

# DATA INSIGHTS OF AIRBNB IN NYC

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# BACKGROUND

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Airbnb is an online platform using which people can rent their unused accommodations.

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During the covid time, Airbnb incurred a huge loss in revenue.

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People have now started travelling again and Airbnb is aiming to bring up the business again and ready to provide services to customers.

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For the past few months, Airbnb has seen a major decline in revenue.

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Now that the restrictions have started lifting and people have started to travel more, Airbnb wants to make sure that it is fully prepared for this change.

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So, analysis has been done on a dataset consisting of various Airbnb listings in New York.

# AIRBNB DATA DESCRIPTION

| Column                         | Description  |
|--------------------------------|--|
| id                             | listing ID   |
| name                           | name of the listing                                  |
| host_id                        | host ID  |
| host_name                      | name of the host                                     |
| neighbourhood_group            | location   |
| neighbourhood                  | area   |
| latitude                       | latitude coordinates                                 |
| longitude                      | longitude coordinates                                |
| room_type                      | listing space type                                   |
| price                          |  |
| minimum_nights                 | amount of nights minimum                             |
| number_of_reviews              | number of reviews                                    |
| last_review                    | latest review  |
| reviews_per_month              | number of reviews per month                          |
| calculated_host_listings_count | amount of listing per host                           |
| availability_365               | number of days when listing is available for booking |

Dataset Description

# DATA ASSUMPTIONS - VARIABLES

## Categorical Variables:

- room\_type
- neighbourhood\_group
- neighbourhood

## Continuous Variables(Numerical):

- Price
- minimum\_nights
- number\_of\_reviews
- reviews\_per\_month
- calculated\_host\_listings\_count
- availability\_365
- Continuous Variables could be binned in to groups too

## Location Variables:

- latitude
- longitude

## Time Variable:

- last\_review

Variable Categories



For the past few months, Airbnb has seen a major decline in revenue. Now that the restrictions have started lifting and people have started to travel more, Airbnb wants to make sure that it is fully prepared for this change.



The different leaders at Airbnb want to understand some important insights based on various attributes in the dataset so as to increase the revenue. Our responsibility is to provide valuable insights to aid in decision making.

# PROBLEM STATEMENT



# DATASET

There are total 48895 rows and 16 columns.  
reviews\_per\_month column is of object Dtype. datetime64 is a better Data type for this column.

RangeIndex: 48895 entries, 0 to 48894

Data columns (total 20 columns):

| #  | Column                         | Non-Null Count | Dtype   |
|----|--------------------------------|----------------|---------|
| 0  | id                             | 48895 non-null | int64   |
| 1  | name                           | 48879 non-null | object  |
| 2  | host_id                        | 48895 non-null | int64   |
| 3  | host_name                      | 48874 non-null | object  |
| 4  | neighbourhood_group            | 48895 non-null | object  |
| 5  | neighbourhood                  | 48895 non-null | object  |
| 6  | latitude                       | 48895 non-null | float64 |
| 7  | longitude                      | 48895 non-null | float64 |
| 8  | room_type                      | 48895 non-null | object  |
| 9  | price                          | 48895 non-null | int64   |
| 10 | minimum_nights                 | 48895 non-null | int64   |
| 11 | number_of_reviews              | 48895 non-null | int64   |
| 12 | last_review                    | 38843 non-null | object  |
| 13 | reviews_per_month              | 38843 non-null | float64 |
| 14 | calculated_host_listings_count | 48895 non-null | int64   |
| 15 | availability_365               | 48895 non-null | int64   |
| 16 | availability_365_categories    | 48895 non-null | object  |
| 17 | minimum_night_categories       | 48895 non-null | object  |
| 18 | number_of_reviews_categories   | 48895 non-null | object  |
| 19 | price_categories               | 48895 non-null | object  |

dtypes: float64(3), int64(7), object(10)

# CREATING FEATURES

## .2 Categorizing the "minimum\_nights" column into 5 categories

```
def minimum_night_categories_function(row):  
    """  
    Categorizes the "minimum_nights" column into 5 categories  
    """  
    if row <= 1:  
        return 'very Low'  
    elif row <= 3:  
        return 'Low'  
    elif row <= 5:  
        return 'Medium'  
    elif (row <= 7):  
        return 'High'  
    else:  
        return 'very High'
```

## 1.1 Categorizing the "availability\_365" column into 5 categories

```
def availability_365_categories_function(row):  
    """  
    Categorizes the "availability_365" column into 5 categories  
    """  
    if row <= 1:  
        return 'very Low'  
    elif row <= 100:  
        return 'Low'  
    elif row <= 200:  
        return 'Medium'  
    elif (row <= 300):  
        return 'High'  
    else:  
        return 'very High'
```

