

COURSERA CAPSTONE PROJECT

Coursera IBM Data Science Certification

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INTRODUCTION

1.1 Scenario and Background

I'm a Data Employee residing in Richmond Town in Bangalore. I currently live within walking distance to Richmond Town Bus-station and in this place, I enjoy many pleasurable places in that area, such as various international restaurants, cafes, malls, Theatres, food shops and entertainment. I have been offered a great opportunity to work in the branch in Manhattan, New York. I am very interested and I want to use this opportunity.

1.2 Problem to be resolved:

In order to make a comparison and evaluation of the rental options in Manhattan New York, I must set some basis, therefore the apartment in Manhattan must meet the following demands:

- Apartment must be 2(or)3 bedrooms.
- The location required is near a metro station in the Manhattan area and within 1.0-mile radius.
- Price of rent not exceed \$7,000 per month.
- Top Facilities in the selected neighbourhood shall be similar to current residence.
- Required to have venues such as coffee shops, restaurants Asian Thai, wine stores, gym and food shops.
- As a reference, I have included a map of venues near current residence in Singapore. (Which is similar to my location in Bangalore didn't get the exact location of my residence in Bangalore so I used Singapore Location).

1.3 Interested Audience

I believe this is a relevant project for a person or entity considering moving to a major city in Europe, US or Asia, since the approach and methodologies used here are applicable in all cases. The use of Foursquare data and mapping techniques combined with data analysis will help resolve the key questions arisen. Lastly, this project is a good practical case toward the development of Data Science skills.

DATA SECTION

2.1 Data of Current Situation

I Currently reside in the neighbourhood of 'MacCallum Street' in Downtown Singapore.(As I'm comparing this with my current residence in Richmond town in Bangalore) I use Foursquare to identify the venues around the area of residence which are then shown in the Singapore map shown in methodology and execution in section 3.0 . It serves as a reference for comparison with the desired future location in Manhattan NY.

2.2 Data Required to resolve the problem

In order to make a good choice of a similar apartment in Manhattan NY, the following data is required: List/Information on neighbourhood's form Manhattan with their Geodata (latitude and longitude. List/Information about the subway metro stations in Manhattan with geodata. Listed apartments for rent in Manhattan area with descriptions (how many beds, price, location, address) Venues and amenities in the Manhattan neighbourhoods (e.g. top 10)

2.3 Sources and Manipulation

The list of Manhattan neighborhoods is worked out during Lab exercise during the course. A csv file was created which will be read in order to create a data frame and its mapping.

A list of Manhattan subway metro stops was compiled in Numbers (Apple excel) and it was complemented with wikipedia data (https://en.wikipedia.org/wiki/List_of_New_York_City_Subway_stations_in_Manhattan) and information from NY Transit authority and Google maps (<https://www.google.com/maps/search/manhattan+subway+metro+stations/@40.7837297,-74.1033043,11z/data=!3m1!4b1>) for a final consolidated list of subway stops names and their address. The geolocation was obtained via an algorithm using Nominatim.

A list of places for rent was collected by web-browsing real estate companies in Manhattan : <http://www.rentmanhattan.com/index.cfm?page=search&state=results> https://www.nestpick.com/search?city=new-york&page=1&order=relevance&district=manhattan&qclid=CjwKCAiAjNigBRAgEiwAGLI2hkP3A-cPxjZYkURqQEswQK2jKQEpv_MvKcrIhRWRzNkc_r-fGi0lxoCA7cQAvD_BwE&type=apartment&display=list https://www.realtor.com/apartments/Manhattan_NY

A csv file was compiled with the rental place that indicated: areas of Manhattan, address, number of beds, area and monthly rental price. The csv file "nnnn.csv" had the following below structure. An algorithm was used to create all the geodata using Nominatim, as shown in section 3.0. The actual algorithm coding may be shown in 'markdown' mode because it takes time to run. With the use of geolocator = Nominatim() , it was possible to determine the latitude and longitude for the subway metro locations as well as for the geodata for each rental place listed. The loop algorithms used are shown in the execution of data in section 3.0 "Great_circle" function from geolocator was used to calculate distances between two points , as in the case to calculate average rent price for units around each subway station and at 1.6 km radius. Foursquare is used to find the avenues at Manhattan neighborhoods in general and a cluster is created to later be able to search for the venues depending of the location shown.

