

Applied Data Science Capstone Project : Final Assignment

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Backgroud

New York City (NYC) and Toronto are located in North America and are major financial hubs in the world. They are made up of different skyscrapers and business centers. Both are very cosmopolitan and have dynamic life style. Apart from the commercial perspective, they also build with many high-rise residential building. Many Global organization around the world have office located in these 2 countries. Many people often relocate from other countries to these 2 cities and working in the central business district (CBD) areas. They may not be aware of the similarities or differences in these 2 cities. One of the examples is related to the ethnic makeups in NYC and Toronto. NYC has a much larger Black and Latino population, whereas Toronto has proportionally more Asians and Indians. Hence the likelihood of NYC having more America or south America Restaurant than Toronto is higher.

The target audience for this project is the expatriate who will move to either cities and will work on the CBD areas. Hence the scopes will focus on the Manhattan New York and East, downtown, central and West Toronto areas

Problem and Interests

Given the diversity of the culture, this project will compare the following neighbourhoods of these two cities and determine how similar or dissimilar they are. In total,

- Manhattan consists of 40 neighbourhoods
- East, downtown, central and West Toronto consists of 39 neighbourhoods

It will focus on 3 topics

- Difference of the venue category between these 2 cities.
- Difference between the food culture based on the type of restaurant.
- Both cities will be independently split into clusters by neighbourhood. And then comparison between clusters will be done and identify similarity based on the venue category

It meant to provide the information for expatriates who plan to live in the neighbourhoods around the CBD areas so that they can choose the neighbourhoods best suit to their life style and needs.

Data (a)

Source of Data

Two data sets, one for Manhattan, one for Toronto, created from the previous labs or projects of the training course will be used as the source of data. These datasets have already populated with the information of the boroughs and neighbourhoods of NYC and Toronto as well as the respective latitudes and longitudes.

Before the data analysis, the neighbourhood candidates (NC) need to be filtered from the source of datasets. The outcome will have 2 datasets.

Data (b)

• Neighbourhood Candidates Set A - represent the 40 neighbourhoods of Manhattan. The sample data is as follows

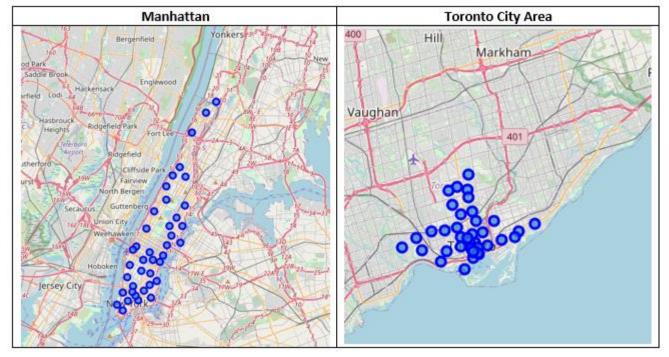
• Neighbourhood Candidates Set B - represent the 39 neighbourhoods of East, downtown, central and West Toronto.

	number	Borough	Neighbourhood	Latitude	Longitude
6	6	Manhattan	Marble Hill	40.876551	-73.910660
100	100	Manhattan	Chinatown	40.715618	-73.994279
101	101	Manhattan	Washington Heights	40.851903	-73.936900
102	102	Manhattan	Inwood	40.867684	-73.921210
103	103	Manhattan	Hamilton Heights	40.823604	-73.949688

	number	Postal Code	Borough	Neighbourhood	Latitude	Longitude
38	37.0	M4E	East Toronto	The Beaches	43.67635739999999	-79.2930312
42	41.0	M4K	East Toronto	The Danforth West, Riverdale	43.6795571	-79.352188
43	42.0	M4L	East Toronto	India Bazaar, The Beaches West	43.6689985	-79.31557159999998
44	43.0	M4M	East Toronto	Studio District	43.6595255	-79.340923
45	44.0	M4N	Central Toronto	Lawrence Park	43.7280205	-79.3887901

Data (c)

Then, the geographical locations of the neighbourhoods will be reviewed to ensure the neighbourhoods are next to each other to ensure they are not scattered too far apart.



