# Coursera Capstone project

Coursera IBM Data Science Certification

### Report Content

- Introduction Section: discussion of the business and the interested audience in this project.
- Data Section: description of the data that will be used to solve the problem and the sources.
- Methodology section: discussion and description of exploratory data analysis carried out, any inferential statistical testing performed, and if any machine learnings were used establishing the strategy and purposes.
- Results section: discussion of the results.
- Discussion section: elaboration and discussion on any observations noted and any recommendations suggested based on the results.
- Conclusion section: report conclusion.

### Introduction

I am currently residing in Rome, within walking distance to Piazza Bologna metro station therefore I have access to good public transportation to work. Likewise, I enjoy many amenities in the neighborhood, such as international cousine restaurants, cafes, food shops and entertainment. I have been offered a great opportunity to work in Manhattan, NY.

#### 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</li>
- Amenities and venues similar to the ones described for current location
- 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

### Data Requirements:

- Geodata for current residence in Rome with venues established using Foursquare
- List of Manhattan (MH) neighborhoods with clustered venues established via Foursquare (https://en.wikipedia.org/wiki/List\_of\_Manhattan\_neighborhoods#Midtown\_neighborhoods)
- A list of Manhattan subway metro stops (https://en.wikipedia.org/wiki/List\_of\_New\_York\_City\_Subway\_stations\_in\_Manhattan and https://www.google.com/maps/search/manhattan+subway+metro+stations/@40.7837297,-74.1033043,11z/data=!3m1!4b1)
- List of apartments for rent in MH area with infos on neighborhood location, address, number of beds, area size, monthly rent price and complemented with geo data via Nominatim (http://www.rentmanhattan.com/index.cfm?page=search&state=results
  https://www.nestpick.com/search?city=new-york&page=1&order=relevance&district=manhattan&gclid=CjwKCAiAjNjgBRAgEiwAGLlf2hkP3A-cPxjZYkURqQEswQK2jKQEpv\_MvKcrlhRWRzNkc\_r-fGi0lxoCA7cQAvD\_BwE&type=apartment&display=list)
- Place to work in MH (Park Avenue and 53rd St) for reference

### Data section

Data sources, data processing and tools used

- Rome data and map is to be created with use of Nominatim, Foursquare and Folium mapping
- MH neighborhoods were obtained from Wikipedia and organized by Neighborhoods with geodata via Nominatim for mapping with Folium
- List of Subways 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 is based on mapping the described data, in order to facilitate the choice of at least two candidate places for rent. The info 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:

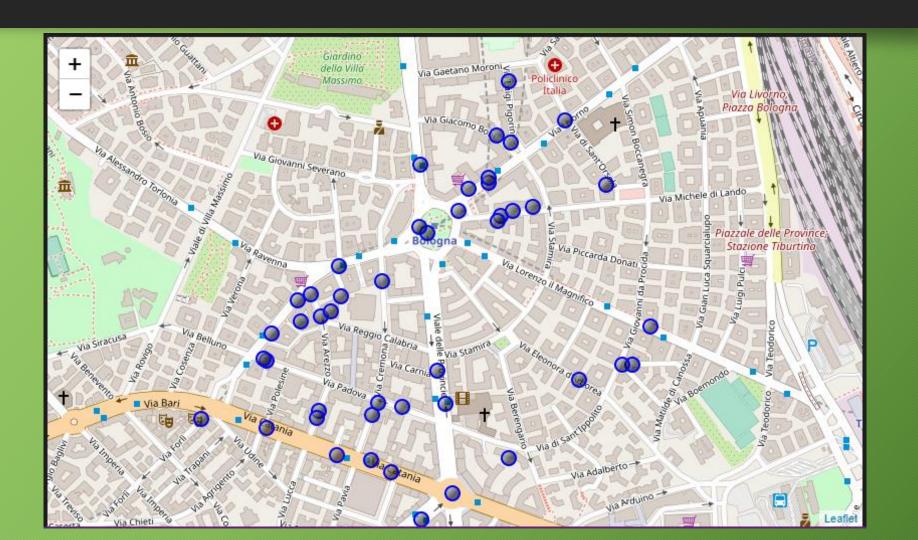
Web-scraping of sites is used to consolidate dataframe infos which was saved as csv files. Geodata was obtained by coding a program to use Nominatim to get lat and long of subway stations and of apartments for rent listed

Geopy\_distance and Nominatim were used to establish relative distances. Seaborn graphic was used for general stats on rental data.

