# APPLIED DATA SCIENCE CAPSTONE

Final Assignment Report

# **Table of Content**

- 1 Introduction
- 2 Data
- 3 Methodology
- 4 Analysis
- 5 Results & Discussion
- 6 Conclusion

# 1. Introduction

# **Background**

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 buildings. 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 choose the neighbourhoods best suit to their life style and needs.

# 2. Data

# Source of Data, Data Acquisition and Cleansing

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.

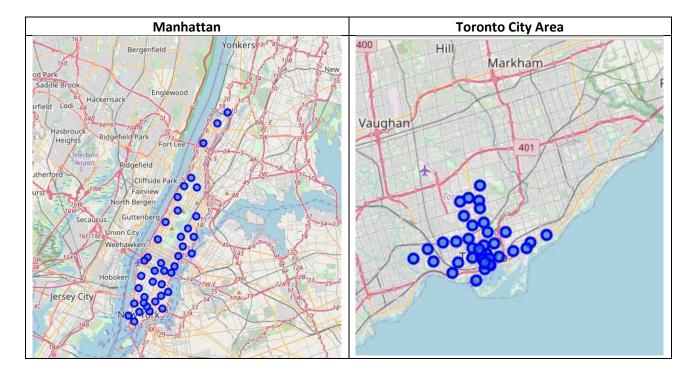
 Neighbourhood Candidates Set A - represent the 40 neighbourhoods of Manhattan. The sample data is as follows

	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

 Neighbourhood Candidates Set B - represent the 39 neighbourhoods of East, downtown, central and West Toronto. The sample data is as follows

	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

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.

# 3. Methodology

After the data source have been loaded into the data frame 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 "**k-means**"; and the comparing the clusters and surface out the similarity based on the venue category

# 4. Analysis

To start the data exploratory analysis, I will explore the venues of Toronto City Center and Manhattan. The Foursquare API will be used to collect the information. To trigger the FourSqure API, the credentials and API parameters will be prepared.

As multiple invocations of FourSquare API are required to extract the complete information, a function called "getNearbyVenues" is defined to get near by venues, latitudes, longitudes and the venue categories. Then this function will be triggered separately to get the near by venues for Toronto City and Manhattan.

The followings show the neighbourhoods, near by venues, latitudes, longitudes and the venue categories for both Toronto City and Manhattan.

|--|

	Neighbourhood	Neighbourhood Latitude	Neighbourhood Longitude		\	/enue	Venue Latitud	e Venue Longitude	Venue Category
0	The Beaches	43.67635739999999	-79.2930312		Glen Manor R	Ravine	43.67682	1 -79.293942	Trail
1	The Beaches	43.67635739999999	-79.2930312	The Big Car	ot Natural Food M	//arket	43.67887	9 -79.297734	Health Food Store
2	The Beaches	43.67635739999999	-79.2930312		Grover Pub and	Grub	43.67918	1 -79.297215	Pub
3	The Beaches	43.67635739999999	-79.2930312		Upper Bea	aches	43.68056	3 -79.292869	Neighborhood
4	The Beaches	43.67635739999999	-79.2930312		Dip	'n Sip	43.67889	7 -79.297745	Coffee Shop
	Neighbourhood	Neighbourhood Latitude	Neighbourhood L	ongitude.	Venue	Venu	e Latitude	Venue Longitude	Venue Category
_	Neighbourhood	Neighbourhood Latitude	Neighbourhood L	ongitude	Venue	Venu	e Latitude	Venue Longitude	Venue Category
0	Marble Hill	40.876551		-73.91066	Arturo's		40.874412	-73.910271	Pizza Place
1	Marble Hill	40.876551		-73.91066	Bikram Yoga		40.876844	-73.906204	Yoga Studio
2	Marble Hill	40.876551		-73.91066	Tibbett Diner		40.880404	-73.908937	Diner
3	Marble Hill	40.876551		-73.91066	Starbucks		40.877531	-73.905582	Coffee Shop
4	Marble Hill	40.876551		-73.91066	Dunkin'		40.877136	-73.906666	Donut Shop

# 4.1. Difference of the venue category between these 2 cities

After the creation of the 2 data frames above, the extraction of the venue categories will be done followed by conversions of the columns to "SET" data type to analyse the similarity and the differences between the neighbourhoods in both cities. The followings show the

