IBM Data Science Professional Certificate

Opening a INDIAN Restaurant in Toronto

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Business Problem

Toronto

- 3 Million inhabitants
- 16th place worldwide with 272 restaurants per 100.000 inhabitants
- **Indian community**
- More than 640,000 people just in Toronto
- Indian cuisine as one of the richest in the world
- If an investor is looking to open a new Indian restaurant where would you recommend it?

Data

List of neighbourhoods in Toronto

- Boroughs and postal codes (Wikipedia)
- Geo coordinates

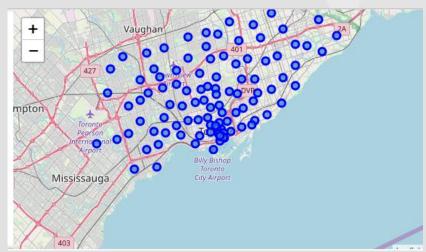
Indian venues data

- Using Foursquare API
 - Category Id for Indian Restaurant

Toronto Neighbourhoods Data Exploratory Analysis

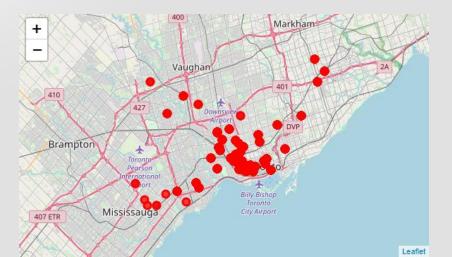
Using Wikipedia data and Coursera CSV

| | PostalCode | Borough | Neighborhood | 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 |



Indian Restaurants Data Exploratory Analysis

Using Foursquare we retrieved all venues per neighbourhood



Data wrangling

Create frequency based data frame per neighbourhood

| | Neighborhood | American Restaurant | Bar | Breakfast Spot | Café | Chinese Restaurant | Coffee Shop | Deli / Bodega | Diner | Event Space |
|---|--|------------------------|---------|-------------------|---------|-----------------------|----------------|------------------|---------|----------------|
| 0 | Adelaide,King,Richmond | 0.0 | 0.02381 | 0.011905 | 0.02381 | 0.000000 | 0.011905 | 0.011905 | 0.02381 | 0.011905 |
| 1 | Agincourt | 0.0 | 0.00000 | 0.000000 | 0.00000 | 0.083333 | 0.000000 | 0.000000 | 0.00000 | 0.000000 |
| 2 | Agincourt North,L'Amoreaux East,Milliken,Steel | 0.0 | 0.00000 | 0.000000 | 0.00000 | 0.058824 | 0.000000 | 0.000000 | 0.00000 | 0.000000 |
| 3 | Albion Gardens,Beaumond Heights,Humbergate,Jam | 0.0 | 0.00000 | 0.000000 | 0.00000 | 0.000000 | 0.000000 | 0.000000 | 0.00000 | 0.000000 |
| 4 | Alderwood,Long Branch | 0.0 | 0.00000 | 0.000000 | 0.00000 | 0.000000 | 0.000000 | 0.000000 | 0.00000 | 0.000000 |

Data wrangling

Create top venues data frame

| | Neighborhood | 1st Most Common Venue | 2nd Most Common Venue | 3rd Most Common Venue | 4th Most Common Venue | 5th Most Common Venue | 6th Most Common Venue | 7th Most Common Venue |
|---|---|--------------------------|----------------------------------|--------------------------|----------------------------------|--------------------------|--------------------------|--------------------------|
| 0 | Adelaide,King,Richmond | Indian Restaurant | Vegetarian / Vegan Restaurant | Café | Diner | Italian Restaurant | Bar | Pizza Place |
| 1 | Agincourt | Indian Restaurant | Chinese Restaurant | Grocery Store | Vegetarian / Vegan Restaurant | Hotel | Bar | Breakfast Spot |
| 2 | Agincourt North,L'Amoreaux East,Milliken,Steel | Indian Restaurant | Indian Chinese Restaurant | Chinese Restaurant | Hotel | Bar | Breakfast Spot | Café |
| 3 | Albion Gardens,Beaumond Heights,Humbergate,Jam | Indian Restaurant | Vegetarian / Vegan Restaurant | Hotel | Bar | Breakfast Spot | Café | Chinese Restaurant |
| 4 | Alderwood,Long Branch | Indian Restaurant | Vegetarian / Vegan Restaurant | Hotel | Bar | Breakfast Spot | Café | Chinese Restaurant |

K-Means

 To perform clustering on the neighbourhoods based on the venue frequency

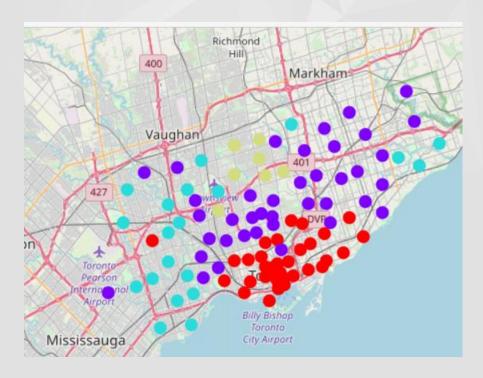
DBSCAN

To perform venue clustering based on concentration

Results

K-Means

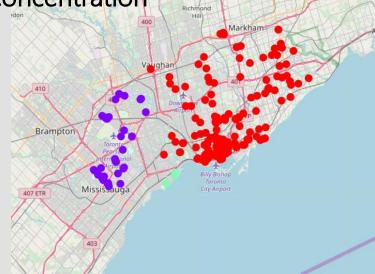
- With 4 clusters
 - Purple (high concentration)
 - Cyan (low concentration)
 - Red (medium concentration)
 - Green (low concentration)



Results

DBSCAN

 Per venue location and applied to neighbourhoods shows density areas and outliers where there is little concentration



Discussion and Conclusion

Recommendation

- Avoid Central area with high density clusters on both K-means and DBSCAN
- Outliers areas in DBSCAN are a good bet
- Red and Green clusters in K-Means have less competition

Conclusion

- Good insight to avoid high competition areas
- More data would be useful (e.g. population density, transports, etc.)