

# The Battle of Neighborhoods



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IBM Data Science Professional Certificate – Capstone Project

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# Introduction

#### **Business problem**

Every city is unique in their own way and give something new. The city of New York is the most populous city in USA. It is diverse and is the financial capital of USA. It is multicultural and rich in heritage and developed enough from a foreign perspective.

Toronto, the capital of Ontario is a major Canadian city. It is the most populous city in Canada and fourth most populous city in North America. Both cities are very diverse and are the financial capitals of their respective countries.

Today Tourism is one of the most essential components of the economy. Most often people would like to visit countries which are having friendly environment. Tourists always eager to travel to different places on the basis of available information. The comparison between two cities always support to choose the specific places according to their choice.

#### Who would be interested?

Jane is a tourist, and she needs to explore how much these two cities, New York and Toronto are similar or dissimilar in aspects from a tourist point of view regarding food, accommodation, beautiful places etc. and to get more helpful information when making decisions about where to travel.

# **Data**

#### **Description of the data**

For this project, Foursquare API will be used to gain the data of two cities, in terms of their neighborhoods. The information we want to focus on are restaurants, hotels, coffee shops, parks, theaters, art galleries, museums around each neighborhood. One Borough from each city will be chosen to analyze their neighborhoods. Therefore, Manhattan from New York and Downtown Toronto from Toronto will be considered.

#### How data will be used to solve the problem

According to the objectives of this project, Machine Learning technique called "Clustering" will be used to segment the neighborhoods with similar objects on the basis of each neighborhood data. Based on foot traffic

in respective neighborhoods tourist's areas and hubs will be located and that information will be used to decide the similarities or dissimilarities between two cities.

# Methodology

Initially we have selected Downtown Toronto and Manhattan to explore their neighborhoods. The data exploration, analysis and visualization for both boroughs are done in the same way but separately.

## **Downtown Toronto**

First Downtown Toronto was considered.

- For Downtown Toronto case, we have extracted table of Toronto's Borough from Wikipedia: http://zims-en.kiwix.campusafrica.gos.orange.com/wikipedia\_en\_all\_nopic/A/List\_of\_postal\_codes\_of\_Canada:\_M
- First the data was arranged according to our requirements.
- In the arrangement phase, which applied multiple steps including but not limited to, eliminating "Not assigned" values,

Neighbourhood	Borough	Postcode	
Not assigned	Not assigned	M1A	0
Not assigned	Not assigned	M2A	1
Parkwoods	North York	МЗА	2
Victoria Village	North York	M4A	3
Harbourfront	Downtown Toronto	M5A	4
Lawrence Heights	North York	M6A	5
Lawrence Manor	North York	M6A	6

```
toronto_df= toronto df[[toronto df["Borough"] != "Not assigned"]
       toronto_df.head()
          Postcode
[96]:
                                       Neighborhood
                            Borough
       2
              МЗА
                           North York
                                           Parkwoods
       3
              M4A
                           North York
                                        Victoria Village
              M5A Downtown Toronto
                                         Harbourfront
       4
       5
              M6A
                           North York Lawrence Heights
              M6A
       6
                           North York
                                       Lawrence Manor
[102]: sum(toronto_df['Neighborhood'] == "Not assigned")
[102]: 0
```

• Since I was not able to get the geographical coordinates of the neighborhoods using the Geocoder package, I have used the csv file that has the geographical coordinates of each postal code.

• Combine neighborhoods which have same geographical coordinates at each borough and sorted against the concerned borough.

[110]:		Postcode	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

• Set index for only Downtown Toronto and eliminate 'Postcode' column.