2.4 Data used to solve problem

The data will be used as follows: Use Foursquare and geopy data to map top 10 venues for all Manhattan neighborhoods and clustered in groups (as per Course LAB) Use foursquare and geopy data to map the location of subway metro stations , separately and on top of the above clustered map in order to be able to identify the venues and ammenities near each metro station, or explore each subway location separately Use Foursquare and geopy data to map the location of rental places, in some form, linked to the subway locations. create a map that depicts, for instance, the average rental price per square ft, around a radius of 1.0 mile (1.6 km) around each subway station - or a similar metrics. I will be able to quickly point to the popups to know the relative price per subway area. Addresses from rental locations will be converted to geodata(lat, long) using Geopy-distance and Nominatim. Data will be searched in open data sources if available, from real estate sites if open to reading, libraries or other government agencies such as Metro New York MTA, etc.

2.5 Mapping of Data

The following maps were created to facilitate the analysis and the choice of the palace to live. Manhattan map of Neighborhoods manhattan subway metro locations Manhattan map of places for rent Manhattan map of clustered venues and neighborhoods Combined maps of Manhattan rent places with subway locations Combined maps of Manhattan rent places with subway locations and venues clusters.

METHODOLOGY

3.1 THE STRATEGY TO FIND THE ANSWER:

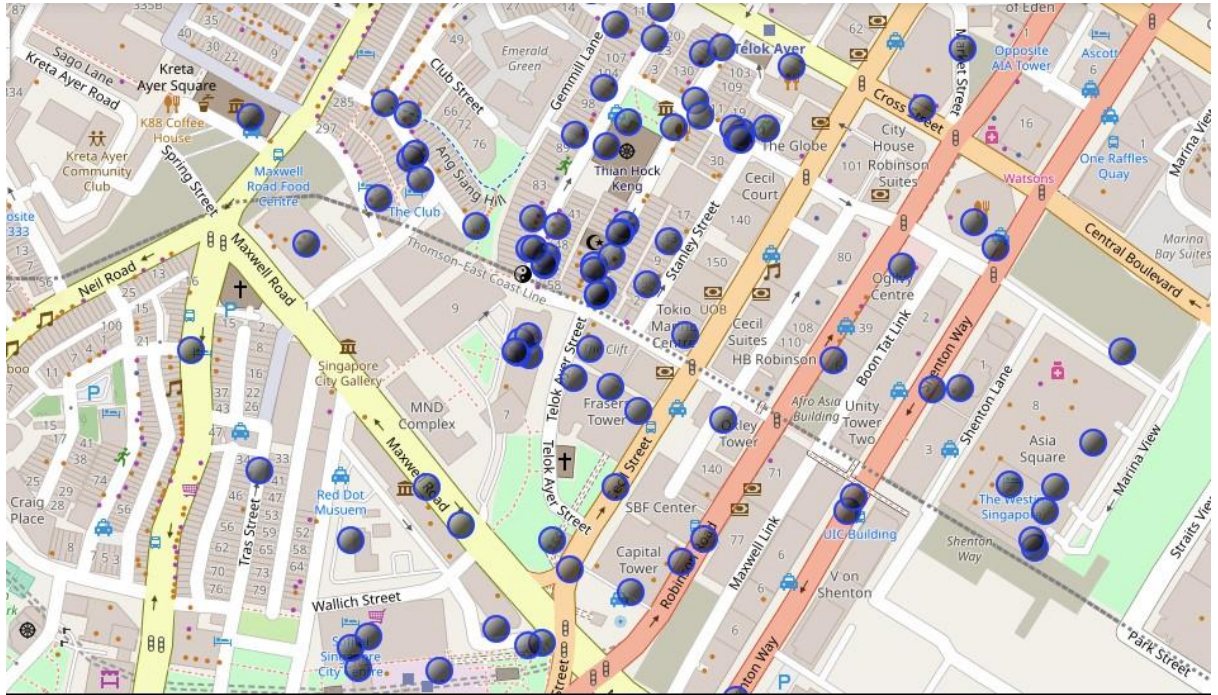
The strategy is based on mapping the described data in section 2.0, in order to facilitate the choice of at least two candidate places for rent. The information will be consolidated in ONE MAP where one can see the details of the apartment, the cluster of venues in the neighborhood and the relative location from a subway station and from work place. A measurement tool icon will also be provided. The popups on the map items will display rent price, location and cluster of venues applicable.

