

**Peer-graded Assignment: Capstone
Project - The Battle of Neighborhoods
(Week 1)**

Dated: July 15, 2021

1. Introduction

1.1 Background:

New York City is not only the most populous city in the United States but also one of the world's largest mega cities with an estimated 2020 population of approximately 8.25 million spread over about 302.6 square miles including 5 boroughs. It is described as the cultural, financial and media capital of the world. It has population from all around the world and exerts significant influence on commerce, entertainment, research, technology, education, politics, art, tourism, sports etc. The pandemic beginning in March 2020 shut down most of the world including New York City. During this time, it has become very evident that some form of exercise is very useful for people not able to go out of the house. People took to doing yoga watching YouTube videos to stay healthy, unable to get to their regular yoga studio. Yoga has become an important part of people's lifestyle during this pandemic to stay healthy or even to recover from illnesses. Given these circumstances, our client would like to open a Yoga studio to teach face to face now that the city is opening back up again. So the basic question that we would like to answer is the best location for our client to open the Yoga studio. Client would like to stay in Manhattan where people are coming back again and offices are also opening up and demand would likely be high. The main criteria would be to find a location with easy access to subway, parking. Location also needs to be close to office buildings so people can step out during

lunch hours for their lessons or their way to and from work. So the question that this final project would like answer is where should our client open their Yoga studio in Manhattan? Main criteria is the location so it's close to transportation as well as office buildings and other entertainment avenues which provide more foot traffic.

1.2 Business Problem

As part of the project, we will try to answer the following questions posed by our client:

1. Where in Manhattan is most optimum location for opening a Yoga studio which has the best chance of succeeding?
2. What location would provide ample traffic in surrounding areas that would also help bring clients to the Yoga studio?

2. Data

The data that will be used to order answer the business questions, is data on New York City neighborhoods, boroughs to include boundaries, latitude, longitude, restaurants, and other yoga studios. From the main data, the borough of Manhattan will be main focus since client wants to stay in Manhattan. New York City data containing the neighborhoods and boroughs, latitudes, and longitudes will be obtained from the data source: https://cocl.us/new_york_dataset All data related to

locations and other entertainment venues and/or other office complexes etc will be obtained via the FourSquare API utilized via the Request library in Python.

3. Methodology

- Data will be collected from https://cocl.us/new_york_dataset and cleaned and processed into a Dataframe.
- Only Manhattan data will be used for the purpose answering the main questions.
- FourSquare app will be utilized via the Request library in Python to locate all venues and then filtered by optimum location by entertainment venues and added to the Dataframe.
- Data will be sorted based on rankings. We will explore the data using the data and ultimately visually assess the data using Python libraries.

4. Analysis

4.1 Install and Import Required Libraries

#Install and Import required libraries

`!pip install beautifulsoup4`

`!pip install lxml`

`import requests # library to handle requests`

```
import pandas as pd # library for data analysis
```

```
import numpy as np # library to handle data in a vectorized manner
```

```
import random # library for random number generation
```

```
!conda install -c conda-forge geopy --yes
```

```
from geopy.geocoders import Nominatim # module to convert an address into  
latitude and longitude values
```

```
# libraries for displaying images
```

```
from IPython.display import Image
```

```
from IPython.core.display import display, HTML
```

```
from IPython.display import display_html
```

```
import pandas as pd
```

```
import numpy as np
```

```
# tranforming json file into a pandas dataframe library
```

```
from pandas.io.json import json_normalize
```

```
!conda install -c conda-forge folium=0.5.0 --yes
```

```
import folium # plotting library
```

```
from bs4 import BeautifulSoup
```

```
from sklearn.cluster import KMeans
```

```
import matplotlib.cm as cm
```

```
import matplotlib.colors as colors
```

```
print('Folium installed')
```

```
print('Libraries imported.')
```

```
Requirement already satisfied: beautifulsoup4 in /home/jupyterlab/conda/envs/python/lib/python3.6/site-packages (4.9.3)
```

```
Requirement already satisfied: soupsieve>1.2; python_version >= "3.0" in /home/jupyterlab/conda/envs/python/lib/python3.6/site-packages (from beautifulsoup4) (2.2.1)
```

```
Requirement already satisfied: lxml in /home/jupyterlab/conda/envs/python/lib/python3.6/site-packages (4.6.3)
```

```
Collecting package metadata (current_repodata.json): done
```

```
Solving environment: done
```

4.2. Data acquisition and cleaning

The data that will be used to order answer the business questions, is data on New York City neighborhoods, boroughs to include boundaries, latitude, longitude, restaurants, and other yoga studios.