- common venue categories in both cities
- venue categories in Manhattan but not in Toronto
- venue categories in Toronto City but not in Manhattan

<b>Common Venue Categories</b>	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',
'Boutique',	'Camera Store',	'Distribution Center',
'Brazilian Restaurant',	'Cantonese Restaurant',	'Doner Restaurant',
'Breakfast Spot',	'Cha Chaan Teng',	'Fish & Chips Shop',
'Bubble Tea Shop',	'Christmas Market',	'Fruit & Vegetable Store',
'Building',	'Circus',	'General Travel',
'Burger Joint',	'Club House',	'Gluten-free Restaurant',
'Burrito Place',	'College Academic Building',	'Home Service',
'Bus Line',	'College Bookstore',	'Hospital',
'Butcher',	'College Theater',	'IT Services',
'Café',	'Comedy Club',	'Lake',
'Cajun / Creole Restaurant',	'Community Center',	'Lawyer',

'Cooking School', 'Light Rail Station', 'Candy Store', 'Caribbean Restaurant', 'Cycle Studio', 'Neighborhood', 'Cheese Shop', 'Czech Restaurant', 'Other Great Outdoors', 'Chinese Restaurant', 'Daycare', 'Plane'. 'Dim Sum Restaurant', 'Chocolate Shop', 'Poutine Place', 'Climbing Gym', 'Dive Bar', 'Sculpture Garden', "Doctor's Office", 'Clothing Store', 'Stadium', 'Cocktail Bar', 'Drugstore', 'Stationery Store', 'Dry Cleaner', 'Coffee Shop', 'Swim School', 'College Arts Building', 'Duty-free Shop', 'Tanning Salon', 'College Cafeteria', 'Empanada Restaurant', 'Theme Restaurant', 'Concert Hall', 'English Restaurant', 'Tibetan Restaurant'} 'Convenience Store'. 'Exhibit'. 'Cosmetics Shop', 'Eye Doctor', 'Creperie', 'Food Stand', 'Cuban Restaurant', 'Golf Course', 'Cupcake Shop', 'Gym Pool', 'Dance Studio', 'Gymnastics Gym', 'Deli / Bodega', 'Hardware Store', 'Department Store', 'Hawaiian Restaurant', 'Dessert Shop', 'Heliport', 'High School', 'Diner', 'Discount Store', 'Hill', 'Himalayan Restaurant', 'Dog Run', 'Hostel', 'Donut Shop', 'Dumpling Restaurant', 'Hot Dog Joint', 'Eastern European Restaurant', 'Hotpot Restaurant', 'Electronics Store', 'Indie Theater'. 'Ethiopian Restaurant', 'Israeli Restaurant', 'Japanese Curry Restaurant', 'Event Space', 'Falafel Restaurant', 'Jewish Restaurant', 'Farmers Market', 'Karaoke Bar'. 'Fast Food Restaurant', 'Kebab Restaurant', 'Filipino Restaurant', 'Kids Store', 'Fish Market', 'Kitchen Supply Store', 'Flea Market', 'Kosher Restaurant', 'Flower Shop', 'Laundry Service', 'Food & Drink Shop', 'Lebanese Restaurant', 'Food Court', 'Library', 'Food Truck'. 'Malay Restaurant', 'Fountain', 'Massage Studio', 'French Restaurant'. 'Mattress Store'. 'Fried Chicken Joint', 'Medical Center', 'Frozen Yogurt Shop', 'Memorial Site', 'Mini Golf', 'Furniture / Home Store', 'Gaming Cafe', 'Mobile Phone Shop', 'Garden', 'Music School', 'Garden Center', 'Music Store', 'Gas Station', 'Nail Salon', 'Gastropub', 'Newsstand', 'Gay Bar', 'Non-Profit'. 'General Entertainment', 'North Indian Restaurant', 'German Restaurant', 'Outdoor Sculpture', 'Gift Shop', 'Outdoors & Recreation', 'Gourmet Shop', 'Paella Restaurant', 'Greek Restaurant', 'Paper / Office Supplies Store', 'Grocery Store', 'Pedestrian Plaza', 'Gym', 'Persian Restaurant', 'Gym / Fitness Center', 'Peruvian Restaurant', 'Harbor / Marina', 'Pet Café', 'Health & Beauty Service', 'Pet Service', 'Health Food Store', 'Photography Studio', 'Historic Site', 'Physical Therapist', 'History Museum', 'Pie Shop', 'Pier', 'Hobby Shop', 'Hookah Bar', 'Pilates Studio', 'Hotel', 'Public Art'. 'Hotel Bar', 'Residential Building (Apartment / Condo) 'Ice Cream Shop', 'Indian Restaurant', 'Resort',