## 1.3 Categorizing the "number\_of\_reviews" column into 5 categories

```
def number_of_reviews_categories_function(row):  
    """  
    Categorizes the "number_of_reviews" column into 5 categories  
    """  
    if row <= 1:  
        return 'very Low'  
    elif row <= 5:  
        return 'Low'  
    elif row <= 10:  
        return 'Medium'  
    elif (row <= 30):  
        return 'High'  
    else:  
        return 'very High'
```

# MISSING VALUES ANALYSIS

last\_review , reviews\_per\_month  
columns have around 20.56% missing  
values  
name and host\_name have 0.03% and  
0.04 % missing values respectively.

```
# To see the sum of missing values for each column  
Airbnb_data.isnull().mean()*100
```

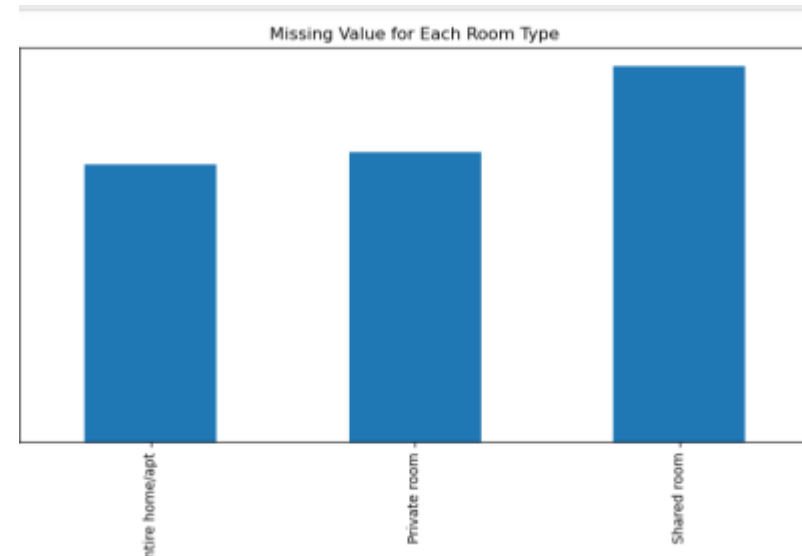
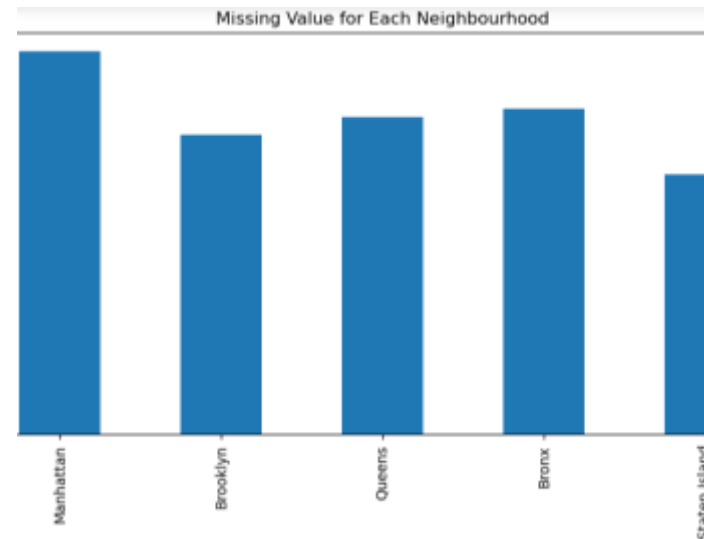
|                                |           |
|--------------------------------|-----------|
| id                             | 0.000000  |
| name                           | 0.032723  |
| host_id                        | 0.000000  |
| host_name                      | 0.042949  |
| neighbourhood_group            | 0.000000  |
| neighbourhood                  | 0.000000  |
| latitude                       | 0.000000  |
| longitude                      | 0.000000  |
| room_type                      | 0.000000  |
| price                          | 0.000000  |
| minimum_nights                 | 0.000000  |
| number_of_reviews              | 0.000000  |
| last_review                    | 20.558339 |
| reviews_per_month              | 20.558339 |
| calculated_host_listings_count | 0.000000  |
| availability_365               | 0.000000  |
| availability_365_categories    | 0.000000  |
| minimum_night_categories       | 0.000000  |
| number_of_reviews_categories   | 0.000000  |
| price_categories               | 0.000000  |
| dtype: float64                 |           |