3.2 THE TOOLS:

- Web-scraping of sites is used to consolidate data-frame information which was saved as csv files for convenience and to simplify the report. Geodata was obtained by coding a program to use Nominatim to get latitude and longitude of subway stations and also for each of (144 units) the apartments for rent listed.
- Geopy_distance and Nominatim were used to establish relative distances. Seaborn graphic was used for general statistics on rental data.
- Maps with popups labels allow quick identification of location, price and feature, thus making the selection very easy.

EXECUTION AND RESULTS

Current Residence Neighborhood in Singapore

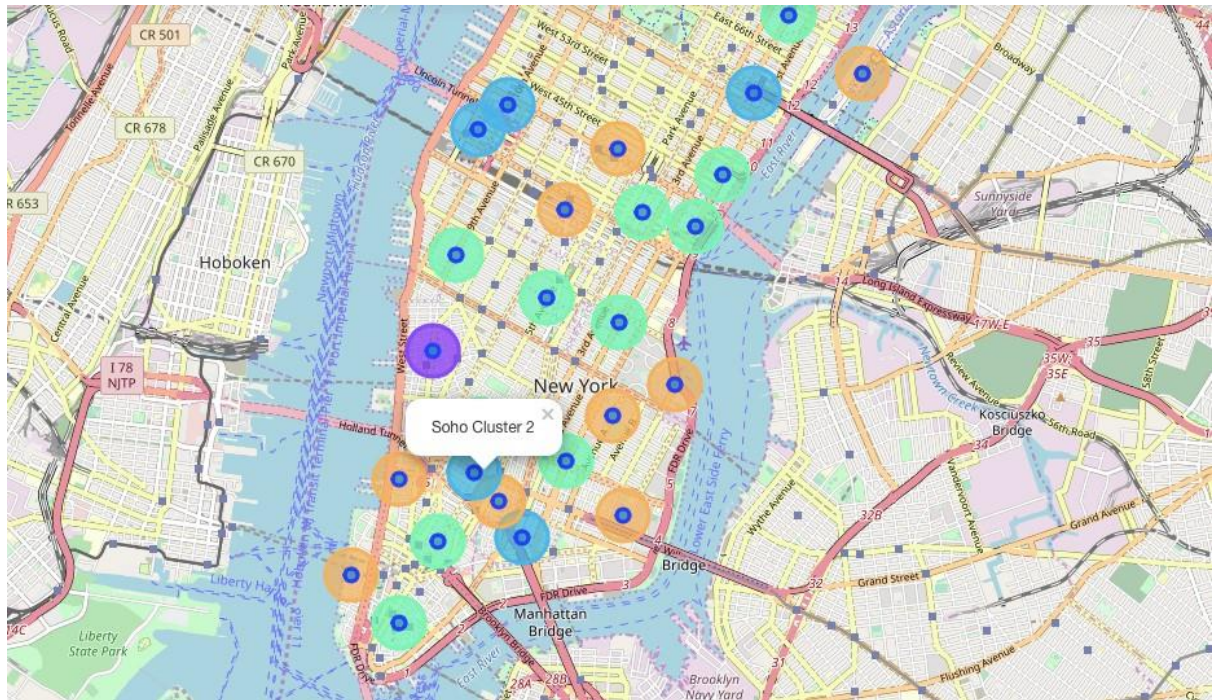


Venues around in Neighborhood

```
# Venues near current Singapore residence place  
SGnearby_venues.head(10)
```

