The Battle of Neighborhoods

Venues comparision – financial hubs of Canada and New York Boroughs

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Coursera – Assignment – Capstone Project Week 4

Author Note

This is for the educational purpose and needing to submit to Coursera Data Science Certification Program.

Analyzing Venues of two boroughs of world financial hubs – Toronto Downtown and Manhattan for Immigrants, student and businessmen

Project Description:

Both the boroughs of Toronto and New York, Toronto Downtown and Manhattan respectively are very big financial hubs of their countries and also facinating to people from different countries outsides.

Students, immigrants, Businessmen from different countries have been targetting these regularly. This case study and Report is to provide the readers required information around the culture and venues which might help them prepare for their decision to persue career or start of a new life in. Having known to venue and traffic information will help make an informed decision in picking right place to travel, stay and begin a new life.

Foursquare API:

This project will use Four-square API as its prime data gathering source as it is one of the rich collection of location info provide for many fortune 500 companies world wide. it has a database of millions of places, especially their places API which provides the ability to perform location search, location sharing and details about a business.

Work Flows:

API data is used to analyze and cluster the venue info from Downtown Toronto and Manhattan broughs. Bascially, this includes -

- Data acquisition and cleansing
- Data preperation
- Feature selection and verification thru visualization
- Clustering

Data acquisition and wrangling -

For Downtown Toronto case, we have extracted table of Toronto's Borough from Wikipedia page. Then we arrange the data according to our requirements. In the arrangement phase, which applied multiple steps including but not limited to, eliminating "Not assigned" values, combine neighborhoods which have same geographical coordinates at each borough and sorted against the concerned borough. For data verification and further exploration, we use Foursquare API to get the coordinates of Downtown Toronto and explore its neighborhoods. The neighborhoods are further characterized as venues and venue categories.

We pulled Downtown Toronto data from the wiki page

- https://en.wikipedia.org/wiki/List_of_postal_codes_of_Canada:_M. And then added long and lattitude using Geospatial_Coordinates.csv. Data wrangling to - eliminate 'not assigned' and merging neighborhoods having same geo space and finally sort them according to their boroughs. used geocoders for long and lat computation for Downtown Toronto. Foursquare API data used for venue location data.

We pulled Manhattan detail using https://cocl.us/new_york_dataset which has all boroughs and then filtered for only Manhattan.

Data preperation-

Taken the venue information of 2 boroughs cities, explore the data, analyze them and visualize them in a seperate section. Then a final conclusion of data analysis in last sections.

Downtown Toronto data is pulled from wesite and has postal code, neighbourhood info. The Geospace info has been fed from the file. And then a data frame has been created for further processing of data. Manhattan data is used from file which is a straight forward radily available info.

Preparation of "Downtown Toronto" data

- Extract data from https://en.wikipedia.org/wiki/List of postal codes of Canada: M
- Add location data from Geospatial_Coordinates.csv
- Cleansing -
- Eliminate neighborhood column (Nan). Throughout the project, Neighbourhood used for clarity
- Drop the postal code column
- Filter for Downtown Toronto borough and set it as index

Preparation of "Manhattan" data

• Extract data from https://cocl.us/new_york_dataset

Geopy is used for location lattitude and longititude calculations for both Downtown toronto and Manhattan.

Feature selection and verification thru visualization –

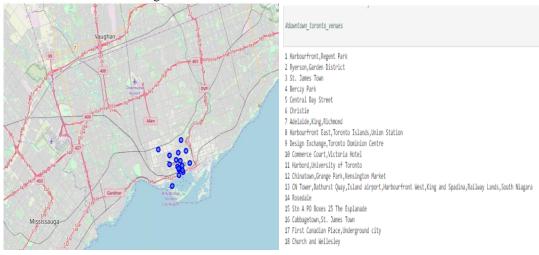
	Borough	Neighbourhood	Latitude	Longitude
0	Downtown Toronto	Harbourfront,Regent Park	43.654260	-79.360636
1	Downtown Toronto	Ryerson, Garden District	43.657162	-79.378937
2	Downtown Toronto	St. James Town	43.651494	-79.375418
3	Downtown Toronto	Berczy Park	43.644771	-79.373306
4	Downtown Toronto	Central Bay Street	43.657952	-79.387383

	Borough	Neighbourhood	Latitude	Longitude
0	Manhattan	Marble Hill	40.876551	-73.910660
1	Manhattan	Chinatown	40.715618	-73.994279
2	Manhattan	Washington Heights	40.851903	-73.936900
3	Manhattan	Inwood	40.867684	-73.921210
4	Manhattan	Hamilton Heights	40.823604	-73.949688

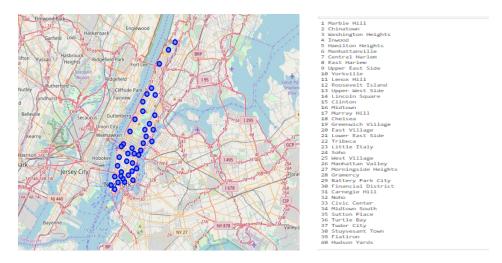
Visualization -

Before clusterting to validate the data on graph for its appearance

Downtown Toronto - Borough



Manhattan-Borough



After clustering using KMeans to cluster and segment the neoghbourhoods

Downtown Toronto: cluster and segmentation

Cluster 1 - Airport Lounge, Coffee shops, Cafe, Restaurants & Grocery Store

downtown_toronto_nerged.shape[1]))]]

Neighbourhood 1st Most Common Venue 2nd Most Common Venue 3nd Most Common Venue 4th Most Common Venue 5th Most Common Venue 6th Most Common Venue 8th Most Common Venue 9th Most Common Venue 9th Most Common Venue 9th Most Common Venue 9th Most Common Venue 1th Most Common Venue 9th Most Common Venue 9th Most Common Venue 1th Most Common Venue 9th Most Common Venue 1th Most Common Venue 1th Most Common Venue 9th Most Common Venue 1th Most Commo