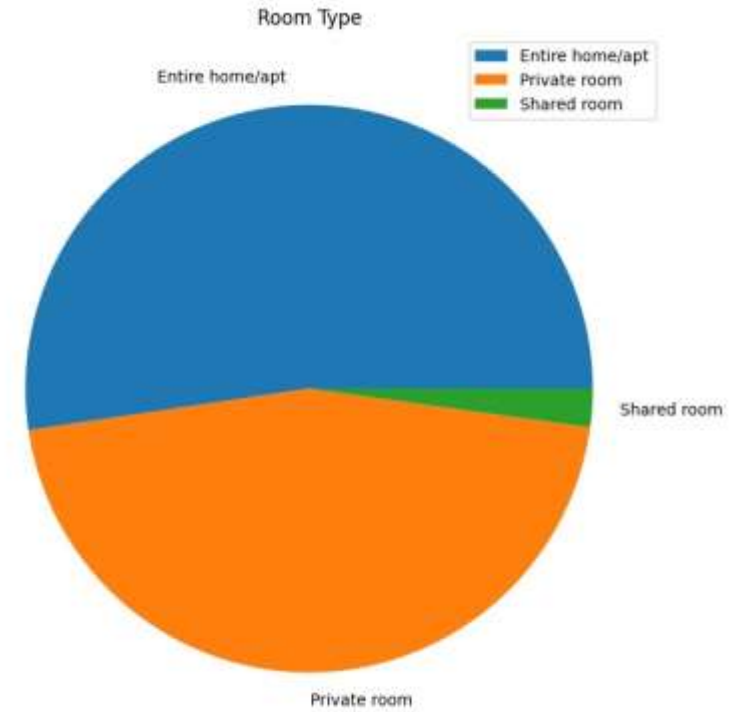
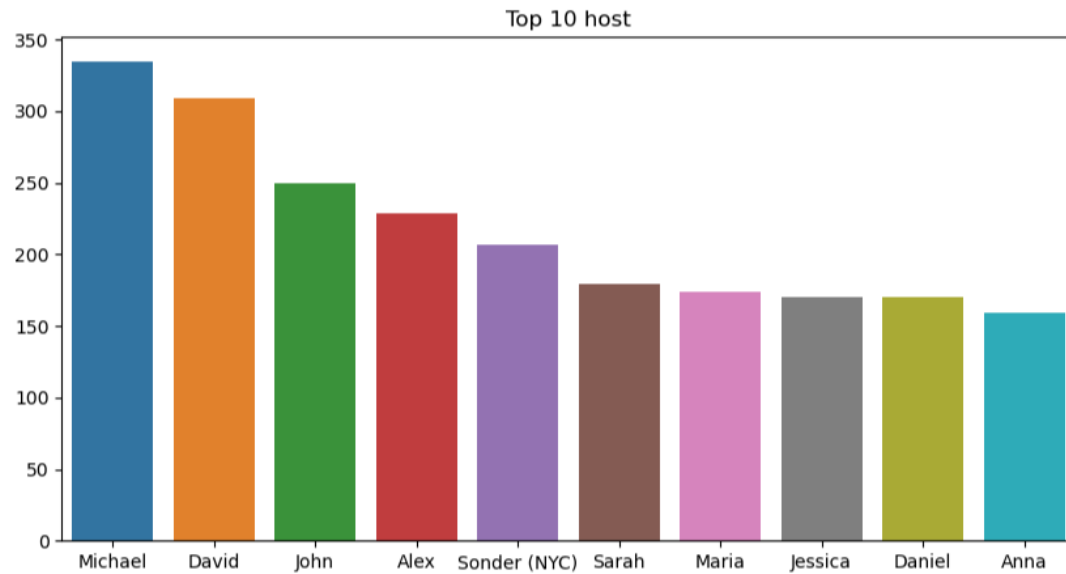


# ANALYZING MISSING VALUE

The Each neighborhood group has 19% missing values in 'last\_review' feature.

The Each neighborhood group has about 22 % missing values in 'last\_review' feature.





# UNIVARIATE ANALYSIS

# LAST REVIEW FEATURE

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The pricing is higher when 'last\_review' feature is missing .

