# Coursera Capstone pro

Coursera IBM Data Science Certificati

## Report Content

- 1. Introduction Section:
- The "business problem" to be solved by this project and who may Data Section:

- Describe Data requirements and Sources needed to solve the pro-Methodology section:
- Main component of the report Execute data processing, describe exploratory data analysis and/or inferential statistical testing performachine 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.

### 1.0 Introduction

#### 1.1 Scenario and Background

I am currently living in Singapore, within walking distance to Downto MRT metro station". I also enjoy great venues and attractions, such cuisine, entertainment and shopping. I have an offer to move to work and I would like to move if I can find a place to live similar with similar

#### 1.2 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 subw Manhattan
- Venues and amenities as in my current residence.

#### 1.3 Interested Audience

I believe the methodology, tools and strategy used in this project is ror entity considering moving to a major city in US, Europe or Asia. Extinguished the Likewise, it can be helpful approach to explore the opening of a new of FourSquare data and mapping techniques combined with data an resolve the key questions arisen. Lastly, this project is a good practic developing Data Science skills.

#### 2.0 Data Section

#### 2.1 Data Requirements

- Geodata for current residence in Singapore with venues established using Foursq

- List of Manhattan (MH) neighborhoods with clustered venues established via Four Lab). https://en.wikipedia.org/wiki/List\_of\_Manhattan\_neighborhoods#Midtown\_neighborhoods#Midt

apartments for rent in Manhattan area with information on neighborhood location, a beds, area size, monthly rent price and complemented with geo data via Nominatim

<u>www.rentmanhattan.com/index.cfm?page=search&state=results</u> <u>https://www.nestpicity=new-</u>

- 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 Manhattan neighborhoods were obtained from Wikipedia and organized by Neigh via Nominatim for mapping with Folium.
- List of Subway stations was obtained via Wikipedia, NY Transit web site and Goog
- List of apartments for rent was consolidated from web-scraping real estate sites for (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 can visualize all details needed to make a selection of apartment

### 3.0 Methodology

The Strategy to find the answer:

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

#### The Tools:

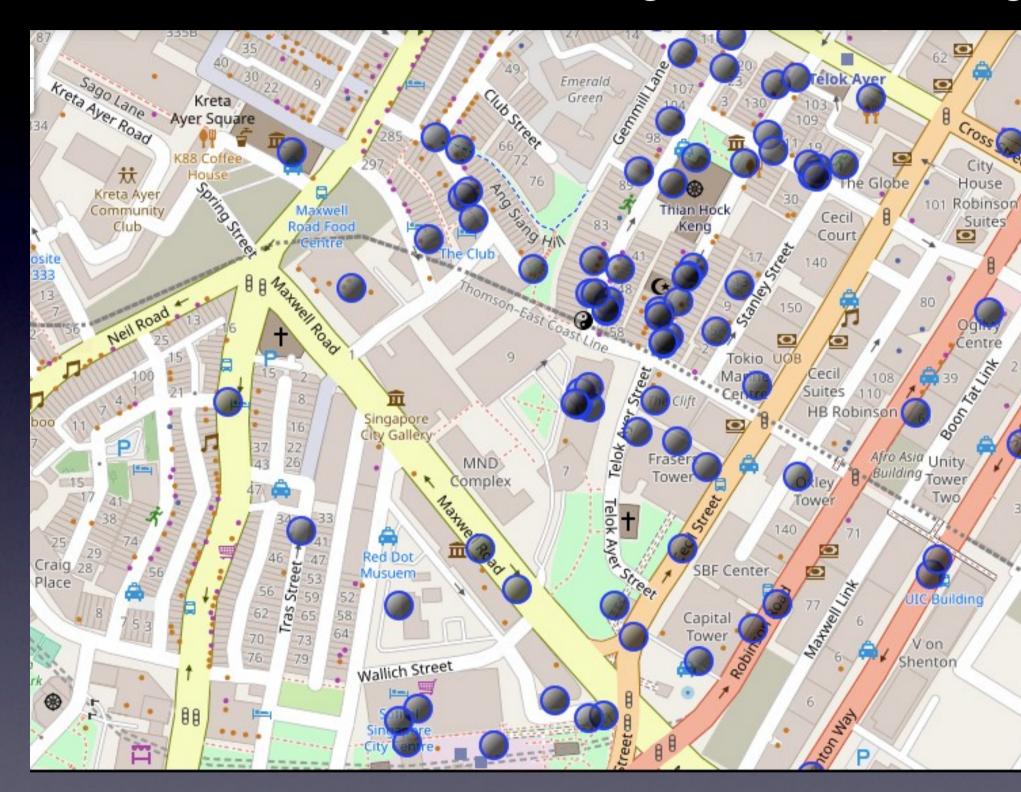
Web-scraping of sites is used to consolidate data-frame information as csv files for convenience and to simply the report. Geodata was a program to use Nominatim to get latitude and longitude of subway for each of (144 units) the apartments for rent listed.

