

# THE BATTLE OF NEIGHBOURHOODS

Applied Data Science Capstone project(IBM)



# CONTENTS

- 1.Introduction (Business Problem)
- 2.Data
- 3.Procedure
- 4.Analysis
- 5.Result and Discussion
- 6.Inference



# 1. INTRODUCTION (BUSINESS PROBLEM)

According the latest NYC&Company release New York City welcomed about 65.2 million tourists in 2018 year - 51.6 million domestic and 13.5 million international visitors. And these numbers are continuously increasing from year to year -

The data set-[https://en.wikipedia.org/wiki/Tourism\\_in\\_New\\_York\\_City](https://en.wikipedia.org/wiki/Tourism_in_New_York_City)

New York City has the largest selection of lodging choices in the country – from the hostels to the luxury hotels. The prices vary from 100\$ till several thousand dollars with average price 292 USD per night.

The Hotel Occupancy rate is also high – in 2018 year it was 88% as seen by the data set given in the below link [https://assets.simpleviewinc.com/simpleview/image/upload/v1/clients/newyorkcity/FYI\\_Hotel\\_reports\\_February\\_2019\\_8607015b-b32a-4c7f-9fbd-84cd2a93cbe6.pdf](https://assets.simpleviewinc.com/simpleview/image/upload/v1/clients/newyorkcity/FYI_Hotel_reports_February_2019_8607015b-b32a-4c7f-9fbd-84cd2a93cbe6.pdf)

Visitors often prefer short stays that are often over weekends - averaging 2.4 nights. This is shown in the link below [https://aka.nyc/content/uploads/2017/12/new\\_york\\_city\\_travel\\_and\\_tourism\\_trend\\_report\\_2017.pdf](https://aka.nyc/content/uploads/2017/12/new_york_city_travel_and_tourism_trend_report_2017.pdf)



## ***Problem Description***

New York city has around 300 hotels with over 75,000 rooms. Airbnb has more than 50,000 apartment listings in New York City as of the year 2018. Due to this it can be hard to find the right fit or to know what one can afford. In this project we will try and find out the most optimal borough in Manhattan where tourists can rent an accommodation through Airbnb service. This will ensure that the tourist has a pleasant stay in New York and a possibility to be near the most visited attractions like Central Park, Times Square etc.

## ***Target Audience***

This analysis will interest visitors or tourists of New York city who prefers short stays and want to select the best neighbourhoods in Manhattan, New York.



## ***Success Criteria***

The success criteria of this project is a recommendation with the set of apartments clusters that have the best score calculated based on

- Cost of accommodation
- Location of accommodation
- Attractions in 100m radius of the accommodation
- Crime rate in a radius of 100m from the accommodation



## 2. DATA

For this analysis we are using free and publicly available datasets.

We will find available Airbnb accommodations in Manhattan, New York and define the most affordable apartment clusters for tourists.

The factors that will contribute to this analysis are:

- Average price of accommodation in the neighbourhood
- Number of tourist attractions near the accommodation
- Crime rate near the accommodation



The datasets for the above listed factors are available on the links given below

- Apartment listing in New York City from *Inside Airbnb* :  
<http://data.insideairbnb.com/united-states/ny/new-york-city/2019-12-04/data/listings.csv.gz>
- Tabulation of New York neighbourhoods :  
<https://data.cityofnewyork.us/api/geospatial/cpf4-rkhq?method=export&format=GeoJSON>
- New York Crime data-  
<https://data.cityofnewyork.us/api/views/5uac-w243/rows.csv?accessType=DOWNLOAD>
- Foursquare API to extract data about venues -museums, restaurants, art galleries, shopping complexes, sightseeing attractions, concert stadiums etc.