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

## Execution and results

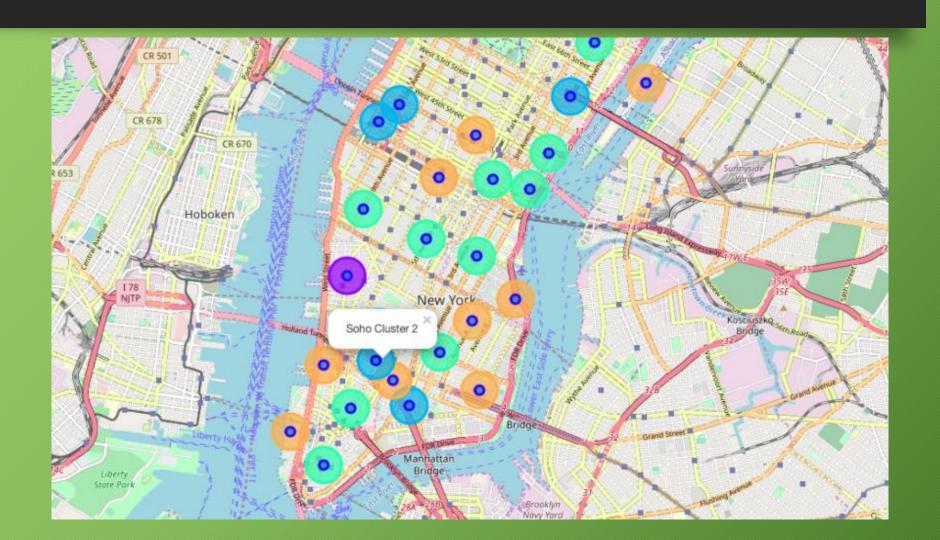
# Current residence in Rome



# Venues around Neighborhood in Rome

|   | name  | categories               | lat       | Ing       |
|---|---|--------------------------|-----------|-----------|
| 0 | Millimetri di Pizza                             | Pizza Place              | 41.914350 | 12.521968 |
| 1 | Mizzica   | Dessert Shop             | 41.912474 | 12.517887 |
| 2 | La Stiva  | Seafood Restaurant       | 41.911363 | 12.520865 |
| 3 | Coffee Pot                                      | Cocktail Bar             | 41.913817 | 12.522222 |
| 4 | Cus Cus   | Mediterranean Restaurant | 41.912225 | 12.518391 |
| 5 | Mercure Roma Piazza Bologna                     | Hotel                    | 41.912311 | 12.518598 |
| 6 | Villa Massimo                                   | Park                     | 41.914248 | 12.521542 |
| 7 | FLOUR farina & cucina (già Pane al Pane) kosher | Food Court               | 41.910857 | 12.519599 |
| 8 | Piazza Bologna                                  | Plaza                    | 41.913544 | 12.520654 |
| 9 | Pico Gelato                                     | Ice Cream Shop           | 41.917847 | 12.519930 |