Geopy\_distance and Nominatim were used to establish relative dist graphic was used for general statistics on rental data.

Maps with popups labels allow quick identification of location, price making the selection very easy

### 4.0 Execution and Results

#### Current residence Neighborhood in Sing

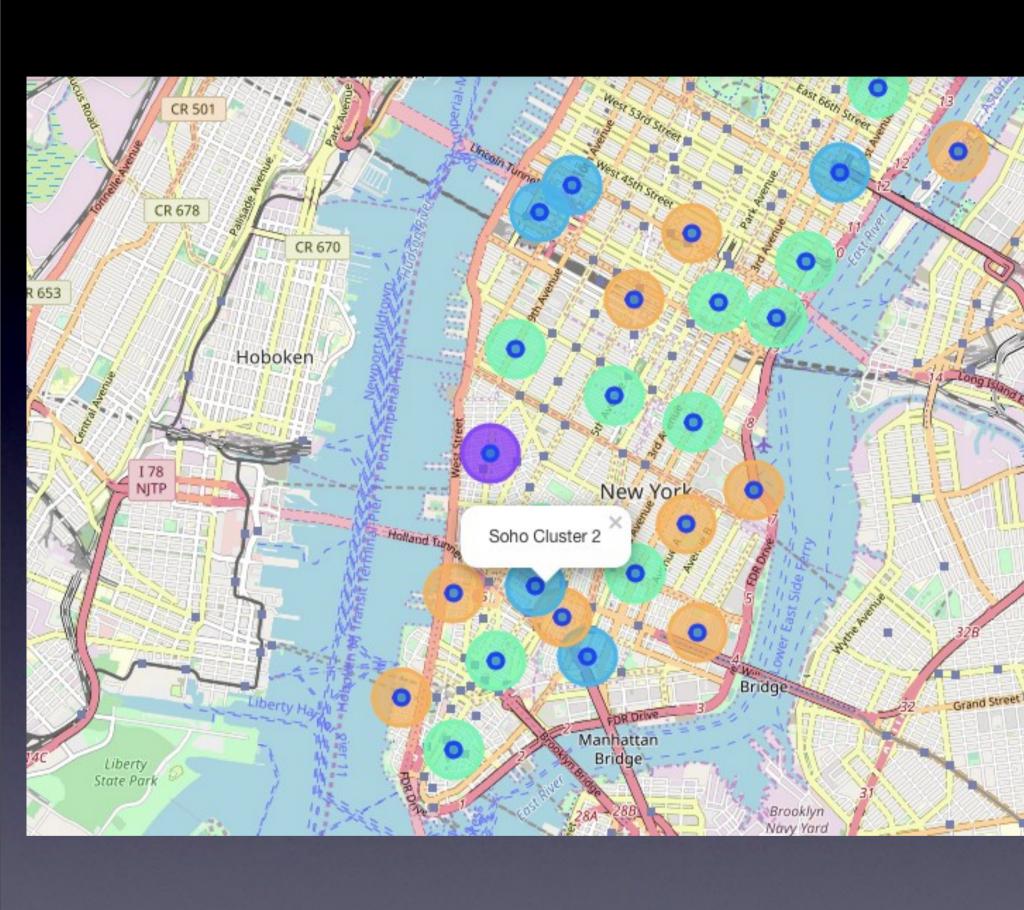


## Venues around Neighborhoo

# Venues near current Singapore residence place SGnearby\_venues.head(10)

	lat	categories	name	
103.847	1.279925	Wine Bar	Napoleon Food & Wine Bar	0
103.847	1.279872	Deli / Bodega	Park Bench Deli	1
103.846	1.280135	Cocktail Bar	Native	2
103.847	1.279175	Burrito Place	Muchachos	3
103.846	1.280462	Dessert Shop	Matt's   The Chocolate Shop	4
103.848	1.281254	Beer Garden	Freehouse	5
103.846	1.280468	Café	PS.Cafe	6
103.847	1.281345	Korean Restaurant	왕대박 Wang Dae Bak Korean BBQ Restaurant	7
103.847	1.280413	Massage Studio	Ancient Therapy	8