### **Airbnb data**

The records were filtered by below given criteria for our project:

- Brough- Manhattan, New York
- Number of reviews  $\geq 10$
- Last review- later than 10/01/2019
- Minimum nights  $\geq 1$
- Availability  $\geq 10$  days/year
- Exclude shared rooms, hostels and RV



## Data Cleaning

### Needed Column Description:

- id: Listing identifier
- name: Accommodation's name
- last\_review: Date accommodation was last reviewed
- listing\_url: Accommodation's url
- picture\_url: Accommodation's picture url
- neighbourhood\_group\_cleansed: New York's borough names e.g Manhattan, Bronx etc. (We will be using accommodations only from Manhattan)
- neighbourhood\_cleansed: Airbnb neighbourhood name (They are not the same as NYC neighbourhood tabulation areas)
- review\_score\_rating: Accommodation's weighted sum of other scores like location
- latitude: Accommodation location's latitude
- longitude: Accommodation location's longitude





- `property_type`: Accommodation's type eg. Private room, entire apartment
- `room_type`: Accommodation's room type
- `bathrooms`: Number of bathrooms
- `bedrooms`: Number of bedrooms
- `square_feet`: Accommodation's area
- `price`: price per night
- `security_deposit`: security deposit amount
- `cleaning_fee`: Additional fee
- `minimum_nights`: Minimum nights for rent
- `number_of_reviews_ltm`: Number of review for the last month
- `reviews_per_month`: Average number of reviews per month
- `number_of_reviews`: Overall number of reviews
- `availability_365`: Available days/year



We should clean different Prices columns :

fill in empty values; convert String to Float, e.g. \$2,100.00 => 2100.00.

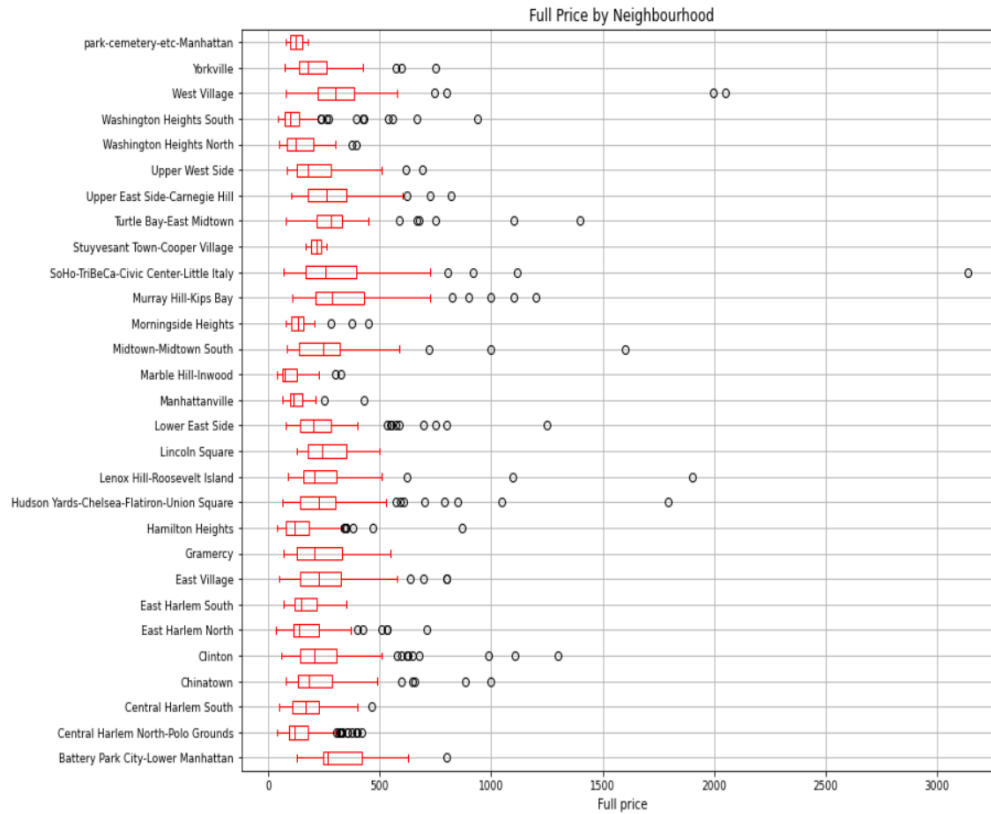
We clean different price columns:

- Fill in empty values
- Convert string to float eg. \$2,100.00- 2,100.00



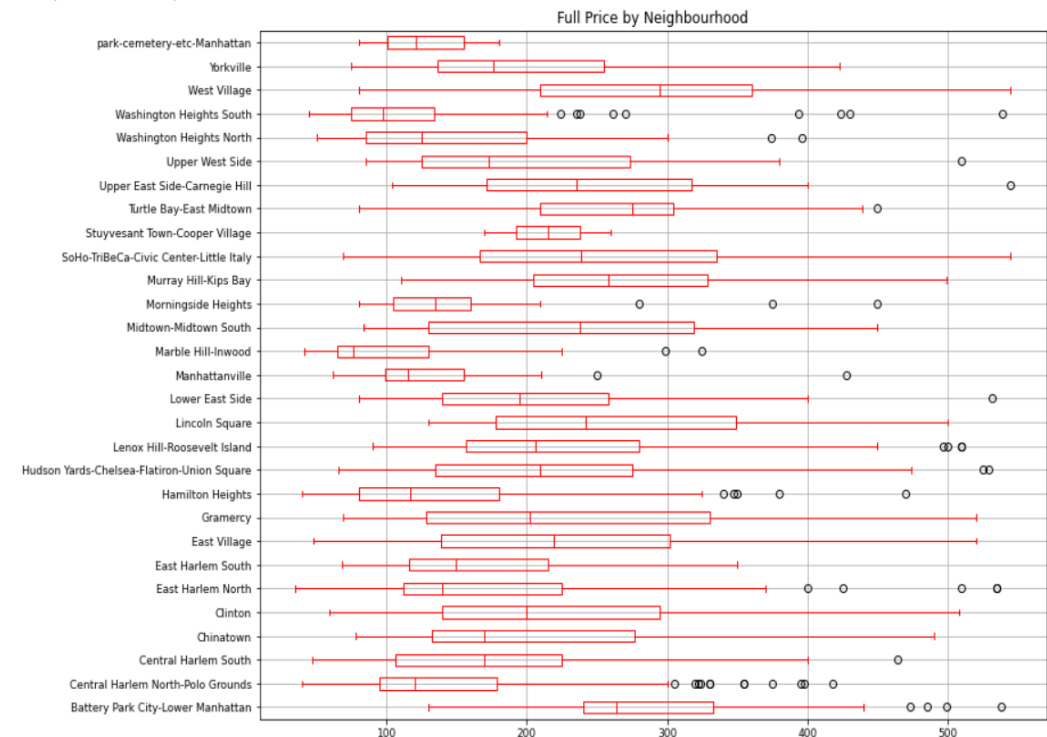
# OUTLINERS

↳ Outliers, also,



In the chart on left we can see some outliers - observation points that are distant from other observations. In our case they are the indication of variance in the data. So we remove them.

Text(0.5, 0.98, '')



# 3. PROCEDURE

In the first step we collected the following data:

- Airbnb Accommodations with their NYC Tabulation Area (official neighbourhood names)
- Airbnb Accommodation's number of crimes nearby
- Defined NYC Tabulation Area (official neighbourhood name) for each Manhattan's crime case.

The second step in our analysis will be a calculation and exploration different neighbourhoods of Manhattan. We will explore the following characteristics:

- number of crimes in the area
- average price per person
- number of accommodations available.



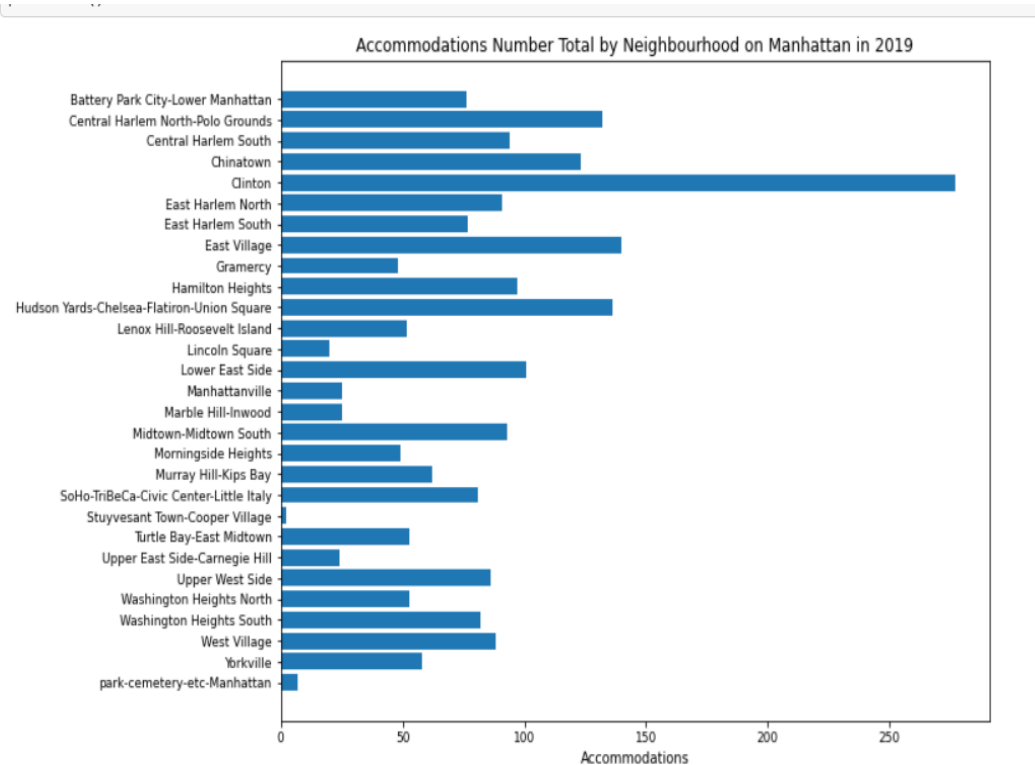
In third step we select Top-100 Airbnb accommodations based on:

- Summary rating, number of crimes and price per person
- Invoke Foursquare API to find Top accommodations' nearby venues create

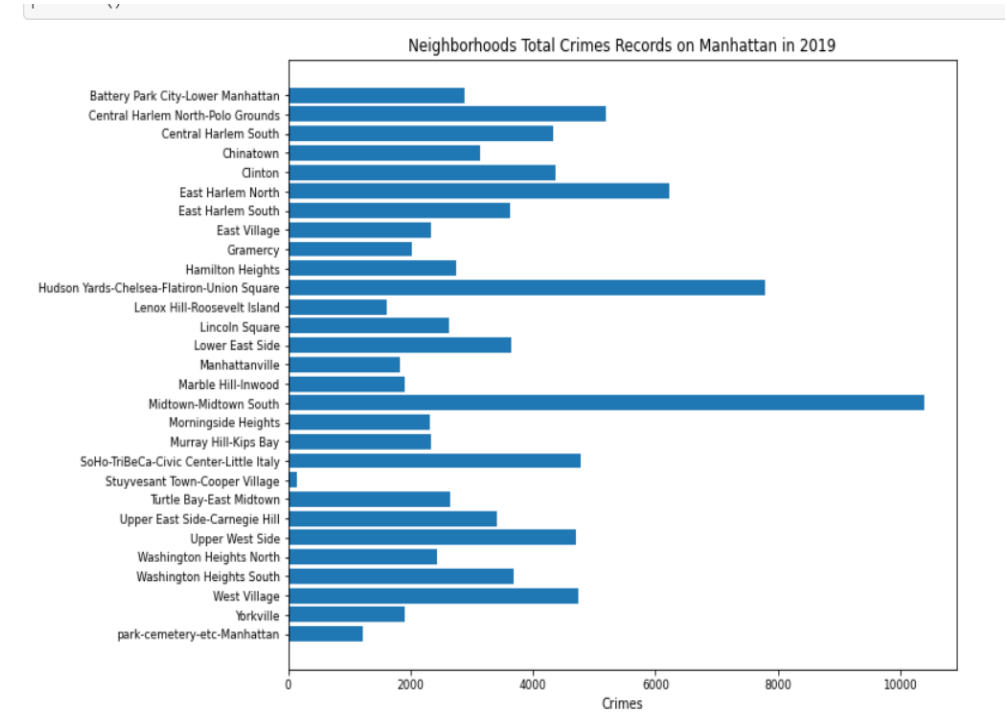


# 4. ANALYSIS

## Apartment Total by Neighbourhood

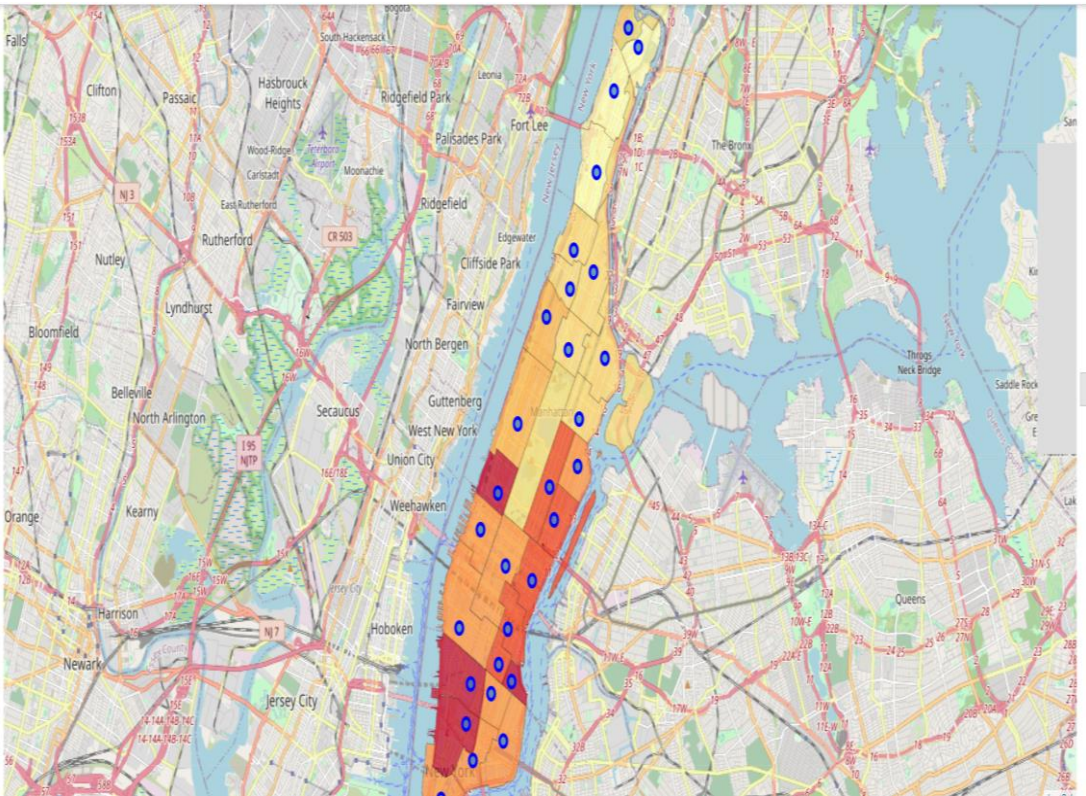


## Neighbourhood Crime Records

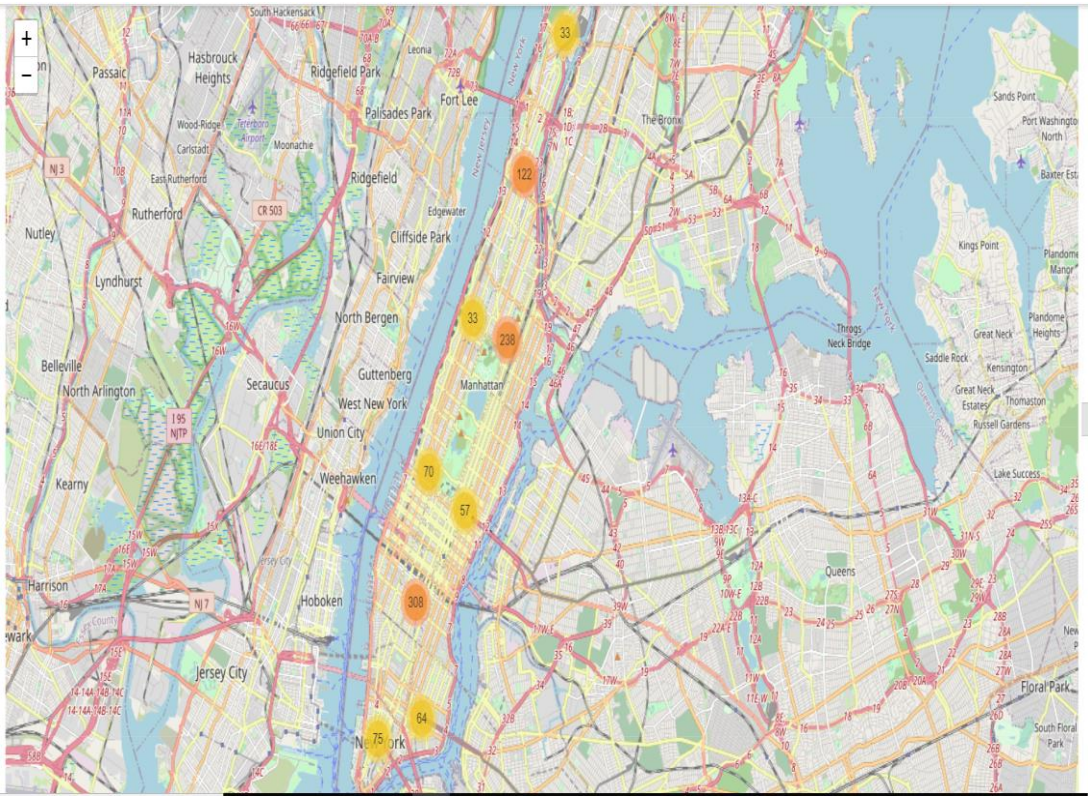




# Analysis Map



# Crimes Cluster Map



### Top-5 Neighbourhoods with the Highest Crime level:

- Midtown-Midtown South - 10,397
- Hudson Yards-Chelsea-Flatiron-Union Square - 7,788
- East Harlem North - 6,221
- Central Harlem North-Polo Grounds - 5,186
- SoHo-TriBeCa-Civic Center-Little Italy - 4,789

### Top-5 Neighbourhoods with the Lowest Crime level in 2019 year :

- Stuyvesant Town-Cooper Village - 145
