

Coursera IBM Data Science Certification

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REPORT CONTENT

- 1. Introduction Section:
- The "business problem" to be solved by this project and who may be interested
- 2. Data Section:
- Describe Data requirements and Sources needed to solve the problem
- 3. Methodology section:
- Main component of the report Execute data processing, describe/discuss any exploratory data analysis and/or inferential statistical testing performed, and/or machine learnings used.
- 4. Results section:
- Discussion of the results and finding of answer
- 5. Discussion section:
- Discussion of observations noted and any recommendations
- 6. Conclusion section:
- Answer chosen and conclusions.

INTRODUCTION

1.1 Scenario and Background

I am currently living in India, within walking distance to Chhatrapati Shivaji Maharaj Terminus. I also enjoy great venues and attractions, such as international cuisine, entertainment and shopping. I have an offer to move to work to Manhattan NY and I would like to move if I can find a place to live similar with similar venues.

2. Problem to be resolved:

How to find an apartment in Manhattan with the following conditions:

- Apartment with min 2 bedrooms
- Monthly rent not to exceed US\$7000/month
- Located within walking distance (<=1.0 mile, 1.6 km) from a subway metro station in Manhattan
- Venues and amenities as in my current residence.

3. Interested Audience

I believe the methodology, tools and strategy used in this project is relevant for a person or entity considering moving to a major city in US, Europe or Asia. Europe, US or Asia, Likewise, it can be helpful approach to explore the opening of a new business. 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 for a person developing Data Science skills.

DATA SECTION

2.1 Data Requirements

- Geodata for current residence in Singapore with venues established using Foursquare.
- -List of Manhattan (MH) neighborhoods with clustered venues established via Foursquare (as in Course
- Lab). https://en.wikipedia.org/wiki/List_of_Manhattan_neighborhoods#Midtown_neighborhoods
- -List of subway metro stations in Manhattan with addresses and geo data (lat,long): https://www.etations.in.Manhattan). (https://www.etations.in.Manhattan). (https://www.etations.in.Manhattan).
- en.wikipedia.org/wiki/List_of_New_York_City_Subway_stations_in_Manhattan), (https://www.google.com/maps/search/manhattan+subway+metro+stations/@40.7837297,-74.1033043,11z/data=!3m1!4b1)
- -List of apartments for rent in Manhattan area with information on neighborhood location, address, number of beds, area size, monthly rent price and complemented with geo data via Nominatim. <a href="http://www.rentmanhattan.com/index.cfm?page=search&state=results_https://www.nestpick.com/search?city=new-cit
- Place to work in Manhattan (Park Avenue and 53rd St) for reference

2.2 Data Sources, Data Processing and Tools used

- Singapore data and map is to be created with use of Nominatim, Foursquare and Folium mapping
- -Manhattan neighborhoods were obtained from Wikipedia and organized by Neighborhoods with geodata via Nominatim for mapping with Folium.
- List of Subway stations was obtained via Wikipedia, NY Transit web site and Google map,
- -List of apartments for rent was consolidated from web-scraping real estate sites for MH. The geolocation (lat,long) data was found with algorithm coding and using Nominatim.
- -Folium map was the basis of mapping with various features to consolidate all data in ONE map where one can visualize all details needed to make a selection of apartment

METHODOLOGY

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.

The Tools:

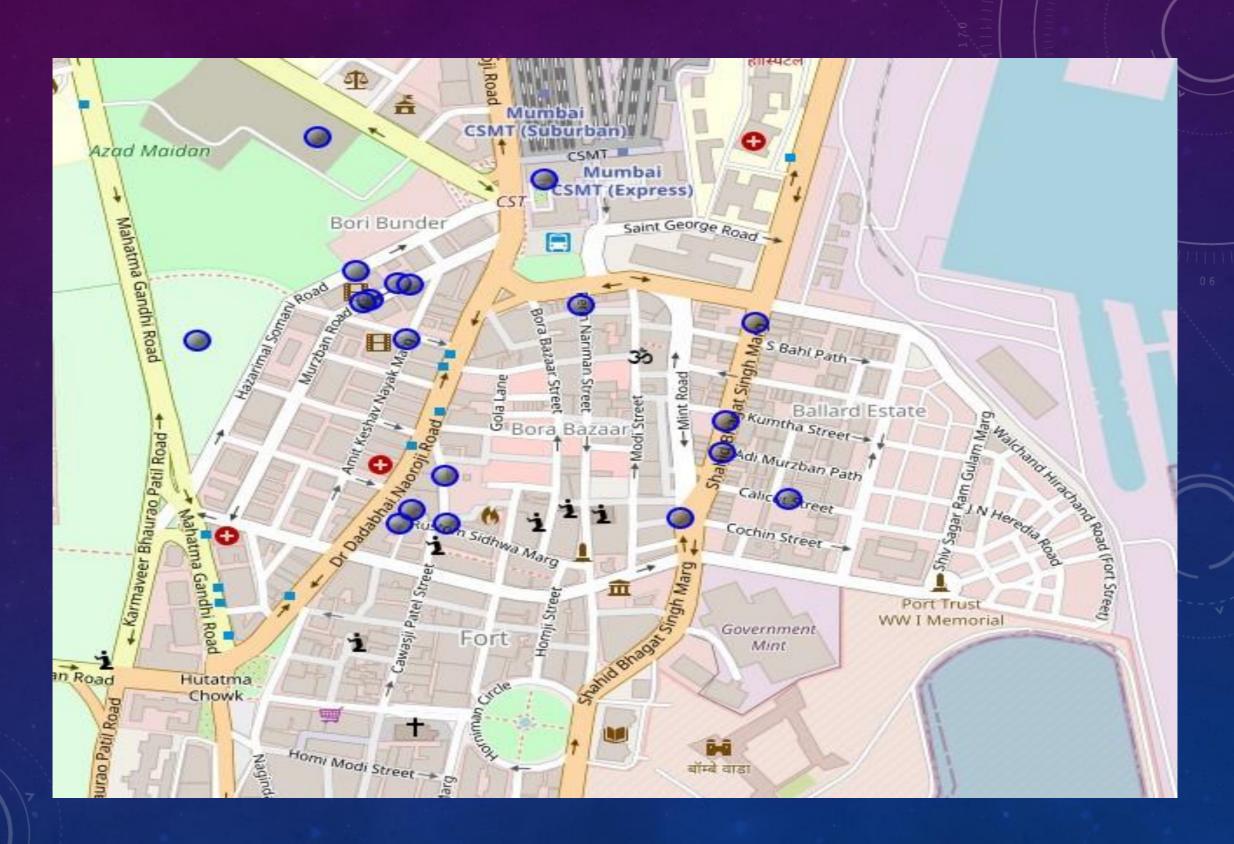
Web-scraping of sites is used to consolidate data-frame information which was saved as csv files for convenience and to simply the report. Geodata was obtained by coding a program to use Nominatim to get latitude and longitude of subway stations and also for each of (144 units) the apartments for rent 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 INDIA