'Indie Movie Theater',

'Rest Area'

'Intersection', 'River', 'Irish Pub', 'Rock Climbing Spot', 'Italian Restaurant', 'Rock Club', 'Japanese Restaurant', 'Scandinavian Restaurant', 'Jazz Club', 'School', 'Jewelry Store', 'Shanghai Restaurant', 'Juice Bar', 'Shipping Store', 'Korean Restaurant', 'Shoe Repair', 'Soba Restaurant', 'Latin American Restaurant', 'Lingerie Store', 'Soccer Field', 'Social Club'. 'Liquor Store', 'Lounge', 'South American Restaurant', 'Market'. 'South Indian Restaurant'. 'Martial Arts School', 'Southern / Soul Food Restaurant', 'Mediterranean Restaurant', 'Spanish Restaurant', "Men's Store", 'Spiritual Center', 'Metro Station', 'Sports Club', 'Mexican Restaurant', 'Stables', 'Middle Eastern Restaurant', 'Street Art', 'Miscellaneous Shop', 'Swiss Restaurant', 'Modern European Restaurant', 'Szechuan Restaurant', 'Molecular Gastronomy Restaurant', 'Tapas Restaurant', 'Monument / Landmark', 'Tech Startup', 'Tennis Court', 'Moroccan Restaurant', 'Movie Theater', 'Tennis Stadium', 'Thrift / Vintage Store', 'Museum', 'Music Venue', 'Tiki Bar', 'New American Restaurant', 'Tourist Information Center', 'Nightclub', 'Turkish Restaurant', 'Noodle House', 'Udon Restaurant', 'Office', 'Used Bookstore', 'Opera House', 'Venezuelan Restaurant', 'Optical Shop', 'Veterinarian', 'Video Store', 'Organic Grocery', 'Park', 'Volleyball Court', 'Performing Arts Venue', 'Waterfront', 'Pet Store', 'Whisky Bar', 'Pharmacy', 'Wings Joint', 'Pizza Place' "Women's Store"} 'Playground', 'Plaza', 'Poke Place'. 'Pool', 'Portuguese Restaurant', 'Post Office', 'Pub', 'Ramen Restaurant', 'Record Shop', 'Rental Car Location', 'Restaurant', 'Roof Deck', 'Sake Bar', 'Salad Place', 'Salon / Barbershop', 'Sandwich Place', 'Scenic Lookout', 'Seafood Restaurant', 'Shoe Store', 'Shopping Mall', 'Skate Park', 'Skating Rink', 'Smoke Shop', 'Smoothie Shop', 'Snack Place', 'Soup Place', 'Spa', 'Speakeasy', 'Sporting Goods Shop', 'Sports Bar', 'Steakhouse', 'Strip Club',

'Supermarket',	
'Supplement Shop',	
'Sushi Restaurant',	
'Taco Place',	
'Tailor Shop',	
'Taiwanese Restaurant',	
'Tea Room',	
'Thai Restaurant',	
'Theater',	
'Toy / Game Store',	
'Trail',	
'Train Station',	
'Vegetarian / Vegan Restaurant',	
'Video Game Store',	
'Vietnamese Restaurant',	
'Wine Bar',	
'Wine Shop',	
'Yoga Studio'}	

- 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.

