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Coursera Capstone Project

Introduction where you discuss the business problem and who would be interested in this project.

The idea of this study is to help my cousins plan a vacation in New York and choose an Airbnb listing based on a criterion of preferences for entertainment, food and other factors which I will define later in this report. My cousins are curious about Manhattan and Brooklyn, so I will scope into these areas for listings that best suits their interests while considering price and location data. My mission is to pinpoint the most happening venues around Airbnb listings to understand what locals from that area are into and what possible events they can attend to get a taste of the culture and vibes in a potential vacation spot. This will allow me to gain a better understanding of Manhattan and Brooklyn and recommend a few listings that meet my cousins needs and wants.

Some of my cousin's requirements include interests around coffee shops, different cuisines of foods, in the Brooklyn and Manhattan area. I will analyze these neighborhoods to recommend a few locations that might be interesting to book an Airbnb around.

Data where you describe the data that will be used to solve the problem and the source of the data.

The two data sources I used to conduct my research and analysis were a Kaggle Airbnb dataset as well as the foursquare API to extract venues in the Brooklyn and Manhattan area. These are the two main data sources I will be using to conduct this locational analysis.

The Airbnb dataset from Kaggle includes important data points such as price, latitude, longitude, borough and neighborhood. I will use this information to extract venues in the locations mentioned above from Foursquare and cross reference the top 10 venues. These will allow me to understand what types of popular venues are in these areas and specific neighborhoods. The kaggle data set can be found here:

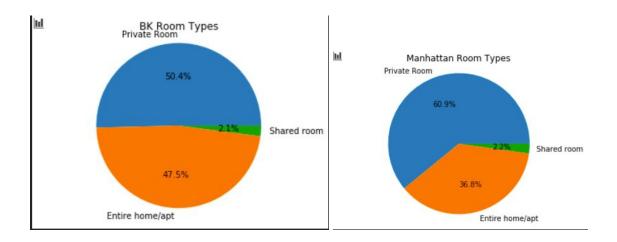
https://www.kaggle.com/dgomonov/new-york-city-airbnb-open-data

Methodology section which represents the main component of the report where you discuss and describe any exploratory data analysis that you did, any inferential statistical testing that you performed, if any, and what machine learnings were used and why.

My first step was to perform exploratory data analysis on the Airbnb dataset as well as the venue information from Foursquare. I found out that the average Airbnb in Manhattan is approximately \$70 more expensive than Brooklyn.

neighbourhood_group Brooklyn 124.38 Manhattan 196.88

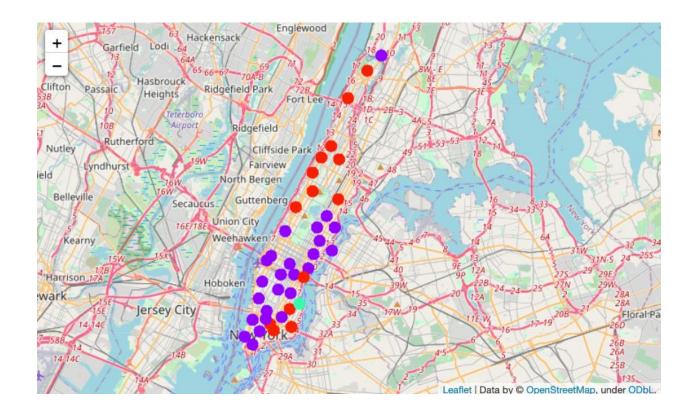
The breakdown of room types are shown below for Brooklyn and Manhattan respectively.



As seen in the two pie charts, there are slightly more private rooms in Manhattan. My cousins mentioned that they are looking for a private room (hotel and apartment) or an entire home/apartment. Basically they are not looking to share any rooms with other parties. I inferred that I will have an abundance of airbnbs to choose from as 90% of the Airbnb listings are private rooms or an entire home in Brooklyn and Manhattan.

After this basic exploratory data analysis, I performed a K-means clustering algorithm model in order to cluster the neighborhoods in both Manhattan and Brooklyn to understand the types of venues in each neighborhood of the cities. I clustered the airbnb listings into 3 clusters for Manhattan and Brooklyn based on distance and the top 10 venue categories that were in the

neighborhoods. The cluster analysis map is shown below for the listings in Brooklyn and Manhattan.





Results section where you discuss the results.