	name	categories	lat	lng
0	Napoleon Food & Wine Bar	Wine Bar	1.279925	103.847333
1	Park Bench Deli	Deli / Bodega	1.279872	103.847287
2	Native	Cocktail Bar	1.280135	103.846844
3	Muchachos	Burrito Place	1.279175	103.847082
4	Matt's The Chocolate Shop	Dessert Shop	1.280462	103.846950
5	Freehouse	Beer Garden	1.281254	103.848513
6	PS.Cafe	Café	1.280468	103.846264
7	왕대박 Wang Dae Bak Korean BBQ Restaurant	Korean Restaurant	1.281345	103.847551
8	Ancient Therapy	Massage Studio	1.280413	103.847481
9	Oven & Fried Chicken	Korean Restaurant	1.280479	103.847522

Manhattan Map – Clusters.



Geodata Manhattan apt for Rent

```
j: mh_rent=pd.read_csv('MH_rent_latlong.csv')
mh_rent.head()
```

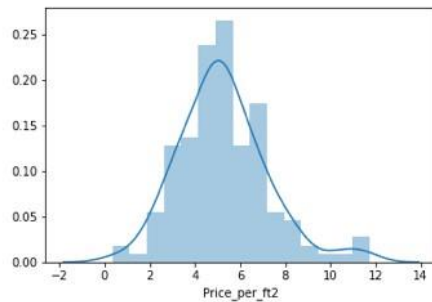
	Address	Area	Price_per_ft2	Rooms	Area-ft2	Rent_Price	Lat	Long
0	West 105th Street	Upper West Side	2.94	5.0	3400	10000	40.799771	-73.966213
1	East 97th Street	Upper East Side	3.57	3.0	2100	7500	40.788585	-73.955277
2	West 105th Street	Upper West Side	1.89	4.0	2800	5300	40.799771	-73.966213
3	CARMINE ST.	West Village	3.03	2.0	1650	5000	40.730523	-74.001873
4	171 W 23RD ST.	Chelsea	3.45	2.0	1450	5000	40.744118	-73.995299

```
j: mh_rent.tail()
```

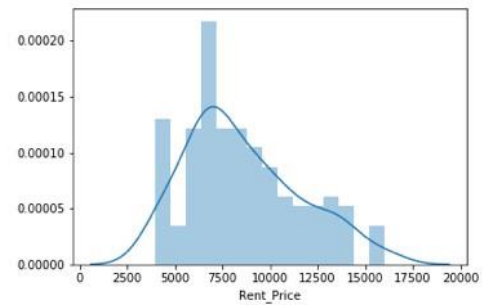
	Address	Area	Price_per_ft2	Rooms	Area-ft2	Rent_Price	Lat	Long
139	200 East 72nd Street	Rental in Lenox Hill	5.15	3.0	1700	8750	40.769465	-73.960339
140	50 Murray Street	No fee rental in Tribeca	7.11	2.0	1223	8700	40.714051	-74.009608
141	300 East 56th Street	No fee rental in Midtown East	3.87	3.0	2100	8118	40.758216	-73.965190
142	1930 Broadway	No fee rental in Central Park West	5.06	2.0	1600	8095	40.772474	-73.981901
143	33 West 9th Street	Rental in Greenwich Village	6.67	2.0	1500	10000	40.733691	-73.997323