[27

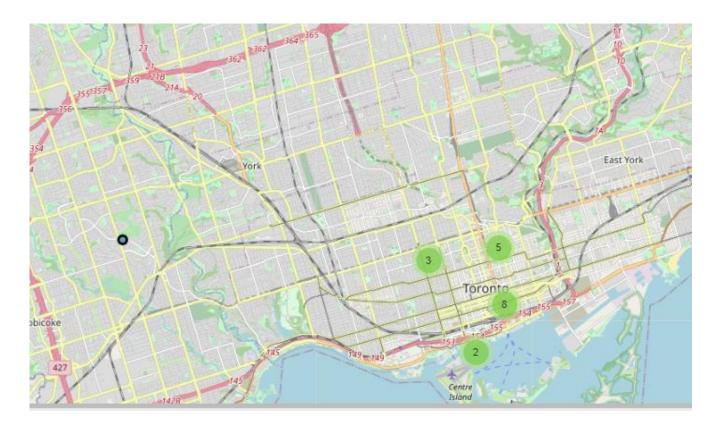
[27]: downtown\_toronto\_data = toronto\_df[toronto\_df['Borough'] == 'Downtown Toronto'].reset\_index(drop=True)
 downtown\_toronto\_data=downtown\_toronto\_data.drop(['Postcode'], axis=1)
 downtown\_toronto\_data

]:		Borough	Neighborhood	Latitude	Longitude
	0	Downtown Toronto	Rosedale	43.679563	-79.377529
	1	Downtown Toronto	Cabbagetown, St. James Town	43.667967	-79.367675
	2	Downtown Toronto	Church and Wellesley	43.665860	-79.383160
	3	Downtown Toronto	Harbourfront	43.654260	-79.360636
	4	Downtown Toronto	Ryerson, Garden District	43.657162	-79.378937
	5	Downtown Toronto	St. James Town	43.651494	-79.375418
	6	Downtown Toronto	Berczy Park	43.644771	-79.373306
	7	Downtown Toronto	Central Bay Street	43.657952	-79.387383
	8	Downtown Toronto	Adelaide, King, Richmond	43.650571	-79.384568
	9	Downtown Toronto	Harbourfront East, Toronto Islands, Union Station	43.640816	-79.381752
	10	Downtown Toronto	Design Exchange, Toronto Dominion Centre	43.647177	-79.381576
	11	Downtown Toronto	Commerce Court, Victoria Hotel	43.648198	-79.379817
	12	Downtown Toronto	Harbord University of Toronto	43 662696	-79 400049

For data verification and further exploration, Foursquare API was used to get the coordinates of Downtown Toronto and explore its neighborhoods.

The neighborhoods are further characterized as venues and venue categories. Put blue dots on Bucharest map to see centers of neighborhoods.





#### Toronto venues

Rosedale Cabbagetown, St. James Town Church and Wellesley Harbourfront Ryerson, Garden District St. James Town Berczy Park Central Bay Street Adelaide, King, Richmond Harbourfront East, Toronto Islands, Union Station Design Exchange, Toronto Dominion Centre Commerce Court, Victoria Hotel Harbord, University of Toronto Chinatown, Grange Park, Kensington Market CN Tower, Bathurst Quay, Island airport, Harbourfront West, King and Spadina, Railway Lands, South Niagara Stn A PO Boxes 25 The Esplanade First Canadian Place, Underground city Christie

• The top 10 venues for each neighborhood were obtained.

	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	8th Most Common Venue	9th Most Common Venue	10th Most Common Venue
0	Adelaide, King, Richmond	Seafood Restaurant	Coffee Shop	Steakhouse	Café	Plaza	Restaurant	Concert Hall	Asian Restaurant	Smoke Shop	Speakeasy
1	Berczy Park	Seafood Restaurant	Farmers Market	Park	Restaurant	Concert Hall	Beer Bar	Basketball Stadium	Cheese Shop	Bakery	Jazz Club
2	CN Tower, Bathurst Quay, Island airport, Harbo	Airport Service	Airport Lounge	Airport Terminal	Boat or Ferry	Bar	Airport	Airport Food Court	Airport Gate	Coffee Shop	Harbor / Marina
3	Cabbagetown, St. James Town	Café	Restaurant	Gift Shop	Italian Restaurant	Park	Caribbean Restaurant	Pet Store	General Entertainment	Butcher	Pub
4	Central Bay Street	Coffee Shop	Gastropub	Art Museum	Café	Bubble Tea Shop	Poke Place	Modern European Restaurant	Middle Eastern Restaurant	Sandwich Place	Japanese Restaurant
5	Chinatown, Grange Park, Kensington Market	Café	Vietnamese Restaurant	Bakery	Fish Market	Farmers Market	Dessert Shop	Mexican Restaurant	Organic Grocery	Coffee Shop	Cocktail Bar
6	Christie	Grocery Store	Café	Park	Baby Store	Italian Restaurant	Nightclub	Coffee Shop	Restaurant	Candy Store	Art Gallery
7	Church and Wellesley	Pizza Place	Creperie	Park	Coffee Shop	Mexican Restaurant	Bubble Tea Shop	Breakfast Spot	Bookstore	Pub	Ramen Restaurant
8	Commerce Court, Victoria Hotel	Café	Gastropub	Coffee Shop	Bakery	Gym / Fitness Center	Japanese Restaurant	Deli / Bodega	Museum	Gluten-free Restaurant	Pub