103.847	1.280479	Korean Restaurant	Oven & Fried Chicken	9

Manhattan Map - Neighborhoods and Cluste



## GeoData Manhattan apts for

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

Address West 105th Street	Area	Price_per_ft2	Rooms	Area-ft2	Rent Price	Lat	Lor
West 105th Street					No. of the last of		
TOOL TOOL OUTOOL	Upper West Side	2.94	5.0	3400	10000	40.799771	-73.9662
East 97th Street	Upper East Side	3.57	3.0	2100	7500	40.788585	-73.9552
West 105th Street	Upper West Side	1.89	4.0	2800	5300	40.799771	-73.9662
CARMINE ST.	West Village	3.03	2.0	1650	5000	40.730523	-74.0018
171 W 23RD ST.	Chelsea	3.45	2.0	1450	5000	40.744118	-73.99529
•	CARMINE ST.		CARMINE ST. West Village 3.03	CARMINE ST. West Village 3.03 2.0	CARMINE ST. West Village 3.03 2.0 1650	CARMINE ST. West Village 3.03 2.0 1650 5000	CARMINE ST. West Village 3.03 2.0 1650 5000 40.730523

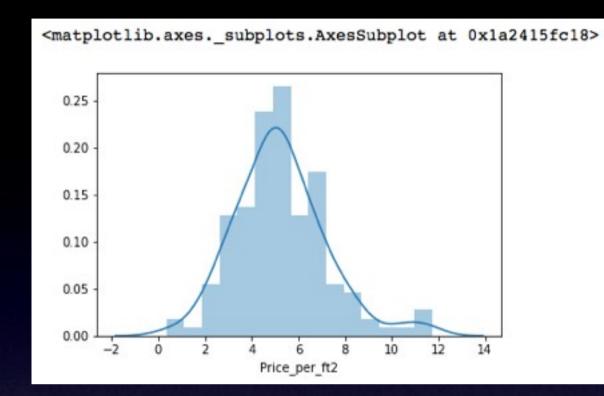
]: mh\_rent.tail()

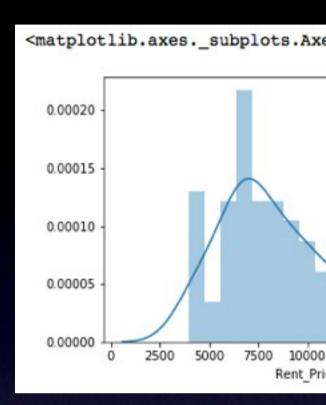
]:

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

#### Rental Price Statistics MH Apartme

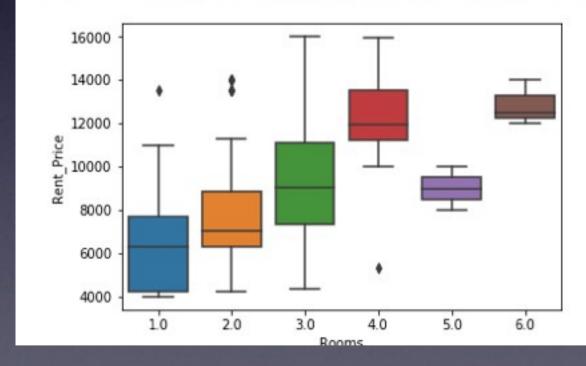
Budget US7000/month is around the mean



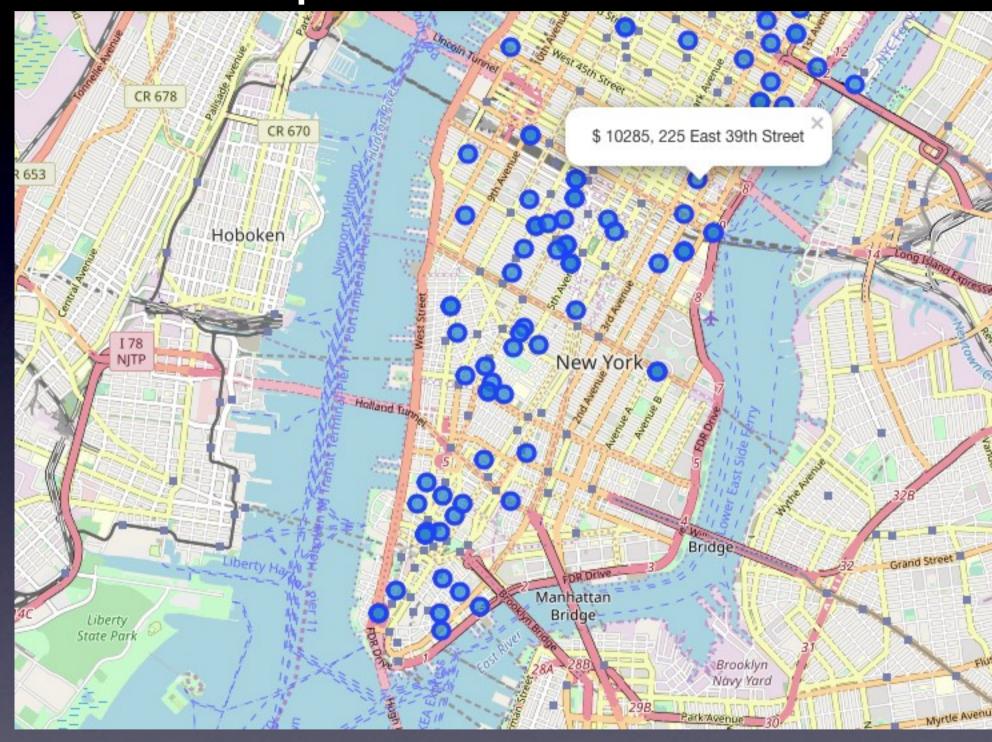




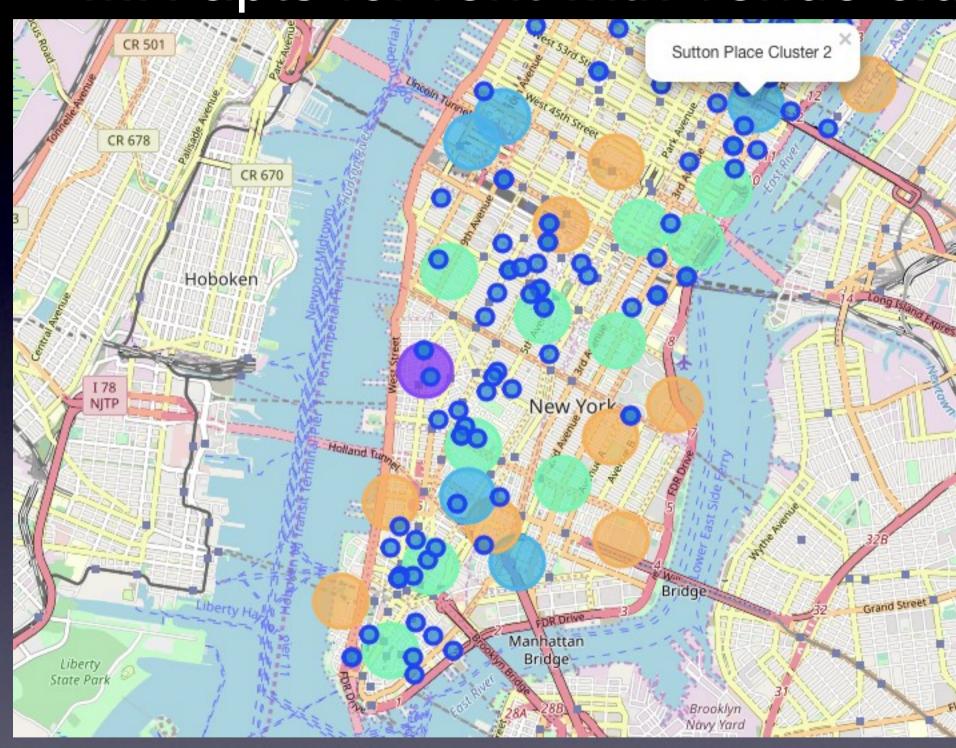
<matplotlib.axes.\_subplots.AxesSubplot at 0x1a25f2a2b0>