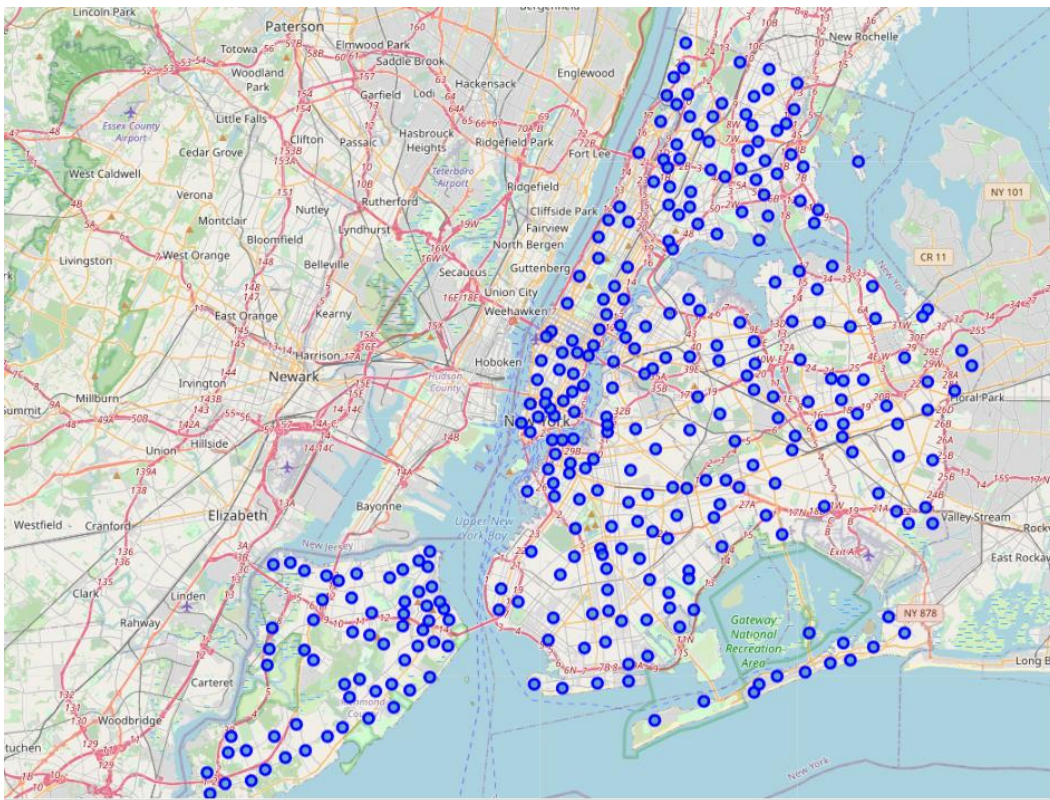
From the main data, the borough of Manhattan will be main focus since client wants to stay in Manhattan.

New York City data containing the neighborhoods and boroughs, latitudes, and longitudes was obtained from the data source: https://cocl.us/new_york_dataset

