

AirBNB Stay Analysis

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Abstract - *This market analysis focuses on Airbnb, a prominent online hospitality marketplace that has revolutionized the way people find accommodation and lodging. Airbnb's unique platform connects hosts offering their properties and guests seeking short-term rentals, creating a vast global network of diverse and unique accommodations. The analysis begins by providing an overview of Airbnb's history, growth trajectory, and its impact on the traditional hospitality sector. Utilizing various market research methodologies and data sources, the study examines Airbnb's market share, geographical expansion, and user demographics. It evaluates the factors that attract hosts to list their properties on the platform, as well as the motivations that drive guests to choose Airbnb over traditional hotels. Ultimately, this market analysis aims to provide valuable insights into Airbnb's position within the hospitality industry, identifying its key success factors and potential areas for improvement. It serves as a valuable resource for industry stakeholders, investors, and policymakers seeking to understand the dynamics of the sharing economy and its impact on the traditional lodging sector. As Airbnb continues to redefine the travel and accommodation landscape, an in-depth analysis of its market presence proves crucial for making informed decisions and predicting its future trajectory.*

Keywords— *Airbnb's market share analysis, future trajectory.*

I. DATASET

In this project we are going to work with New York City Airbnb open Data.

Summary of the key attributes:-

- A. Listing Information: Details about individual listings, such as unique identifiers, property type (e.g., entire home, private room, shared room, number of bedrooms and bathrooms, and amenities provided (e.g., Wi-Fi, kitchen).
- B. Information: Information about hosts.
- C. Geographical Information: Location-related data, such as latitude and longitude coordinates, neighborhood, city, and county of the listings.
- D. Price and Availability: Pricing details for renting the property, including the base price, cleaning fees, and additional charges.
- E. Reviews and Ratings: Information on guest reviews and ratings, including overall rating, cleanliness etc.
- F. Booking Details: Data related to the number of reviews, bookings, and sometimes the minimum and maximum length of stay.
- G. Calendar Data: Detailed information about the availability of the listing over time, including dates that are available for booking and dates that are already booked.

II. PROJECT FLOW

The main idea of this project is to develop a filtering system and analyze the stays available in New York City, for the AirBNB dataset that we have obtained from Kaggle. In this particular section of the report, we shall be discussing both these aspects one after the other.

A. Filter System:

In this model, we have developed a 4-stage / 4-layer filter for the dataset that we have. The filters modify the dataset on these 4 grounds: Location, Room Type, Price & Number of Reviews.

The first filter is a ***location filter*** that reduces the existing dataset into a smaller one on the basis of the location the user wishes to reside at for their vacation. The filtering is done on the basis of the *'neighbourhood_group'* attribute. The user can simply enter the location that they wish to reside in and the dataset gets filtered accordingly.

Location - wise filtering

```
def location_filter(loc_name):  
    return df[df['neighborhood'] == loc_name, columns[1:]]  
  
loc_df = location_filter("Upper Westside")  
loc_df
```