Cluster 2 - Gastropubs

downtown_toronto_merged.loc[downtown_toronto_merged['Cluster Labels'] == 1, downtown_toronto_merged.columns[[1] + list(range(5, downtown_toronto_merged.shape[1]))]]

	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	10th Most Common Venue
8 [esign Exchange, Toronto Dominion Centre	Café	Caffee Shap	Deli / Bodega	Restaurant	Bakery	Gym	Gym / Fitness Center	Hotel	Japanese Restaurant	Gastropub
9	Commerce Court, Victoria Hotel	Café	Gastropub	Coffee Shop	Deli / Bodega	Bakery	Gym	Gym / Fitness Center	Beer Bar	American Restaurant	Museum
10	Harbord, University of Toronto	Bookstore	Bakery	Japanese Restaurant	Restaurant	Dessert Shop	Sandwich Place	Beer Bar	Chinese Restaurant	Italian Restaurant	Comfort Food Restaurant
16	First Canadian Place, Underground city	Café	Coffee Shap	Restaurant	Steakhouse	Gastropub	Pub	Deli / Bodega	Bakery	Pizza Place	Seafood Restaurant

Cluster 3 - Cafes

downtown toronto merged.loc[downtown toronto merged['Cluster Labels'] == 2, downtown toronto merged.columns[[1] + list(range[5, downtown toronto merged.shape[1]))]]

	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	10th Most Common Venue
4	Central Bay Street	Coffee Shop	Italian Restaurant	Bubble Tea Shop	Park	Japanese Restaurant	Seafood Restaurant	Spa	Ramen Restaurant	Sushi Restaurant	Sandwich Place
13	Rosedale	Park	Trail	Playground	Building	Wine Bar	Convenience Store	Chinese Restaurant	Chocolate Shop	Church	Clothing Store

Cluster 4 - Coffee Shops, Cafe, Park & Japanese Restaurants

downtown_toronto_merged.loc[downtown_toronto_merged['Cluster Labels'] == 3, downtown_toronto_merged.columns[[1] + list(range(5, downtown_toronto_merged.shape[1])))]]

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	10th Most Common Venue
14 Stn A PO Boxes 25 The Esplanade	Cocktail Bar	Beer Bar	Farmers Market	French Restaurant	Steakhouse	Park	Museum	Clothing Store	Jazz Club	Concert Hall

Cluster 5 - Seafood, steakhouse, Hotel & Cafe

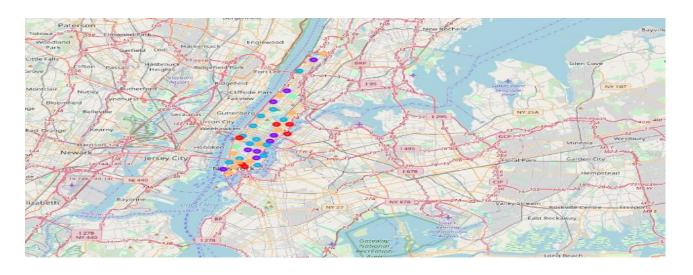
downtown_toronto_merged.loc[downtown_toronto_merged['Cluster Labels'] == 4, downtown_toronto_merged.columns[[1] + list(range(5, downtown_toronto_merged.shape[1]))]

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	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	10th Most Common Venue
3	Berczy Park	Seafood Restaurant	Farmers Market	Steakhouse	Museum	Park	Cocktail Bar	Liquor Store	Caffee Shap	Bistro	Beer Bar
5	Christie	Grocery Store	Café	Park	Coffee Shop	Baby Store	Italian Restaurant	Athletics & Sports	Restaurant	Diner	Nightdub
6	Adelaide,King,Richmond	Asian Restaurant	Coffee Shop	Steakhouse	Seafood Restaurant	Lounge	Pizza Place	Plaza	Concert Hall	Bar	Hotel
7 Harbou	urfront East, Toronto Islands, Union Station	Park	Plaza	Café	Sporting Goods Shop	New American Restaurant	Performing Arts Venue	Bubble Tea Shop	Lake	Salad Place	Italian Restaurant
11 Ch	hinatown, Grange Park, Kensington Market	Café	Caribbean Restaurant	Vietnamese Restaurant	Mexican Restaurant	Bakery	Farmers Market	Dessert Shop	Coffee Shop	Cocktail Bar	Organic Grocery
12 CN Towe	er,Bathurst Quay,Island airport,Harbourf	Airport Lounge	Airport Service	Airport	Airport Food Court	Airport Gate	Airport Terminal	Boat or Ferry	Caffee Shap	Sculpture Garden	Harbor / Marina
15	Cabbagetown, St. James Town	Café	Restaurant	Park	Bakery	Caribbean Restaurant	Butcher	Pet Store	Pub	Jewelry Store	Italian Restaurant



Manhaatan: cluster and segmentation





Observations & Recommendations

Historical places are situated in Downtown Toronto neighborhoods and monument venues are in Manhattan neighborhoods. Venues like Airport lounges, Harbor, Gym, museums and night clubs are available for tourists in Downtown Toronto.

At first sight we can primarily recommend, Downtown Toronto neighborhoods as they have easy access to Airports to save time and save some money. Easy access to traveling always leads to convenience for tourists, students and emigrants.

Conclusion

The downtown Toronto and Manhattan neighborhoods have more like similar venues. Most of the times choices and needs also drive the tourists so Asian might more incline towards the Toronto while European may towards Manhattan due to some cultural background and immigration benefits.