

Capstone: The Battle of Neighborhood

Recommending Indian restaurant in Toronto, CA
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Introduction

- This project deals with segmenting the neighborhood of Toronto on the basis of the similarities of the locations and clustering the locations on the basis of the competition with other Indian restaurants.

Problem Statement

Given the location data of the neighborhood of the Toronto, Canada, one would like to cluster the similar neighborhoods according to the frequency of Indian restaurants and on the basis of analysis we would like to recommend best neighborhoods to start a new Indian restaurant.

Data

To solve this problem, I will need below data:

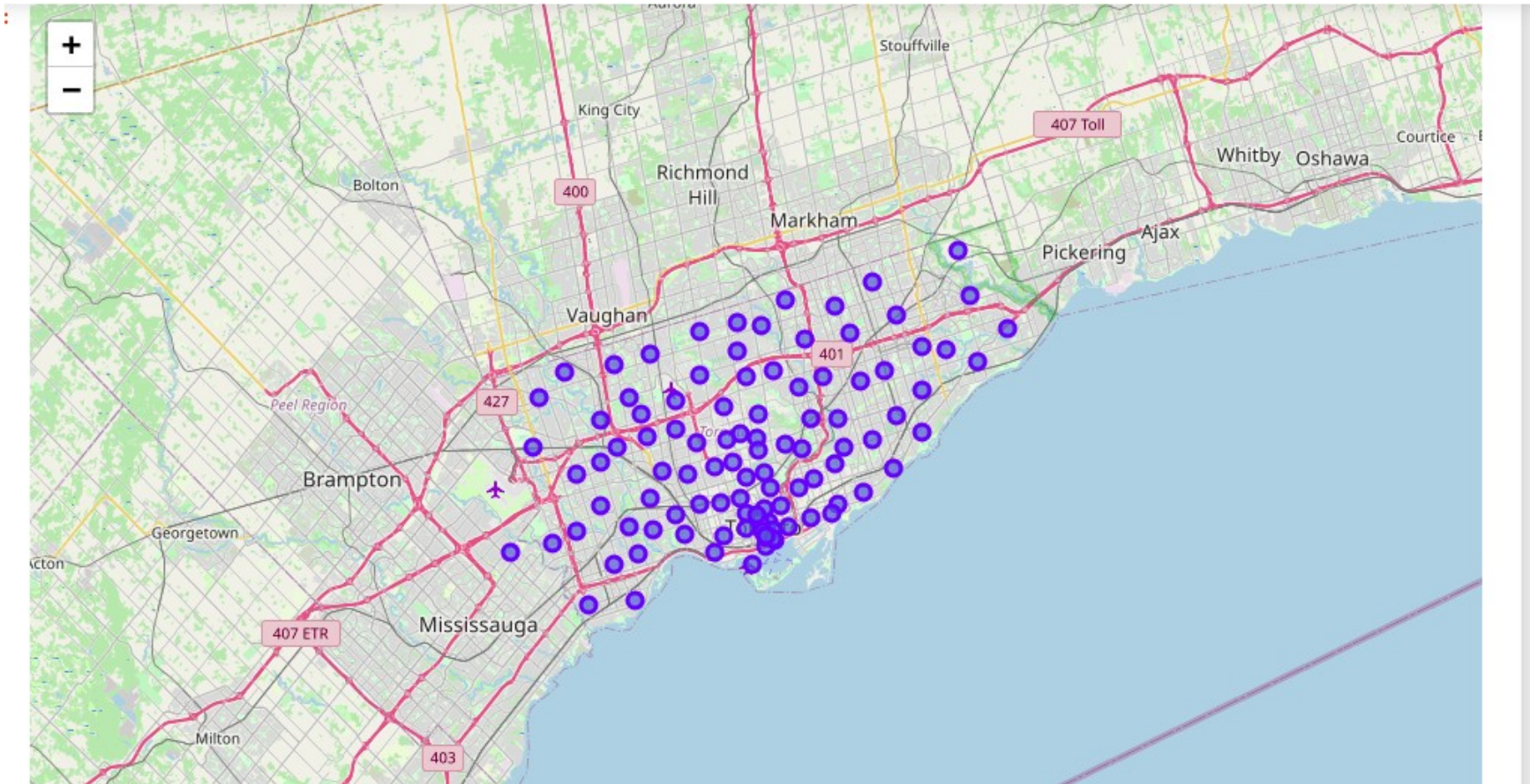
- List of neighborhoods in Toronto, Canada. “
https://en.wikipedia.org/wiki/List_of_postal_codes_of_Canada:_M”
- Latitude and Longitude of these neighborhoods.
- Venue data related to Indian restaurants. This will help us find the neighborhoods that are most suitable to open a Indian restaurant.
- Extracting the data
 - Scrapping of Toronto neighborhoods via Wikipedia
 - Getting Latitude and Longitude data of these neighborhoods via Geocoder package
 - Using Foursquare API to get venue data related to these neighborhoods.