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reviews are less likely to be given for shared rooms

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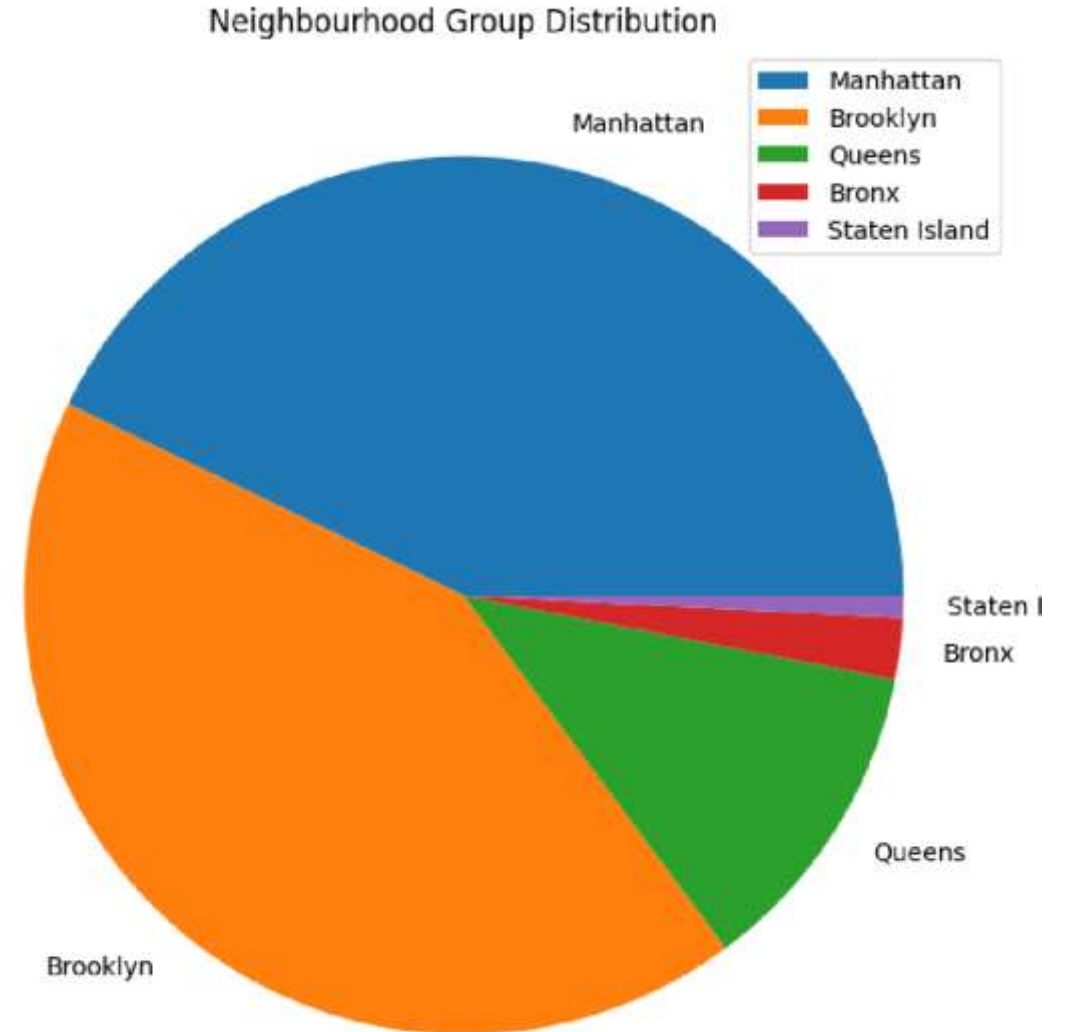
When the prices are high reviews are less likely to be given