# MH Map - Neighborhoods and cluster of venues



# GeoData MH apts for rent

```
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 |

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 |

# Rental price stats MH apts Budget \$7000/month is around the mean

12000

8000

6000

4000

1.0

2.0

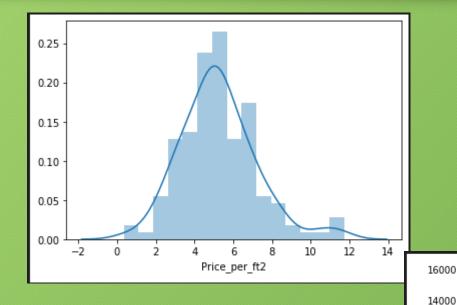
3.0

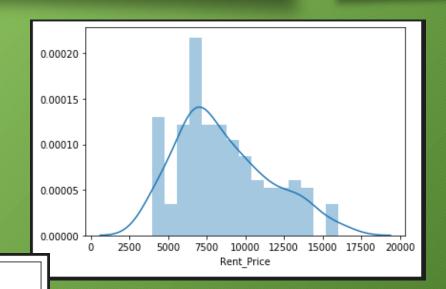
Rooms

5.0

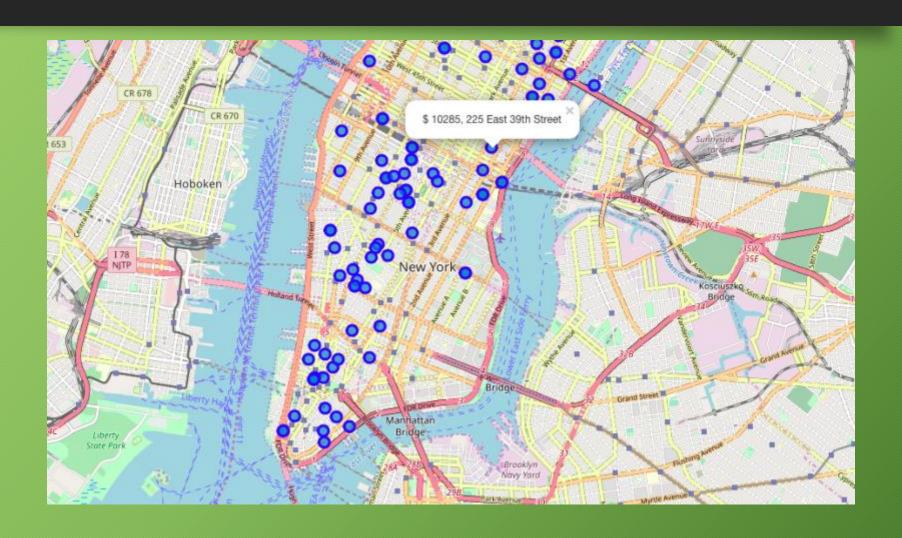
6.0

Rent Price

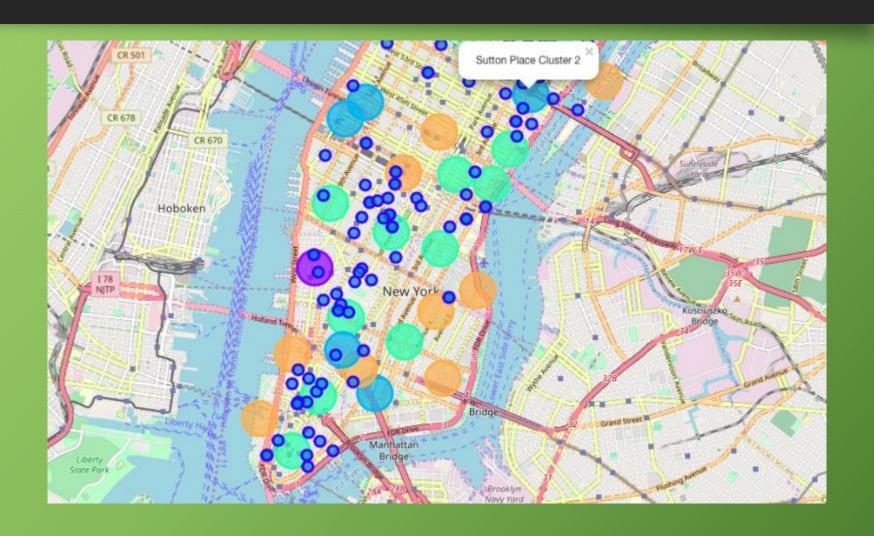




# Apts for rent in MH



# MH apts for rent with venue clusters



### Venues of cluster 3

Financial

District

Noho

29

31

Coffee Shop

Italian

Hotel

French

Gym

Cocktail Bar

Wine Shop

Gift Shop

## 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 10th Most 1st Most 2nd Most 3rd Most 4th Most 5th Most 6th Most 7th Most 8th Most 9th Most Neighborhood Common Venue Mexican American Frozen Spanish Pizza Place Café Wine Bar 3 Inwood Lounge Bakery Restaurant Restaurant Yogurt Shop Restaurant Deli / Italian Seafood Mexican Sushi Falafel Other Manhattanville Beer Garden Bike Trail Coffee Shop Nightlife Restaurant Restaurant Restaurant Bodega Restaurant Restaurant Gym / Fitness Sushi Italian Deli / Sporting Thai Lenox Hill Coffee Shop Pizza Place Burger Joint Gym 10 Goods Shop Restaurant Restaurant Center Bodega Restaurant Vegetarian / Sushi Upper West Italian Indian Cosmetics Mexican Wine Bar 12 Bar Bakery Vegan Coffee Shop Side Restaurant Restaurant Restaurant Shop Restaurant Restaurant Gym / Fitness Salon / Sandwich Japanese French Italian Murray Hill Hotel Coffee Shop **Burger Joint** Bar 16 Center Barbershop Restaurant Restaurant Place Restaurant Ice Cream Italian Seafood American Coffee Shop Nightclub Art Gallery Bakery Theater 17 Hotel Chelsea Restaurant Shop Restaurant Restaurant Sushi Greenwich Italian French Chinese Indian Seafood Electronics Café 18 Clothing Store Bakery Restaurant Village Restaurant Restaurant Restaurant Restaurant Store Restaurant Thrift / Mexican Grocery Italian Wine Shop Restaurant Cocktail Bar Pizza Place Gramercy Bagel Shop Coffee Shop 27 Restaurant Vintage Store Restaurant Store

Steakhouse