# Apartments for Rent in Mh



# MH apts for rent with venue clu



### Venues of cluster 3

## kk is the cluster number to explore

kk = 3

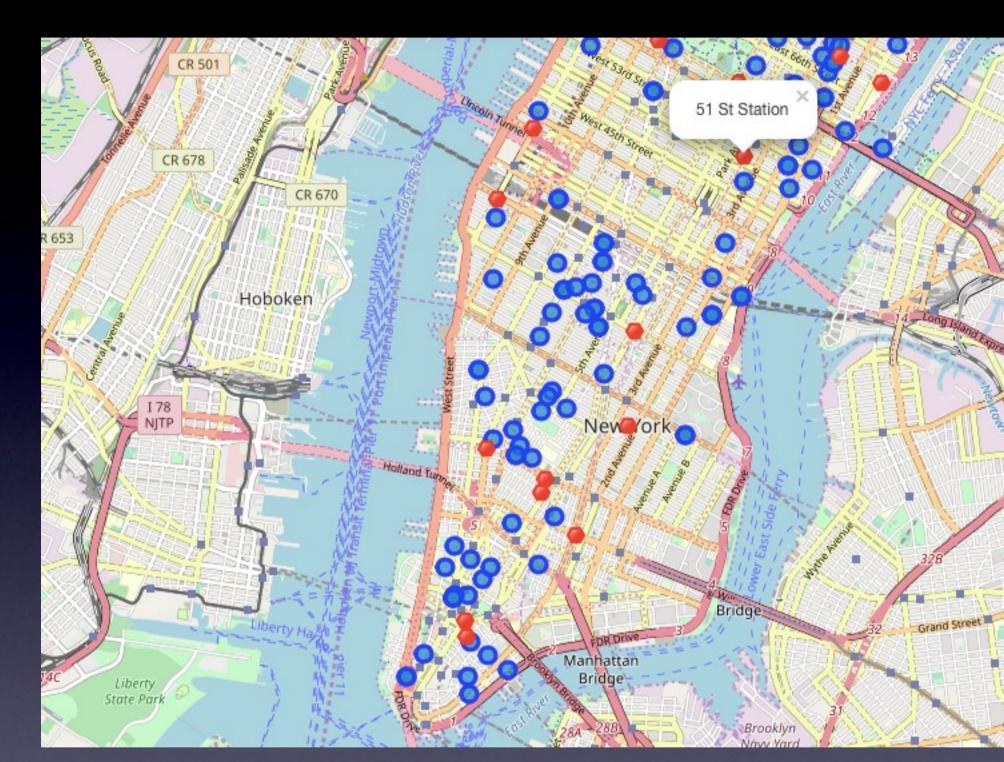
manhattan\_merged.loc[manhattan\_merged['Cluster Labels'] == kk, manhattan\_merged.columns[[]