In this section I will dive into two neighborhoods in each Manhattan and Brooklyn that have the most frequent venues of coffee shops, diverse restaurants and shopping. In Manhattan, I have scoped into Chelsea and Greenwich Village neighborhoods. In Brooklyn, I have scoped into the neighborhoods of Williamsburg and Kensington. The top 10 most common venues in Chelsea, Greenwich Village, Williamsburg and Kensington are shown below in dataframes.

| Neighborh | 1st Mo ood Comm Ven | ion Com | | d Most 4 ¹ Common Venue | th Most Common Venue | 5th Most Common Venue | 6th Most Common Venue | Co | Most mmon enue | 8th Most Common Venue | C | Most 10th Common (Venue |
|-------------------|---------------------------|-------------------------------|-----------------------------|--|-----------------------------|-----------------------------|-----------------------------|-----------------------------------|----------------------------|--------------------------------|--------------------------------|--------------------------------|
| Greenw: Villa | | | | Sushi aurant | Café _F | American Restaurant | Coffee Shop | | dian Irant | Boutique | Cosm | etics Ch Shop Resta |
| Cluster Labels | Neighborhood | 1st Most d Common Venue | 2nd Most Common Venue | Commo | n Commo | n Commo | n Com | lost 7 [.] mon nue | th Most Common Venue | 8th Most Common Venue | 9th Most Common Venue | Most Common |
| 3 1 | Chelsea | a Coffee Shop | Bakery | , America Restauran | | | | | French taurant | Ice Cream Shop | Wine Shop | |
| Cluster Labels | 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 M Com Ve | mon Co | Most ommon Venue | 10th Most Common Venue |
| 10 1 | Kensington | Thai Restaurant | Grocery Store | Ice Cream Shop | Pizza Place | Sandwich Place | Donut Shop | Gas Station | Japan Restaur | | eli / odega | Furniture / Home Store |
| | | | | A . I W I | 411 | | | 7.1. 11 | A.I. H. | 76 VALUE | 24 13 17 | |

Event Japanese Tapas Space Restaurant Restaurant

Cluster Labels Neighborhood

1 Williamsburg

Coffee

Pizza Place

Pet Store

Bagel Shop

The top 10 most common venues in Chelsea are shown above which includes coffee shops, bakery, american restaurant, art gallery, Italian restaurant, seafood restaurant, French restaurant, Ice Cream shop, wine shop and hotel. The top 10 most common venues in Greenwich are Italian, Clothing store, Sushi, Cafe, American Restaurant, Coffee Shop, Indian Restaurant, Boutique, cosmetics shop and Chinese restaurant. The top 10 most common venues in Kensington are Thai restaurant, grocery store, ice cream shop, pizza place, sandwich shop, donut shop, gas station, Japanese restaurant, Deli/Bodega, and Home stores. The top 10 most common venues in Williamsburg are bagel shop, coffee shop, pizza place, pet store, event space, Japanese restaurant, Tapas restaurant, clothing store, steakhouse and Latin American restaurant. After conversations with my cousins, at least 75% of these venues are of interest to them. There is a variety of cuisines in these neighborhoods as well as a bunch of dessert places, cafes and shopping venues. After applying the clustering algorithm I was able to group listings based on location and recommend ten Airbnbs based on a price range of 100-200\$ and number of reviews being greater than 10. This constraint is to ensure that the listings have been used before by other individuals and are safe to purchase. In the next section, I will dig into the recommendations made per neighborhood.

Discussion section where you discuss any observations you noted and any recommendations you can make based on the results.

As displayed in the last section of my Jupyter notebook, I have created a list of recommendations for each neighborhood based on the requirements:

- 1. price is between \$100 \$200
- 2. Number of reviews is greater than 10 reviews

| | name | neighbourhood_group | neighbourhood | room_type | price | minimum_nights | number_of_reviews r |
|-------|---|---------------------|---------------|--------------------|-------|----------------|---------------------|
| 3960 | New, spacious 1BD in Williamsburg | Brooklyn | Williamsburg | Entire home/apt | 102 | 7 | 26 |
| 9225 | HUGE Entire Flr. Private Entry/Bath | Brooklyn | Williamsburg | Private room | 103 | 100 | 128 |
| 17247 | Luxe & Spacious Williamsburg Studio | Brooklyn | Williamsburg | Entire home/apt | 105 | 5 | 11 |
| 8441 | Wburg 1 bedroom – Spacious & Sunny | Brooklyn | Williamsburg | Entire home/apt | 105 | 5 | 33 |
| 16903 | Sunny Williamsburg w Balcony – Long Term Rental | Brooklyn | Williamsburg | Entire home/apt | 105 | 15 | 30 |
| 14755 | Apartment in Williamsburg – Long term lease | Brooklyn | Williamsburg | Entire home/apt | 105 | 180 | 10 |
| 27357 | Sun Drenched Williamsburg 1 BR Prime Location | Brooklyn | Williamsburg | Entire home/apt | 107 | 4 | 17 |
| 21971 | *PRIVATE APARTMENT L TRAIN LORIMER STOP | Brooklyn | Williamsburg | Entire home/apt | 108 | 28 | 17 |
| 33946 | Charming Room with 2 Beds+ Laundry in Williams | Brooklyn | Williamsburg | Private room | 109 | 30 | 43 |
| 19762 | Apartment with backyard in East Williamsburg | Brooklyn | Williamsburg | Entire home/apt | 109 | 4 | 91 |