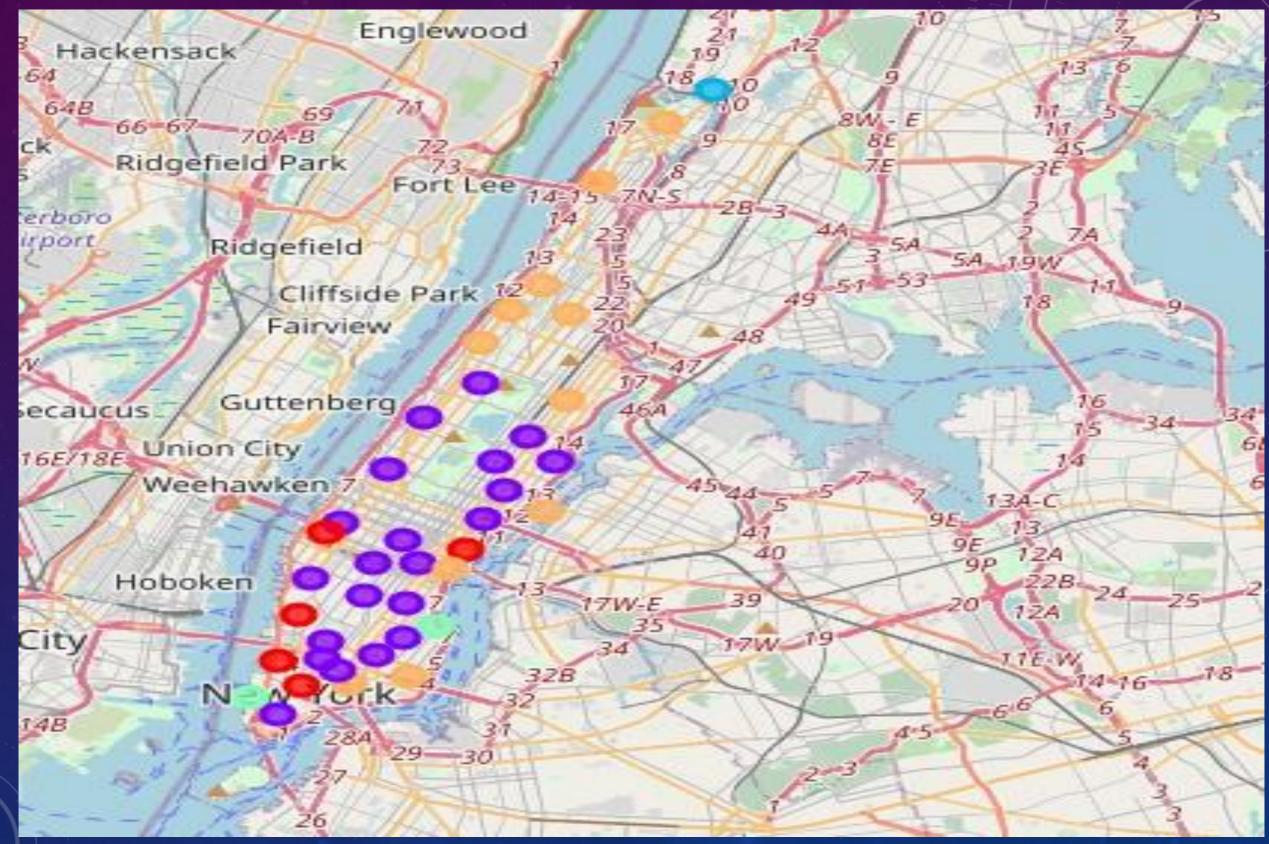
VENUES AROUND NEIGHBORHOOD IN INDIA

Venues near current Singapore residence place IGnearby_venues.head(10)

1

	name	categories	lat	Ing
0	Town House Cafe	Bar	18.938550	72.833464
1	Royal China	Chinese Restaurant	18.938715	72.832933
2	Sher-E-Punjab	Indian Restaurant	18.937944	72.837853
3	Cafe Excelsion	Café	18.937701	72.833566
4	Chhatrapati Shivaji Maharaj Terminus	Train Station	18.940088	72.835257
5	Pancham Puriwala	Indian Restaurant	18.938214	72.835697
6	Sterling Cineplex	Multiplex	18.938296	72.833104
7	Ideal Corner	Indian Restaurant	18.934961	72.834050
8	Hotel Fountain Plaza	Indian Restaurant	18.935671	72.834030
9	Café Universal	Irani Cafe	18.936021	72.837453

MANHATTAN MAP - NEIGHBORHOODS AND CLUSTER OF VENUES



GEODATA MANHATTAN APTS FOR RENT

mh_rent=pd.read_csv(r'C:\Users\HP\Desktop\MH_rent_latlong.csv')
mh_rent.head()

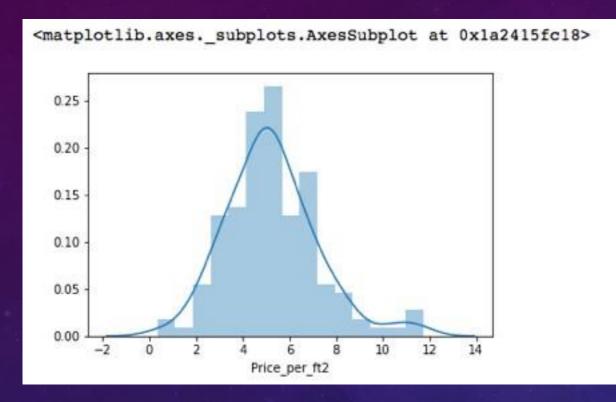
Address	Area	Price_per_ft2	Rooms	Area-ft2	Rent_Price	Lat	Long
West 105th Street	Upper West Side	2.94	5.0	3400	10000	40.799771	-73.966213
East 97th Street	Upper East Side	3.57	3.0	2100	7500	40.788585	-73.955277
West 105th Street	Upper West Side	1.89	4.0	2800	5300	40.799771	-73.966213
CARMINE ST.	West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

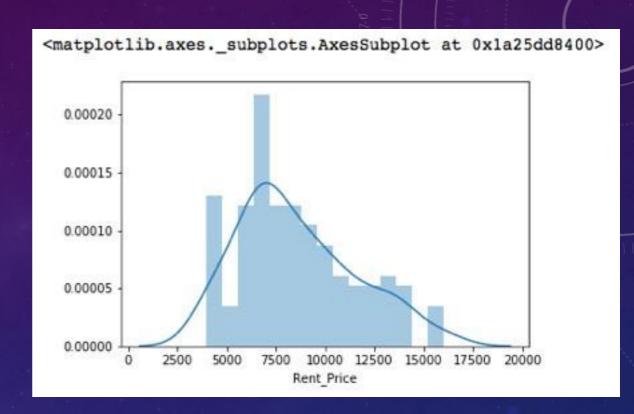
mh_rent.tail()

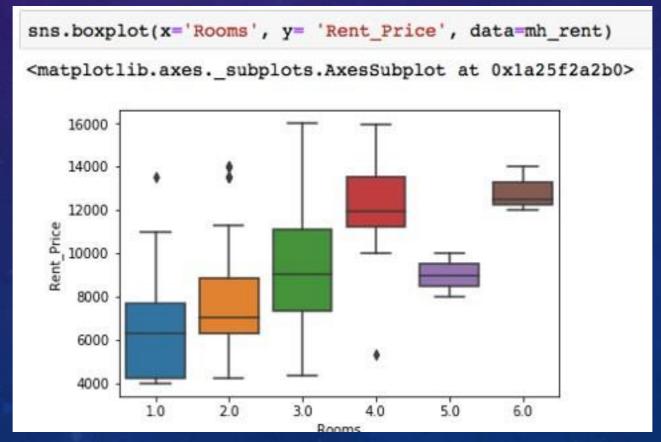
03	Address	Area	Price_per_ft2	Rooms	Area-ft2	Rent_Price	Lat	Long
139	200 East 72nd Street	Rental in Lenox Hill	5.15	3.0	1700	8750	40.769465	-73.960339
140	50 Murray Street	No fee rental in Tribeca	7.11	2.0	1223	8700	40.714051	-74.009608
141	300 East 56th Street	No fee rental in Midtown East	3.87	3.0	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