	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	
3	Inwood	Mexican Restaurant	Lounge	Pizza Place	Café	Wine Bar	Bakery	American Restaurant	
5	Manhattanville	Deli / Bodega	Italian Restaurant	Seafood Restaurant	Mexican Restaurant	Sushi Restaurant	Beer Garden	Coffee Shop	ı
10	Lenox Hill	Sushi Restaurant	Italian Restaurant	Coffee Shop	Gym / Fitness Center	Pizza Place	Burger Joint	Deli / Bodega	
12	Upper West Side	Italian Restaurant	Bar	Bakery	Vegetarian / Vegan Restaurant	Indian Restaurant	Coffee Shop	Cosmetics Shop	
16	Murray Hill	Sandwich Place	Hotel	Japanese Restaurant	Gym / Fitness Center	Coffee Shop	Salon / Barbershop	Burger Joint	ı
17	Chelsea	Coffee Shop	Italian Restaurant	Ice Cream Shop	Bakery	Nightclub	Theater	Art Gallery	ı
18	Greenwich Village	Italian Restaurant	Sushi Restaurant	French Restaurant	Clothing Store	Chinese Restaurant	Café	Indian Restaurant	
27	Gramercy	Italian Restaurant	Restaurant	Thrift / Vintage Store	Cocktail Bar	Bagel Shop	Coffee Shop	Pizza Place	ı
29	Financial District	Coffee Shop	Hotel	Gym	Wine Shop	Steakhouse	Bar	Italian Restaurant	F
31	Noho	Italian Restaurant	French Restaurant	Cocktail Bar	Gift Shop	Bookstore	Grocery Store	Mexican Restaurant	

## Manhattan subway stations ged

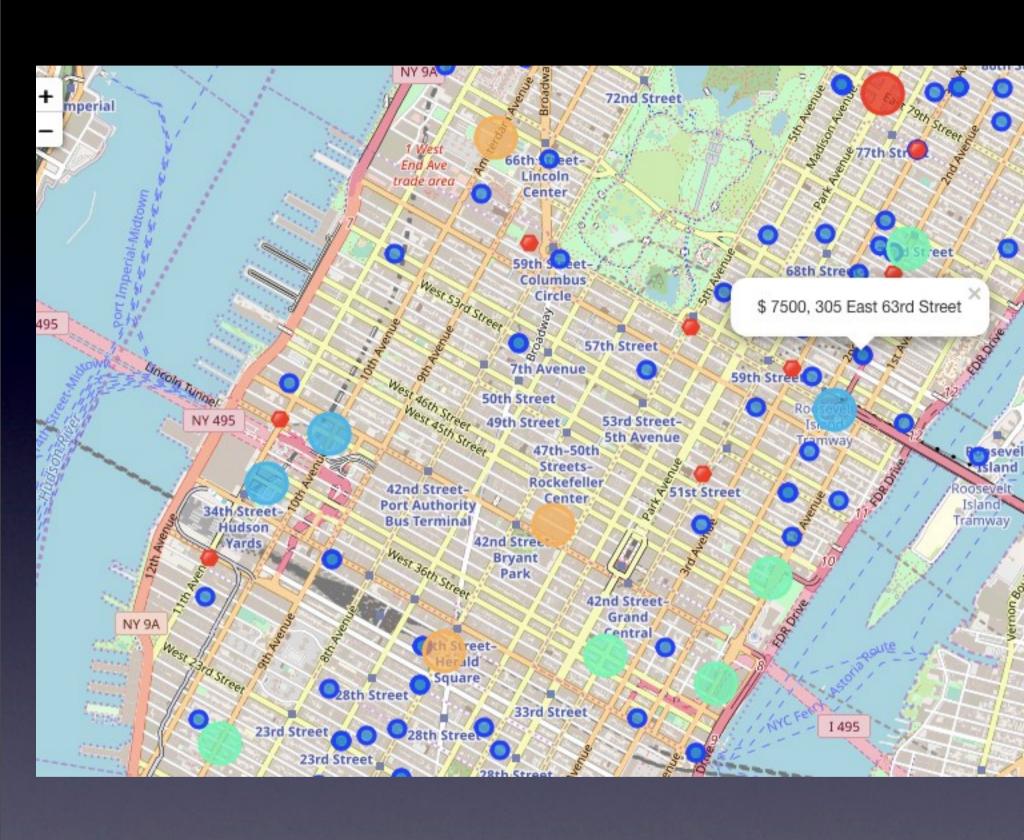
click to	scroll output; double click to hid	de sub_addr	ress	lat	long
0 0	yckman Street Subway Station	170 Nagle Ave, New York, NY 10034, L	USA 40.861	1857 -73.92	2450
1	57 Street Subway Station	New York, NY 10106, U	USA 40.764	1250 -73.95	5452
2	Broad St	New York, NY 10005, U	USA 40.730	0862 -73.98	3715
3	175 Street Station	807 W 177th St, New York, NY 10033, U	USA 40.847	7991 -73.93	3978
4	5 Av and 53 St	New York, NY 10022, U	USA 40.764	1250 -73.95	5452
mhsu mhsu (22,	abl=mh.drop_duplicate abl.shape 4)	s and creating new set mhs s(subset=['lat','long'], k		t").rese	t_i
mhsu mhsu (22,	ubl=mh.drop_duplicate ubl.shape			t").rese	t_i
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mhsu mhsu (22,	abl=mh.drop_duplicate abl.shape 4) subl.tail() sub_station 190 Street Subway Station	s(subset=['lat','long'], k  sub_address  Bennett Ave, New York, NY 10040, USA	lat 40.858113 40.762259	long -73.932983 -73.966271	<u>,</u>
mhsumhsumhsumhsumhsumhsumhsumhsumhsumhsu	abl=mh.drop_duplicate abl.shape 4) subl.tail() sub_station 190 Street Subway Station 59 St-Lexington Av Station	sub_address  Bennett Ave, New York, NY 10040, USA  E 60th St, New York, NY 10065, USA	lat 40.858113 40.762259 40.764250	-73.932983 -73.966271 -73.954525	3