	name	lat	lon	best name	neighborhood	group	neighborhood	latitude	longitude	room type	price	minimum price	number of reviews	last review	reviews per month	calculated host response rate	hostings count	availability
1	294	33.94	-84.38	Stella McKenna Cafe	2041	coffee	Neighborhood	40.75312	-73.0377	Neighborhood	225	9	43	2016-03-21	0.86	2	0	
232	3002	34.0	-84.38	Starbucks Coffee Co. NYC	198531	coffee	Neighborhood	40.75287	-73.03881	Neighborhood	225	20	42	2016-03-01	0.79	1	0	
340	41598	34.0	-84.38	Upper West Side restaurant on 145th	198531	Breakfast/Brunch	Neighborhood	40.75249	-73.03887	Neighborhood	100	10	54	2016-04-09	0.9	1	140	
373	110129	34.0	-84.38	Harold's Chicken Shop	198531	Fast Food	Neighborhood	40.75249	-73.03876	Neighborhood	160	10	54	2016-07-02	0.87	1	140	
382	114988	34.0	-84.38	Neighborhood Studio, Perfect Location	198531	Bed and Breakfast	Neighborhood	40.75259	-73.04008	Neighborhood	140	6	54	2016-03-10	0.45	7	0	
40744	3043682	34.0	-84.38	West Studio 4 private w/ 2 queen beds	198531243	Condo	Neighborhood	40.75253	-73.03977	Neighborhood	240	2	0	N/A	N/A	1	70	
40745	3447329	34.0	-84.38	100% Clean Studio	198531243	Condo	Neighborhood	40.75249	-73.03943	Neighborhood	240	2	0	N/A	N/A	1	70	
40751	3447330	34.0	-84.38	West Studio 4 private w/ 2 queen beds	198531243	Condo	Neighborhood	40.75259	-73.03984	Neighborhood	240	2	0	N/A	N/A	1	70	
40752	3447331	34.0	-84.38	West Studio 4 private w/ 2 queen beds	198531243	Condo	Neighborhood	40.75259	-73.03984	Neighborhood	240	2	0	N/A	N/A	1	70	
40753	3447332	34.0	-84.38	West Studio 4 private w/ 2 queen beds	198531243	Condo	Neighborhood	40.75259	-73.03984	Neighborhood	240	2	0	N/A	N/A	1	70	
40754	3447333	34.0	-84.38	West Studio 4 private w/ 2 queen beds	198531243	Condo	Neighborhood	40.75259	-73.03984	Neighborhood	240	2	0	N/A	N/A	1	70	
40755	3447334	34.0	-84.38	West Studio 4 private w/ 2 queen beds	198531243	Condo	Neighborhood	40.75259	-73.03984	Neighborhood	240	2	0	N/A	N/A	1	70	
40756	3447335	34.0	-84.38	West Studio 4 private w/ 2 queen beds	198531243	Condo	Neighborhood	40.75259	-73.03984	Neighborhood	240	2	0	N/A	N/A	1	70	
40757	3447336	34.0	-84.38	West Studio 4 private w/ 2 queen beds	198531243	Condo	Neighborhood	40.75259	-73.03984	Neighborhood	240	2	0	N/A	N/A	1	70	
40758	3447337	34.0	-84.38	West Studio 4 private w/ 2 queen beds	198531243	Condo	Neighborhood	40.75259	-73.03984	Neighborhood	240	2	0	N/A	N/A	1	70	
40759	3447338	34.0	-84.38	West Studio 4 private w/ 2 queen beds	198531243	Condo	Neighborhood	40.75259	-73.03984	Neighborhood	240	2	0	N/A	N/A	1	70	
40760	3447339	34.0	-84.38	West Studio 4 private w/ 2 queen beds	198531243	Condo	Neighborhood	40.75259	-73.03984	Neighborhood	240	2	0	N/A	N/A	1	70	
40761	3447340	34.0	-84.38	West Studio 4 private w/ 2 queen beds	198531243	Condo	Neighborhood	40.75259	-73.03984	Neighborhood	240	2	0	N/A	N/A	1	70	
40762	3447341	34.0	-84.38	West Studio 4 private w/ 2 queen beds	198531243	Condo	Neighborhood	40.75259	-73.03984	Neighborhood	240	2	0	N/A	N/A	1	70	
40763	3447342	34.0	-84.38	West Studio 4 private w/ 2 queen beds	198531243	Condo	Neighborhood	40.75259	-73.03984	Neighborhood	240	2	0	N/A	N/A	1	70	
40764	3447343	34.0	-84.38	West Studio 4 private w/ 2 queen beds	198531243	Condo	Neighborhood	40.75259	-73.03984	Neighborhood	240	2	0	N/A	N/A	1	70	
40765	3447344	34.0	-84.38	West Studio 4 private w/ 2 queen beds	198531243	Condo	Neighborhood	40.75259	-73.03984	Neighborhood	240	2	0	N/A	N/A	1	70	
40766	3447345	34.0	-84.38	West Studio 4 private w/ 2 queen beds	198531243	Condo	Neighborhood	40.75259	-73.03984	Neighborhood	240	2	0	N/A	N/A	1	70	
40767	3447346	34.0	-84.38	West Studio 4 private w/ 2 queen beds	198531243	Condo	Neighborhood	40.75259	-73.03984	Neighborhood	240	2	0	N/A	N/A	1	70	