Features selection

The venues and venue categories will be the key features for the analysis. Hence, Foursquare API will be used to extract the revenues and revenue categories of all the neighbourhoods for these 2 cities. These data will combine with the datasets Set A and Set B to create new datasets that have the neighbourhoods and the revenue categories.

Methodology

After the data source have been loaded into the dataframe with data cleansing and filtering, Foursquare API will be used to collect the venues, latitudes, longitudes and venue categories for the neighbourhoods of Manhattan and Toronto City area.

To address the 1st audience interest, multiple datasets will be created to store venue categories followed by using "SET" operations to identify

- The common venue categories for both cities.
- The venue categories existed in Manhattan but not in Toronto City Area.
- The venue categories existed in Toronto City but not in Manhattan.

Difference between the food culture based on the type of restaurant will be the 2nd part of interest in this project. The "Restaurant" will be the key word to extract the records from the previous datasets and conduct an analysis or comparison.

Finally, the similarity of neighbourhood based on the venue category will be assessed. To do that, one hot encoding will be use to split the column which contains numerical categorical data to many columns depending on the number of categories present in that column. Both cities will be independently split into clusters by neighbourhood using cluster algorithm "kmeans"; and the comparing the clusters and surface out the similarity based on the venue category

Analysis (a)

Explore the venues of Toronto City Center and Manhattan

Use Foursquare API to extract near by venues, latitudes, longitudes and the venue categories.

Difference of the venue category between these 2 cities

- The number of same venues categories for both Manhattan and Toronto is 192.
- Manhattan has 141 venues categories different from Toronto City.
- Toronto City has 44 venues categories different from Manhattan.

Common Venue Categories	venue categories in	venue categories in Toronto City but			
	Manhattan but not in Toronto	not in Manhattan			
{'Adult Boutique',	{'Accessories Store',	'Airport',			
'American Restaurant',	'Afghan Restaurant',	'Airport Food Court',			
'Antique Shop',	'African Restaurant',	'Airport Gate',			
'Art Gallery',	'Arepa Restaurant',	'Airport Lounge',			
'Art Museum',	'Argentinian Restaurant',	'Airport Service',			
'Arts & Crafts Store',	'Auditorium',	'Airport Terminal',			
'Asian Restaurant',	'Australian Restaurant',	'Aquarium',			
'Athletics & Sports',	'Austrian Restaurant',	'Auto Workshop',			
'BBQ Joint',	'Badminton Court',	'Baseball Stadium',			
'Baby Store',	'Baseball Field',	'Basketball Stadium',			
'Bagel Shop',	'Basketball Court',	'Beach',			
'Bakery',	'Beer Garden',	'Belgian Restaurant',			
'Bank',	'Big Box Store',	'Brewery',			
'Bar',	'Bike Shop',	'Church',			
'Bed & Breakfast',	'Bike Trail',	'College Auditorium',			
'Beer Bar',	'Board Shop',	'College Gym',			
'Beer Store',	'Boxing Gym',	'College Rec Center',			
'Bike Rental / Bike Share',	'Bridal Shop',	'Colombian Restaurant',			
'Bistro',	'Bridge',	'Comfort Food Restaurant',			
'Boat or Ferry',	'Bus Station',	'Comic Shop',			
'Bookstore',	'Cafeteria',	'Coworking Space',			

Analysis (b)

Difference between the food culture based on the type of restaurant

In order to compare the food culture based on the type of restaurant, the "Sets" created in previous slide will be converted back to data frames and filter all rows with "Restaurant" in the venue categories. Please see the sample outcome in the table.