Apts for rent (blue) and subway station



## Selected Apartment!

The ONE consolidated map shows all information for de Apartments address, price, neighborhood, cluster of venues and subvenue Blue dots=apts, Red dots=Subway station, Bubbles=Cluster



### **Apartment Selection**

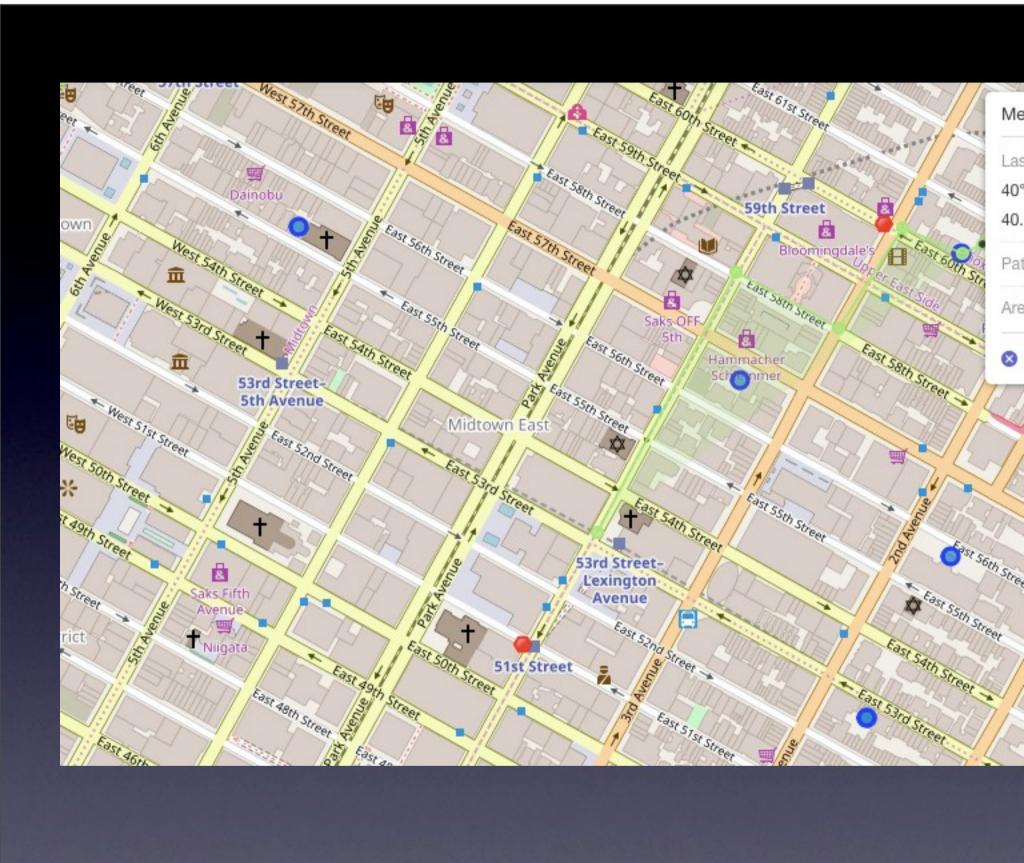
Using the "one map" above, I was able to explore all possibilities s provide the information needed for a good decision.