40768	3447347	34.0	-84.38	West Studio 4 private w/ 2 queen beds	198531243	Condo	Neighborhood	40.75259	-73.03984	Neighborhood	240	2	0	N/A	N/A	1	70	
40769	3447348	34.0	-84.38	West Studio 4 private w/ 2 queen beds	198531243	Condo	Neighborhood	40.75259	-73.03984	Neighborhood	240	2	0	N/A	N/A	1	70	
40770	3447349	34.0	-84.38	West Studio 4 private w/ 2 queen beds	198531243	Condo	Neighborhood	40.75259	-73.03984	Neighborhood	240	2	0	N/A	N/A	1	70	
40771	3447350	34.0	-84.38	West Studio 4 private w/ 2 queen beds	198531243	Condo	Neighborhood	40.75259	-73.03984	Neighborhood	240	2	0	N/A	N/A	1	70	
40772	3447351	34.0	-84.38	West Studio 4 private w/ 2 queen beds	198531243	Condo	Neighborhood	40.75259	-73.03984	Neighborhood	240	2	0	N/A	N/A	1	70	
40773	3447352	34.0	-84.38	West Studio 4 private w/ 2 queen beds	198531243	Condo	Neighborhood	40.75259	-73.03984	Neighborhood	240	2	0	N/A	N/A	1	70	
40774	3447353	34.0	-84.38	West Studio 4 private w/ 2 queen beds	198531243	Condo	Neighborhood	40.75259	-73.03984	Neighborhood	240	2	0	N/A	N/A	1	70	
40775	3447354	34.0	-84.38	West Studio 4 private w/ 2 queen beds	198531243	Condo	Neighborhood	40.75259	-73.03984	Neighborhood	240	2	0	N/A	N/A	1	70	
40776	3447355	34.0	-84.38	West Studio 4 private w/ 2 queen beds	198531243	Condo	Neighborhood	40.75259	-73.03984	Neighborhood	240	2	0	N/A	N/A	1	70	
40777	3447356	34.0	-84.38	West Studio 4 private w/ 2 queen beds	198531243	Condo	Neighborhood	40.75259	-73.03984	Neighborhood	240	2	0	N/A	N/A	1	70	
40778	3447357	34.0	-84.38	West Studio 4 private w/ 2 queen beds	198531243	Condo	Neighborhood	40.75259	-73.03984	Neighborhood	240	2	0	N/A	N/A	1	70	
40779	3447358	34.0	-84.38	West Studio 4 private w/ 2 queen beds	198531243	Condo	Neighborhood	40.75259	-73.03984	Neighborhood	240	2	0	N/A	N/A	1	70	
40780	3447359	34.0	-84.38	West Studio 4 private w/ 2 queen beds	198531243	Condo	Neighborhood	40.75259	-73.03984	Neighborhood	240	2	0	N/A	N/A	1	70	
40781	3447360	34.0	-84.38	West Studio 4 private w/ 2 queen beds	198531243	Condo	Neighborhood	40.75259	-73.03984	Neighborhood	240	2	0	N/A	N/A	1	70	
40782	3447361	34.0	-84.38	West Studio 4 private w/ 2 queen beds	198531243	Condo	Neighborhood	40.75259	-73.03984	Neighborhood	240	2	0	N/A	N/A	1	70	
40783	3447362	34.0	-84.38	West Studio 4 private w/ 2 queen beds	198531243	Condo	Neighborhood	40.75259	-73.03984	Neighborhood	240	2	0	N/A	N/A	1	70	
40784	3447363	34.0	-84.38	West Studio 4 private w/ 2 queen beds	198531243	Condo	Neighborhood	40.75259	-73.03984	Neighborhood	240	2	0	N/A	N/A	1	70	
40785	3447364	34.0	-84.38	West Studio 4 private w/ 2 queen beds	198531243	Condo	Neighborhood	40.75259	-73.03984	Neighborhood	240	2	0	N/A	N/A	1	70	
40786	3447365	34.0	-84.38	West Studio 4 private w/ 2 queen beds	198531243	Condo	Neighborhood	40.75259	-73.03984	Neighborhood	240	2	0	N/A	N/A	1	70	
40787	3447366	34.0	-84.38	West Studio 4 private w/ 2 queen beds	198531243	Condo	Neighborhood	40.75259	-73.03984	Neighborhood	240	2	0	N/A	N/A	1	70	
40788	3447367	34.0	-84.38	West Studio 4 private w/ 2 queen beds	198531243	Condo	Neighborhood	40.75259	-73.03984	Neighborhood	240	2	0	N/A	N/A	1	70	
40789	3447368	34.0	-84.38	West Studio 4 private w/ 2 queen beds	198531243	Condo	Neighborhood	40.75259	-73.03984	Neighborhood	240	2	0	N/A	N/A	1	70	
40790	3447369	34.0	-84.38	West Studio 4 private w/ 2 queen beds	198531243	Condo	Neighborhood	40.75259	-73.03984	Neighborhood	240	2	0	N/A	N/A	1	70	
40791	3447370	34.0	-84.38	West Studio 4 private w/ 2 queen beds	198531243	Condo	Neighborhood	40.75259	-73.03984	Neighborhood	240	2	0	N/A	N/A	1	70	