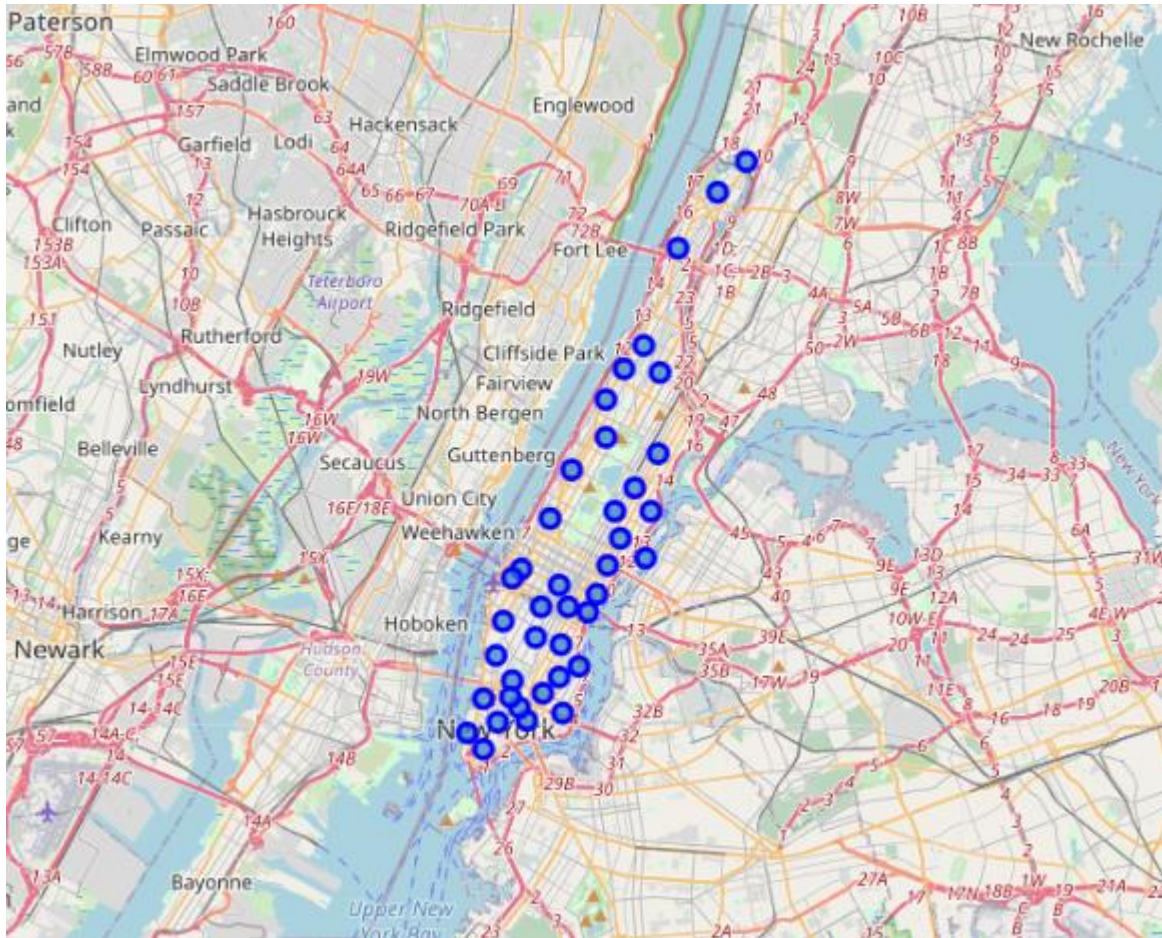
All data related to locations and other entertainment venues and/or other office complexes etc was obtained via the FourSquare API utilized via the Request library in Python.

4.3. Exploratory Data Analysis

Using geolocator and Folium packages of Python, we create a map of New York City showing all the neighborhoods.



Since client wants to open the studio in the borough of Manhattan, we subset the data to only show the borough of Manhattan.



Using FourSquare data we can create the top ten venues for each neighborhood. Using pandas, we create and explore the dataframe.

Data is grouped by their neighborhoods and their means are taken to calculate top neighborhoods for various Venue categories.

As can be seen from figure below:

| | Neighborhood | 1st Most Common Venue | 2nd Most Common Venue | 3rd Most Common Venue | 4th Most Common Venue | 5th Most Common Venue |
|----|--------------------|-----------------------|-----------------------|-------------------------|-----------------------|---------------------------|
| 0 | Battery Park City | Park | Coffee Shop | Hotel | Clothing Store | Women's Store |
| 1 | Carnegie Hill | Coffee Shop | Café | Wine Shop | Yoga Studio | Bar |
| 2 | Central Harlem | African Restaurant | Chinese Restaurant | Public Art | French Restaurant | American Restaurant |
| 3 | Chelsea | Coffee Shop | Bakery | Art Gallery | American Restaurant | Italian Restaurant |
| 4 | Chinatown | Chinese Restaurant | Bakery | Cocktail Bar | Dessert Shop | American Restaurant |
| 5 | Civic Center | Coffee Shop | Spa | Gym / Fitness Center | American Restaurant | French Restaurant |
| 6 | Clinton | Italian Restaurant | Theater | Gym / Fitness Center | Sandwich Place | Coffee Shop |
| 7 | East Harlem | Bakery | Mexican Restaurant | Thai Restaurant | Sandwich Place | Latin American Restaurant |
| 8 | East Village | Bar | Mexican Restaurant | Speakeasy | Pizza Place | Ice Cream Shop |
| 9 | Financial District | Coffee Shop | Pizza Place | Cocktail Bar | Park | Bar |
| 10 | Flatiron | Italian Restaurant | Japanese Restaurant | New American Restaurant | American Restaurant | Mediterranean Restaurant |

