

Battle of the Neighborhoods

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- Introduction/ Business Problem

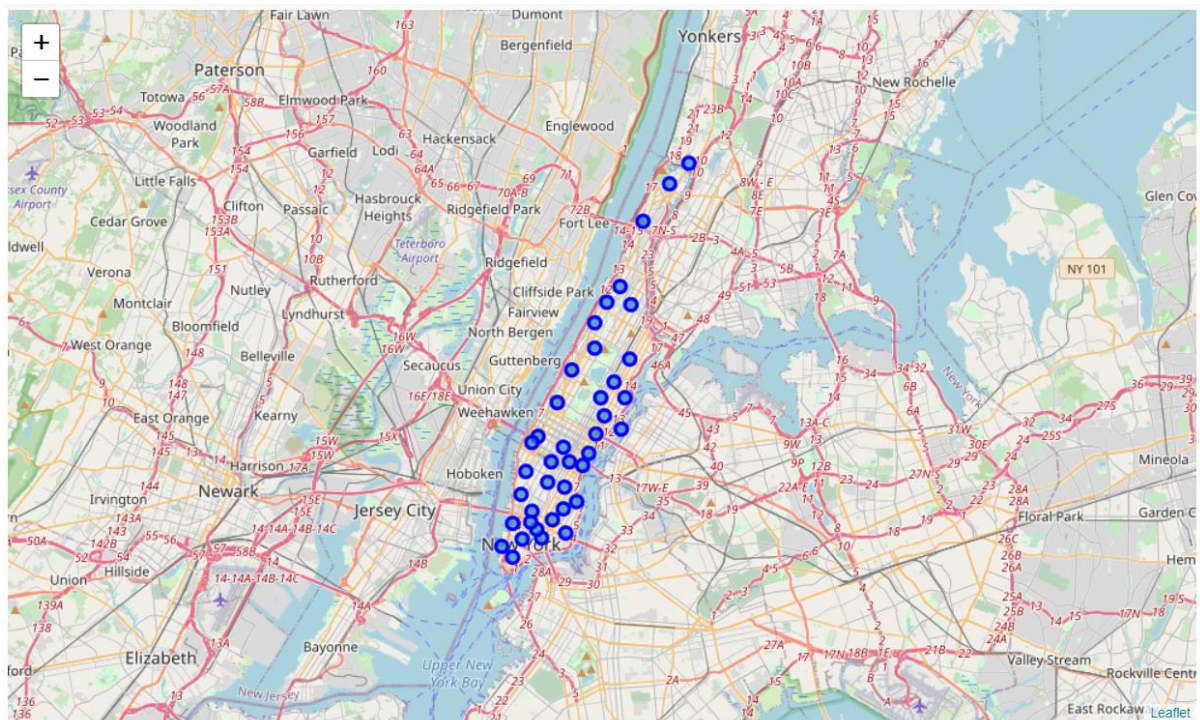
The right location to open a **supplement store** in Manhattan, NYC is crucial to the success of supplement store franchises looking to expand. Location analysis can play a crucial role in determining the right spot for opening up a franchise.

Many sports enthusiasts and gym goers from various ethnicities reside in New York. With a population north of 8 million, this is a prime city for business.

The **right location** to open up a supplement store would be a place that is frequently visited by sports enthusiasts and gym goers. The right spot would be one where the most number of such people are found. Additionally, we'd look for a spot that has no competition so we'll take other supplement stores into account as well.

- Data

Neighborhood data can be found on this JSON file:
(https://geo.nyu.edu/catalog/nyu_2451_34572)



The venues in each neighborhood can be found using Foursquare's API.

We'll use the locations of the following venue categories to clearly define the right location to open a supplement store:

- Athletics & Sports Venues
- Baseball Fields
- Basketball Court
- Boxing Gyms
- Climbing Gyms
- Cycle Studios
- Gyms, Gym/Fitness Centers, Gym Pools
- Soccer Fields
- Sports clubs, Sporting Goods Shops
- Tennis court

All venue categories will be grouped by their neighborhood for further inspection.