RENTAL PRICE STATISTICS MH APARTMENTS

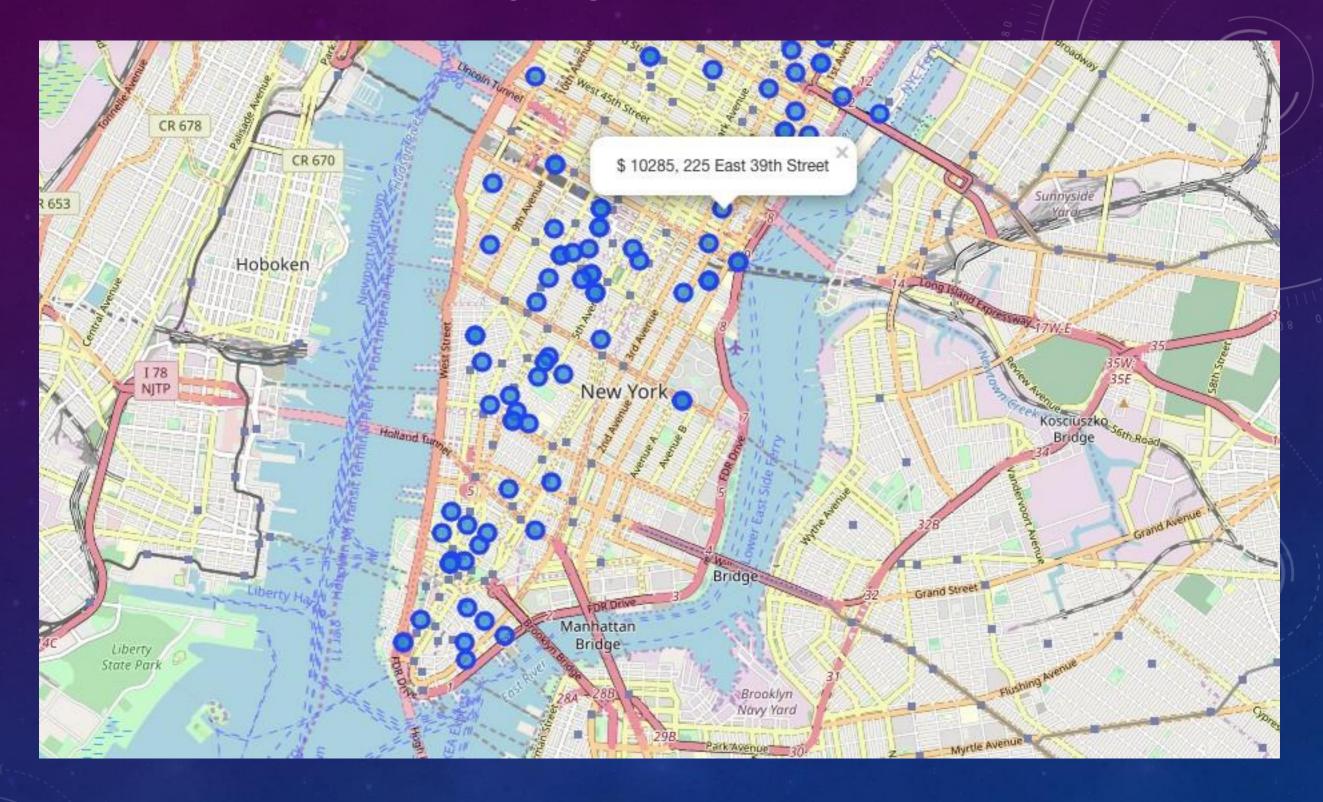
BUDGET US7000/MONTH IS AROUND THE MEAN



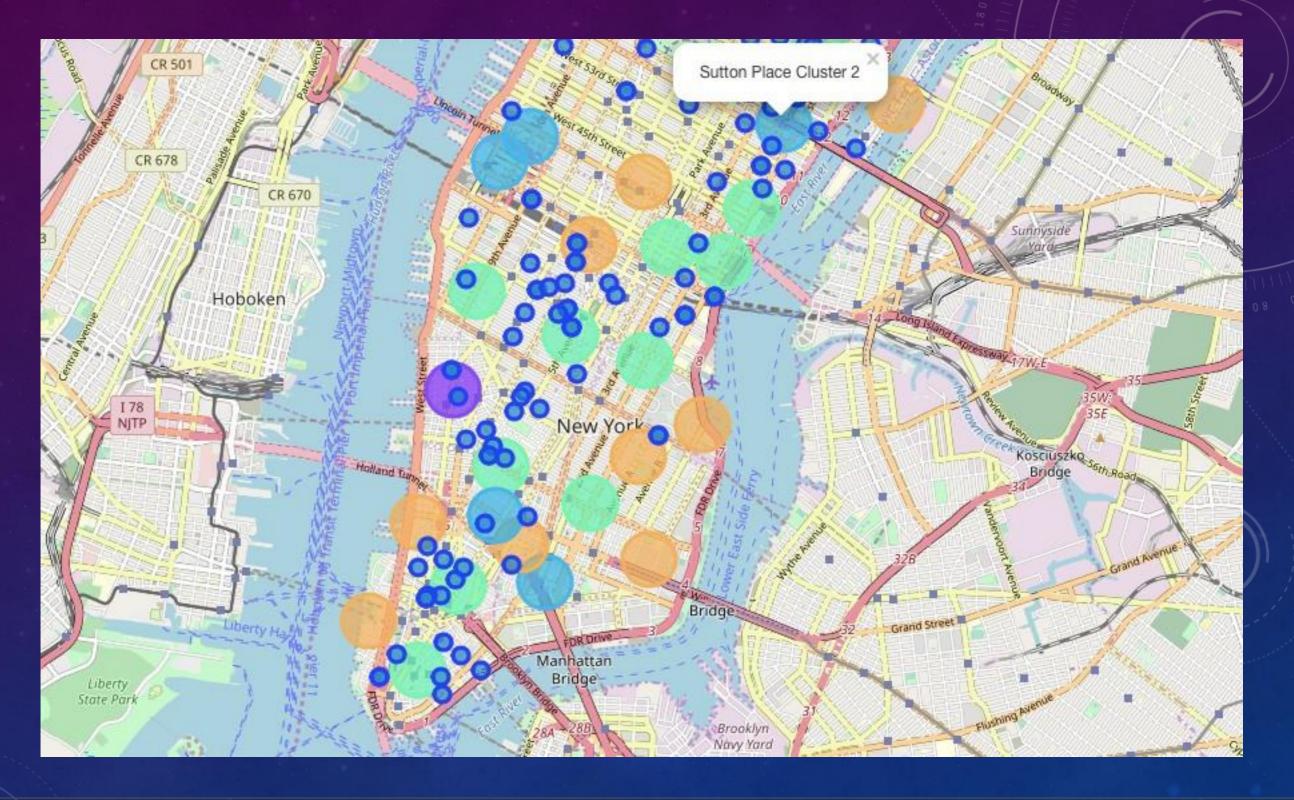




APARTMENTS FOR RENT IN MH



MH APTS FOR RENT WITH VENUE CLUSTERS



VENUES OF CLUSTER3

kk is the cluster number to explore
kk = 3

manhattan_merged.loc[manhattan_merged['Cluster Labels'] == kk, manhattan_merged.columns[[1] + list(range(5, manhattan_merged.shap

	Neighborhood	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	Inwood	Mexican Restaurant	Lounge	Pizza Place	Café	Wine Bar	Bakery	American Restaurant	Park	Frozen Yogurt Shop	Spanish Restaurant
5	Manhattanville	Deli / Bodega	Italian Restaurant	Seafood Restaurant	Mexican Restaurant	Sushi Restaurant	Beer Garden	Coffee Shop	Falafel Restaurant	Bike Trail	Other Nightlife
10	Lenox Hill	Sushi Restaurant	Italian Restaurant	Coffee Shop	Gym / Fitness Center	Pizza Place	Burger Joint	Deli / Bodega	Gym	Sporting Goods Shop	Thai Restaurant
12	Upper West Side	Italian Restaurant	Bar	Bakery	Vegetarian / Vegan Restaurant	Indian Restaurant	Coffee Shop	Cosmetics Shop	Wine Bar	Mexican Restaurant	Sushi Restaurant