Final Data

	Postal code	Borough	Neighborhood	Latitude	Longitude
0	M1B	Scarborough	Malvern , Rouge	43.8067	-79.1944
1	M1C	Scarborough	Rouge Hill , Port Union , Highland Creek	43.7845	-79.1605
2	M1E	Scarborough	Guildwood , Morningside , West Hill	43.7636	-79.1887
3	M1G	Scarborough	Woburn	43.771	-79.2169
4	M1H	Scarborough	Cedarbrae	43.7731	-79.2395
5	M1J	Scarborough	Scarborough Village	43.7447	-79.2395
6	M1K	Scarborough	Kennedy Park , Ionview , East Birchmount Park	43.7279	-79.262
7	M1L	Scarborough	Golden Mile , Clairlea , Oakridge	43.7111	-79.2846
8	M1M	Scarborough	Cliffside , Cliffcrest , Scarborough Village West	43.7163	-79.2395
9	M1N	Scarborough	Birch Cliff , Cliffside West	43.6927	-79.2648
10	M1P	Scarborough	Dorset Park , Wexford Heights , Scarborough To...	43.7574	-79.2733
11	M1R	Scarborough	Wexford , Maryvale	43.7501	-79.2958
12	M1S	Scarborough	Agincourt	43.7942	-79.262
13	M1T	Scarborough	Clarks Corners , Tam O'Shanter , Sullivan	43.7816	-79.3043
14	M1V	Scarborough	Milliken , Agincourt North , Steeles East , L'...	43.8153	-79.2846
15	M1W	Scarborough	Steeles West , L'Amoreaux West	43.7995	-79.3184
16	M1X	Scarborough	Upper Rouge	43.8361	-79.2056
17	M1Y	Scarborough	Milliken , Agincourt North , Steeles East , L'...	43.8153	-79.2846

Methodology

- Using Folium to map all the neighborhoods in Toronto



Methodology

- Majorly relied on Foursquare API to retrieve all venues of each neighborhoods, then group by each neighborhoods and to count how many venues before filter top 10 most common venue types of each neighborhoods

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(4193, 7)
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Out[22]:

	Neighborhood	Neighborhood Latitude	Neighborhood Longitude	Venue	Venue Latitude	Venue Longitude	Venue Category
0	Malvern , Rouge	43.806686	-79.194353	Wendy's	43.807448	-79.199056	Fast Food Restaurant
1	Rouge Hill , Port Union , Highland Creek	43.784535	-79.160497	Royal Canadian Legion	43.782533	-79.163085	Bar
2	Rouge Hill , Port Union , Highland Creek	43.784535	-79.160497	SEBS Engineering Inc. (Sustainable Energy and ...	43.782371	-79.156820	Construction & Landscaping
3	Guildwood , Morningside , West Hill	43.763573	-79.188711	G & G Electronics	43.765309	-79.191537	Electronics Store
4	Guildwood , Morningside , West Hill	43.763573	-79.188711	Big Bite Burrito	43.766299	-79.190720	Mexican Restaurant

Methodology

- One hot coding for presence and absence of venues in the neighborhood.