- park-cemetery-etc-Manhattan - 1,213
- Lenox Hill-Roosevelt Island - 1,604
- Manhattanville - 1,832
- Yorkville - 1,898





# Foursquare API Neighbourhoods Analysis

We choose Top accommodations by

- review\_scores\_rating - overall accommodations rating - from maximum 100 to lower values
- full\_price - from lower price to higher
- price\_per\_person - from lower price to higher
- crimes - from lower number to higher

|   | name   | latitude | longitude | Venue              | Venue Latitude | Venue Longitude | Venue Category |
|---|--|----------|-----------|--------------------|----------------|-----------------|----------------|
| 0 | Private Bedroom in Cozy Hamilton Heights Apartment | 40.82749 | -73.94461 | Branson Got Juice! | 40.825525      | -73.943681      | Juice Bar      |
| 1 | Private Bedroom in Cozy Hamilton Heights Apartment | 40.82749 | -73.94461 | Harlem Public      | 40.828803      | -73.948685      | Bar            |
| 2 | Private Bedroom in Cozy Hamilton Heights Apartment | 40.82749 | -73.94461 | Buddha Taco Bar    | 40.829007      | -73.948089      | Bar            |
| 3 | Private Bedroom in Cozy Hamilton Heights Apartment | 40.82749 | -73.94461 | Monkey Cup         | 40.825694      | -73.947234      | Coffee Shop    |
| 4 | Private Bedroom in Cozy Hamilton Heights Apartment | 40.82749 | -73.94461 | The Honeywell      | 40.828549      | -73.948739      | Speakeasy      |