# 4.2. 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 section 4.1. will be converted back to data frames and filter all rows with "Restaurant" in the venue categories. The outcome is as follows

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
		iviaililattaii
Fast Food Restaurant	Swiss Restaurant	Comfort Food Restaurant
Latin American Restaurant	African Restaurant	Theme Restaurant
Molecular Gastronomy Restaurant	Malay Restaurant	Gluten-free Restaurant
Modern European Restaurant	Scandinavian Restaurant	Belgian Restaurant
Korean Restaurant	Jewish Restaurant	Doner Restaurant
Ramen Restaurant	Cantonese Restaurant	Tibetan Restaurant
Middle Eastern Restaurant	Spanish Restaurant	Colombian Restaurant
Mexican Restaurant	Czech Restaurant	
Vietnamese Restaurant	Japanese Curry Restaurant	
Chinese Restaurant	Kebab Restaurant	
Mediterranean Restaurant	Dim Sum Restaurant	
Taiwanese Restaurant	Shanghai Restaurant	
Falafel Restaurant	Lebanese Restaurant	
Filipino Restaurant	Szechuan Restaurant	
Portuguese Restaurant	Hotpot Restaurant	
Greek Restaurant	Arepa Restaurant	
Vegetarian / Vegan Restaurant	Afghan Restaurant	
Indian Restaurant	Empanada Restaurant	
Japanese Restaurant	Paella Restaurant	
Dumpling Restaurant	Hawaiian Restaurant	
Seafood Restaurant	Turkish Restaurant	
German Restaurant	South Indian Restaurant	
Thai Restaurant	Tapas Restaurant	
Cajun / Creole Restaurant	Austrian Restaurant	
Eastern European Restaurant	Persian Restaurant	
Cuban Restaurant	Soba Restaurant	
Caribbean Restaurant	English Restaurant	
Restaurant	Israeli Restaurant	
Brazilian Restaurant	Australian Restaurant	
Moroccan Restaurant	South American Restaurant	
French Restaurant	Himalayan Restaurant	
Sushi Restaurant	Southern / Soul Food Restaurant	

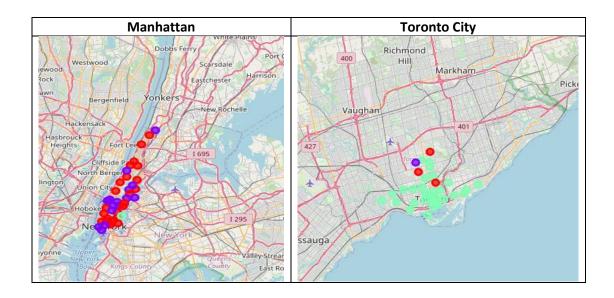
American Restaurant Asian Restaurant New American Restaurant Ethiopian Restaurant Italian Restaurant North Indian Restaurant Argentinian Restaurant Kosher Restaurant Peruvian Restaurant Udon Restaurant Venezuelan Restaurant

# 4.3. 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. The sample data is as follows

						Manl	hattan						
	Neighbourhood	Accessori Sto	es Ad ore Boutio		fghan nurant R	African estaurant	American Restaurant		Arepa Restaurant	Argentinian Restaurant	Art Gallery	Video Store	
0	Battery Park City	(	0.0	0.0	0.0	0.000000	0.000000	0.0	0.0	0.000000	0.000000	0.0	0.
1	Carnegie Hill	(	0.0	0.0	0.0	0.000000	0.011364	0.0	0.0	0.011364	0.000000	0.0	0.
2	Central Harlem	(	0.0	0.0	0.0	0.065217	0.043478	0.0	0.0	0.000000	0.043478	0.0	0.
3	Chelsea	(	0.0	0.0	0.0	0.000000	0.030000	0.0	0.0	0.000000	0.060000	0.0	0.
4	Chinatown	(	0.0	0.0	0.0	0.000000	0.030000	0.0	0.0	0.000000	0.000000	0.0	0.
						Toron	to City						
	Neighbourhood	Adult Boutique	Airport	Airport Food Court	Airpor Gate			Airport Terminal			Tibeta ' Restauran		Trail
0	Berczy Park	0.0	0.000000	0.000000	0.00000	0.00000	0.000000	0.000000	0.0	0.0	. 0.	0.0	0.0
1	Brockton, Parkdale Village, Exhibition Place	0.0	0.000000	0.000000	0.00000	0.00000	0.000000	0.000000	0.0	0.0	. 0.	0.0	0.0
2	Business reply mail Processing Centre, South C	0.0	0.000000	0.000000	0.00000	0.00000	0.000000	0.000000	0.0	0.0	. 0.	0.0	0.0
3	CN Tower, King and Spadina, Railway Lands, Har	0.0	0.055556	0.055556	0.05555	6 0.11111	1 0.166667	0.111111	0.0	0.0	. 0.	0.0	0.0
4	Central Bay Street	0.0	0.000000	0.000000	0.00000	0.00000	0.000000	0.000000	0.0	0.0	. 0.	0.0	0.0