Bookstore

Italian

Restaurant

Mexican

Pizza Place

Hotel

Bar

Grocery Store

Gym / Fitness

Coffee Shop

Center

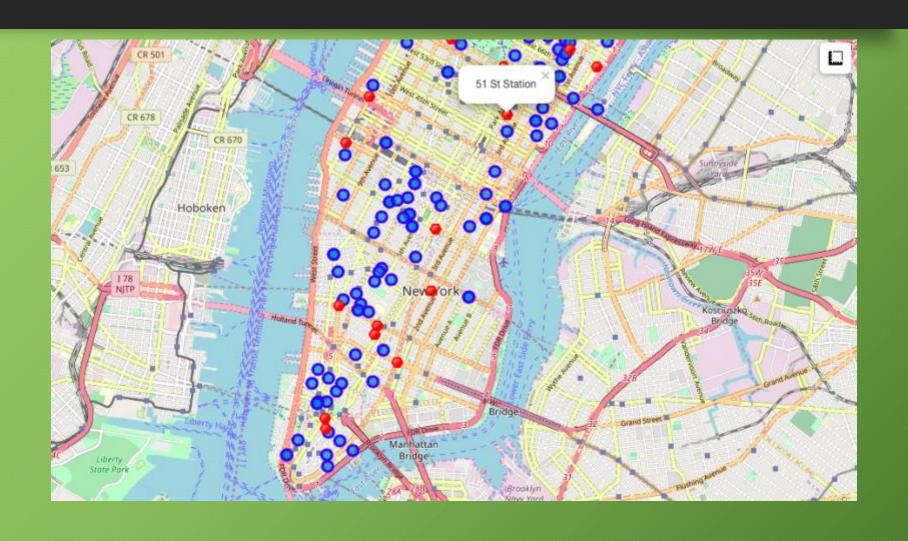
Park

Sushi

# MH subs geodata

| lick to                                  | scroll output; double click to hid   | e sub_addres  | ss                                      | lat I  | long |
|--|--|---|---|--|------|
| 0 [                                      | Dyckman Street Subway Station  | 170 Nagle Ave, New York, NY 10034, US   | A 40.861                                | 857 -73.924                                    | 4509 |
| 1  | 57 Street Subway Station   | New York, NY 10106, US  | SA 40.764                               | 250 -73.954                                    | 4525 |
| 2  | Broad St   | New York, NY 10005, US  | A 40.730                                | 862 -73.987                                    | 7156 |
| 3  | 175 Street Station   | 807 W 177th St, New York, NY 10033, US  | A 40.847                                | 991 -73.939                                    | 9785 |
| 4  | 5 Av and 53 St   | New York, NY 10022, US  | SA 40.7642                              | 250 -73.954                                    | 4525 |
| mhsi                                     | ubl.shape  | s(subset=['lat','long'], ke   | ep="last                                | ").reset                                       | t_in |
| mhsi<br>mhsi<br>(22)                     | ubl=mh.drop_duplicate  | s(subset=['lat','long'], ke   | ep="last                                | ").reset                                       | t_in |
| nhsi<br>nhsi<br>(22)                     | ubl=mh.drop_duplicate ubl.shape , 4)   | s(subset=['lat','long'], ked sub_address  | ep="last                                | long   |      |
| mhsumhsumhsumhsumhsumhsumhsumhsumhsumhsu | ubl=mh.drop_duplicate ubl.shape , 4) subl.tail() sub_station   |   | lat                                     | long   |      |
| mhsumhsumhsumhsumhsumhsumhsumhsumhsumhsu | ubl=mh.drop_duplicates ubl.shape , 4) subl.tail() sub_station 190 Street Subway Station  | sub_address   | lat<br>0.858113                         | long<br>-73.932983                             |      |
