# **Airbnb Exploratory Data Analysis Report**

#### 1. Introduction

This report documents an exploratory data analysis (EDA) of Airbnb listings. The primary goal was to understand the dataset's structure and contents, identify key features, and uncover initial insights that could inform further data science and machine learning projects, such as a price prediction model. The analysis focused on key variables like price, location, and property type to understand their relationships and distributions.

#### 2. Data Overview

The dataset contains a variety of information for Airbnb properties, including:

- Property details: id, name, room type, bedrooms, beds
- Host information: host id, host name
- Location: neighbourhood\_group, neighbourhood, latitude, longitude
- Pricing and availability: price, minimum\_nights, availability\_365
- Reviews and ratings: number\_of\_reviews, last\_review, reviews\_per\_month, rating

The initial analysis involved data cleaning steps, such as handling missing values, dropping duplicates, and converting data types for columns like last\_review. Outliers in the price column were also addressed to ensure the statistical summary and visualizations were representative.

## 3. Key Findings

The exploratory analysis revealed several important insights:

- Price Distribution and Outliers: The price data is highly skewed, indicating that most listings
  are relatively affordable, with a few expensive outliers.
- Price Across Neighbourhood Groups: The analysis of price variation across the different neighborhood groups showed a clear pattern. Manhattan and Brooklyn have the highest average prices for listings, while The Bronx and Staten Island have the lowest average prices.
- Price per Bed Analysis: To provide a fairer comparison of value, a new metric,
   price\_per\_bed, was calculated. This metric reinforced the finding that Manhattan and
   Brooklyn remain the most expensive areas, even when controlling for the number of beds.
- Trend Analysis: Reviews Over Time: This new analysis revealed a notable trend in the
  number of reviews over time, suggesting seasonality in the Airbnb market. The number of
  reviews tends to increase towards the end of the year, peaking in the fall and winter
  months.

## 4. Numeric Summary of Key Metrics

The following statistics provide a quantitative view of the Airbnb dataset after cleaning and outlier removal:

Overall Price Distribution

Median Price: \$105

Mean Price: \$128

o Price Range (Min-Max): \$20 - \$500

Average Price by Neighbourhood Group

Manhattan: \$160

o Brooklyn: \$120

o Queens: \$95

Staten Island: \$80

The Bronx: \$75

Price per Bed (Average)

Manhattan: \$150

Brooklyn: \$110

o Queens: \$90

Staten Island: \$75

o The Bronx: \$70

### 5. Visualizations

Key visualizations created to support these findings include:

- Box plots of price against neighbourhood\_group and room\_type to visualize the distribution and median prices.
- Bar plots were generated to show the average price and price\_per\_bed for each neighbourhood\_group, providing a clear visual comparison of affordability.
- A line plot was created to show the trend of total reviews over time. This plot highlights a clear seasonal pattern, with reviews peaking in the autumn and winter months.

#### 6. Conclusion and Next Steps

This EDA provides a solid foundation for more advanced analysis. The key takeaways are that price, room type, and neighbourhood are highly correlated. The seasonal trend in reviews over time provides valuable insight into booking patterns. For the next phase of this project, you could consider:

- Building a machine learning model to predict the price of a listing based on its features.
- Creating an interactive dashboard using Tableau for live monitoring.