Density of existing supplement shop densities have to be taken into consideration to avoid competition which would affect revenue.

	Neighborhood	Supplement Shop	Athletics & Sports	Baseball Field	Basketball Court	Boxing Gym	Gym	Cycle Studio	Gym / Fitness Center	Gym Pool	Soccer Field	Sports Club	Sporting Goods Shop	Tennis Court
0	Battery Park City	0.0	0.015873	0.0	0.0	0.0	0.047619	0.000000	0.000000	0.0	0.0	0.00	0.0	0.0
1	Carnegie Hill	0.0	0.000000	0.0	0.0	0.0	0.034091	0.000000	0.034091	0.0	0.0	0.00	0.0	0.0
2	Central Harlem	0.0	0.000000	0.0	0.0	0.0	0.021739	0.021739	0.043478	0.0	0.0	0.00	0.0	0.0
3	Chelsea	0.0	0.000000	0.0	0.0	0.0	0.010000	0.020000	0.000000	0.0	0.0	0.00	0.0	0.0
4	Chinatown	0.0	0.000000	0.0	0.0	0.0	0.000000	0.000000	0.000000	0.0	0.0	0.01	0.0	0.0

• Methodology

We'll analyze each neighborhood and take a **mean** of all sports facilities to quantify the **concentration** of each sports facility in each neighborhood. We have ignored all stadiums and other facilities which are used infrequently throughout the year and are reserved for tournaments.

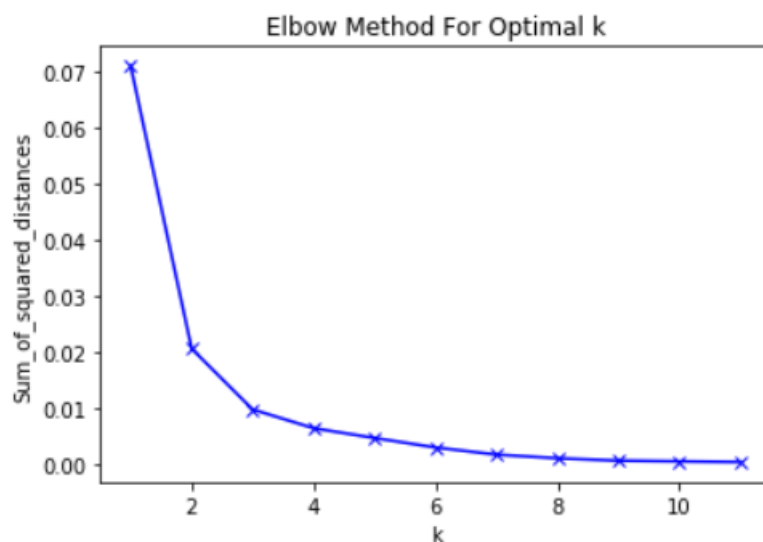
Further, we'll sum all column mean values to get a value on the **total** sports facility concentration in **each** neighborhood. This step will exclude supplement shops.

	Neighborhood	Supplement Shop	Athletics & Sports	Baseball Field	Basketball Court	Boxing Gym	Gym	Cycle Studio	Gym / Fitness Center	Gym Pool	Soccer Field	Sports Club	Sporting Goods Shop	Tennis Court	Sum
	Flatiron	0.000000	0.000000	0.000000	0.000000	0.000000	0.040000	0.020000	0.070000	0.000000	0.000000	0.01	0.020000	0.000000	0.160000
	Hudson Yards	0.000000	0.017544	0.000000	0.000000	0.000000	0.035088	0.017544	0.052632	0.000000	0.000000	0.00	0.000000	0.000000	0.122807
	Marble Hill	0.040000	0.000000	0.000000	0.000000	0.000000	0.080000	0.000000	0.000000	0.000000	0.000000	0.00	0.000000	0.000000	0.120000
	Roosevelt Island	0.000000	0.000000	0.000000	0.000000	0.000000	0.038462	0.000000	0.038462	0.000000	0.038462	0.00	0.000000	0.000000	0.115385
	Stuyvesant Town	0.000000	0.000000	0.055556	0.000000	0.000000	0.000000	0.000000	0.055556	0.000000	0.000000	0.00	0.000000	0.000000	0.111111

Afterwards, we'll use the **KMeans** Machine Learning algorithm to cluster neighborhoods based on the sports facilities and supplement shops present. We'll create a new Dataframe for this purpose. We'll drop all sports venue category columns and retain the neighborhood, sum, and supplement shop columns.