# Clustering Neighborhoods - Downtown, Toronto

• A new data frame was created including the cluster labels as well as the top 10 venues for each neighborhood.

4]:		Borough	Neighborhood	Latitude	Longitude	Cluster Labels	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	Con \
	0	Downtown Toronto	Rosedale	43.679563	-79.377529	1	Park	Playground	Trail	Dance Studio	Cheese Shop	Chocolate Shop	Clothing Store	Cocktail Bar	Coffee Shop	Cı
	1	Downtown Toronto	Cabbagetown, St. James Town	43.667967	-79.367675	3	Café	Restaurant	Gift Shop	Italian Restaurant	Park	Caribbean Restaurant	Pet Store	General Entertainment	Butcher	
	2	Downtown Toronto	Church and Wellesley	43.665860	-79.383160	3	Pizza Place	Creperie	Park	Coffee Shop	Mexican Restaurant	Bubble Tea Shop	Breakfast Spot	Bookstore	Pub	R Resta
	3	Downtown Toronto	Harbourfront	43.654260	-79.360636	3	Coffee Shop	Park	Breakfast Spot	Distribution Center	Chocolate Shop	Pub	Restaurant	Performing Arts Venue	Bakery	
	4	Downtown Toronto	Ryerson, Garden District	43.657162	-79.378937	0	Café	Burger Joint	Sandwich Place	Electronics Store	Diner	Comic Shop	Middle Eastern Restaurant	Clothing Store	Music Venue	В
	4															-

## Manhattan, New York

- For Manhattan, we used a saved data file which is already explored through foursquare API in which we have extracted all the boroughs of New York and then sorted against the concerned borough.
- Then we explored the Manhattan neighborhoods as venues and venue categories

```
Access the data

[81]: |wget -q -0 'newyork_data.json' https://cf-courses-data.s3.us.cloud-object-storage.appdomain.cloud/IBMDeveloperSkillsNetwork-DS0701EN-Sprint('Data downloaded!')

Data downloaded!
```

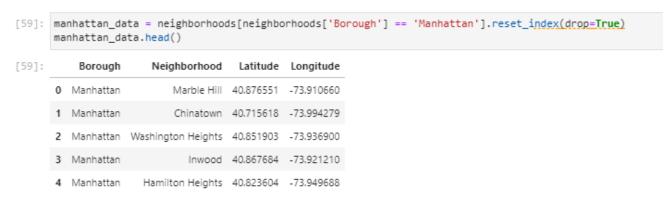
#### examine the resulting dataframe.



#### Make sure that the dataset has all 5 boroughs and 306 neighborhoods

The dataframe has 5 boroughs and 306 neighborhoods.

#### Creating new Dataframe manhattan\_data



## Get the geographical coordinates of Manhattan.

```
[60]: address = 'Manhattan, NY'

geolocator = Nominatim(user_agent="ny_explorer")
location = geolocator.geocode(address)
latitude = location.latitude
longitude = location.longitude
print('The geograpical coordinate of Manhattan are {}, {}.'.format(latitude, longitude))
The geograpical coordinate of Manhattan are 40.7896239, -73.9598939.
```

• The neighborhoods are further characterized as venues and venue categories. Put blue dots on Bucharest map to see centers of neighborhoods.





