'M1E')
toronto_merged[f]
```

```
Out [29]:
```

	PostalCode	Borough	Neighbourhood	Latitude	\
2	M1E	Scarborough	Guildwood, Morningside, West Hill	43.763573	
96	M9L	North York	Humber Summit	43.756303	
	Longitude	Cluster Labels	1st Most Common Venue	2nd Most Common Venue	\
2	-79.188711	1.0	Breakfast Spot	Electronics Store	
96	-79.565963	1.0	Pizza Place	Empanada Restaurant	
	3rd Most Common Venue	4th Most Common Venue	5th Most Common Venue	\	
2	Rental Car Location	Intersection	Mexican Restaurant		
96	Yoga Studio	Eastern European Restaurant	Dive Bar		
	6th Most Common Venue	7th Most Common Venue	8th Most Common Venue	\	
2	Medical Center	Pizza Place	Empanada Restaurant		
96	Dog Run	Doner Restaurant	Donut Shop		
	9th Most Common Venue	10th Most Common Venue			
2	Ethiopian Restaurant	Eastern European Restaurant			
96	Drugstore	Dumpling Restaurant			