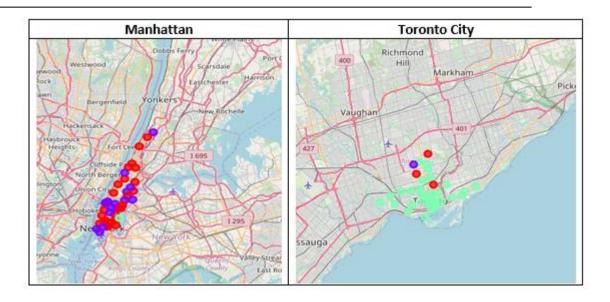
1.			
	List of common Restaurants	List of Restaurants in Manhattan	List of Restaurants in
		but not in Toronto City area	Toronto City area but not in
			Manhattan
			0.5.5.15
		Swiss Restaurant	Comfort Food Restaurant
	Fast Food Restaurant	African Restaurant	Theme Restaurant
	Latin American Restaurant	Malay Restaurant	Gluten-free Restaurant
	Molecular Gastronomy Restaurant	Scandinavian Restaurant	Belgian Restaurant
	Modern European Restaurant	Jewish Restaurant	Doner Restaurant
	Korean Restaurant	Cantonese Restaurant	Tibetan Restaurant
	Ramen Restaurant	Spanish Restaurant	Colombian Restaurant
	Middle Eastern Restaurant	Czech Restaurant	
	Mexican Restaurant	Japanese Curry Restaurant	
	Vietnamese Restaurant	Kebab Restaurant	
	Chinese Restaurant	Dim Sum Restaurant	
	Mediterranean Restaurant	Shanghai Restaurant	
	Taiwanese Restaurant	Lebanese Restaurant	
	Falafel Restaurant	Szechuan Restaurant	
	Filipino Restaurant	Hotpot Restaurant	
	Portuguese Restaurant	Arepa Restaurant	
	Greek Restaurant	Afghan Restaurant	
	Vegetarian / Vegan Restaurant	Empanada Restaurant	
	Indian Restaurant	Paella Restaurant	

Analysis (c)

Both cities will be independently split into clusters by neighbourhood. And then comparison between clusters will be done and identify similarity based on the venue category

Further analyse the neighbourhood of Manhattan & Toronto City is required. To recreate the clusters of neighbourhoods for both cities, **One hot encoding** will be used to split the column which contains numerical categorical data to many columns depending on the number of categories present in that column. In this case, the categorical data of venue categories will split to multiple columns. The clustering labels will be added.

Once the data frames are ready, K-means clustering machine learning algorithm will be applied to cluster the neighbourhoods for both cities. Folium maps will be prepared to show the clusters for both cities.



Finally, new data frames will be built with the top 10 most common venue categories, which will provide the better understanding about the facilities available in each neighbourhood. This can provide better ways to do the comparison. The clusters are as follows