#### Manhattan venues

Marble Hill Chinatown Washington Heights Inwood Hamilton Heights Manhattanville Central Harlem East Harlem Upper East Side Yorkville Lenox Hill Roosevelt Island Upper West Side Lincoln Square Clinton Midtown Murray Hill Chelsea Greenwich Village

East Village Lower East Side Tribeca Little Italy Soho West Village Manhattan Valley Morningside Heights Gramercy Battery Park City Financial District Carnegie Hill Noho Civic Center Midtown South Sutton Place Turtle Bay Tudor City Stuyvesant Town Flatiron Hudson Yards

```
[66]: # How many unique categories can be curated from all the returned venues
print('There are {} uniques categories.'.format(len(manhattan_venues['Venue Category'].unique())))
```

There are 204 uniques categories.

• The top 10 venues for each neighborhood were obtained.

[71]:		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	8th Most Common Venue	9th Most Common Venue	10th Most Common Venue
	0	Battery Park City	Park	Food Court	Memorial Site	Sandwich Place	BBQ Joint	Gym	Coffee Shop	Cooking School	Shopping Mall	Performing Arts Venue
	1	Carnegie Hill	Italian Restaurant	Gym	Bookstore	Bagel Shop	Dance Studio	Community Center	Coffee Shop	French Restaurant	Café	Pizza Place
	2	Central Harlem	French Restaurant	Jazz Club	Beer Bar	Dessert Shop	Cycle Studio	Ethiopian Restaurant	Music Venue	Cocktail Bar	Spa	Gym / Fitness Center
	3	Chelsea	Cupcake Shop	Theater	Indian Restaurant	Hotel	French Restaurant	Fish Market	New American Restaurant	Nightclub	Coffee Shop	Chinese Restaurant
	4	Chinatown	Chinese Restaurant	Sandwich Place	Spa	Bakery	New American Restaurant	Furniture / Home Store	Cocktail Bar	Noodle House	Greek Restaurant	Bike Shop
	5	Civic Center	Gym / Fitness Center	Falafel Restaurant	French Restaurant	Spa	Yoga Studio	Dance Studio	Park	Monument / Landmark	Molecular Gastronomy Restaurant	Sushi Restaurant

# Clustering Neighborhoods – Manhattan, New York

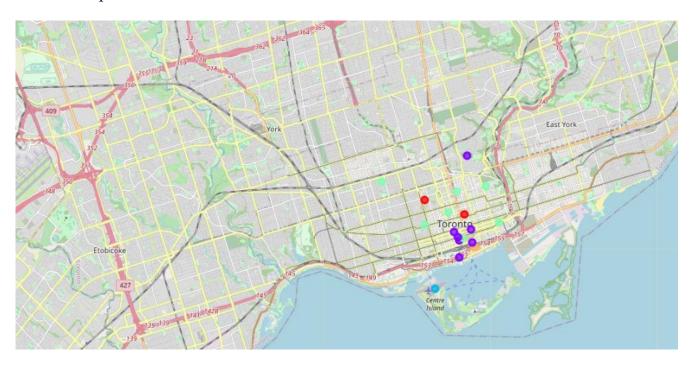
• A new data frame was created including the cluster labels as well as the top 10 venues for each neighborhood.

3]:		Borough	Neighborhood	Latitude	Longitude	Cluster Labels	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	
	0	Manhattan	Marble Hill	40.876551	-73.910660	4	Gym	Sandwich Place	Yoga Studio	Discount Store	Pizza Place	Steakhouse	Supplement Shop	Seafood Restaurant	Te Stac
	1	Manhattan	Chinatown	40.715618	-73.994279	3	Chinese Restaurant	Sandwich Place	Spa	Bakery	New American Restaurant	Furniture / Home Store	Cocktail Bar	Noodle House	G Restau
	2	Manhattan	Washington Heights	40.851903	-73.936900	3	Café	Pizza Place	Wine Shop	Park	Market	Deli / Bodega	Pet Café	Ramen Restaurant	Restau
	3	Manhattan	Inwood	40.867684	-73.921210	3	Deli / Bodega	Bakery	Wine Bar	Park	Diner	Pet Store	Farmers Market	Frozen Yogurt Shop	
	4	Manhattan	Hamilton Heights	40.823604	-73.949688	1	Yoga	Cocktail	Caribbean	Mexican	Bakery	Italian Restaurant	Coffee	Smoke	