Rents in Manhattan Apartments

<matplotlib.axes._subplots.AxesSubplot at 0x1a2415fc18>

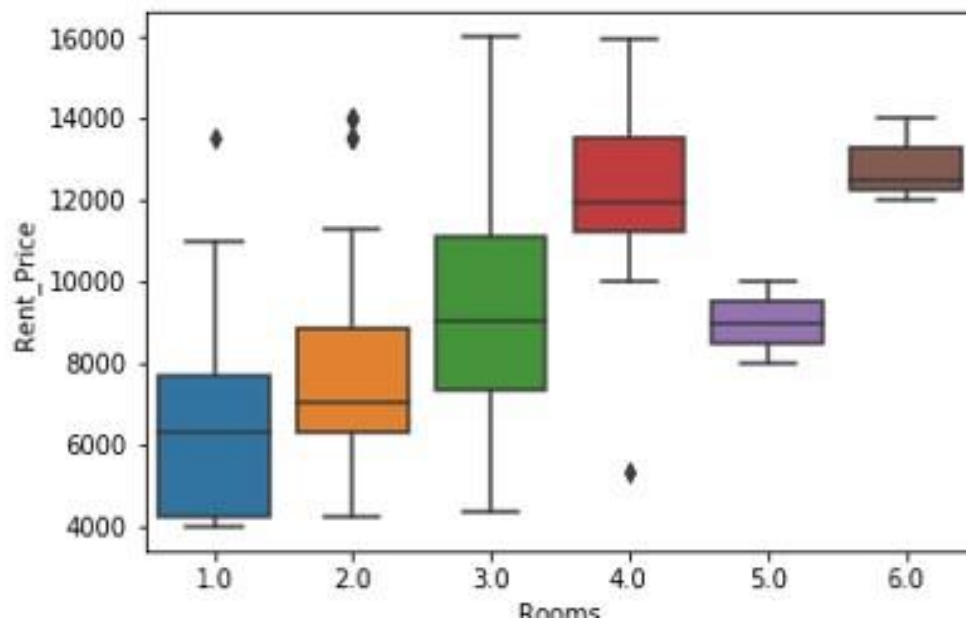


<matplotlib.axes._subplots.AxesSubplot at 0x1a25dd8400>

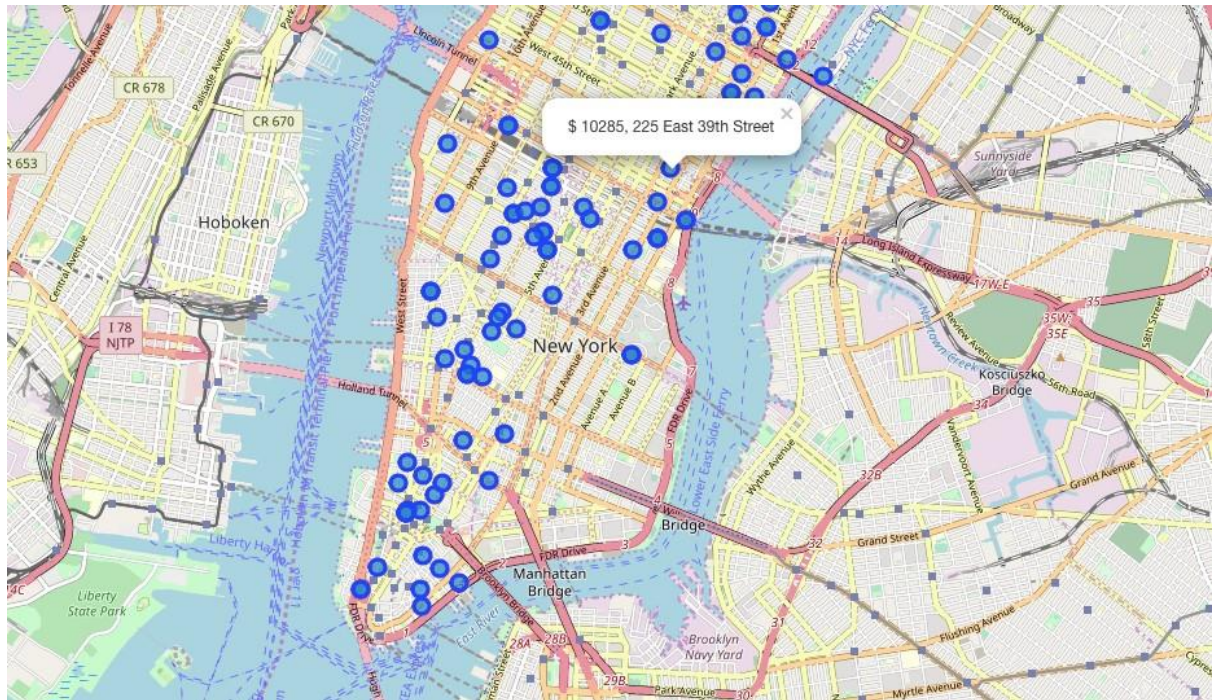


```
sns.boxplot(x='Rooms', y='Rent_Price', data=mh_rent)
```