16	Murray Hill	Sandwich Place	Hotel	Japanese Restaurant	Gym / Fitness Center	Coffee Shop	Salon / Barbershop	Burger Joint	French Restaurant	Bar	Italian Restaurant
17	Chelsea	Coffee Shop	Italian Restaurant	Ice Cream Shop	Bakery	Nightclub	Theater	Art Gallery	Seafood Restaurant	American Restaurant	Hotel
18	Greenwich Village	Italian Restaurant	Sushi Restaurant	French Restaurant	Clothing Store	Chinese Restaurant	Café	Indian Restaurant	Bakery	Seafood Restaurant	Electronics Store
27	Gramercy	Italian Restaurant	Restaurant	Thrift / Vintage Store	Cocktail Bar	Bagel Shop	Coffee Shop	Pizza Place	Mexican Restaurant	Grocery Store	Wine Shop
29	Financial District	Coffee Shop	Hotel	Gym	Wine Shop	Steakhouse	Bar	Italian Restaurant	Pizza Place	Park	Gym / Fitness Center
31	Noho	Italian Restaurant	French Restaurant	Cocktail Bar	Gift Shop	Bookstore	Grocery Store	Mexican Restaurant	Hotel	Sushi Restaurant	Coffee Shop
20	Civile Cooker	Gym /	Delver	Italian	Cooldoil Doc	French	Sandwich	0-# Ch	C	Vana Challa	Ded