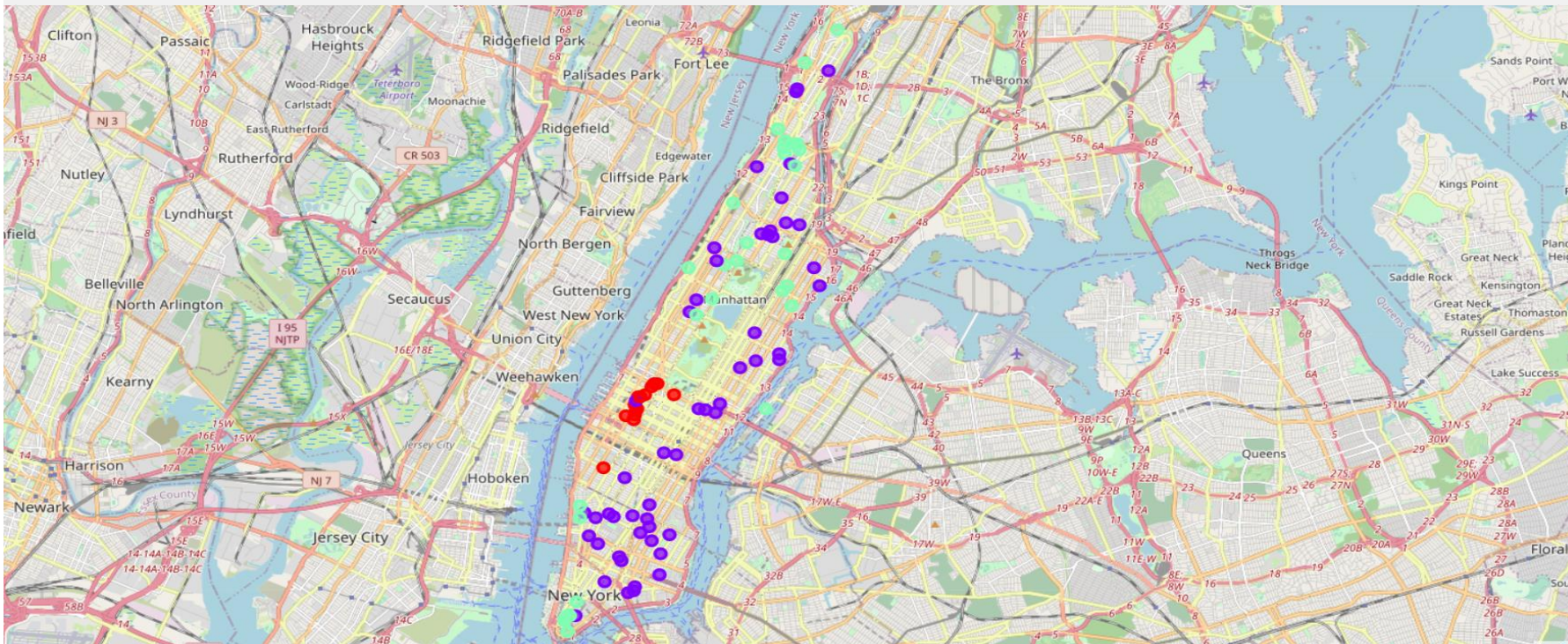
## Custom top level categories

|    | name   | latitude | longitude | Venue                        | Venue Latitude | Venue Longitude | Venue Category       |
|----|--|----------|-----------|------------------------------|----------------|-----------------|----------------------|
| 5  | Private Bedroom in Cozy Hamilton Heights Apartment | 40.82749 | -73.94461 | Wat's On Your Plate          | 40.825632      | -73.948494      | Caribbean Restaurant |
| 6  | Private Bedroom in Cozy Hamilton Heights Apartment | 40.82749 | -73.94461 | Barepas                      | 40.827717      | -73.946173      | Arepa Restaurant     |
| 7  | Private Bedroom in Cozy Hamilton Heights Apartment | 40.82749 | -73.94461 | Bono Trattoria               | 40.830180      | -73.947420      | Italian Restaurant   |
| 8  | Private Bedroom in Cozy Hamilton Heights Apartment | 40.82749 | -73.94461 | Convent Garden               | 40.828560      | -73.942531      | Garden               |
| 12 | Private Bedroom in Cozy Hamilton Heights Apartment | 40.82749 | -73.94461 | Riverside Drive Running Path | 40.827738      | -73.942124      | Scenic Lookout       |



# K-means Clustering

|   | Cluster Labels | name   | 1st Most Common Venue | 2nd Most Common Venue | 3rd Most Common Venue | 1st Most Common Venue Share | 2nd Most Common Venue Share | 3rd Most Common Venue Share |
|---|----------------|--|-----------------------|-----------------------|-----------------------|-----------------------------|-----------------------------|-----------------------------|
| 0 | 0              | **Stylish, Quiet, Centrally Located (9th & 52nd) | Food Place            | Entertainment         | Sightseeing           | 0.64                        | 0.36                        | 0                           |
| 1 | 2              | 157-C  | Food Place            | Sightseeing           | Fine Art              | 0.58                        | 0.25                        | 0.12                        |
| 2 | 2              | A neat bedroom in a cozy 3-bedroom apartment     | Sightseeing           | Food Place            | Shopping              | 0.45                        | 0.45                        | 0.03                        |
| 3 | 1              | Art filled peaceful paradise EV Union Square     | Food Place            | Entertainment         | Sightseeing           | 0.67                        | 0.17                        | 0.08                        |
| 4 | 1              | Artsy Parisian Apt in Greenwich Village          | Food Place            | Entertainment         | Sightseeing           | 0.65                        | 0.25                        | 0.1                         |