40792	3447371	34.0	-84.38	West Studio 4 private w/ 2 queen beds	198531243	Condo	Neighborhood	40.75259	-73.03984	Neighborhood	240	2	0	N/A	N/A	1	70	
40793	3447372	34.0	-84.38	West Studio 4 private w/ 2 queen beds	198531243	Condo	Neighborhood	40.75259	-73.03984	Neighborhood	240	2	0	N/A	N/A	1	70	
40794	3447373	34.0	-84.38	West Studio 4 private w/ 2 queen beds	198531243	Condo	Neighborhood	40.75259	-73.03984	Neighborhood	240	2	0	N/A	N/A	1	70	
40795	3447374	34.0	-84.38	West Studio 4 private w/ 2 queen beds	198531243	Condo	Neighborhood	40.75259	-73.03984	Neighborhood	240	2	0	N/A	N/A	1	70	
40796	3447375	34.0	-84.38	West Studio 4 private w/ 2 queen beds	198531243	Condo	Neighborhood	40.75259	-73.03984	Neighborhood	240	2	0	N/A	N/A	1	70	
40797	3447376	34.0	-84.38	West Studio 4 private w/ 2 queen beds	198531243	Condo	Neighborhood	40.75259	-73.03984	Neighborhood	240	2	0	N/A	N/A	1	70	
40798	3447377	34.0	-84.38	West Studio 4 private w/ 2 queen beds	198531243	Condo	Neighborhood	40.75259	-73.03984	Neighborhood	240	2	0	N/A	N/A	1	70	
40799	3447378	34.0	-84.38	West Studio 4 private w/ 2 queen beds	198531243	Condo	Neighborhood	40.75259	-73.03984	Neighborhood	240	2	0	N/A	N/A	1	70	
40800	3447379	34.0	-84.38	West Studio 4 private w/ 2 queen beds	198531243	Condo	Neighborhood	40.75259	-73.03984	Neighborhood	240	2	0	N/A	N/A	1	70	
40801	3447380	34.0	-84.38	West Studio 4 private w/ 2 queen beds	198531243	Condo	Neighborhood	40.75259	-73.03984	Neighborhood	240	2	0	N/A	N/A	1	70	
40802	3447381	34.0	-84.38	West Studio 4 private w/ 2 queen beds	198531243	Condo	Neighborhood	40.75259	-73.03984	Neighborhood	240	2	0	N/A	N/A	1	70	
40803	3447382	34.0	-84.38	West Studio 4 private w/ 2 queen beds	198531243	Condo	Neighborhood	40.75259	-73.03984	Neighborhood	240	2	0	N/A	N/A	1	70	
40804	3447383	34.0	-84.38	West Studio 4 private w/ 2 queen beds	198531243	Condo	Neighborhood	40.75259	-73.03984	Neighborhood	240	2	0	N/A	N/A	1	70	
40805	3447384	34.0	-84.38	West Studio 4 private w/ 2 queen beds	198531243	Condo	Neighborhood	40.75259	-73.03984	Neighborhood	240	2	0	N/A	N/A	1	70	
40806	3447385	34.0	-84.38	West Studio 4 private w/ 2 queen beds	198531243	Condo	Neighborhood	40.75259	-73.03984	Neighborhood	240	2	0	N/A	N/A	1	70	
40807	3447386	34.0	-84.38	West Studio 4 private w/ 2 queen beds	198531243	Condo	Neighborhood	40.75259	-73.03984	Neighborhood	240	2	0	N/A	N/A	1	70	
40808	3447387	34.0	-84.38	West Studio 4 private w/ 2 queen beds	198531243	Condo	Neighborhood	40.75259	-73.03984	Neighborhood	240	2	0	N/A	N/A	1	70	
40809	3447388	34.0	-84.38	West Studio 4 private w/ 2 queen beds	198531243	Condo	Neighborhood	40.75259	-73.03984	Neighborhood	240	2	0	N/A	N/A	1	70	
40810	3447389	34.0	-84.38	West Studio 4 private w/ 2 queen beds	198531243	Condo	Neighborhood	40.75259	-73.03984	Neighborhood	240	2	0	N/A	N/A	1	70	
40811	3447390	34.0	-84.38	West Studio 4 private w/ 2 queen beds	198531243	Condo	Neighborhood	40.75259	-73.03984	Neighborhood	240	2	0	N/A	N/A	1	70	
40812	3447391	34.0	-84.38	West Studio 4 private w/ 2 queen beds	198531243	Condo	Neighborhood	40.75259	-73.03984	Neighborhood	240	2	0	N/A	N/A	1	70	
40813	3447392	34.0	-84.38	West Studio 4 private w/ 2 queen beds	198531243	Condo	Neighborhood	40.75259	-73.03984	Neighborhood	240	2	0	N/A	N/A	1	70	
40814	3447393	34.0	-84.38	West Studio 4 private w/ 2 queen beds	198531243	Condo	Neighborhood	40.75259	-73.03984	Neighborhood	240	2	0	N/A	N/A	1	70	
40815	3447394	34.0	-84.38	West Studio 4 private w/ 2 queen beds	198531243	Condo	Neighborhood	40.75259	-73.03984	Neighborhood	240	2	0	N/A	N/A	1	70	
40816	3447395	34.0	-84.38	West Studio 4 private w/ 2 queen beds	198531243	Condo	Neighborhood	40.75259	-73.03984	Neighborhood	240	2	0	N/A	N/A	1		

