

The Battle of Neighborhoods

1. Introduction/Business Problem

There are many people who are working in various cities (say New York and Toronto) across the world. Let's say a person got a job offer from a great company with great career prospects in other city or other borough of his/her current city. If person accept the job offer then he/she has to shift to new location. I think a person would love to shift a location which is exactly or almost similar to his/her last location because he/she loves the great amenities and other types of venues that exist in his/her current neighbourhood like school, gym, swimming pool, Amusement park, restaurants, coffee-shops, spencer etc. So I'll find out what are borough-neighbourhoods are very similar to give current location. A person can shift within the city or from one city to another city. If person is shifting within the city then he/she must be changing the borough and my task would be finding neighbourhoods in that borough which are very similar to his/her last location. If a person is shifting from one city to another city then my task would be finding the similar neighbourhoods.

2. Data

2.1 New York Data

I'll be using the borough and neighbourhood data of two cities. My first city would be New York and data of this can be downloaded from this [link](#). This data would be *json* format and it can be very easily transformed into *pandas dataframe*. I would take only those are important for me like borough, neighbourhood, latitude and longitude. For more information of this data you can visit this [site](#).

	Borough	Neighbourhood	Latitude	Longitude
0	Bronx	Wakefield	40.894705	-73.847201
1	Bronx	Co-op City	40.874294	-73.829939
2	Bronx	Eastchester	40.887556	-73.827806
3	Bronx	Fieldston	40.895437	-73.905643
4	Bronx	Riverdale	40.890834	-73.912585

Figure - 1: New York Data

2.2 Toronto Data

Second data of Toronto city can not be directly downloaded. Postal Code, borough and neighbourhood can be scrapped from [wikipedia page](#). There are some rows where "Not assigned" is written. Drop the rows where borough is "Not assigned" (only). If Neighbourhood is 'Not assigned' but Borough is assigned then make corresponding Borough as Neighbourhood. Merge the rows if Postal Code and Borough of two or more rows same and merged neighbourhood will be separated by comma ",". Latitude and longitude information can be downloaded from [here](#). Finally this data and Toronto data can be merged together. Once both data are ready we'll use Foursquare API to get the venues near each neighbourhood. To get the nearby venues data we must have Foursquare account and app which will give the credentials.

	Postal Code	Borough	Neighbourhood	Latitude	Longitude
0	M1B	Scarborough	Rouge, Malvern	43.806686	-79.194353
1	M1C	Scarborough	Highland Creek, Rouge Hill, Port Union	43.784535	-79.160497
2	M1E	Scarborough	Guildwood, Morningside, West Hill	43.763573	-79.188711
3	M1G	Scarborough	Woburn	43.770992	-79.216917
4	M1H	Scarborough	Cedarbrae	43.773136	-79.239476

Figure - 2: New York Data

3. Methodology

After getting above data, we'll find the nearby venues to each borough and neighbourhood pair in both cities. Foursquare API will be used to get the nearby venue. To use the Foursquare API we'll need the developer account and an app which will give the *client id* and *client secrets*. Both city may yield in different number of venue categories, but we'll take only common venue categories. Now we can find a borough and neighbourhood of a city is how much similar to the boroughs of another city. In other words we can find top similar boroughs and neighbourhoods in another city. We'll use cosine similarity to find the similarity between two city

4. Getting Venues Using Foursquare API

Foursquare API will be used to get the nearby venues. We're going to find the venues in 1km range from a borough and neighbourhood pair.

4.1 For New York City

The shape of this data is (306, 475). 302 columns are venue categories column.

	Borough	Neighbourhood	Latitude	Longitude	ATM	Accessories Store	Adult Boutique	Afghan Restaurant	African Restaurant	Airport Lounge	Airport Service	Airport Terminal	American Restaurant	
0	Staten Island	St. George	40.644982	-74.079353	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.014085	
1	Staten Island	New Brighton	40.640615	-74.087017	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.022222	
2	Staten Island	Stapleton	40.626928	-74.077902	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.027027	
3	Staten Island	Rosebank	40.615305	-74.069805	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.000000	
4	Staten Island	West Brighton	40.631879	-74.107182	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.0	0.016129	

Figure - 3: New York Data with Venue Categories

4.2 For Toronto City

The shape of this data is (103, 339). 334 columns are venue categories columns.