	Neighborhood	Sum	Supplement Shop
0	Battery Park City	0.063492	0.0
1	Carnegie Hill	0.068182	0.0
2	Central Harlem	0.086957	0.0
3	Chelsea	0.030000	0.0
4	Chinatown	0.010000	0.0

The **elbow method** will be deployed to calculate the optimal k value.



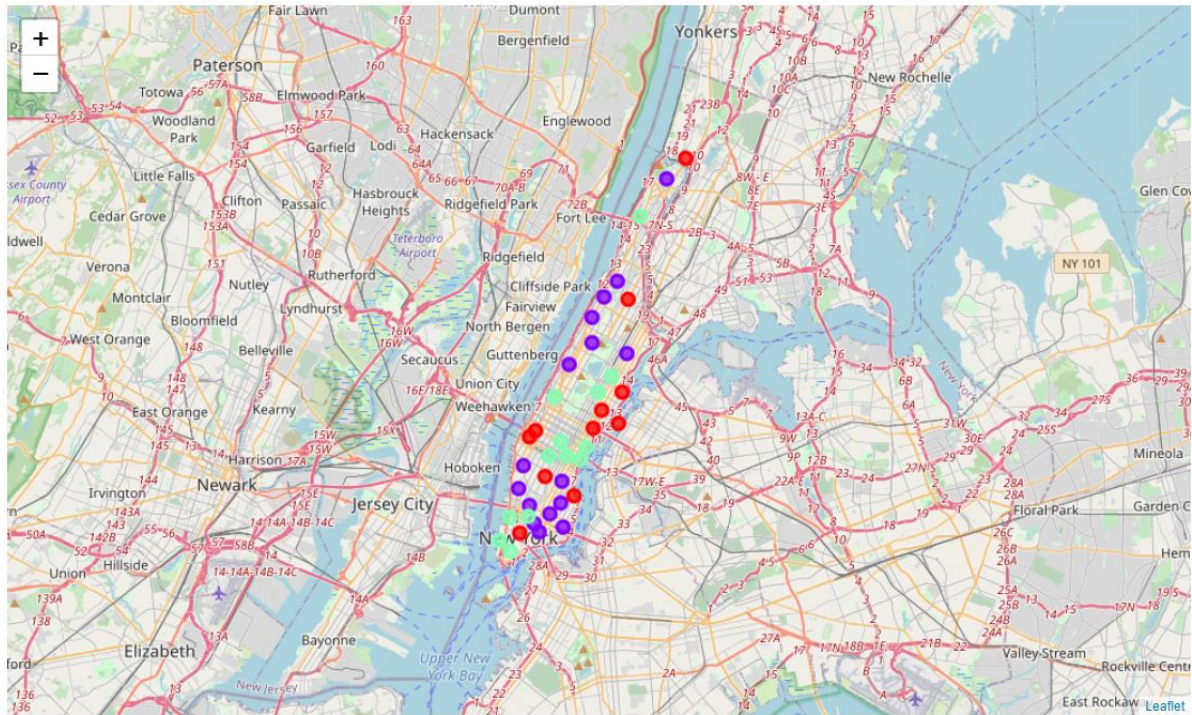
The optimal number of clusters for this dataset is 3.

We'll drop the neighborhood column and train the KMeans algorithm with this dataset and form 3 clusters.

We'll include the cluster labels in a Dataframe that includes the coordinates of each neighborhood.