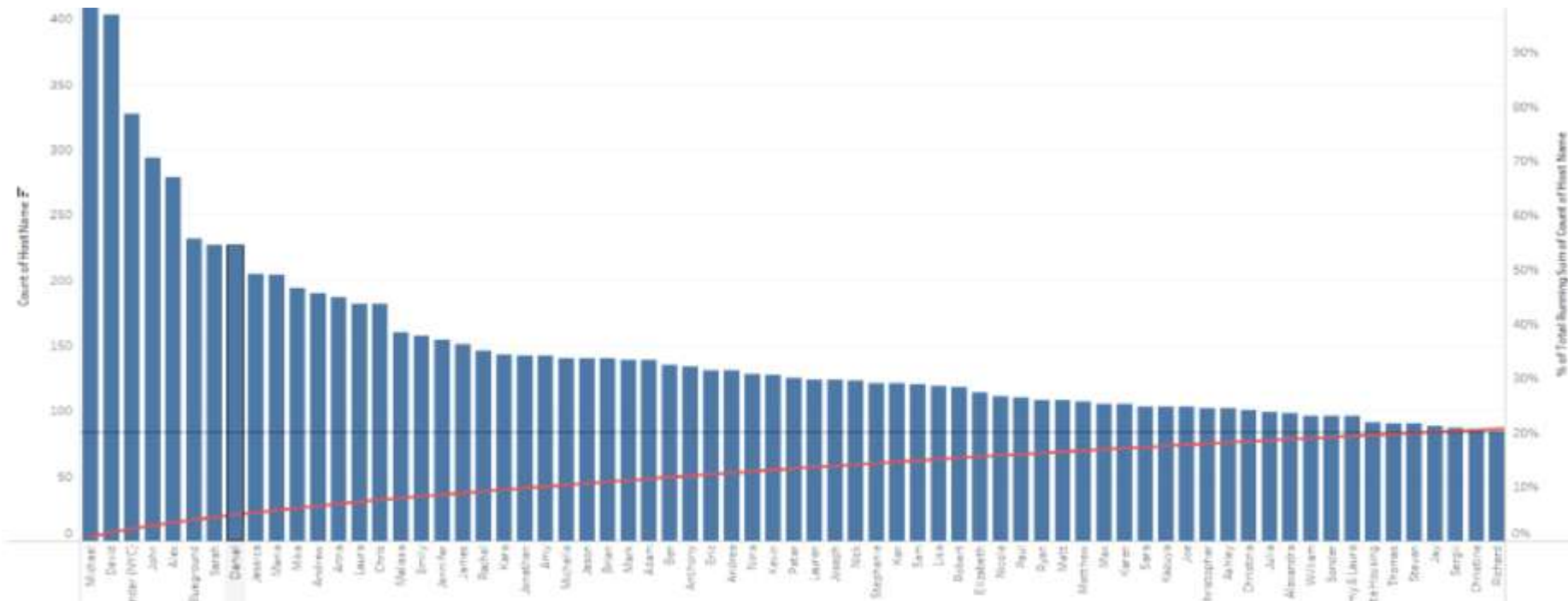
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The above analysis seems to show that the missing values here are not MCAR (missing completely at random)

# MOST CONTRIBUTING NEIGHBOURS

What are the neighborhoods needed to target?  
81% of the listing are Manhattan and Brooklyn neighborhood group