Apartment 1 rent cost is US7500 slightly above the US7000 budge 400 meters from subway station at 59th Street and work place (Pa another 600 meters way. I can walk to work place and use subway around. Venues for this apt are as of Cluster 2 and it is located in a East side of Manhattan.

Apartment 2 rent cost is US6935, just under the US7000 budget. A meters from subway station at Fulton Street, but I will have to ride work, possibly 40-60 min ride. Venues for this apt are as of Cluste

Based on current Singapore venues, I feel that Cluster 2 type of veresemblance to my current place. That means that APARTMENT since the extra monthly rent is worth the conveniences it provides.

I will walk to work Walk from home to work is less than 1 km!



### Venus in Cluster 2 near future

## kk is the cluster number to explore

kk = 2

manhattan\_merged.loc[manhattan\_merged['Cluster Labels'] == kk, manhattan\_merged.columns[[1

	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	
0	Marble Hill	Coffee Shop	Discount Store	Yoga Studio	Steakhouse	Supplement Shop	Tennis Stadium	Shoe Store	
1	Chinatown	Chinese Restaurant	Cocktail Bar	Dim Sum Restaurant	American Restaurant	Vietnamese Restaurant	Salon / Barbershop	Noodle House	
6	Central Harlem	African Restaurant	Seafood Restaurant	French Restaurant	American Restaurant	Cosmetics Shop	Chinese Restaurant	Event Space	Liq
9	Yorkville	Coffee Shop	Gym	Bar	Italian Restaurant	Sushi Restaurant	Pizza Place	Mexican Restaurant	Deli
14	Clinton	Theater	Italian Restaurant	Coffee Shop	American Restaurant	Gym / Fitness Center	Hotel	Wine Shop	
23	Soho	Clothing Store	Boutique	Women's Store	Shoe Store	Men's Store	Furniture / Home Store	Italian Restaurant	Medi R
26	Morningside Heights	Coffee Shop	American Restaurant	Park	Bookstore	Pizza Place	Sandwich Place	Burger Joint	
34	Sutton Place	Gym / Fitness Center	Italian Restaurant	Furniture / Home Store	Indian Restaurant	Dessert Shop	American Restaurant	Bakery	
39	Hudson Yards	Coffee Shop	Italian Restaurant	Hotel	Theater	American Restaurant	Café	Gym / Fitness Center	R

# 5.0 Discussion

- In general, I am positively impressed with organization, content and lab works pres the Coursera IBM Certification Course
- I feel this Capstone project presented monoportunity to practice and apply the Datools and methodologies learned. I have good project that I can present as an examy potential.

 I feel I have acquired a good starting poin a professional Data Scientist and I will co exploring to creating examples of practic

# 6.0 Conclusions

- I feel rewarded with the efforts, time and mo believe this course with all the topics covered worthy of appreciation.
- This project has shown me a practical applica a real situation that has impacting personal ar impact using Data Science tools.

- The mapping with Folium is a very powerful to consolidate information and make the analysi thoroughly and with confidence. I would recouse in similar situations.
- One must keep abreast of new tools for DS to to appear for application in several business f