eg : If we search neighborhood group as Manhattan we will find all the listings of Manhattan.

The second filter filters the dataset further on the ground of what **room type** the user wishes to stay in. Once the user enters the required room type, they can view all such rooms available in the selected location.

Room - wise filtering

```
def room_filters(
    min_sq_ft=400, max_sq_ft = None, max_bathrooms=1
):
    """
    min_sq_ft: minimum square feet required
    max_sq_ft: maximum square feet required
    max_bathrooms: maximum number of bathrooms required
    """
    return room_list[
        (room_list['sq_ft'] >= min_sq_ft) &&
        (room_list['sq_ft'] <= max_sq_ft) &&
        (room_list['bathrooms'] <= max_bathrooms)
    ]

room_list = room_filters(
    min_sq_ft=400, max_sq_ft=None, max_bathrooms=1
)

room_list
```

	ID	name	host id	host name	neighbourhood group	neighbourhood	latitude	longitude	room type	price	minimum nights	number of reviews	last review	reviews per month	calculated host listings count	availability 30 days
418	233662	Large Sunny Loft with 2 Bedrooms	1222866	Yibeyi	Neighborhood	Midtown	40.76275	-73.8462	Private room	110	2	81	2019-09-02	0.65	1	264
543	304040	Beautiful Apartment in Manhattanville	1388093	Caroline	Neighborhood	Midtown	40.74333	-73.9842	Private room	110	3	234	2019-09-02	2.45	2	167
397	402384	Luxury room in West Village	1388093	Caroline	Neighborhood	Midtown	40.74233	-73.99262	Private room	140	3	102	2019-09-02	1.08	2	107
1438	648262	Cozy Studio in West Village	1747374	Kyle	Neighborhood	Midtown	40.74640	-73.98429	Private room	100	3	41	2019-09-05	0.22	1	0
139	688084	10th floor Apartment in Midtown	2488582	Kenan	Neighborhood	Midtown	40.75211	-73.97877	Private room	120	4	31	2019-09-06	0.38	1	0
47507	3685652	City Condo in the West Village	2046356	Dianna	Neighborhood	Midtown	40.73584	-73.98268	Private room	125	1	6	2019-07-01	6.00	1	142
69906	36073362	Luxury One-Bedroom, Top Floor, Full Kitchen	20066461	James	Neighborhood	Midtown	40.75126	-73.97311	Private room	118	30	0	N/A	N/A	12	360
43232	36030030	Spacious Sunny Apartment (City & Co.)	24837034	Farrel	Neighborhood	Midtown	40.75669	-73.84848	Private room	100	2	0	N/A	N/A	1	88
40576	36170186	Luxury Loft in West Village	19071773	Leah Luu	Neighborhood	Midtown	40.73269	-73.99267	Private room	130	2	1	2019-07-06	1.00	1	7
40775	36476761	Spacious Midtown Cozy Loft w/ City View	17701238	Kelly	Neighborhood	Midtown	40.75268	-73.97389	Private room	100	3	0	N/A	N/A	1	2