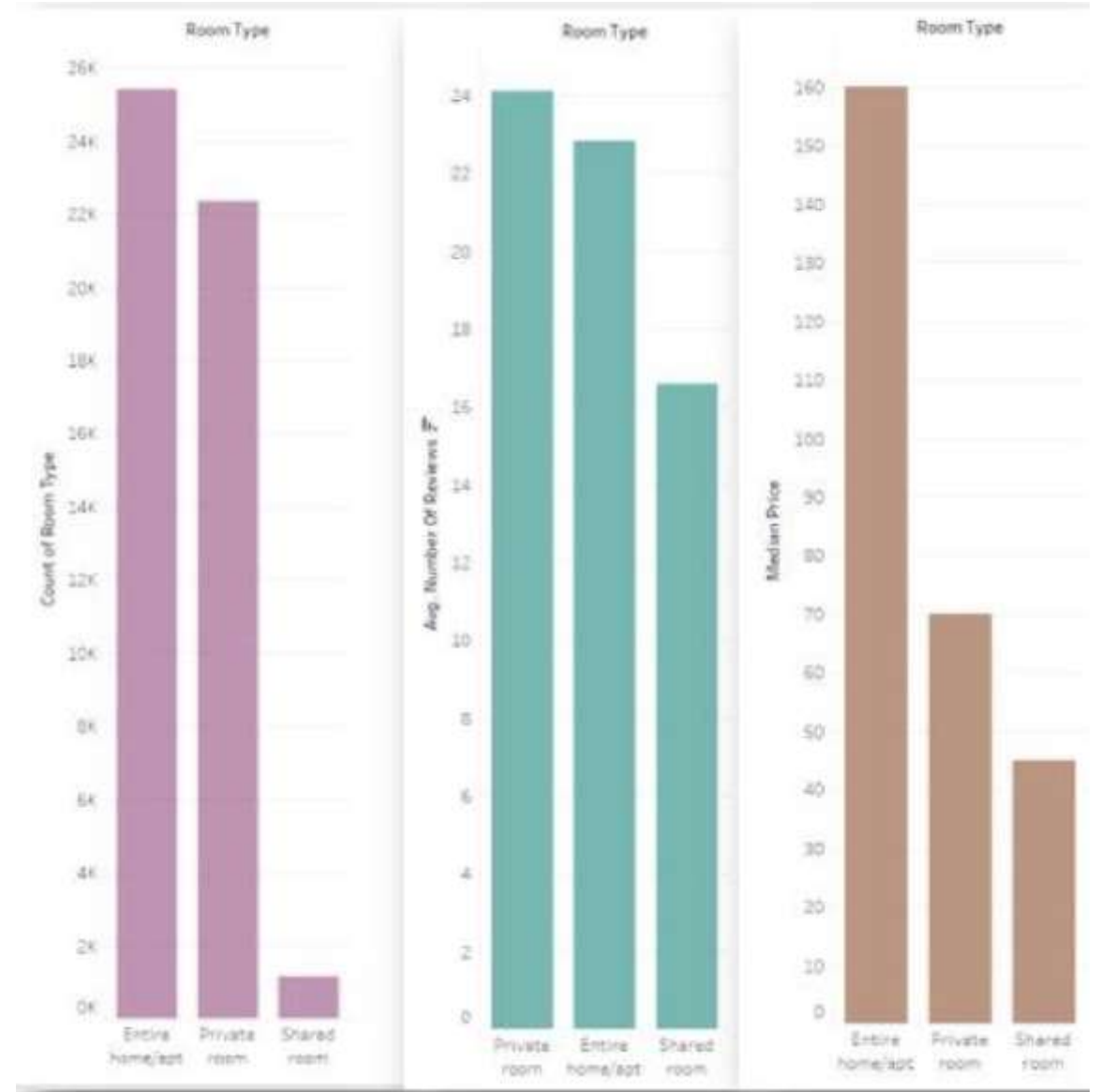
# EVERY HOST MATTER

The top 60 hosts only make up 20% of the total host count.