	Borough	Neighborhood	Latitude	Longitude	Cluster Labels	Sum	Supplement Shop
0	Manhattan	Marble Hill	40.876551	-73.910660	0	0.120000	0.040000
1	Manhattan	Chinatown	40.715618	-73.994279	1	0.010000	0.000000
2	Manhattan	Washington Heights	40.851903	-73.936900	2	0.05814	0.011628
3	Manhattan	Inwood	40.867684	-73.921210	1	0.000000	0.000000
4	Manhattan	Hamilton Heights	40.823604	-73.949688	1	0.000000	0.000000

We'll use this Dataframe to plot a map with the clusters overlayed on top.



• Analysis

Analyzing each cluster reveals that clusters 0 and 2 have a significant portion of sports facilities and closer inspection reveals that cluster 0 has a significant chunk of sports facilities.

Meanwhile, cluster 1 has to be avoided like the plague as it has very few sports facilities; therefore, it's in the best interest of stakeholders to avoid cluster 1 and focus on cluster 0.

Cluster 0

	Neighborhood	Sum	Supplement Shop
0	Marble Hill	0.120000	0.04
6	Central Harlem	0.086957	0.00
9	Yorkville	0.090000	0.00
10	Lenox Hill	0.100000	0.00
11	Roosevelt Island	0.115385	0.00
14	Clinton	0.100000	0.00
32	Civic Center	0.103093	0.00
34	Sutton Place	0.100000	0.00
37	Stuyvesant Town	0.111111	0.00
38	Flatiron	0.160000	0.00
39	Hudson Yards	0.122807	0.00

Cluster 1

	Neighborhood	Sum	Supplement Shop
1	Chinatown	0.010000	0.0
3	Inwood	0.000000	0.0
4	Hamilton Heights	0.000000	0.0
5	Manhattanville	0.000000	0.0
7	East Harlem	0.024390	0.0
12	Upper West Side	0.022727	0.0
17	Chelsea	0.030000	0.0
18	Greenwich Village	0.020000	0.0
19	East Village	0.000000	0.0
20	Lower East Side	0.020000	0.0
22	Little Italy	0.000000	0.0
24	West Village	0.020000	0.0
25	Manhattan Valley	0.023256	0.0
26	Morningside Heights	0.024390	0.0
27	Gramercy	0.000000	0.0
31	Noho	0.020000	0.0

Cluster 2

	Neighborhood	Sum	Supplement Shop
2	Washington Heights	0.058140	0.011628
8	Upper East Side	0.057471	0.000000
13	Lincoln Square	0.084211	0.000000
15	Midtown	0.080000	0.000000
16	Murray Hill	0.063158	0.000000
21	Tribeca	0.064103	0.000000
23	Soho	0.040000	0.000000
28	Battery Park City	0.063492	0.000000
29	Financial District	0.050000	0.000000
30	Carnegie Hill	0.068182	0.000000
33	Midtown South	0.050000	0.000000
35	Turtle Bay	0.040000	0.000000
36	Tudor City	0.066667	0.000000

• Conclusion

Closer analysis of cluster 0 reveals that there is a dearth of supplement shops but ample presence of sports facilities that would attract prospective customers.

	Neighborhood	Sum	Supplement Shop
38	Flatiron	0.160000	0.00
39	Hudson Yards	0.122807	0.00
0	Marble Hill	0.120000	0.04
11	Roosevelt Island	0.115385	0.00
37	Stuyvesant Town	0.111111	0.00
32	Civic Center	0.103093	0.00
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9	Yorkville	0.090000	0.00
6	Central Harlem	0.086957	0.00

Clearly, Flatiron and Hudson Yards have the highest concentrations of sports facilities and surprisingly, no supplement shops. It would be ideal to focus on the neighborhoods on the list above for further inspection.

This capstone project is a part of the IBM Data Science Professional Course.

The purpose of this project is to extract meaningful data about Manhattan, NYC and apply data science techniques to identify the right locations to open up a shop.

Final decision has to be taken by stakeholders to determine which of the locations described above are ideal depending on the vicinity of the surroundings itself.