MANHATTAN SUBWAY STATIONS GEODATA

	sub_station		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

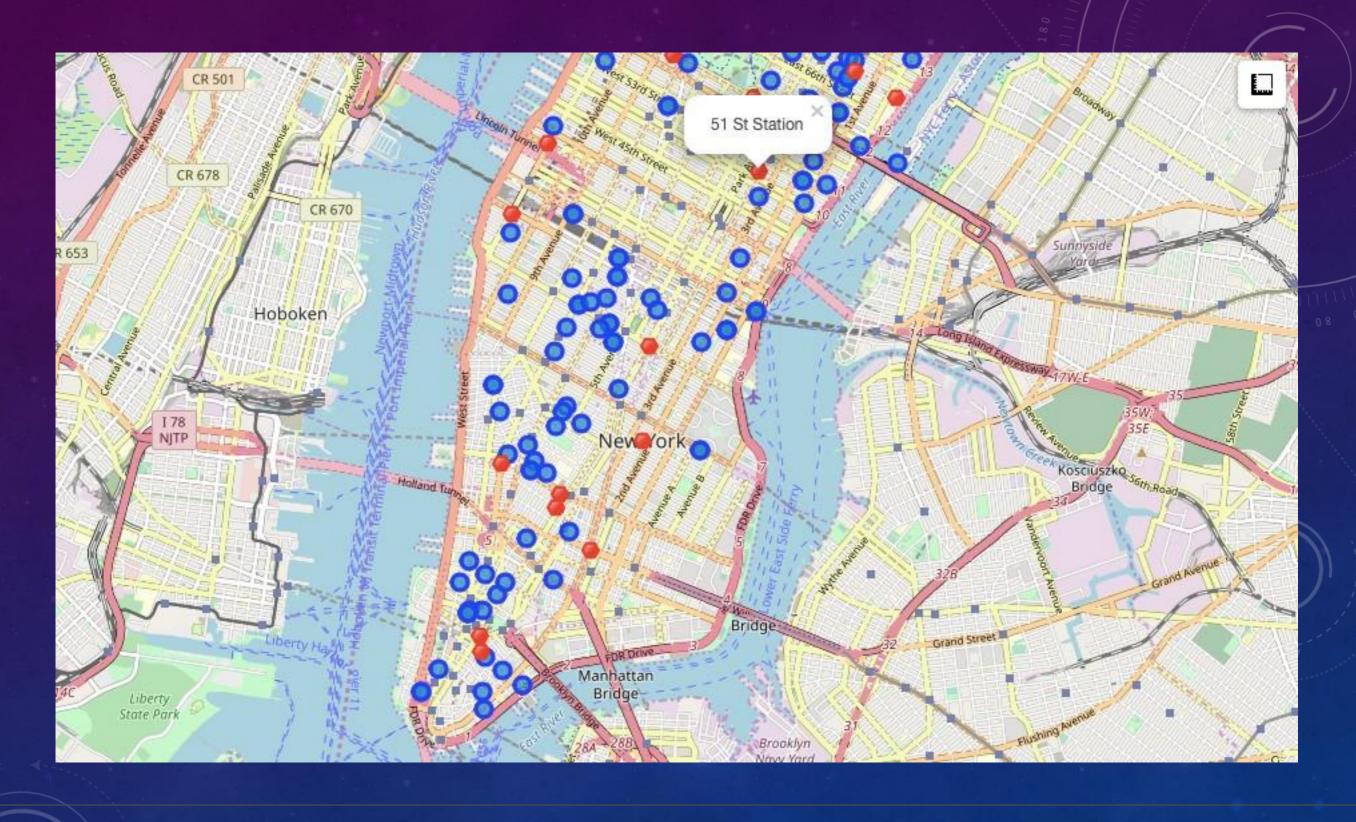
removing duplicate rows and creating new set mhsub1
mhsub1=mh.drop_duplicates(subset=['lat','long'], keep="last").reset_index(drop=True)
mhsub1.shape# removing duplicate rows and creating new set mhsub1

(22, 4)

mhsub1.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

APTS FOR RENT (BLUE) AND SUBWAY STATIONS (RED)