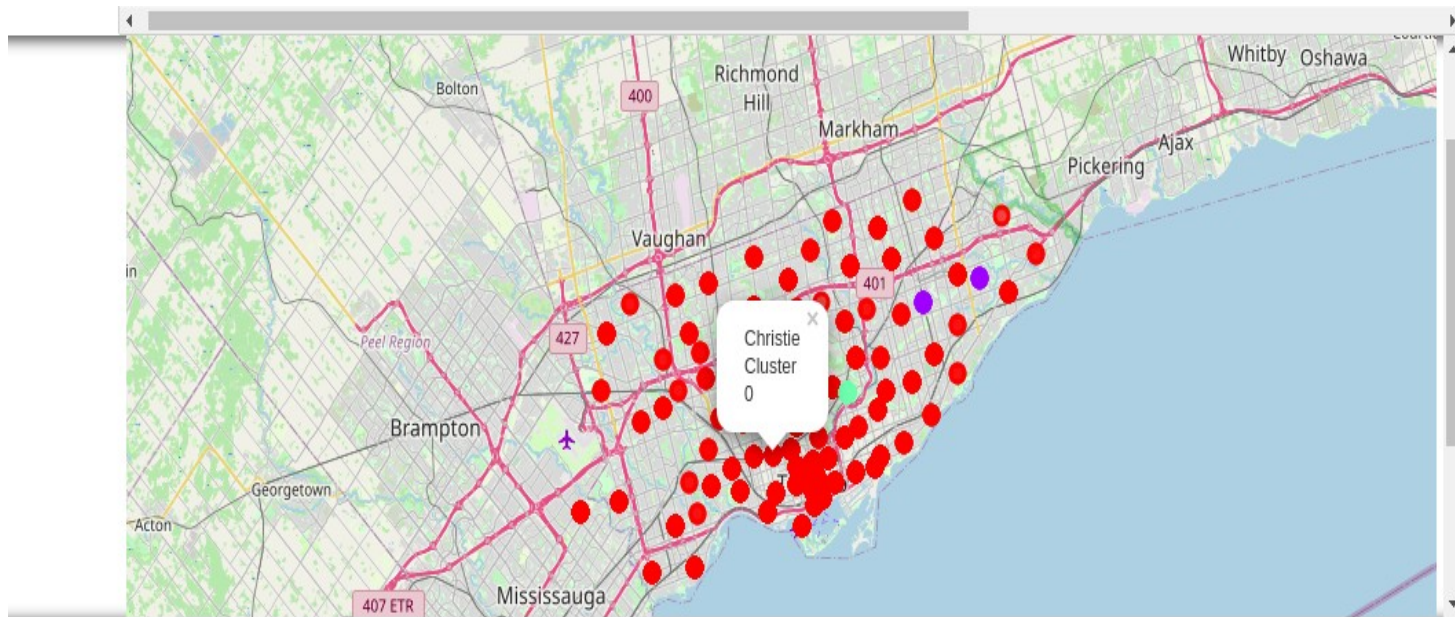
Neighborhood	New American Restaurant	Nightclub	Noodle House	Office	Opera House	Optical Shop	Organic Grocery	Other Great Outdoors	Outdoor Sculpture	Park	Performing Arts Venue	Pet Store	Pharmacy	Pizza Place	Plane	Playground	Plaza
Malvern , Rouge	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
Rouge Hill , Port Union , Highland Creek	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
Rouge Hill , Port Union , Highland Creek	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0
Guildwood , Morningside , West Hill	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0	0

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In [26]: toronto_onehot.shape
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Out[26]: (2143, 269)
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Results

- K mean clustering is used to cluster the neighborhood based on dissimilarities. Three clusters were made