Manhattan Clusters

				IVIGII	hattan C	iustei u	-			
	Neighbourhood	tst Most Common Wines	2nd Most Common Welce	Ord Most Common Writes	4th Most Common Wenge	5th Most Common Venue	6th Most Common Venue	7th Most Common Venue	Om Most Common Venue	9th Most Common Vimpe
100	Chinatown	Chinese Restaurant	Baracy	Cocital Bar	Bubble fee Shop	Ice Cream Shop	Restaurant.	Salen / Eartechop	Ostical Shee	Descert Strep
101	Washington meights	Cafe	Estary	Grocery Store	Dell / Bodega	Chinese Restaurant	Mobile Phone Shop	New American Restautent	Latin American Restaurant	Park
102	Invest	Mexican Rectaurant	Lounge	Restaurant	Care	Balany	Spanish Restaurant	Frazen Yogurt Shop	Carbbean Restaurant	Chinese Restaurant
103	Hamilton Heights	Page Place	Coffee Shop	Cefé	Merican Restautant	Dak/ Bottega	Cootel Ser	Letin American Restaurant	Sushi Restaurant	Park
164	Manhatlanville	Septood Reclayant	Coffee Strop	Restaurant	Mexican Rectaurant	Chorese Restaurest	Det / Exdeps	Suchi Restaurant	Climbing Oym	Supernanut
105	Certral Harten	African Reclaurant	Connetics Shop	French Restaurant	American Restaurant	Tar.	Chinese Restaurant	Act Gallery	Seafood Restaurant	Too
105	East Harlem	Mexican Restaurant	Enlary	Thai Restaurant	Del i Bodeșa	Spe	Letin American Rectaurant	Sandwich Place	Taco Place	Oym
100	Yorkella	Total Rectaured	Qym	Coffee Shop	Det / Botega	Suoni Resignment	6ar	Wine Shop	Ower	Japanese Restaurant
109	Lensu HE	turian Rectaurent	Dusti Restaured	Coffee Shop	Cocital Ear	Picza Place	cini	Gym./ Ethets Center	Gym	Burger Joint
991	Upper West Side	Wine Bar	Eurory	Bar	Restaurant	Indian Restaurant	Carls	Coffee Strop	Pizza Piace	loe Cream Shop
113	Lincoln Square	Cattle	Plaza		Oym / Fitness	Thostor	Concert Hair	Performing	Inde Movie Thester	Qym
				Man	hattan C	luster 1	3	Arts Veture	Sheeter	
Ī	Neighbourhood	1st Most Common Wesse	2nd Most Common Winne	Ord Most Common Wesse	4th Wost Common Venue	Sth West Conumon Venue	6th Most Common Venue	7th Most Common Venue	Sth Most Common Vanue	9th Mo Comes Van
	Martin Hit	Oyes	Discount Store	Sendunch Place	Coffee Shop	Yoga Studio	Pizze Place	Steakhouse	Shopping Mail	Seeto Restaux
107	Upper East Side	Restaurant	Coffee Shop	Einbe	Ballers	Gym / Fibrees Center	American Ressevant	500	Franch Restaurant	Ho
110	Received island	Park	Byn	Dry Cleaner	Bucole Tea Shop	Doccer Field	Farmers Market	Supermanut	Matro Station	Sale
113	Carton	Theater	Hallan Restaurant	Gym / Fitness Center	Coffee Shop	American Restaurant	Gym	Spa	Wine Shop	Ho
154	Melove	Hotel	Clothing Store	Coffee Shop	Sporting Geods Shop	Theater	Exceptive	Cate	Stephouse	0)
113	Muntay Will	Coffee Shop	Standarion Place	Bar	Japanese Restaurant	American Restaurant	Gym / Fitness Center	Burger Joint	Hotel	Mediterrane Restaura
125	Morningside Heights	Coffee Shop	Pan	American Restaurant	Bookstore	Burger Josef	Care	Ice Cream Shop	New American Restaurant	Supernari
127	Battery Park City	Coffee Shop	Park	Potel	Clothing Store	Sym	Memorial Site	Shopping Mail	Wine Shop	Burger Jo
128	Financial District	Coffee Shop	Pizza Place	Bar	Hatel	Oym	Cocital Bar	Park	Merican Restaurant	Gym / Fitne Cen
				Man	hattan C	luster 2	Š			
	Neighbourhood	Tal Most Common Venue	Common	Common	Common	Commo	es Commo	n Commo	n Common	Comm