Grouping the data of Manhattan and after one hot encoding the top ten neighborhoods with maximum number of Yoga studios, gyms and cafes emerges.

| Neighborhood | Yoga Studio | Gym | Gym / Fitness Center | Gymnastics Gym | Café |
|--------------|------------------|----------|----------------------|----------------|--------------|
| 22 | Marble Hill | 0.043478 | 0.086957 | 0.000000 | 0.0 0.000000 |
| 1 | Carnegie Hill | 0.033708 | 0.022472 | 0.033708 | 0.0 0.067416 |
| 13 | Hamilton Heights | 0.033333 | 0.016667 | 0.000000 | 0.0 0.066667 |
| 35 | Upper East Side | 0.030612 | 0.000000 | 0.040816 | 0.0 0.000000 |
| 5 | Civic Center | 0.030000 | 0.010000 | 0.050000 | 0.0 0.010000 |
| 10 | Flatiron | 0.030000 | 0.010000 | 0.020000 | 0.0 0.020000 |
| 20 | Manhattan Valley | 0.021739 | 0.000000 | 0.000000 | 0.0 0.021739 |
| 19 | Lower East Side | 0.020408 | 0.020408 | 0.000000 | 0.0 0.040816 |
| 15 | Inwood | 0.018182 | 0.000000 | 0.000000 | 0.0 0.054545 |
| 33 | Tudor City | 0.012500 | 0.037500 | 0.012500 | 0.0 0.050000 |

We plot the data to see the top neighborhoods better:

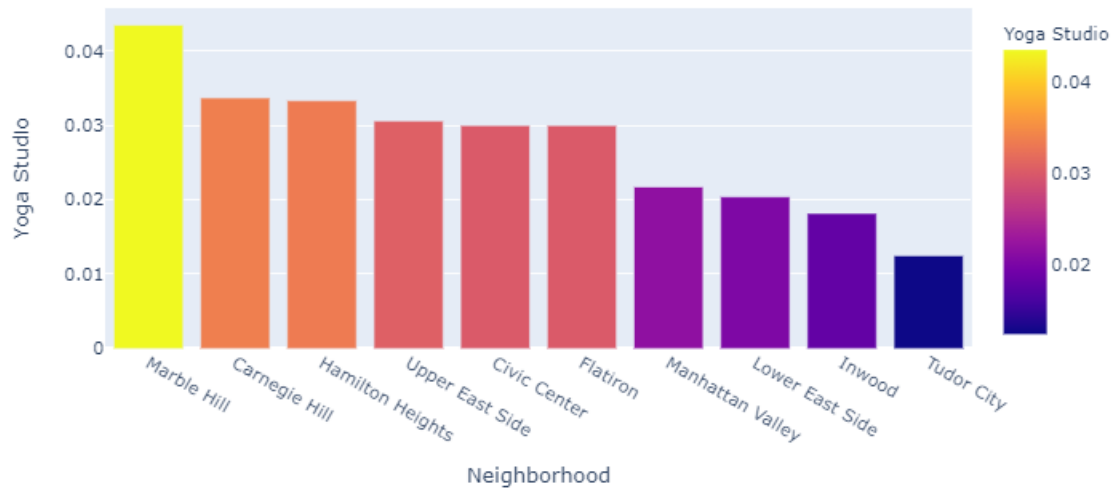


Figure 3. Bar plot showing highest number of Yoga studios in the neighborhoods of Manhattan