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**

					luster 0	nhattan C	<u>M</u> ar				
on	9th Mos Common Venue	8th Most Common Venue	7th Most Common Venue	6th Most Common Venue	5th Most Common Venue	4th Most Common Venue	3rd Most Common Venue	2nd Most Common Venue	1st Most Common Venue	Neighbourhood	
ор	Dessert Shop	Optical Shop	Salon / Barbershop	Hotpot Restaurant	Ice Cream Shop	Bubble Tea Shop	Cocktail Bar	Bakery	Chinese Restaurant	Chinatown	100
ırk	Pari	Latin American Restaurant	New American Restaurant	Mobile Phone Shop	Chinese Restaurant	Deli / Bodega	Grocery Store	Bakery	Café	Washington Heights	101
	Chinese Restauran	Caribbean Restaurant	Frozen Yogurt Shop	Spanish Restaurant	Bakery	Café	Restaurant	Lounge	Mexican Restaurant	Inwood	102
ırk	Pari	Sushi Restaurant	Latin American Restaurant	Cocktail Bar	Deli / Bodega	Mexican Restaurant	Café	Coffee Shop	Pizza Place	Hamilton Heights	103
et	Supermarke	Climbing Gym	Sushi Restaurant	Deli / Bodega	Chinese Restaurant	Mexican Restaurant	Italian Restaurant	Coffee Shop	Seafood Restaurant	Manhattanville	104
pa	Spa	Seafood Restaurant	Art Gallery	Chinese Restaurant	Bar	American Restaurant	French Restaurant	Cosmetics Shop	African Restaurant	Central Harlem	105
m	Gyn	Taco Place	Sandwich Place	Latin American Restaurant	Spa	Deli / Bodega	Thai Restaurant	Bakery	Mexican Restaurant	East Harlem	106
	Japanese Restauran	Diner	Wine Shop	Bar	Sushi Restaurant	Deli / Bodega	Coffee Shop	Gym	Italian Restaurant	Yorkville	108
int	Burger Join	Gym	Gym / Fitness Center	Café	Pizza Place	Cocktail Bar	Coffee Shop	Sushi Restaurant	Italian Restaurant	Lenox Hill	109
	Ice Crean Shop	Pizza Place	Coffee Shop	Café	Indian Restaurant	Italian Restaurant	Bar	Bakery	Wine Bar	Upper West Side	111
m	Gyn	Indie Movie Theater	Performing Arts Venue	Concert Hall	Theater	Gym / Fitness Center	Italian Restaurant	Plaza	Café	Lincoln Square	112

	Neighbourhood	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	8th Most Common Venue	9th Most Common Venue
6	Marble Hill	Gym	Discount Store	Sandwich Place	Coffee Shop	Yoga Studio	Pizza Place	Steakhouse	Shopping Mall	Seafood Restaurant
107	Upper East Side	Italian Restaurant	Coffee Shop	Exhibit	Bakery	Gym / Fitness Center	American Restaurant	Spa	French Restaurant	Hotel
110	Roosevelt Island	Park	Gym	Dry Cleaner	Bubble Tea Shop	Soccer Field	Farmers Market	Supermarket	Metro Station	School
113	Clinton	Theater	Italian Restaurant	Gym / Fitness Center	Coffee Shop	American Restaurant	Gym	Spa	Wine Shop	Hotel
114	Midtown	Hotel	Clothing Store	Coffee Shop	Sporting Goods Shop	Theater	Bookstore	Café	Steakhouse	Gym
115	Murray Hill	Coffee Shop	Sandwich Place	Bar	Japanese Restaurant	American Restaurant	Gym / Fitness Center	Burger Joint	Hotel	Mediterranean Restaurant
125	Morningside Heights	Coffee Shop	Park	American Restaurant	Bookstore	Burger Joint	Café	Ice Cream Shop	New American Restaurant	Supermarket
127	Battery Park City	Coffee Shop	Park	Hotel	Clothing Store	Gym	Memorial Site	Shopping Mall	Wine Shop	Burger Joint
128	Financial District	Coffee Shop	Pizza Place	Bar	Hotel	Gym	Cocktail Bar	Park	Mexican Restaurant	Gym / Fitness Center
				Manl	nattan C	Cluster 2				
	Neighbourhood	1st Most Common Venue	Common	Common	4th Mos Commo Venu	n Commo	on Comm	on Comm	on Common	Common
275	Stuyvesant Town	Park	Bar	. Boat or Ferry	Coffee Sho	p Helipo	ort Food Tru	ick Gas Stati	on Bistro	Skating Rink