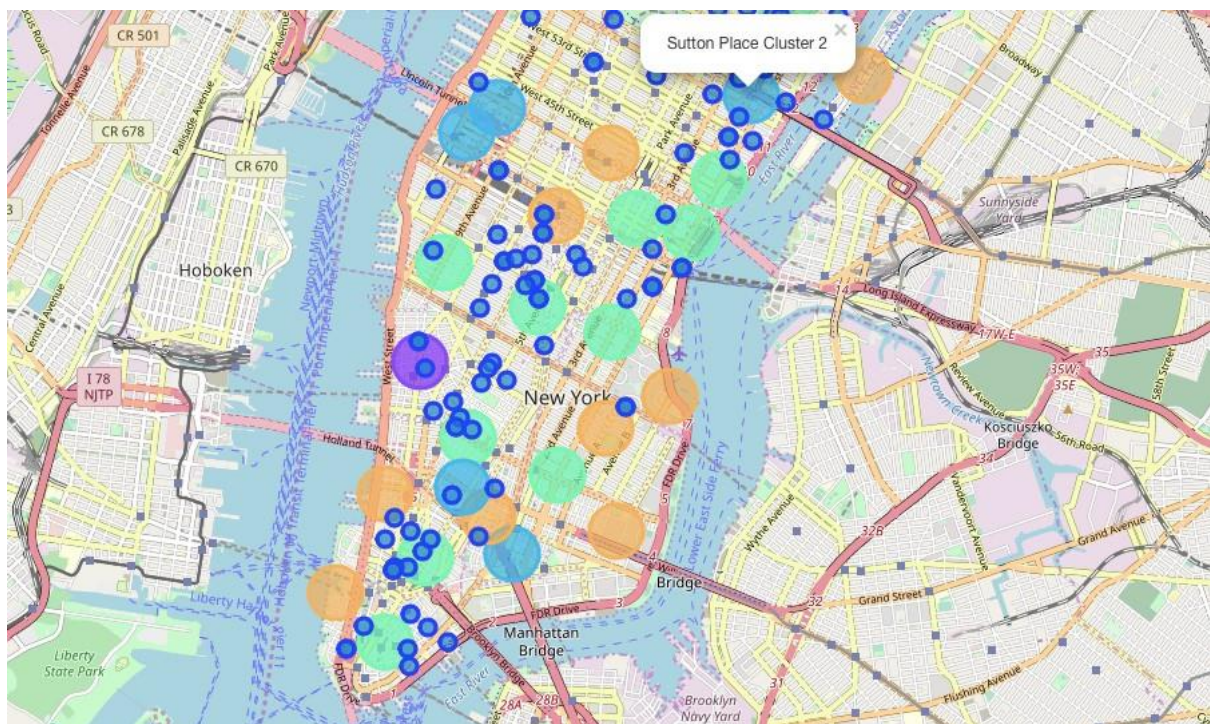
<matplotlib.axes._subplots.AxesSubplot at 0x1a25f2a2b0>



Apartments for Rent in Manhattan



Manhattan apt for Rent-Clusters



Manhattan Subway Stations GeoData

click to scroll output; double click to hide				
	sub_address	lat	long	
0	Dyckman Street Subway Station	170 Nagle Ave, New York, NY 10034, USA	40.861857	-73.924509
1	57 Street Subway Station	New York, NY 10106, USA	40.764250	-73.954525
2	Broad St	New York, NY 10005, USA	40.730862	-73.987156
3	175 Street Station	807 W 177th St, New York, NY 10033, USA	40.847991	-73.939785
4	5 Av and 53 St	New York, NY 10022, USA	40.764250	-73.954525