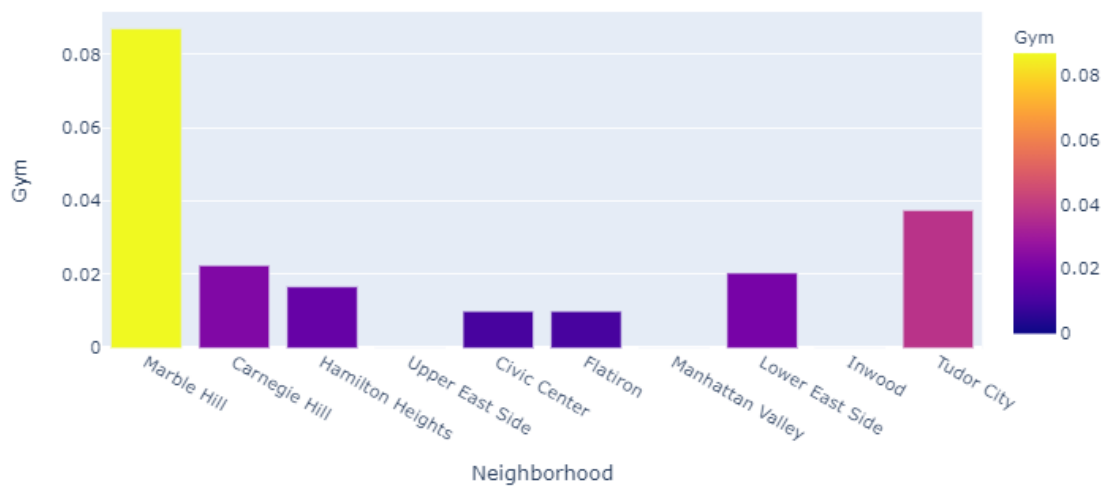


Figure 4. Bar plot showing highest number of Gyms in the neighborhoods of Manhattan

4.4 Cluster and Visualize the Neighborhoods

Using kmeans cluster, the neighborhoods of Manhattan were divided into 5 clusters.

Examine Clusters

Cluster 1

```
manhattan_merged.loc[manhattan_merged['Cluster Labels'] == 0, manhattan_merged.columns[[1] + list(range(5, manhattan_merged.shape[1]))]]
```

| | Neighborhood | 1st Most Common Venue | 2nd Most Common Venue | 3rd Most Common Venue | 4th Most Common Venue | 5th Most Common Venue |
|----|--------------------|-----------------------|-----------------------|---------------------------|-----------------------|---------------------------|
| 2 | Washington Heights | Café | Bakery | Bank | Mobile Phone Shop | Pizza Place |
| 3 | Inwood | Mexican Restaurant | Café | Restaurant | Lounge | Chinese Restaurant |
| 4 | Hamilton Heights | Pizza Place | Café | Mexican Restaurant | Deli / Bodega | Coffee Shop |
| 7 | East Harlem | Bakery | Mexican Restaurant | Thai Restaurant | Sandwich Place | Latin American Restaurant |
| 20 | Lower East Side | Chinese Restaurant | Art Gallery | Latin American Restaurant | Pharmacy | Pizza Place |

Cluster 2

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

| | Neighborhood | 1st Most Common Venue | 2nd Most Common Venue | 3rd Most Common Venue | 4th Most Common Venue | 5th Most Common Venue |
|----|--------------------|-----------------------|-----------------------|-------------------------|--------------------------|--------------------------|
| 1 | Chinatown | Chinese Restaurant | Bakery | Cocktail Bar | Dessert Shop | American Restaurant |
| 6 | Central Harlem | African Restaurant | Chinese Restaurant | Public Art | French Restaurant | American Restaurant |
| 8 | Upper East Side | Exhibit | Italian Restaurant | American Restaurant | Juice Bar | Gym / Fitness Center |
| 14 | Clinton | Italian Restaurant | Theater | Gym / Fitness Center | Sandwich Place | Coffee Shop |
| 15 | Midtown | Hotel | Sporting Goods Shop | Clothing Store | Coffee Shop | Bookstore |
| 16 | Murray Hill | Hotel | Coffee Shop | American Restaurant | Sandwich Place | Japanese Restaurant |
| 17 | Chelsea | Coffee Shop | Bakery | Art Gallery | American Restaurant | Italian Restaurant |
| 19 | East Village | Bar | Mexican Restaurant | Speakeasy | Pizza Place | Ice Cream Shop |
| 22 | Little Italy | Bakery | Café | Chinese Restaurant | Mediterranean Restaurant | Hotel |
| 25 | Manhattan Valley | Mexican Restaurant | Bar | Coffee Shop | Vietnamese Restaurant | Playground |
| 27 | Gramercy | Italian Restaurant | Bar | Pizza Place | Bagel Shop | Wine Shop |
| 28 | Battery Park City | Park | Coffee Shop | Hotel | Clothing Store | Women's Store |
| 29 | Financial District | Coffee Shop | Pizza Place | Cocktail Bar | Park | Bar |
| 30 | Carnegie Hill | Coffee Shop | Café | Wine Shop | Yoga Studio | Bar |
| 32 | Civic Center | Coffee Shop | Spa | Gym / Fitness Center | American Restaurant | French Restaurant |
| 33 | Midtown South | Korean Restaurant | Hotel | Dessert Shop | Gym / Fitness Center | Hotel Bar |
| 34 | Sutton Place | Italian Restaurant | Pizza Place | Gym / Fitness Center | Coffee Shop | Indian Restaurant |
| 38 | Flatiron | Italian Restaurant | Japanese Restaurant | New American Restaurant | American Restaurant | Mediterranean Restaurant |
| 39 | Hudson Yards | Gym / Fitness Center | Italian Restaurant | Hotel | American Restaurant | Thai Restaurant |