# **Results**

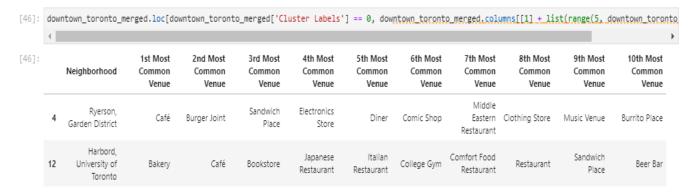
## **Downtown Toronto**

## Cluster Map



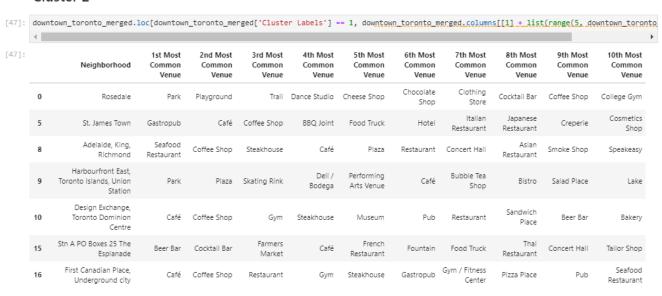
Now, we can examine each cluster and determine the discriminating venue categories that distinguish each cluster. Based on the defining categories, we can then assign a name to each cluster.

# Cluster 1 (Restaurant, Café, Bar, Gym)



# Cluster 2 (Park, Shops, Playground, Gym)

#### Cluster 2



## **Cluster 3 (Airport, Harbor)**



# Cluster 4 (Restaurant, Spa, Pub, Museum, Night Club, Park)

#### Cluster 4



## Cluster 5 (Basketball Stadium, Market, Bar, Seafood)

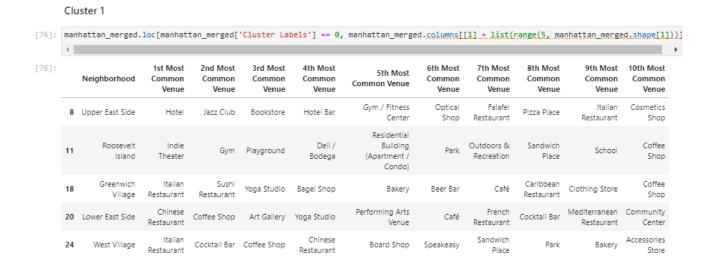


## **Manhattan New York**

## Cluster map



# Cluster 1 (Restaurant, Café, Theater, Gym, Bar, Coffee Shop)



# Cluster 2 (Restaurant, Seafood, Bar, Gym, Spa, Park, Museum)

+																			
	Neighborh	ood	1st N Comr Ve		2nd M Comn Ver	non Con	Most nmon enue	4th Me Comm Ven	ion (	5th Most Common Venue	Con	Most nmon /enue	7th M Comi		Con	Most nmon /enue	Con	Most nmon /enue	10th Mo Commo Ven
4	Ham Hei	ilton ights	Yoga St	udio	Cocktail	Bar	bean urant	Mexic Restaur		Bakery		talian urant	Coffee S	Shop	Smoke	Shop		Café	Pizza Pla
5	Manhattar	nville	Coffee S	hop	Ita Restaur	lian rant	Bar	Climb	ing iym	Café	Gastr	opub	Mex Restau	cican urant	Superm	narket		afood surant	Bike Tr
6	Central Ha	rlem	Fre Restau	ench Irant	Jazz C	Club Be	er Bar	Dess Sh	sert Cyc	le Studio		opian urant	Music Ve	enue	Cockta	ail Bar		Spa	Gyn Fitne Cent
7	East Ha	rlem	Mex Restau		Restaur	rant	Latin erican urant	Food Tru	uck Ste	eakhouse		New erican urant	Taco F	Place		Gym	Ве	er Bar	Sandwi Pla
16	Murray	/ Hill	Japar Restau		Burger Jo	oint Coffee	Shop	Muse	um Re	Sushi estaurant		Gym	Tea R	oom	Sandwich	Place		waiian aurant	Е
22	Little	Italy	Wine	Bar	Ice Cre SI		dwich Place	С	afé Re	Asian estaurant	Go	urmet Shop	Sa Barbers	lon / shop	Salad	Place	Resta	Thai aurant	Optio Sh
23	5	Soho	Bout	ique	Salo Barbersi	Shoe	Store	Cupca Sh	ake nop	Bakery		Dance itudio		hing Store	Music	Store	Miscella	neous Shop	Desse Sh
Λ	Morningside Heights	Coffe	ee Shop	Вос	okstore	American Restaurant		Park	P	'ub Re	Mexican	Res	Greek taurant		Outdoor Sculpture	:	Sandwich Place	Salad	Place
ı	Battery Park City		Park	Food	d Court	Memorial Site	Sa	ndwich Place	BBQ Jo	int	Gym	Coffe	e Shop	Cooki	ng School	Shop	ping Mall	Perfor Arts V	-
	Financial District	Coffe	ee Shop	Jewelr	y Store	Gym / Fitness Center	Event	t Space	Monumen Landma	. 5a	ad Place	Stea	khouse		American Restaurant		terranean estaurant		Gym
	Noho	Win	ne Shop		French taurant	Coffee Shop	Cock	tail Bar	Itali Restaura		Boutique	G	ourmet Shop		ern / Soul Food Restaurant		Sandwich Place	Hima Resta	
	Turtle Bay	Tenni	s Court	Du	ity-free Shop	Café	M	luseum	Seafo Restaura	BO)	ing Gym	Cock	tail Bar		ffee Shop	Groo	cery Store	Resta	Sushi