# 5. RESULT & DISCUSSION

The analysis defined 3 clusters All clusters have a 'Food Place' category as the first Common Venues. This is common among the clusters defines clusters. But they are distinguished by the other characteristics as

- Average price per person
- Average crimes Rate
- The second common Venues
- Number of available Airbnb accommodations
- Neighbourhood location



**Cluster 0** - This is the smallest cluster with the following particular qualities (Nightclub, Stadium, Pub, Theatre, Concert and so on):

- highest average price\_per\_person among all clusters - \$111
- highest average crimes rate among all clusters - 102
- is 1st and the 2nd Top Common Venue Categories
- contains 15% from all top accommodations (Top-100 Airbnb accommodations)
- spreads on Chelsea, Hell's Kitchen, and Midtown Airbnb's Neighbourhoods.

**Cluster 1** : It contains

- average price\_per\_person - \$110
- average crimes rate - 67
- mix of all Venue Categories
- contains 58% from all accommodations selected from analysis (Top-100 Airbnb accommodations)
- spreads almost on all Manhattan's areas.



## **Cluster 2** Sightseeing is the cheapest one with many Sightseeing attractions nearby

- lowest average price\_per\_person - \$59
- lowest crimes rate among all clusters - 65
- Sightseeing is the second top Common Venue Category
- contains 27% from all top accommodations (Top-100 Airbnb accommodations)
- spreads on East Harlem, Financial District, Harlem, Inwood, Morningside Heights, Roosevelt Island, Upper West Side, Washington Heights, West Village.
- We identified three clusters from which a visitor could choose an appropriate accommodation based on his/her preferences or needs.





## Top Neighborhoods Statistics

Top-5 Manhattan Tab Area (Airbnb Neighborhoods) with Lowest Average Price per Person in 2019 year:

- Marble Hill-Inwood (Marble Hill, Inwood)- 45.48 USD - 25 accommodations
- Washington Heights South (Washington Heights)- 46.79 USD - 82 accommodations
- Washington Heights North (Inwood, Washington Heights)- 54.74 USD - 53 accommodations
- Central Harlem North-Polo Grounds (Harlem, East Harlem)- 57.00 USD - 132 accommodations
- Manhattanville (Harlem)- 59.75 USD - 25 accommodations

Top-5 Manhattan Tab Area (Airbnb Neighborhoods) with the Lowest Crime level in 2019 year :

- Stuyvesant Town-Cooper Village (Stuyvesant Town) -145
- park-cemetery-etc-Manhattan (Inwood, Washington Heights) - 1,213
- Lenox Hill-Roosevelt Island (Roosevelt Island, Upper East Side) - 1,604
- Manhattanville (Harlem) - 1,832
- Yorkville (Upper East Side) - 1,898



## Note

- Our analysis was limited to Manhattan Brugh's only
- Foursquare developer account has a limitation of 950 calls/day so maybe it's worth to upgrade our free account to analyze Top-1000 Airbnb accommodations instead of Top-100.



# 6. CONCLUSION

The basic data analysis was performed to identify Manhattan's Neighborhoods clusters for a short stay visit. During the analysis, we cleaned and investigated Manhattan Neighborhoods' datasets, found some statistical characteristics and visualize them.

The aim of this project is to help Manhattan visitors select the Airbnb neighborhoods where to stay based on the most common venues, price policy, and safety characteristics:

- If a person is interested in Entertainment (Nightlife, Pubs, Concerts, Movies) we recommend paying attention for accommodations from the Cluster 0 - \* Entertainment: Chelsea, Hell's Kitchen, and Midtown Airbnb's Neighborhoods. But the person should take into the consideration the high prices and crime rate for this location
- if a person is looking for a neighborhood with lower prices and nice views nearby we recommend looking at Cluster 2 - Sightseeing: Chelsea, Hell's Kitchen, and Midtown Airbnb's Neighborhoods
- if a person does not have any preferences - investigate proposals from Cluster 1 - Mix. It has average prices and spreads over almost all Manhattan's neighborhoods.





# THANK YOU

Author: Ieesha Deshmukh