Cluster 3

```
manhattan_merged.loc[manhattan_merged['Cluster Labels'] == 2, manhattan_merged.columns[[1] + list(range(5, manhattan_merged.shape[1]))]]
```

| | Neighborhood | 1st Most Common Venue | 2nd Most Common Venue | 3rd Most Common Venue | 4th Most Common Venue | 5th Most Common Venue |
|----|---------------------|-----------------------|-------------------------|-----------------------|--------------------------|-----------------------|
| 5 | Manhattanville | Coffee Shop | Mexican Restaurant | Italian Restaurant | Bar | Seafood Restaurant |
| 9 | Yorkville | Italian Restaurant | Gym | Coffee Shop | Deli / Bodega | Sushi Restaurant |
| 10 | Lenox Hill | Italian Restaurant | Coffee Shop | Sushi Restaurant | Pizza Place | Café |
| 11 | Roosevelt Island | Deli / Bodega | Coffee Shop | Park | Bus Line | Bubble Tea Shop |
| 12 | Upper West Side | Italian Restaurant | Bakery | Wine Bar | Bar | Café |
| 13 | Lincoln Square | Plaza | Performing Arts Venue | Concert Hall | Café | Theater |
| 18 | Greenwich Village | Italian Restaurant | Clothing Store | Boutique | Sushi Restaurant | Dessert Shop |
| 21 | Tribeca | American Restaurant | Park | Italian Restaurant | Wine Bar | Café |
| 23 | Soho | Clothing Store | Boutique | Italian Restaurant | Mediterranean Restaurant | Shoe Store |
| 24 | West Village | Italian Restaurant | New American Restaurant | Cocktail Bar | American Restaurant | Park |
| 26 | Morningside Heights | Coffee Shop | Park | Bookstore | American Restaurant | Burger Joint |
| 31 | Noho | Italian Restaurant | Cocktail Bar | French Restaurant | Pizza Place | Hotel |
| 35 | Turtle Bay | Italian Restaurant | Coffee Shop | Sushi Restaurant | Deli / Bodega | Café |
| 36 | Tudor City | Park | Mexican Restaurant | Café | Pizza Place | Gym |