# **Cluster 3 (Restaurant, Night Club)**

	Clu	ster 3										
[78]:	manh	nattan_merged.]	loc[manhatta	n_merged['C	luster Label	s'] == 2, m	anhattan_me	rged. <u>columns</u>	[[1] + list(	range(5, ma	anhattan_merg	ed.shape[1]))]
[78]:		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	8th Most Common Venue	9th Most Common Venue	10th Most Common Venue
	17	Chelsea	Cupcake Shop	Theater	Indian Restaurant	Hotel	French Restaurant	Fish Market	New American Restaurant	Nightclub	Coffee Shop	Chinese Restaurant

## Cluster 4 (Restaurant, Café, Bar, Gym, Spa)

Restaurant

Home Store

Cluster 4 [79]: manhattan\_merged.loc[manhattan\_merged['Cluster Labels'] == 3, manhattan\_merged.columns[[1] + list(range(5, manhattan\_merged.shape[1]))] [79]: 3rd Most 6th Most 7th Most 1st Most 2nd Most 4th Most 5th Most 8th Most 9th Most 10th Most Neighborhood Common Venue Chinese Greek New American Furniture / Sandwich Chinatown Spa Cocktail Bar Noodle House Bike Shop Restaurant Restaurant Place Restaurant Home Store Deli / Washington Ramen 2 Café Pizza Place Wine Shop Park Market Pet Café Restaurant Coffee Shop Restaurant Heights Bodega Deli / Farmers Frozen Yogurt Mexican 3 Inwood Bakery Wine Bar Park Diner Pet Store Café Bodega Market Restaurant Shop Middle College Thai Health Food Gourmet Taco Place Liquor Store Cocktail Bar 10 Lenox Hill Eastern Gym Burger Joint Academic Restaurant Store Shop Restaurant Building Southern / Soul Upper West Italian American Cupcake Movie Seafood Mediterranean Greek 12 Tiki Bar Food Bookstore Side Restaurant Restaurant Shop Theater Restaurant Restaurant Restaurant Restaurant Indie Movie Performing College Arts Opera 13 Lincoln Square Theater Concert Hall Plaza Library Gift Shop Circus Theater Arts Venue Building House Vietnamese Korean Dessert Moroccan 19 East Village Pet Café Coffee Shop Beer Store Beer Bar Bar Dog Run Restaurant Restaurant Shop Restaurant Italian American Sushi New American Cycle 21 Tribeca Wine Shop Park Poke Place Hotel Playground Restaurant Restaurant Restaurant Restaurant Studio Thrift / Dim Sum 27 Gramercv Pizza Place Coffee Shop Vintage Playground Irish Pub Cocktail Bar Restaurant Shop Store Gym / Grocery Greek 34 Sutton Place Beer Garden Yoga Studio Bakery Food Court Coffee Shop Steakhouse Fitness Restaurant Store Center Seafood Spanish 36 Tudor City Park Yoga Studio Salad Place Café Boxing Gym Restaurant Restaurant Restaurant Bodega Gym / Japanese Furniture / Miscellaneous 38 Flatiron Yoga Studio Women's Store Fitness Donut Shop Salad Place Gym