| | name | neighbourhood_group | neighbourhood | room_type | price | minimum_nights | number_of_reviews | rev |
|-------|---|---------------------|----------------------|--------------------|-------|----------------|-------------------|-----|
| 24305 | Large, Queen Bedroom in heart of Greenwich Vil | Manhattan | Greenwich Village | Private room | 106 | 8 | 25 | |
| 5305 | Sunny, Charming Boho Apt Near all in the Village! | Manhattan | Greenwich Village | Private room | 115 | 25 | 70 | |
| 18277 | ◆ of Greenwich Village ~ Walk/Transit Score 100 | Manhattan | Greenwich Village | Private room | 115 | 5 | 65 | |
| 1856 | Clean cozy room central historic gay best area | Manhattan | Greenwich Village | Private room | 120 | 21 | 79 | |
| 1943 | Cute Studio Greenwich Village NYC | Manhattan | Greenwich Village | Entire home/apt | 120 | 30 | 17 | : |
| 6835 | Stylish Apartment with Spacious Private Room | Manhattan | Greenwich Village | Private room | 120 | 4 | 19 | |
| 7364 | CHARMING NYC Studio Loft NEAR SOHO! | Manhattan | Greenwich Village | Entire home/apt | 136 | 30 | 43 | |
| 14735 | Luxurious SOHO Apartment | Manhattan | Greenwich Village | Entire home/apt | 142 | 30 | 10 | |
| 2512 | Dream Apartment – Greenwich Village | Manhattan | Greenwich Village | Entire home/apt | 150 | 12 | 16 | |
| 26492 | SOHO / GREENWICH VILLAGE\nPrime location- 1 Be | Manhattan | Greenwich Village | Entire home/apt | 150 | 80 | 18 | |

| name | neighbourhood_group | neighbourhood | room_type | price | minimum_nights | number_of_reviews re |
|---|---------------------|---------------|--------------------|-------|----------------|----------------------|
| Modern, Spacious Hideaway in Historic Brooklyn | Brooklyn | Kensington | Entire home/apt | 110 | 6 | 12 |

| | пате | neighbourhood_group | neighbourhood | room_type | price | minimum_nights | number_of_reviews |
|-------|---|---------------------|---------------|--------------------|-------|----------------|-------------------|
| 21623 | Relaxing Stylish Room near Train and Attractions | Manhattan | Chelsea | Private room | 103 | 5 | 77 |
| 5921 | Spacious room with view in Chelsea | Manhattan | Chelsea | Private room | 110 | 6 | 20 |
| 11226 | Perfect 8th Avenue Apt in Chelsea | Manhattan | Chelsea | Private room | 112 | 5 | 31 |
| 34171 | Cozy & Private Bedroom near Highline- CHELSEA | Manhattan | Chelsea | Private room | 115 | 55 | 11 |
| 2515 | Room in Great Loft - Fun/Convenient | Manhattan | Chelsea | Private room | 119 | 4 | 106 |
| 42887 | Unique & Stylish Apartment in Chelsea | Manhattan | Chelsea | Entire home/apt | 120 | 5 | 13 |
| 1057 | Lovely Chelsea 1 Bedroom | Manhattan | Chelsea | Entire home/apt | 125 | 9 | 10 |
| 30051 | Sunny Chelsea Studio + Balcony (Central Location) | Manhattan | Chelsea | Entire home/apt | 127 | 5 | 30 |
| 24406 | 1 Bedroom/1 Livingroom on 21st and 8th in Chelsea | Manhattan | Chelsea | Entire home/apt | 130 | 26 | 39 |
| 16371 | Convenient & comfortable studio near Penn Station | Manhattan | Chelsea | Entire home/apt | 130 | 6 | 63 |

I have listed the 10 cheapest Airbnbs in Chelsea, Greenwich Village, Kensington and Williamsburg. These lists include information about listing description, price, room type, price and number of reviews. Kensington location only included 1 listing according to the requirements mentioned above, but I believe the other neighborhoods are more suitable for my cousins' interests.

Conclusion section where you conclude the report.

In conclusion, this project has allowed me to perform a cluster analysis to understand two specific cities in New York and recommend Airbnb's in these cities that best meet my cousins' needs and requirements. I conducted initial research through conversation to understand what type of entertainment and requirements they have while looking for when travelling somewhere. After understanding their needs and wants, I dove deeper into the neighborhoods and used the Kaggle Airbnb data set to recommend listings based on price and other factors that I mentioned above in the report.