Cluster 4

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

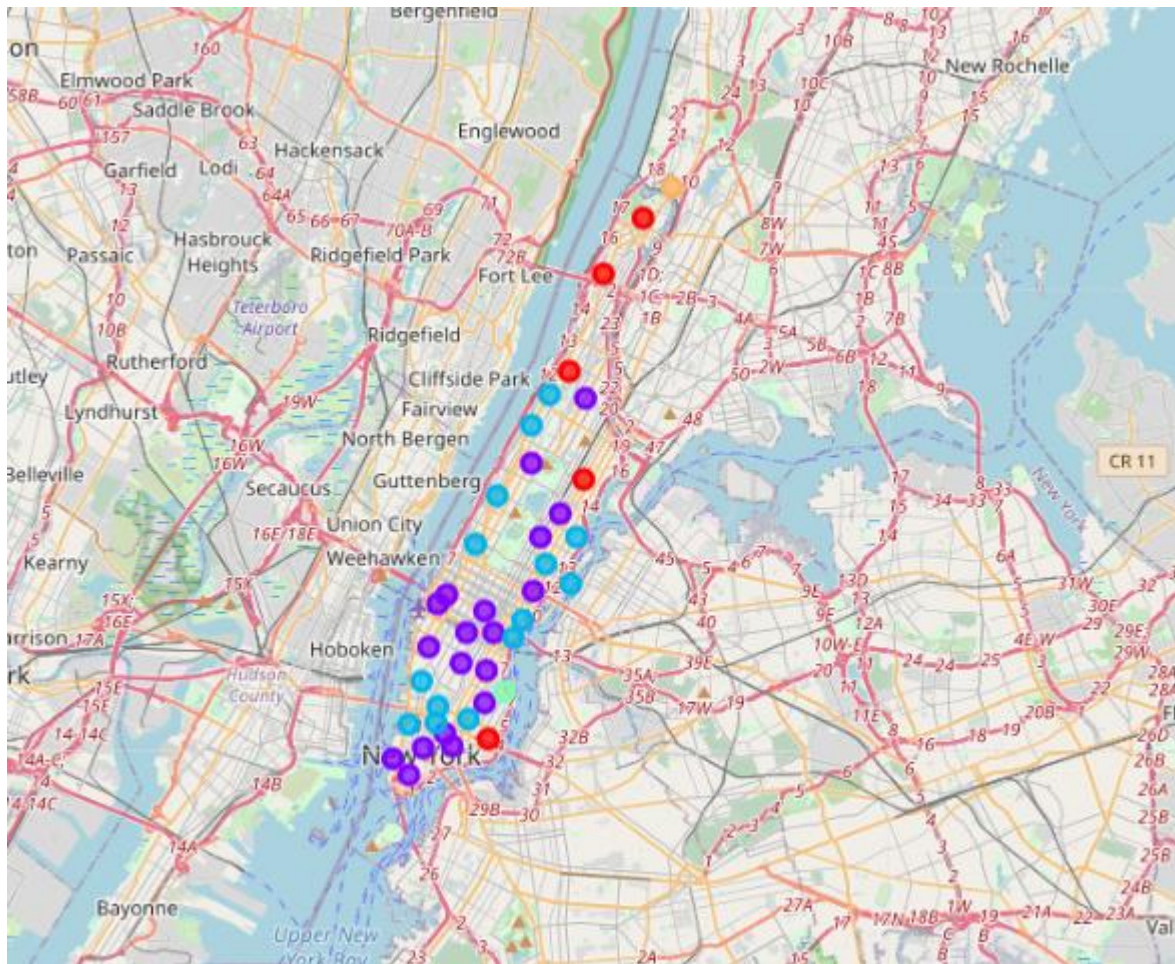
| | Neighborhood | 1st Most Common Venue | 2nd Most Common Venue | 3rd Most Common Venue | 4th Most Common Venue | 5th Most Common Venue |
|----|-----------------|-----------------------|-----------------------|-----------------------|-----------------------|-----------------------|
| 37 | Stuyvesant Town | Bar | Park | Coffee Shop | Cocktail Bar | Playground |

Cluster 5

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

| | Neighborhood | 1st Most Common Venue | 2nd Most Common Venue | 3rd Most Common Venue | 4th Most Common Venue | 5th Most Common Venue |
|---|--------------|-----------------------|-----------------------|-----------------------|-----------------------|-----------------------|
| 0 | Marble Hill | Coffee Shop | Sandwich Place | Gym | Yoga Studio | Bank |

4.5 Visualizing the clusters:



5. Results/Conclusions:

As can be seen from the above data processing and analysis that the most optimum locations for a new Yoga Studio would be the neighborhoods of Marble Hill, Carnegie Hill, Upper East Side and Hamilton Heights in Manhattan. The analysis has shown that these are the areas with most number of Gyms and Fitness centers. This shows that these areas are populated with people who are interested in health studios. These 4 areas also have many cafes situated in the area. This is also another avenue for foot traffic for the the new yoga studio. Of these four neighborhood areas

the most likely area would be Carnegie Hill. This neighborhood is adjacent to Upper East Side so it also has potential for more spill over traffic. Upper East Side also is also fairly residential and that increases the customer base.

This analysis needs more data for transportation issues. Any new business needs accessibility and it's proximity to subway stations, bus stations or parking areas is highly desired. So further comprehensive analysis with external transportation data would be an asset. This result is based on existing data showing existing gyms, fitness centers, cafes and restaurants in the area. The assumption being that since these businesses are established here there is scope for another fitness business to open and thrive. With the city opening back up again after the pandemic, people are eager to come out of homes and work out in the company of other people face to face. And this is where the new Yoga studio will make its mark. The pandemic had forced the closure of many small boutique businesses like yoga studios so with everything opening up there is definitely potential for a new yoga studio to thrive!