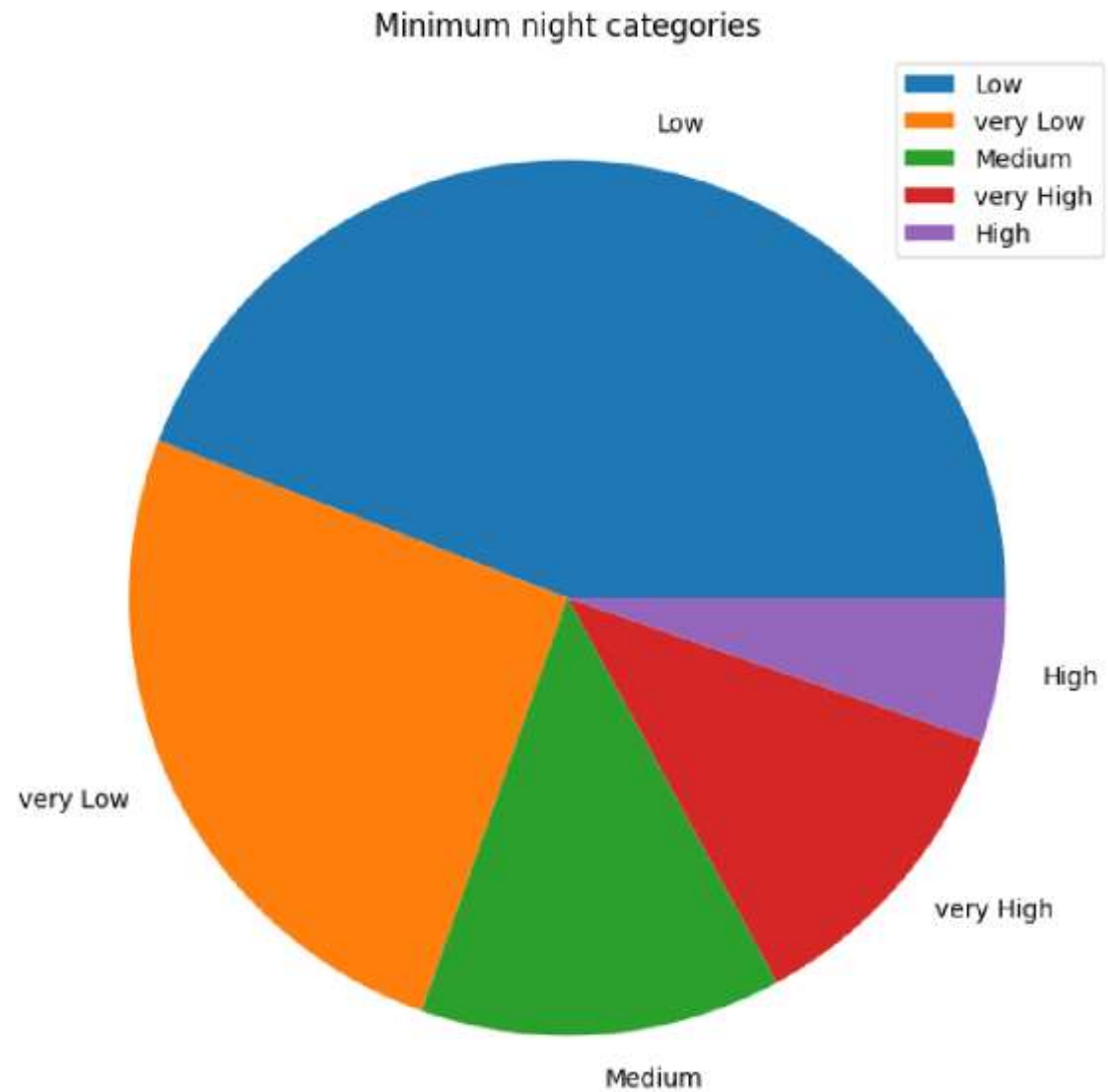
# THE PROBLEMS OF SHARED ROOMS

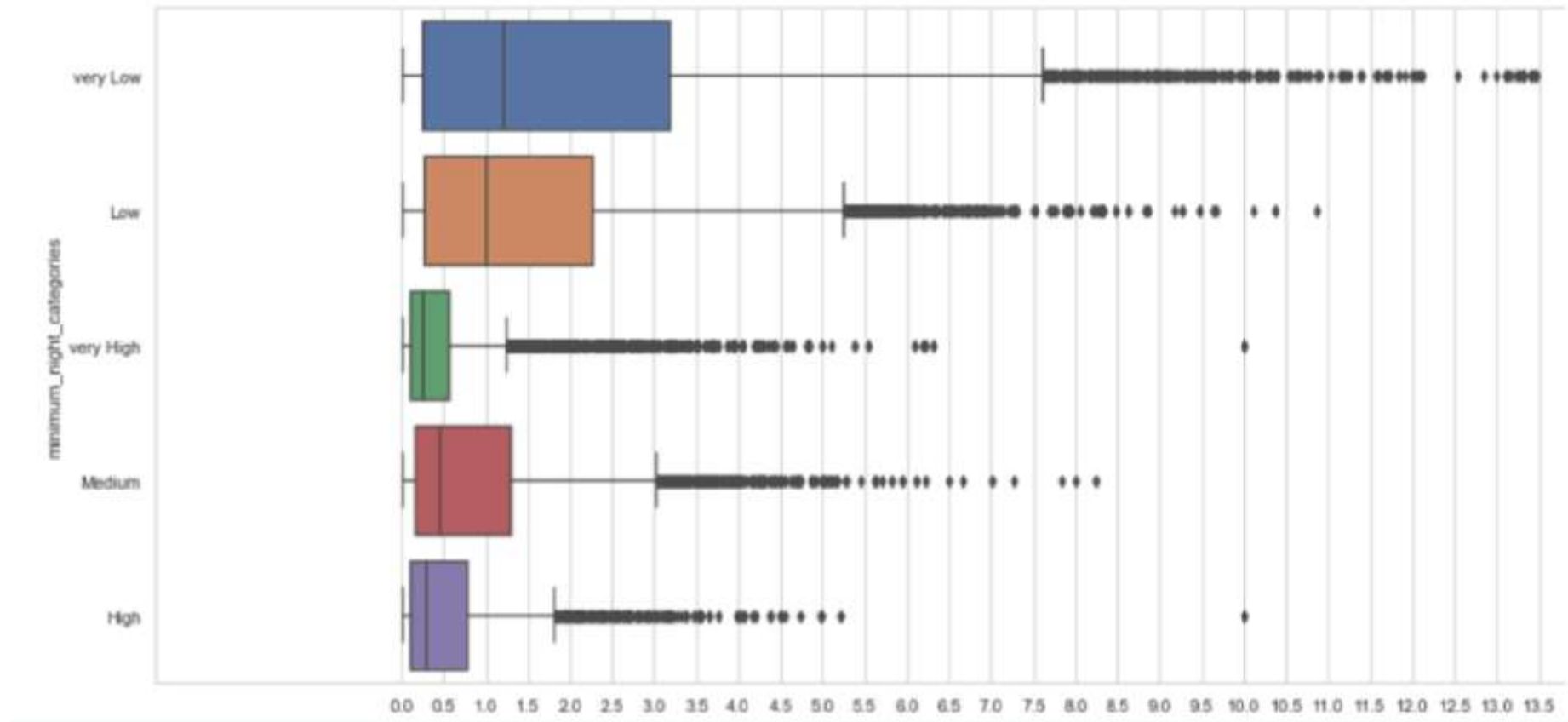
Median rates for shared rooms are significantly lower. They are less likely to be reviewed. Shared rooms only accounts for 2% of the total types of rooms.



# MINIMUM NIGHT CATEGORIES

Low Category in minimum nights feature contribute 40%.