402 rows x 16 columns

eg : Here we have searched the room_type as Private room, so we find all the corresponding listings under this category.

The third filter reduces the dataset on the basis of the *budget* of the user. This is the **price filter** that generates a dataset according to the upper and lower budget bounds of the user. In this way, the user can obtain a much more sophisticated view of the options available to them according to their preferences of stay.

[illegible]

eg : Here we have searched for the listings having price ranging from 70-100.

Last but not the least, we have a filter that reorders the dataset in a descending order on the basis of the number of reviews a stay has got. This is the *review filter*. The reason why we have taken the number of reviews as a ground of categorization is because higher the number of reviews, the better the stay is likely to be. The user definitely wants to have the best options available for them at the top of their screen and hence we decided the sorting to be in a descending fashion, thus generating the best possible filtering system.

Sorting existing dataset by number of reviews

highest to lowest

```
sort_values(review_count)
sort_values(review_count, ascending=False)
sort_values(review_count)
```

View

Yellow (Cider)

View

id	name	host_id	host_name	neighbourhood_group	neighbourhood	latitude	longitude	room_type	price	minimum_nights	number of reviews	last review	reviews per month	calculated host listings count	availability_365	
5676	Madison	146364001	Jeffrey	Neighborhood	Lower East Side	40.7161	-73.9819	Private room	39	2	140	2019-07-06	6.65	5	176	
1596	100013	City Room - Midtown East	229481	Carol	Neighborhood	Midtown East	40.7599	-73.9647	Private room	72	5	451	2019-09-03	5.26	2	10
1155	100013	Private Room in Queens, Midtown East	277373	Agnes	Neighborhood	Midtown East	40.7527	-73.9450	Private room	85	1	439	2019-07-05	5.12	2	106
7	115	Large furnished room in Midtown East	887	Danica	Neighborhood	Midtown East	40.7589	-73.9585	Private room	70	2	439	2019-08-24	5.47	1	200
3862	100101	East Village Loft in Midtown East	561654	Sarah	Neighborhood	East Village	40.7336	-73.9813	Private room	100	1	434	2019-06-22	5.19	2	105
11882	100101	Beautiful Spacious Apt in Midtown East	661348	Morgan	Neighborhood	Midtown East	40.7517	-73.9520	Private room	25	0	0	NaN	NaN	1	8
2526	100101	Stunning, Bright Room in Midtown East	139366	Sharon	Neighborhood	Midtown East	40.7599	-73.9592	Private room	15	0	0	NaN	NaN	2	6
34622	2100107	Big City Condo in Midtown East	1070030	John	Neighborhood	Midtown East	40.7593	-73.9580	Private room	100	2	0	NaN	NaN	1	10
6881	1000015	Modern Apartment in Midtown East	2417126	John	Neighborhood	Midtown East	40.7581	-73.9520	Private room	150	1	0	NaN	NaN	0	109
14902	100101	Queen's Village Apartment	4321436	Julia	Neighborhood	Upper West Side	40.7390	-73.9812	Private room	60	0	0	NaN	NaN	1	8

522 rows × 16 columns

eg : To sort an existing Airbnb dataset by the number of views, we extracted the "number of views" column from the dataset, sorted it based on the "number of views" column in descending order. The sorted dataset will now display listings with the highest number of views at the top, allowing you to identify the most viewed properties on AirBNB.

```

(base) C:\Users\jaiga\Test Folder\DMND Lab\python terminal_display.py