	Postal Code	Borough	Neighbourhood	Latitude	Longitude	Accessories Store	Adult Boutique	Afghan Restaurant	Airport	Airport Lounge	American Restaurant	Amphitheater	Animal Shelter
0	M8V	Etobicoke	Humber Bay Shores, Mimico South, New Toronto	43.605647	-79.501321	0.0	0.000000	0.0	0.0	0.0	0.052632	0.0	0.0
1	M8W	Etobicoke	Alderwood, Long Branch	43.602414	-79.543484	0.0	0.000000	0.0	0.0	0.0	0.000000	0.0	0.0
2	M8X	Etobicoke	The Kingsway, Montgomery Road, Old Mill North	43.653654	-79.506944	0.0	0.000000	0.0	0.0	0.0	0.000000	0.0	0.0
3	M8Y	Etobicoke	Humber Bay, King's Mill Park, Kingsway Park So...	43.636258	-79.498509	0.0	0.000000	0.0	0.0	0.0	0.000000	0.0	0.0
4	M8Z	Etobicoke	Kingsway Park South West, Mimico NW, The Queen...	43.628841	-79.520999	0.0	0.016667	0.0	0.0	0.0	0.033333	0.0	0.0

Figure - 4: Toronto Data with Venue Categories

5. Analysis

5.1 Common Venue Categories

There are some common venue categories in both data. Figure-5 is showing that there are total 304 common venue categories in both city. So we'll take only these common categories of data. Other than common categories of the data we'll be dropped.

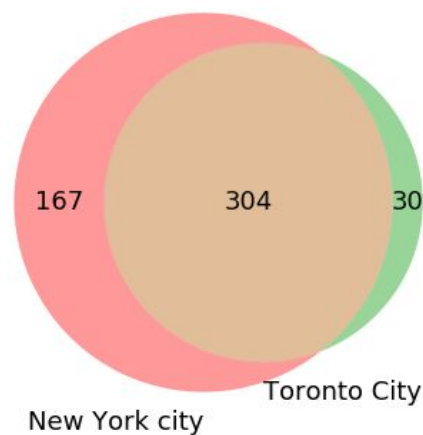


Figure - 5: Venn-Diagram of Venue Categories
Of both cities

5.2 Top 7 similar borough in Toronto City

Let's say current location is Bronx, Riverdale, New York. We'll get the index, latitude and longitude of this location from the New York city data. We can get the all common venue categories data of corresponding index and then we'll multiply with Toronto City data. This will give how much each borough is similar to current location. We'll sort these values and get the top 7 boroughs. Then we'll visualize it using the *folium*.

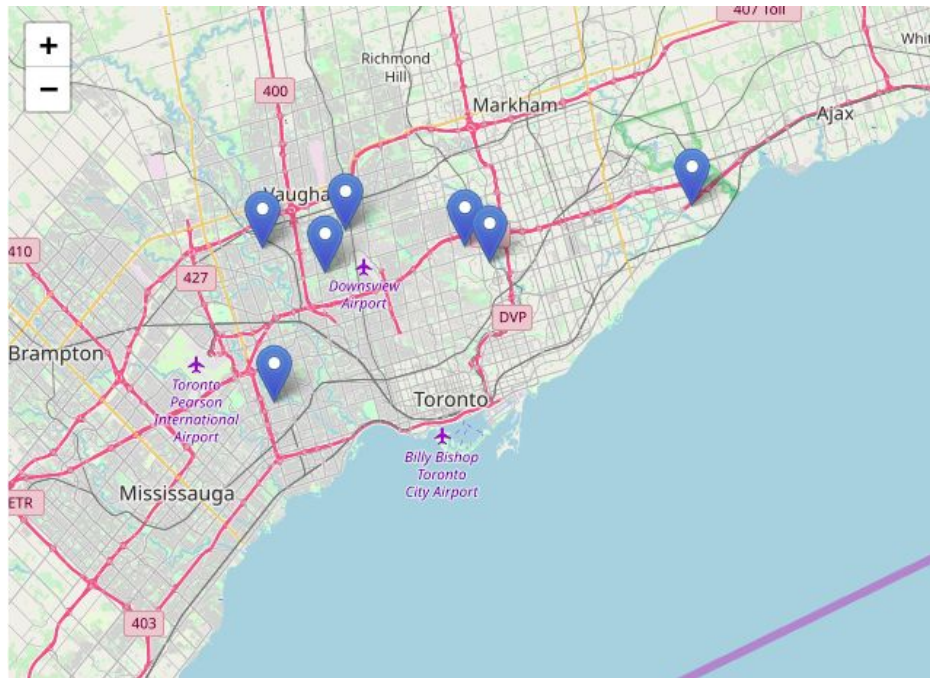


Figure - 6: Top 7 boroughs in Toronto city to the current location

6. Conclusion

Cosine similarity is used here to find how similarity between two boroughs. This model can be implemented within a city also. This will help to find a suitable place for people. This will give most similar borough and neighbourhood.