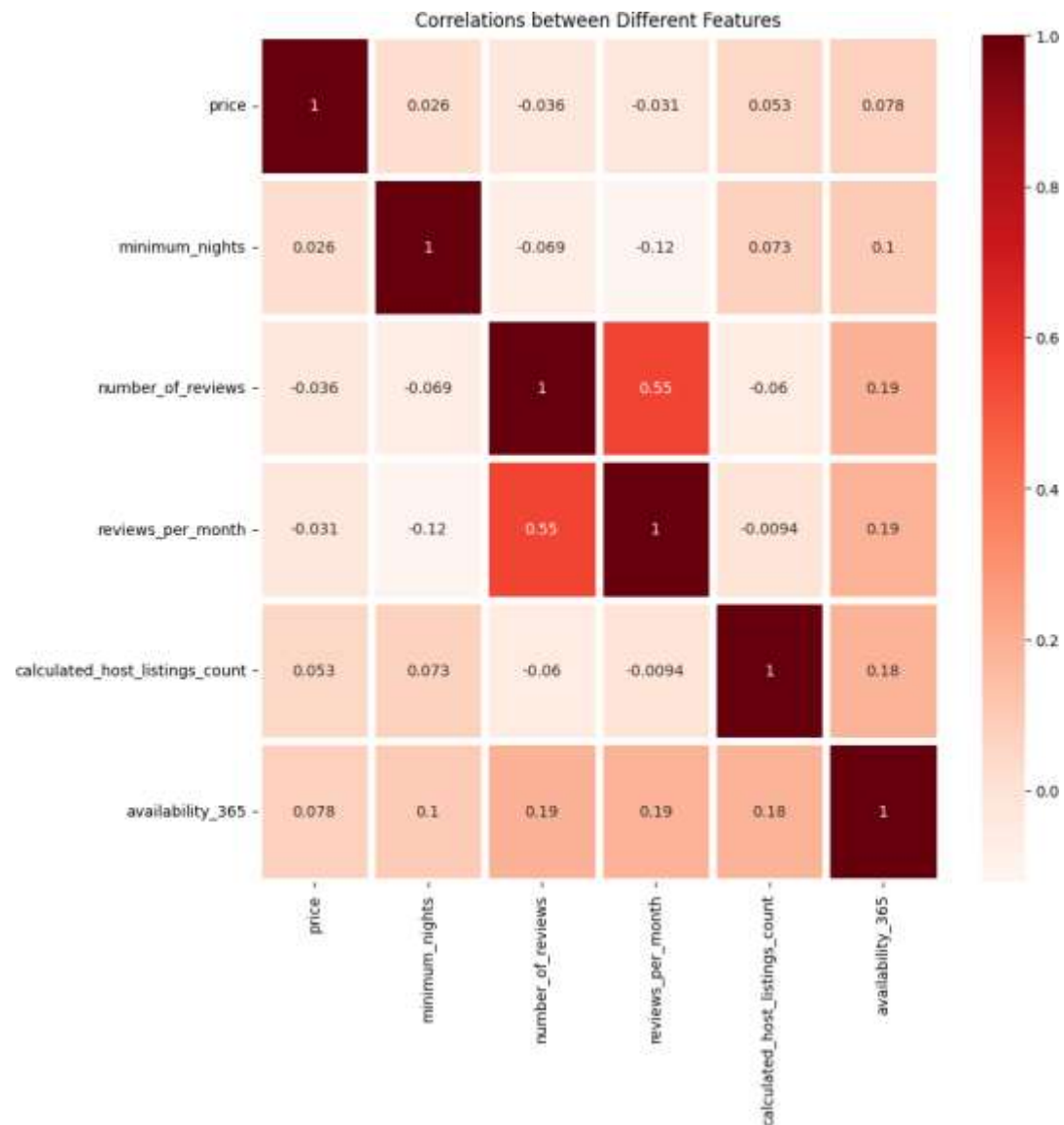




# EFFECT OF MINIMUM NIGHT CATEGORIES

Customers are more likely to leave reviews for lower number of minimum nights

# BIVARIATE AND MULTIVARIATE ANALYSIS



# DATA METHODOLOGY

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Conducted a thorough analysis of New York Airbnb's Dataset.

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Cleaned the data set using python.

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Derived the necessary features.

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Used group aggregation , pivot table and other statistical methods.

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Created charts and visualization using Tableau.



# CONCLUSION

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Strong significant insights are delivered based on various attributes in the dataset.

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Ample amount and variate of visuals have can used in the presentations for the stake-holders.

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Data collection team should collect data about review scores so that it can strengthen the later analysis.



THANK YOU