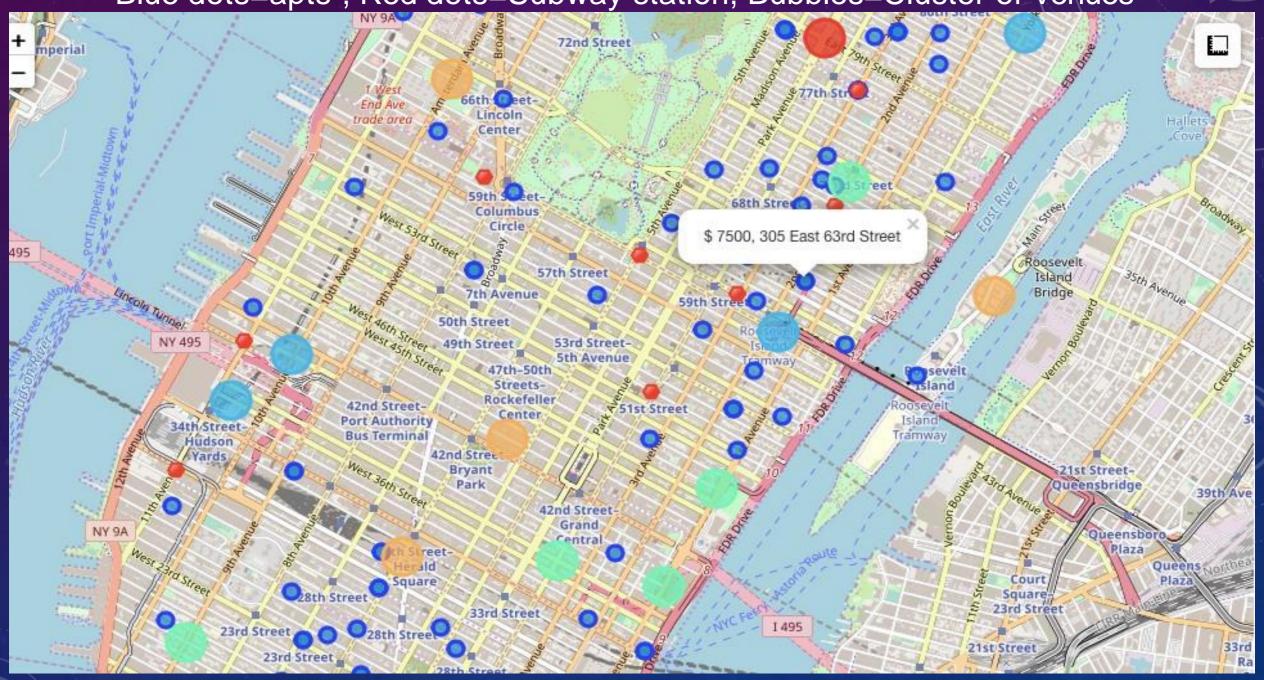


SELECTED APARTMENT

The ONE consolidated map shows all information for decision:

Apartments address, price, neighborhood, cluster of venues and subway station nearby.

Blue dots=apts, Red dots=Subway station, Bubbles=Cluster of Venues



APARTMENT SELECTION

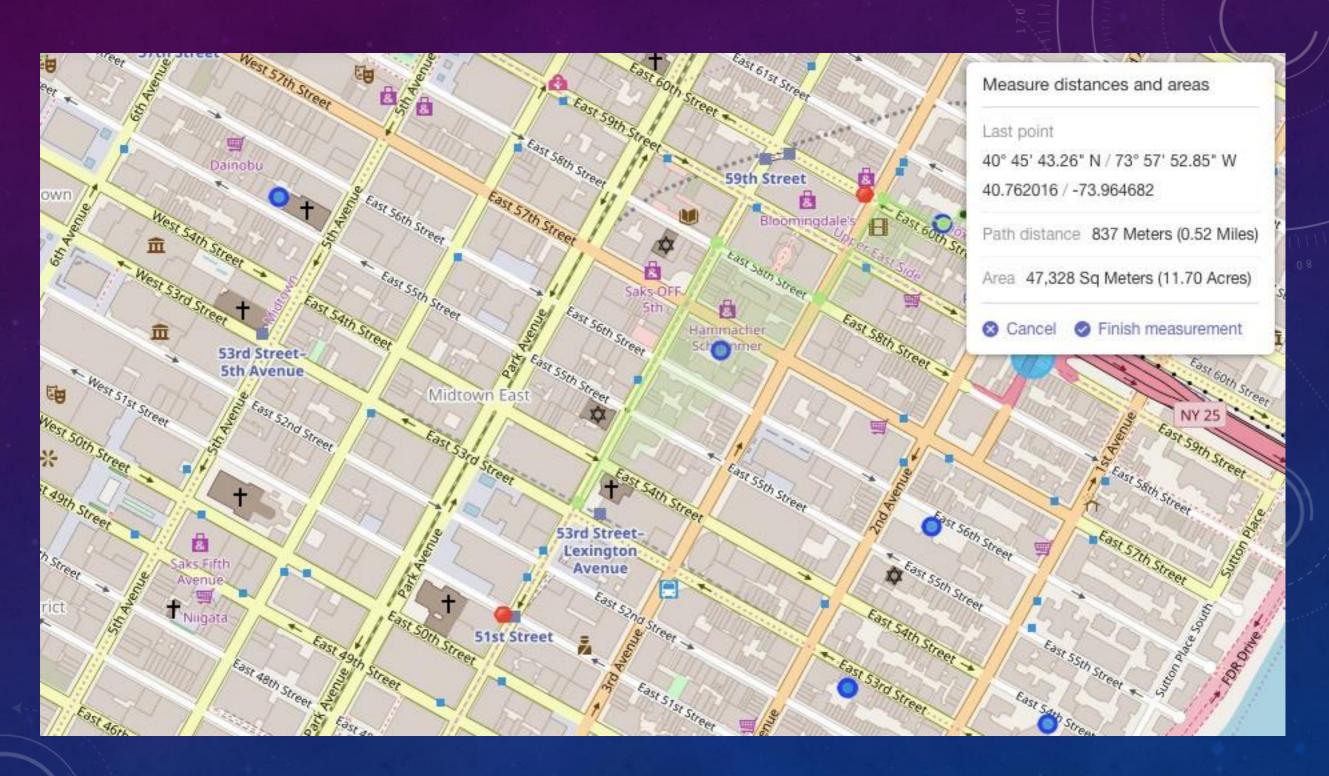
Using the "one map" above, I was able to explore all possibilities since the popups provide the information needed for a good decision.