# **Toronto Clusters**

				Tord	nto Clu	ıster 0				
	Neighbourhoo	1st Most Common Venue	Commo	n Commo	Commo	on Common	Common	7th Most Common Venue	8th Most Common Venue	9th Most Common Venue
45	Lawrence Pa	rk Park	Swir Scho		e Yoga Stud	lio Diner	. Event Space	Ethiopian Restaurant	Electronics Store	Eastern European Restaurant
51	Roseda	le Park	Playgroun	nd Tra	il Yoga Stud	Department Store		Electronics Store	Eastern European Restaurant	Dumpling Restaurant
65	Forest Hill North West, Forest H Road Pa	ill Park	Sus Restaurar		Jewel Sto		Donut Shop	Discount Store	Distribution Center	Dog Run
				Tord	nto Clu	ıster 1				
ı	Neighbourhood		2nd Most Common Venue	3rd Most Common Venue	4th Most Common Venue	5th Most Common Venue	6th Most Common Venue	7th Most Common Venue	8th Most Common Venue	9th Most Common Venue
64	Roselawn	Home Service	Garden	Yoga Studio	Department Store	Event Space	Ethiopian Restaurant	Electronics Store	Eastern European Restaurant	Dumpling Restaurant
				Tord	nto Clu	ıster 2				

	Neighbourhood	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	8th Most Common Venue	9th Most Common Venue
38	The Beaches	Coffee Shop	Health Food Store	Neighborhood	Trail	Pub	Yoga Studio	Dog Run	Diner	Discount Store
42	The Danforth West, Riverdale	Greek Restaurant	Coffee Shop	Italian Restaurant	Ice Cream Shop	Furniture / Home Store	Liquor Store	Indian Restaurant	Spa	Bookstore
43	India Bazaar, The Beaches West	Sandwich Place	Park	Fast Food Restaurant	Coffee Shop	Food & Drink Shop	Light Rail Station	Restaurant	Italian Restaurant	Fish & Chips Shop
44	Studio District	Coffee Shop	American Restaurant	Bakery	Brewery	Café	Gastropub	Yoga Studio	Fish Market	Pet Store
46	Davisville North	Gym / Fitness Center	Hotel	Breakfast Spot	Food & Drink Shop	Sandwich Place	Department Store	Park	Convenience Store	Distribution Center
47	North Toronto West, Lawrence Park	Clothing Store	Coffee Shop	Yoga Studio	Fast Food Restaurant	Italian Restaurant	Café	Mexican Restaurant	Salon / Barbershop	Metro Station
48	Davisville	Sandwich Place	Dessert Shop	Pizza Place	Coffee Shop	Sushi Restaurant	Café	Italian Restaurant	Gym	Gas Station
49	Moore Park, Summerhill East	Lawyer	Restaurant	Yoga Studio	Dessert Shop	Event Space	Ethiopian Restaurant	Electronics Store	Eastern European Restaurant	Dumpling Restaurant
50	Summerhill West, Rathnelly, South Hill, Forest	Coffee Shop	American Restaurant	Liquor Store	Restaurant	Bank	Bagel Shop	Supermarket	Sushi Restaurant	Fried Chicken Joint
52	St. James Town, Cabbagetown	Coffee Shop	Café	Pizza Place	Restaurant	Italian Restaurant	Bakery	Pub	Beer Store	Chinese Restaurant

# 5. 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.

# 6. 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.