```

Name	Location	Lodging	Price	Reviews
Beautiful 1br on Upper West Side	Upper West Side, Manhattan	Home/Apt	135	53
CBG Helps Haiti Rm #2	Park Slope, Brooklyn	Private Room	110	71
Clean and Quiet in Brooklyn	Bedford-Stuyvesant, Brooklyn	Private Room	35	0
Fort Greene Broomstone	Fort Greene, Brooklyn	Private Room	88	135
2 bedroom - Upper East Side-great for kids	Upper East Side, Manhattan	Home/Apt	250	66
Charming East Village One Bedroom Flat	East Village, Manhattan	Home/Apt	190	21
Great Location for NYC	East Village, Manhattan	Private Room	58	2
House On Henry (3rd FLR Suite)	Carroll Gardens, Brooklyn	Home/Apt	175	233
2 BR / 2 Bath Duplex Apt with patio East Village	East Village, Manhattan	Home/Apt	350	7
Secluded Master Bedroom in Beautiful Huge Apt	Washington Heights, Manhattan	Private Room	85	36

```

Close

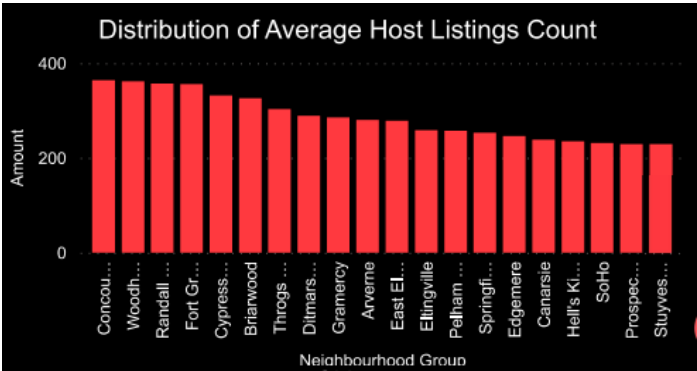
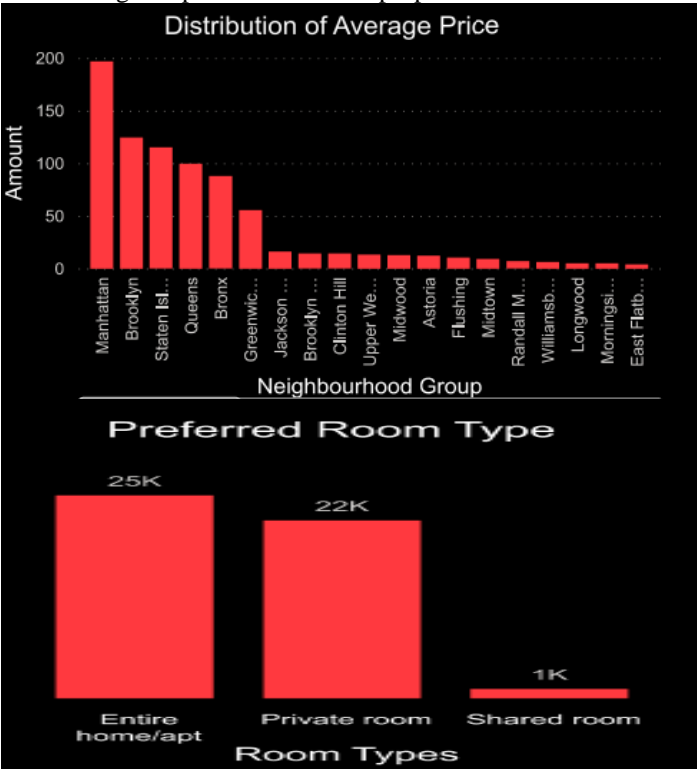
```

The finally filtered dataset contains information on location (city or neighborhood), lodging type (e.g., entire home, private room), price, number of reviews and a brief

description of each listing(name of listing). This comprehensive dataset is ideal for conducting detailed analysis, visualizations.

B. Stay Analysis (Exploratory Data Analysis):

For analyzing price trends, room types and availability at different locations, etc. we have performed exploratory analysis on our dataset. Using PowerBI, we have generated a few of the given plots for the same purpose.



III. PROJECT PROGRESS

The entire project has been successfully accomplished, thanks to the committed efforts of every team member.