| mhsi<br>(22)<br>mhsi                     | abl=mh.drop_duplicates abl.shape  4)  subl.tail()  sub_station  190 Street Subway Station  59 St-Lexington Av Station                    | sub_address<br>Bennett Ave, New York, NY 10040, USA 4   | lat<br>0.858113<br>0.762259             | long<br>-73.932983<br>-73.966271               |      |
| mhs:<br>(22,<br>mh:                      | abl=mh.drop_duplicates abl.shape  4)  sub1.tail()  sub_station  190 Street Subway Station  59 St-Lexington Av Station  57 Street Station | sub_address<br>Bennett Ave, New York, NY 10040, USA 4<br>E 60th St, New York, NY 10065, USA 4 | lat<br>0.858113<br>0.762259<br>0.764250 | long<br>-73.932983<br>-73.966271<br>-73.954525 |      |

# Apts for rent (blue) and subs stations (red)

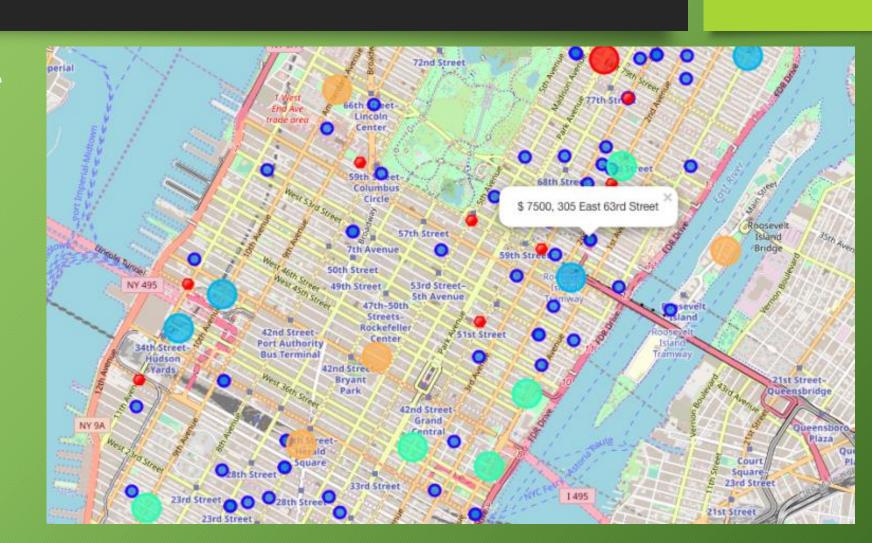


## Selected Apt

The consolidated map shows all the <u>informations used</u> for the decision:

- Apts address
- Price
- Neighborhood
- Cluster of venues
- Sub station nearby

Blue dots = apts Red dots = subs Bubble = cluster of venues



### **APTS** selection

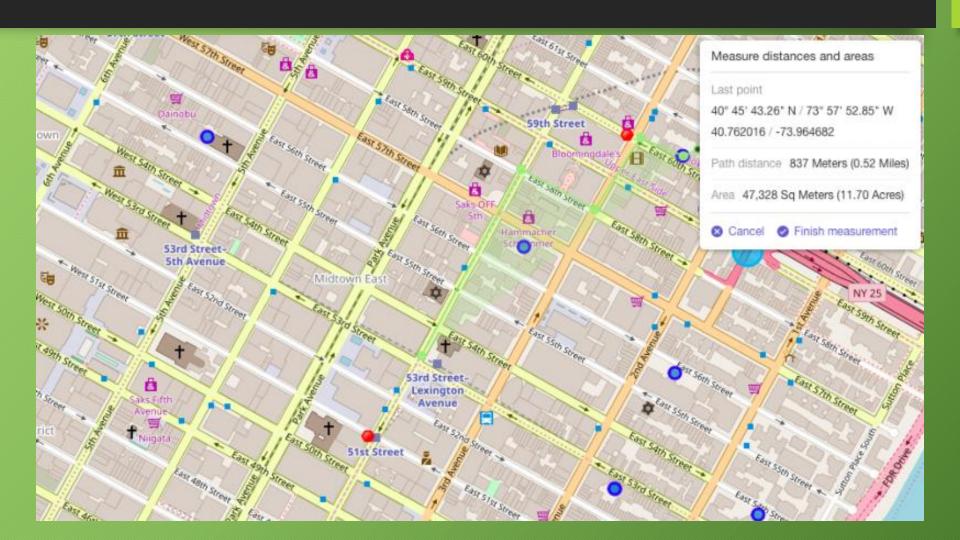
Using the map above I was able to explore all the possibilities since the popups provide the info needed for a good decision.

Apt #1: rent cost is \$7500, slightly above the budget. It is located 400 m from subway station at 59th Street and work place is another 600 m away. Venues for this apt are as of Cluster 2 and it is located in a fine district in the East side of MH

Apt #2: rent cost is \$6935, slightly under the budget. It is located 60 m from subway station at Fulton Street but to go to work place it will need 40-60 min subway ride. Venues for this apt are as of Cluster 3.

Based on current Rome venues, Cluster 2 type of venues is a closer to my current living place, consequently Apt #1 is a better choice: extra monthly rent is worth the conveniences it provides.