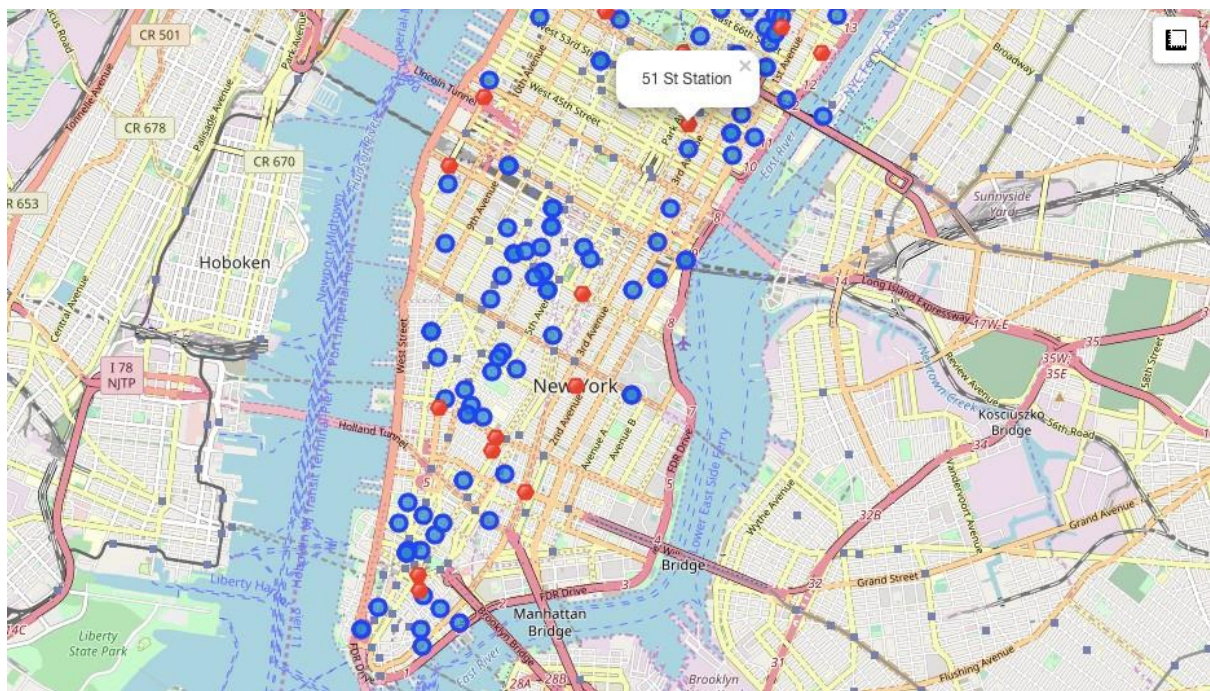
```
# removing duplicate rows and creating new set mhsubl
mhsubl=mh.drop_duplicates(subset=['lat','long'], keep="last").reset_index(drop=True)
mhsubl.shape
```

(22, 4)