Apartment 1 rent cost is US7500 slightly above the US7000 budget. Apt 1 is located 400 meters from subway station at 59th Street and work place (Park Ave and 53rd) is another 600 meters way. 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 US6935, just under the US7000 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 Indian 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.

WALK TO WORK HOME TO OFFICE IS LESS THAN 1 KM!



VENUS IN CLUSTER 2: MY FUTURE HOME

kk is the cluster number to explore

ck = 2

manhattan_merged.loc[manhattan_merged['Cluster Labels'] == kk, manhattan_merged.columns[[1] + list(range(5, manhattan_merged.shap

0	Neighborhood	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
0	Marble Hill	Coffee Shop	Discount Store	Yoga Studio	Steakhouse	Supplement Shop	Tennis Stadium	Shoe Store	Gym	Bank	Seafood Restaurant
1	Chinatown	Chinese Restaurant	Cocktail Bar	Dim Sum Restaurant	American Restaurant	Vietnamese Restaurant	Salon / Barbershop	Noodle House	Bakery	Bubble Tea Shop	Ice Cream Shop
6	Central Harlem	African Restaurant	Seafood Restaurant	French Restaurant	American Restaurant	Cosmetics Shop	Chinese Restaurant	Event Space	Liquor Store	Beer Bar	Gym / Fitness Center
9	Yorkville	Coffee Shop	Gym	Bar	Italian Restaurant	Sushi Restaurant	Pizza Place	Mexican Restaurant	Deli / Bodega	Japanese Restaurant	Pub
14	Clinton	Theater	Italian Restaurant	Coffee Shop	American Restaurant	Gym / Fitness Center	Hotel	Wine Shop	Spa	Gym	Indie Theater
23	Soho	Clothing Store	Boutique	Women's Store	Shoe Store	Men's Store	Furniture / Home Store	Italian Restaurant	Mediterranean Restaurant	Art Gallery	Design Studio
26	Morningside Heights	Coffee Shop	American Restaurant	Park	Bookstore	Pizza Place	Sandwich Place	Burger Joint	Café	Deli / Bodega	Tennis Court
34	Sutton Place	Gym / Fitness Center	Italian Restaurant	Furniture / Home Store	Indian Restaurant	Dessert Shop	American Restaurant	Bakery	Juice Bar	Boutique	Sushi Restaurant
39	Hudson Yards	Coffee Shop	Italian Restaurant	Hotel	Theater	American Restaurant	Café	Gym / Fitness Center	Thai Restaurant	Restaurant	Gym

DISCUSSION

- I am greatly indebted to Coursera and IBM for providing a wonderful platform to learn from basics to professional level
- I was able to learn different aspects of data science and machine learning from this course
- I was able to learn many new libraries in python and effectively utilize them
- The final project enlighten me and brought to a new level and changed the perspective of my thinking
- With these knowledge/skills I will continue to learn more about data science and try to acquire the practical knowledge

CONCLUSIONS

- I feel rewarded with the efforts, time spent to complete course with full satisfaction and I could cover all the topics well worthy of appreciation.
- The final project gave me apractical application to resolve aproblem in real situation using Data Science tools.
- The mapping with Folium is avery powerful technique to consolidate information and make the analysis and decision thoroughly and with confidence.
- One must keep abreast of new tools for DS that continue to appear for application in several technology/ business fields.
- I will try my level best to build my career as a data scientist.