	Clust er 1	Clust er 2	Clust er 3
Mean	0.003	0.3	0.11

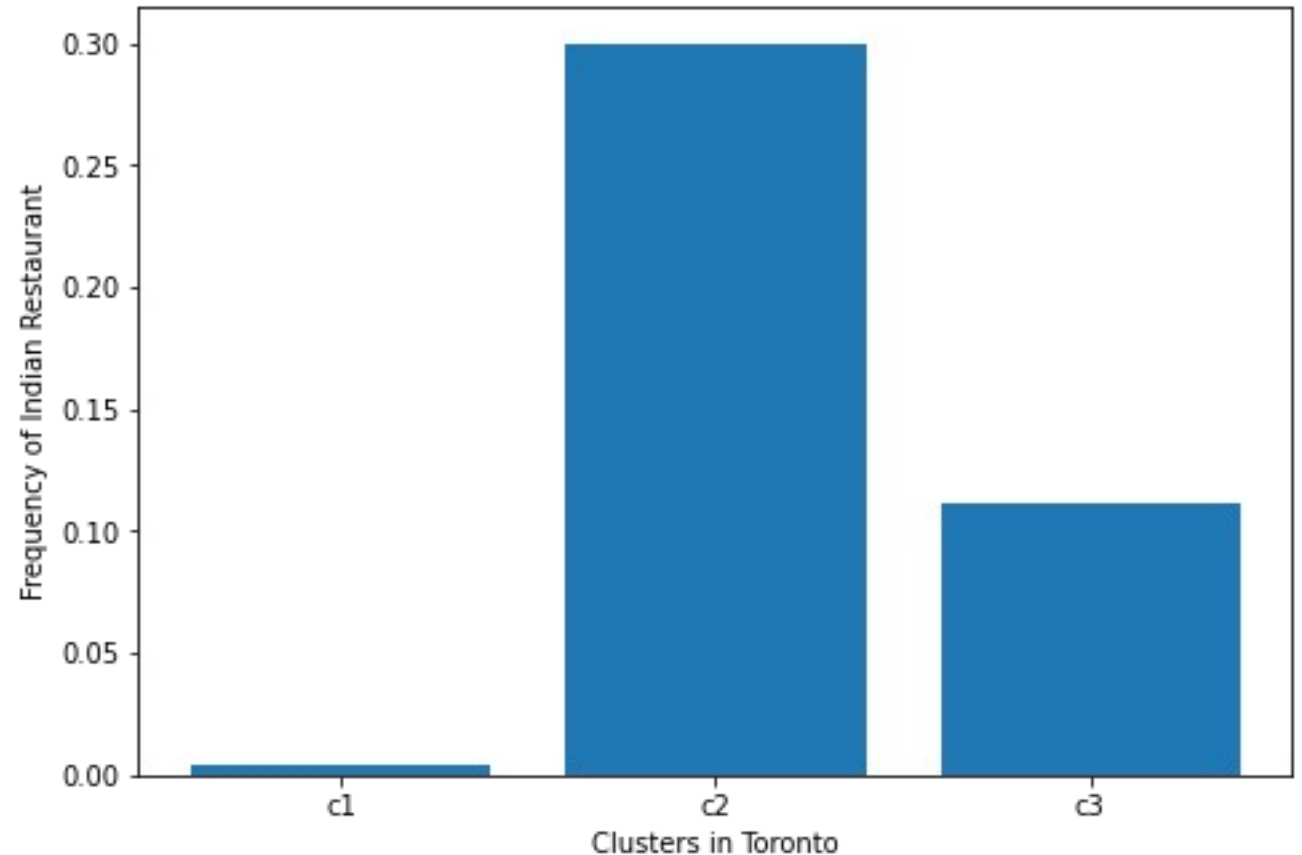
In []:

Results

Cluster 1 : Has minimum number of Indian restaurants

Cluster 2: Has highest amount of Indian Restaurants

Cluster 3: In between



Discussion and Conclusion

Cluster 1 has the least amount of indian restaurants, whereas **Cluster 2** has the highest amount of indian restaurants. Through this data analysis, we can recommend to choose any neighborhood from the **cluster 1** to start a new Indian restaurant, as there is very less competition and a demand for Indian restaurant.