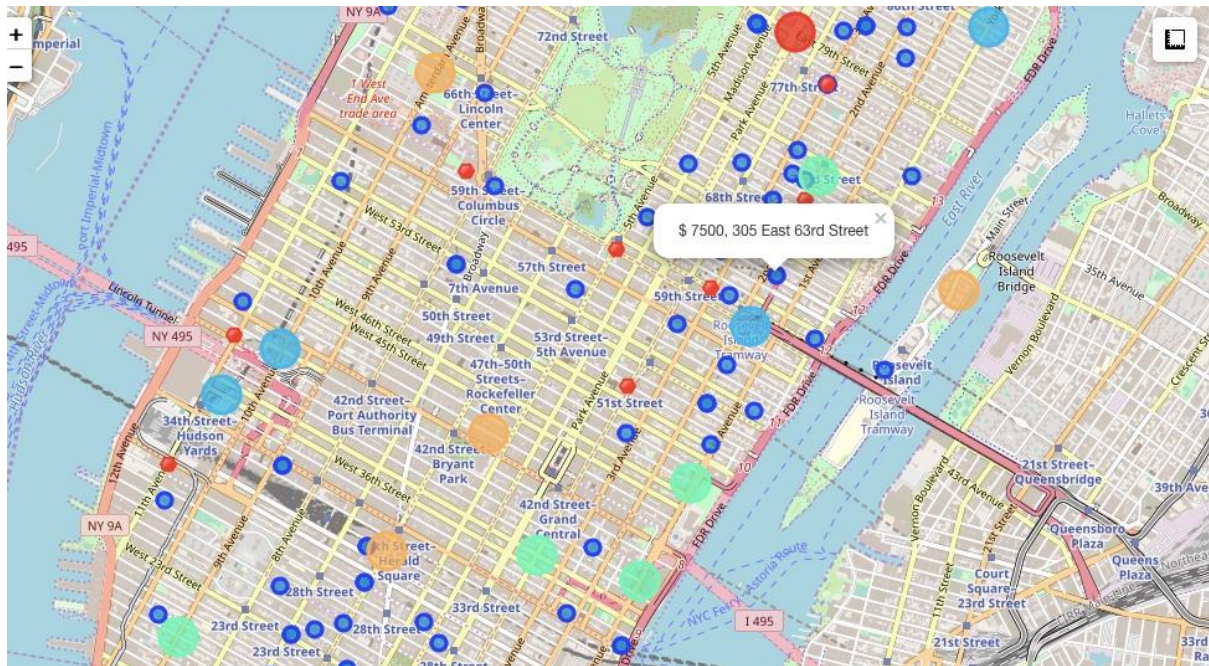
```
: mhsubl.tail()
```

	sub_station	sub_address	lat	long
17	190 Street Subway Station	Bennett Ave, New York, NY 10040, USA	40.858113	-73.932983
18	59 St-Lexington Av Station	E 60th St, New York, NY 10065, USA	40.762259	-73.966271
19	57 Street Station	New York, NY 10019, United States	40.764250	-73.954525
20	14 Street / 8 Av	New York, NY 10014, United States	40.730862	-73.987156
21	MTA New York City	525 11th Ave, New York, NY 10018, USA	40.759809	-73.999282

APT for Rent



Selected Apartment



The one which is shown in the Graph is the selected Apartment.

Apartment Selection

- Apartment 1 rent cost is US\$7500 slightly above the US\$7000 budget. Apt 1 is located 400 meters from subway station at 59th Street and work place (Park Ave and 53rd) is another 600 meters away. I can walk to work place and use subway for other places around. Venues for this apt are as of Cluster 2 and it is located in a fine district in the East side of Manhattan.
- Apartment 2 rent cost is US\$6935, just under the US\$7000 budget. Apt 2 is located 60 meters from subway station at Fulton Street, but I will have to ride the subway daily to work , possibly 40-60 min ride. Venues for this apt are as of Cluster 3
- Based on current Singapore venues, I feel that Cluster 2 type of venues is a closer resemblance to my current place. That means that APARTMENT 1 is a better choice since the extra monthly rent is worth the conveniences it provides.

DISCUSSION

- I'm impressed with the overall content and lab works presented during the Coursera IBM Certification Course.**
- I feel this Capstone project presented me a great opportunity to practice and apply the Data Science tools and methodologies learned in this Course.**
- I feel I have acquired a good starting point to become a professional Data Scientist and I will continue exploring to creating examples of practical cases.**

CONCLUSION

- I feel rewarded with the efforts and money spent. I believe this course with all the topics covered is well worthy of appreciation.**
- This project has shown me a practical application to resolve a real situation that has impacting personal and financial impact using Data Science tools.**
- The mapping with Folium is a very powerful technique to consolidate information and make the analysis and decision thoroughly and with confidence. I would recommend for use in similar situations.**

Thanks COURSERA/IBM who helped me in this Course.

I also thank each and every friend who involved in my success by corrections.