Determining the optimum location to set up a gym in Manhattan

IBM Capstone Project

Business Problem

The problem I have chosen to analyse involves the decision regarding where to open a gym in Manhattan, New York City, USA. There are several factors which will play a role in determining this optimal location. The gym needs to be centrally located and needs to be easily accessible from all parts of the city.

The gym should also have complementary venues near it. For example, the presence of a spa nearby will encourage customers to purchase gym membership of said gym. The gym should also not exist in a place where several gyms already exist which would generate unnecessary competition. Clusters will be generated to assess each neighbourhood.

I will attempt to generate a list of neighbourhoods most suitable which will result in maximum revenue for the stakeholder wishing to set up the gym.

DATA

Based on the factors I discussed above, the following data will be required:

- The distance of each neighbourhood from the centre of the city.
- The venues nearby which are complementary to the gym.
- The number of gyms that exist in each neighbourhood.

The neighbourhoods have been identified from the source: https://cocl.us/new_york_dataset. All the features will be extracted from this data source. The other data will be extracted using the Foursquare API. The location of each neighbourhood will be obtained from the GeoPy GeoCoder package in Python.

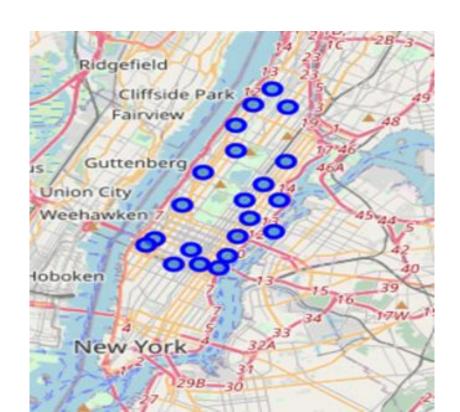
Data frame

[10]:		Borough	Neighborhood	Latitude	Longitude
	0	Bronx	Wakefield	40.894705	-73.847201
	1	Bronx	Co-op City	40.874294	-73.829939
	2	Bronx	Eastchester	40.887556	-73.827806
	3	Bronx	Fieldston	40.895437	-73.905643
	4	Bronx	Riverdale	40.890834	-73.912585

	301	Manhattan	Hudson Yards	40.756658	-74.000111
	302	Queens	Hammels	40.587338	-73.805530
	303	Queens	Bayswater	40.611322	-73.765968
	304	Queens	Queensbridge	40.756091	-73.945631
	305	Staten Island	Fox Hills	40.617311	-74.081740

306 rows × 4 columns

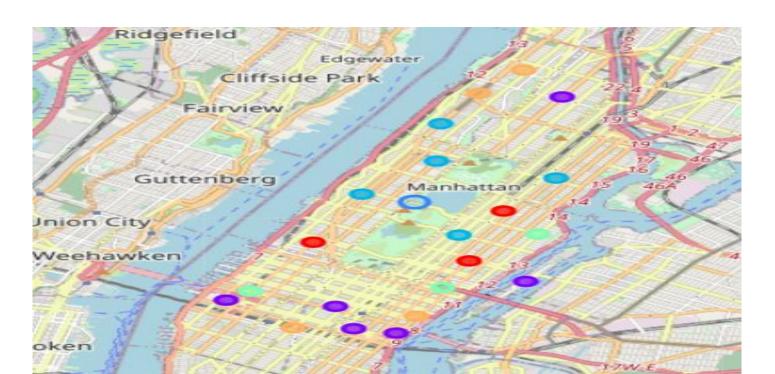
Map of Manhattan containing the relevant neighborhoods