Restaurant

Shop

Center

# Cluster 5 (Market, Park, Gym, Coffee)

]:	manh	attan_mer	ged.	loc[manha	attan_mer@	ged['Cluste	r Label	s'] ==	4, manhattar	_merged. <u>co</u>	lumns[[1] +	list(range(5	, manhattan_mer	ged.shape[1
]:		Neighborho	bod	1st Mos Commo	n Comn	non Com		4th Most Common Venue	Common	6th Most Common Venue	7th Most Common Venue	8th Most Common Venue	9th Most Common Venue	10th Most Common Venue
	0	Marble	Hill	Gyn	n Sandv	vich Yoga St ace	udio	Discount Store	Pizza Place	Steakhouse	Supplement Shop	Seafood Restaurant	Tennis Stadium	Donut Shop
	9	York	ville	Deli Bodeg	Wine S	nop It Restau	alian Ho ırant	bby Shop	Diner	Dog Run	Park	Coffee Shop	Café	Gyr
	14	Clin	iton	Gym Fitnes Cente	s The	ater Sport	s Bar	Park	Sporting Goods Shop	Building	Cocktail Bar	Mediterranean Restaurant	Gym	Comed Clui
	15	Midto	own	Hote	el Cu Restaur		cery	Plaza	Park	Salad Place	Coffee Shop	Clothing Store	Smoke Shop	Frenc Restauran
	25	Manhat Va	ttan illey	Mexica Restauran		Bar Yoga St	udio	Cosmetics Shop		Pizza Place	Coffee Shop	Grocery Store	Hawaiian Restaurant	Bike Sho
	30	Carnegie	Hill	Italia: Restauran		iym Books	tore B	agel Shop	Dance Studio	Community Center	Coffee Shop	French Restaurant	Café	Pizza Plac
	Civ	vic Center		Gym / Fitness Center	Falafel Restaurant	French Restaurant		Spa \	'oga Studio	Dance Studio	Park	Monument / Landmark	Molecular Gastronomy Restaurant	Sush Restauran
	Midto	wn South		Korean taurant	Dessert Shop	Coffee Shop	Ch	Fried nicken Joint	Scenic Lookout	Grocery Store	Hotel	Wine Bar	Cosmetics Shop	Lingeri Stor
	Si	tuyvesant Town	E	Boat or Ferry	Park	Bar	Coffee	Shop	Gym / Fitness Center	Heliport	Harbor / Marina	Baseball Field	Farmers Market	Bistr
	Hud	son Yards		nerican taurant	Hotel	Gym / Fitness Center		Park	Public Art	Theater	Roof Deck	Pedestrian Plaza	Residential Building (Apartment / Condo)	Supermarke

# **Discussion**

- After clustering the data of the respective neighborhoods, in Downtown and Manhattan, It can be seen that there are venues which can attract the Tourists.
- The neighborhoods are much similar in features like Theaters, Gyms, Parks, food places, Clubs, museums, Spas etc.
- As far as concern to dissimilarity, it differs in terms of some unique places like historical places, Airports and Harbors.
- When comparing the tourist places, it can be observed that the historical place is only situated in Downtown Toronto and the Monument or landmark venue is in Manhattan neighborhoods.
- Similarly, Airport facility, Harbor, Sculpture garden and Boat or ferry services are also available in Downtown Toronto while venues like Nightlife, climbing gym and Museums are present in Manhattan.
- As far as concern to recommendations, we recommend Downtown Toronto Neighborhoods will be considered first to visit. The tourists have an easily travelling access due to Airport facility, which not only saves time but also helps to save money. This saved money can be utilized to explore more, the attracting venues.

# Conclusion

- The downtown Toronto and Manhattan neighborhoods have more like similar venues.
- As we know that every place is unique in its own way, so that's argument is present in both neighborhoods.
- The dissimilarity exists in terms of some different venues and facilities but not on a larger extent.