# I will walk to work, home-work distance is less than 1 km



### Venues in Cluster 2 near future home

## kk is the cluster number to explore kk = 2 manhattan merged.loc[manhattan merged['Cluster Labels'] == kk, manhattan merged.columns[[1] + list(range(5, manhattan merged) 1st Most 10th Most 2nd Most 3rd Most 4th Most 5th Most 6th Most 7th Most 8th Most 9th Most Common Neighborhood Common Common Common Common Common Common Common Common Common Venue Supplement Tennis Seafood Discount Coffee Shop Yoga Studio Shoe Store Marbie Hill Steakhouse Gym Bank Store Shop Stadium Restaurant Chinese Dim Sum Vietnamese Salon / Bubble Tea Ice Cream American Noodle Chinatown Cocktall Bar Bakery Restaurant House Restaurant Restaurant Restaurant Barbershop Shop Shop African Seafood French Cosmetics Gym / Fitness American Chinese **Event Space** Beer Bar Central Harlem Liquor Store Restaurant Restaurant Restaurant Restaurant Shop Restaurant Center Italian Sushi Mexican Japanese Bar Yorkville Coffee Shop Gym Pizza Place Deli / Bodega Pub 9 Restaurant Restaurant Restaurant Restaurant Gym / American Italian Indie Theater 14 Clinton Theater Coffee Shop Fitness Hotel Wine Shop Spa Gym Restaurant Restaurant Center Clothing Women's Furniture / Italian Mediterranean Men's Store 23 Soho Boutique Shoe Store Art Gallery Design Studio Store Store Home Store Restaurant Restaurant Morningside American Sandwich Deli. Coffee Shop Park Café 26 Bookstore Pizza Place Burger Joint Tennis Court Heights Restaurant Place Bodega Gym / Italian Furniture / Indian American Sushi Sutton Place Dessert Shop Bakery Juice Bar Fitness Boutique Restaurant Restaurant Home Store Restaurant Restaurant Center Gym / Italian American Thai Hudson Yards Coffee Shop Hotel Theater Café Fitness Restaurant Gvm Restaurant Restaurant Restaurant Center

### Discussion

In general, I am positively impressed with the overall organization, 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. I have created a good project that I can present as an example to show my potential. 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, time 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.

One must keep abreast of new tools for DS that continue to appear for application in several business fields.