Analysis using FOURSQUARE API

5th Most Common Venue	4th Most Common Venue	3rd Most Common Venue	2nd Most Common Venue	1st Most Common Venue	Neighborhood	
Gym	Gym / Fitness Center	Bakery	Yoga Studio	Coffee Shop	Carnegie Hill	0
French Restaurant	Bar	Seafood Restaurant	African Restaurant	Cosmetics Shop	Central Harlem	1
Gym	Wine Shop	Gym / Fitness Center	Coffee Shop	Theater	Clinton	2
Deli / Bodega	Latin American Restaurant	Thai Restaurant	Bakery	Mexican Restaurant	East Harlem	3
Mexican Restaurant	Deli / Bodega	Coffee Shop	Café	Pizza Place	Hamilton Heights	4
Italian Restaurant	Café	Gym / Fitness Center	American Restaurant	Hatel	Hudson Yards	5
Cocktail Ban	Café	Coffee Shop	Pizza Place	Italian Restaurant	Lenox Hill	6
Performing Arts Venue	Gym / Fitness Center	Concert Hall	Café	Plaza	Lincoln Square	7
Mexican Restaurant	Coffee Shop	Pizza Place	Bar	Yoga Studio	Manhattan Valley	8
Mexican Restaurant	Park	Italian Restaurant	Coffee Shop	Seafood Restaurant	Manhattanville	9
Spa	Theater	Clothing Store	Hotel	Coffee Shop	Midtown	10
Dessert Shop	Burger Joint	Japanese Restaurant	Hotel	Korean Restaurant	Midtown South	11
Burger Joint	Bookstore	Coffee Shop	American Restaurant	Park	Morningside Heights	12
Gym / Fitness Center	Pizza Place	Hotel	Coffee Shop	Sandwich Place	Murray Hill	13
Farmers Market	Soccer Field	Japanese Restaurant	Plaza	Park	Roosevelt Island	14
Gym	Coffee Shop	Park	Gym / Fitness Center	Italian Restaurant	Sutton Place	15
Pizza Place	Deli / Bodega	Mexican Restaurant	Park	Café	Tudor City	16
Wine Bar	Deli / Bodega	Park	Coffee Shop	Italian Restaurant	Turtle Bay	17
Exhibit	Gym / Fitness Center	Bakery	Juice Bar	Italian Restaurant	Upper East Side	18
Mediterranean Restaurant	Coffee Shop	Wine Bar	Bakery	Italian Restaurant	Upper West Side	19
Bar	Deli / Bodega	Gym	Italian Restaurant	Coffee Shop	Yorkville	20

KNN CLUSTERS ON THE MAP OF MANHATTAN



INFORMATION ABOUT THE CLUSTERS

Cluster 1- It represents the neighbourhoods that are at a greater distance from the centre of Manhattan. Each neighbourhood as at least 1 gym. I would not recommend setting up a gym here as it is far from Central Park. These neighbourhoods are marked in **red**.

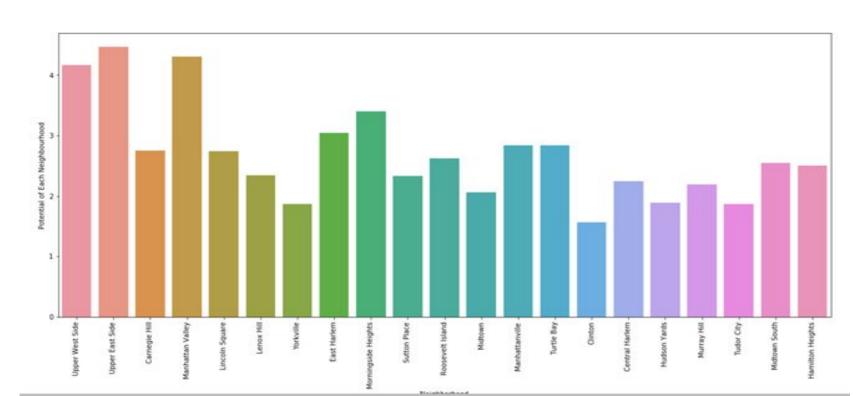
Cluster 2 - These are neighbourhoods with plenty of gyms already. There will be fierce competition here and that may impact revenue. These points are marked in **purple**.

Cluster 3 - These are the points closest to Central Park and are greatly untapped. These neighbourhoods offer great potential. I would recommend these neighbourhoods. However, one must keep in mind that these neighbourhoods are expensive and there will be significant overhead costs to setting up a gym here. These points are marked in light blue.

Cluster 4 - These are points that are slightly further away from the centre. However, no gym exists in the neighbourhood of these vicinities so it may result in high revenue. The cost of setting up the gym might also be much lower than cluster 2. They are marked in light green.

Cluster 5- This cluster shows points that are somewhat between the outskirts and centre of Manhattan. There exist gyms already which may offer some competition. However, that number is far lesser than that of cluster 1. They can also be promising as prices will be lower. These are shown in orange.

BAR GRAPH OF POTENTIAL VS NEIGHBOURHOOD



CONCLUSION

The purpose of this project was to identify a suitable location for stakeholders looking to set up a gym. Through the application of clustering and Foursquare API, I have identified the best neighbourhoods for this project. They are

- Manhattan Valley
- Upper East Side
- Manhattanville
- Turtle Bay

Manhattan Valley would require higher capital but would generate the maximum revenue. It is the closest to the centre and has complementary businesses. Upper East Side is a good prospect as well but will be even more expensive. Manhattanville and Turtle Bay are ideal prospects for stakeholders who would not wish to play the excessive charges of location. The competition in this region is higher as most gym owners would prefer this region.