Toronto Clusters

				Toro	nto Clu	ster 0				
	Neighbourhoo	fel Mor Commo Venu	e Commi	on Common	Commo	n Common	6th Most Common Venue	Common	88 Most Common Venue	9th Mor Commo Venu
45	Lavrence Par	n Fa	n Sen		Yoga Stud	o Diner	Event Space		Electronics Store	Europea Europea Restauran
51	Rocata	is Pa	n. Paygrou	rd Trai	Yoga Stud	o Department Store	Elfriopian Restaurant		Eastern European Reclaurant	Dumplin Restaurar
165	Forest Hill North West, Forest H Road Pai	iii Pa	n Sectors		Jewel Stor		Donal Street	Discount Store	Destributeen Center	Dog Ru
				Toro	nto Clu	ster 1				
, je	Neighbourhood	fat Most Common Venue	2nd Mont Common Venue	3rd Most Common Venue	4th Most Common Venue	5th Most Common Venue	Gth Most Common Venue	7th Most Common Venue	Eth Most Common Venue	9th Most Common Venue
64	Resident	Home Berrice	Gerden	Yoge Studio	Department Store	Event Space	Ethiopian Restaurant	Electronics Store	European Restaurant	Dumpling Restaurant
				Toro	nto Clu	ster 2				
	Neighbourhood	Tat Most Common Venue	2nd Most Common Vanue	3rd Most Common Venue	4th Mos Common Venue	Common	6th Most Common Venue	7th Most Commos Venue	Eth Most Common Witner	5th Most Common Vanue
36	The Beaches	Coffee Shap	Health Food Store	Neigreomood	he	n Pa	Yoge Studio	Dog Run	Diner	Discount
42	The Danforth West, Rivertale	Greek Restaurant	Coffee Shop	Restaurant	toe Crean Shop		Liquor Store	Indian Restaurant	Spe	Sociators
43	India Babasi. The Beaches West		Park	Fast Food Rectaurant	Coffee Shop	Food & Drink Shap	Light Ruil Station	Restaurant	Italian Restaurant	Fish & Chips Shop
44	Studio District	Coffee Stop	American Restourant	Barary	Braver	Cata	Gestropub	Yoga Studio	Fish Market	Pet Store
46	Oaveville North	Oym / Fitness Center	Hatel	Breakfast Spot	Food 6 Drink Shop		Department Store	Park	Convenience Store	Distribution Certier
42	North Toronto West, Lawrence Park	Clothing Store	Coffee Shop	Yoga thutio	Fact Food Restauran		Cure	Mexican Restaurant	Salon / Barbershop	Metro Station
48	Davisville	Sandwich Place	Dessert Shop	Ficta Place	Coffee Shap	Sustri Restaurant	Cure	Totion Restaurant	Oym	Gen Station
49	Woore Fark. Summerhill East	Lauryer	Restaurant	Yoge Studio	Desser Shop	Evert Space	Ethiopian Postaurarti	Electronics Store	Eastern European Restaurant	Dumping Restaurant
50	Summerhill West, Rathnelly, South Hill, Forest.	Coffee Shop	American Restaurent	Liquor Store	Restauran	f Dark	Slapel Shop	Supermarket	Subi Restaurant	Fried Chicten Joint
92	St. James Tovin, Cabbagetouri	Coffee	CW	Fizza Place	Restauran	teller. Restaurant	Barary	Pus	Beer Store	Chinese Restaurant

Results & Discussion

It is interesting to see the different shapes of the 2 cities; Manhattan is rectangular shaped while Toronto City Area is more squarish shaped.

The number of same venues categories for both Manhattan and Toronto is 192. Manhattan has 141 venues categories different from Toronto City. Toronto City has 44 venues categories different from Manhattan.

The food culture for both cities covered almost all regions: Asia, LATAM, European, Australia, Africa, North and South America. Manhattan has many restaurants, which are related to different provinces of different countries. Toronto City area has some interesting theme restaurants, which not exist in Manhattan.

The neighbourhood of cluster 0 and 1 of Manhattan are similar to cluster 2 of Toronto City, given that these clusters have good mix of venues and facilities such as cafe / coffeeshop, Gym, Spa, restaurants, Part and Hotel.

Cluster 0 of Toronto City has more parks, trails, playgrounds, restaurants, shops that seem to be more suitable for family living.

Both cluster 2 of Manhattan and cluster 0 of Toronto City have only 1 neighbourhood, the comparison is inconclusive, given that the venues are quite different.

Conclusion

Although some differences have been surfaced out from this study, the majority of venue categories are similar. Expatriates stayed in either one of the cities should have no problem to adapt on the other.