AIRBNB STORYTELLING CASE STUDY

AGENDA

- Objective
- Background
- Key findings
- Recommendations
- Appendix
 - Data methodology
 - Data model Assumptions

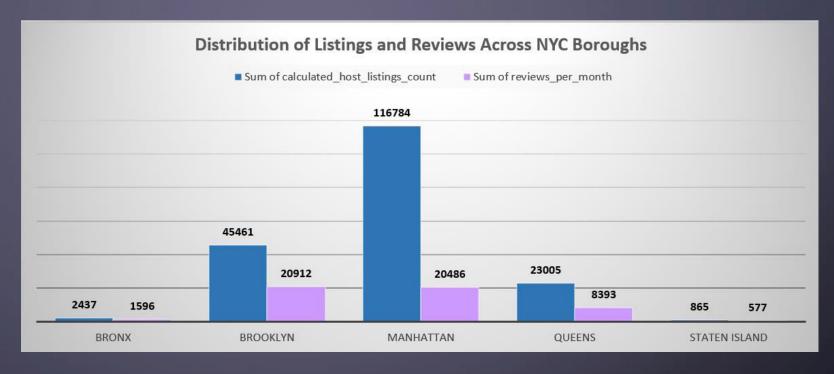
OBJECTIVE

- Analyze Airbnb's listing data and identify which types of hosts to acquire and target specific areas for expansion.
- Identify trends in Airbnb listing performance across NYC's boroughs
- Provide recommendations for improving revenue generation based on insights derived from data analysis.

BACKGROUND

- Airbnb has seen a significant decline in revenue over the past few months. There is
 a need to better understand how listings vary in performance across different
 boroughs and neighborhoods of NYC.
- Leaders want to understand trends in customer preferences, host acquisitions, and property performance.
- Insights are needed to increase bookings and ensure an optimal mix of listings.

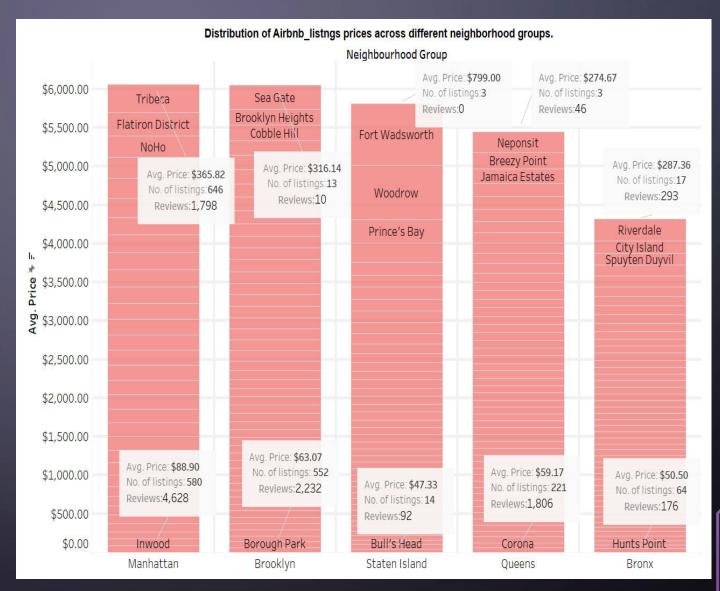
IMPACT ON MARKET THROUGH LISTINGS AND ENGAGEMENT



- Manhattan has a strong market presence but Brooklyn has higher customer engagement.
- With 23,005 listings and 8,393 reviews per month, **Queens** demonstrates a good balance, suggesting a stable market.
- Customer Engagement for Staten Island is **not significantly worse** than the Bronx, even though it has a **smaller market presence**.

MIN TO MAX AVERAGE PRICE DISTRIBUTION ACROSS NEIGHBORHOODS

- •Budget-friendly areas such as Inwood (Manhattan), Borough Park (Brooklyn), and Corona (Queens) shows significant customer engagement.
- •Manhattan has extreme neighborhood pricing variations, from high priced (Tribeca) to more affordable (Inwood), with solid listing and customer presence throughout.
- •Bronx has a **dynamic range** of property pricing, showing moderate customer presence in both premium (Riverdale) and budget listings (Hunts Point) areas.



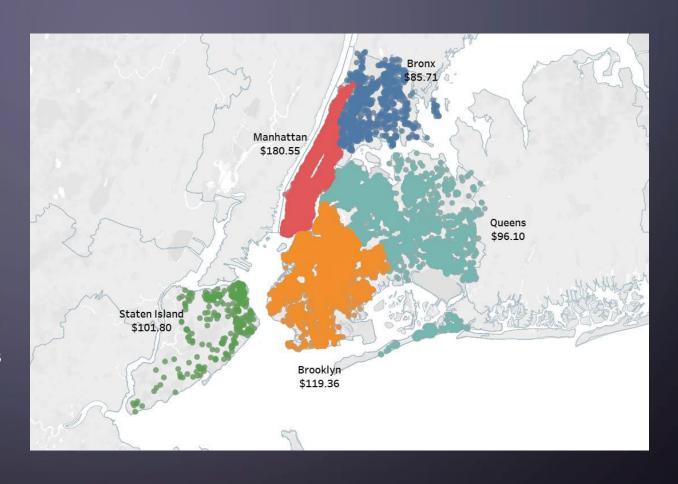
TOP-PERFORMING LISTINGS OF EACH NEIGHBOURHOOD GROUPS WITH HIGHEST REVIEWS



- In Bronx and Staten Island, specific hosts have **multiple top-performing listings** in the different neighborhoods indicating their dominance in these neighborhood groups.
- Hosts with properties in neighborhoods like Prospect Heights (Brooklyn) and Hell's Kitchen (Manhattan)
 dominate the market, reflecting strong booking trends.
- In Brooklyn, Manhattan and Queens, the highest-reviewed listings are concentrated in just a few neighborhoods i.e. Bedford-Stuyvesant, Hell's Kitchen and East Elmhurst, highlighting **key areas of high demand** within these boroughs.

PRICE AND LISTINGS ANALYSIS ACROSS NYC BOROUGHS

- **Manhattan leads** with the highest prices and maximum presence all over the location.
- Brooklyn and Queens with larger geographic area shows balanced pricing and substantial listings suggesting a stable market.
- Despite its large area, Staten Island's few listings demand higher prices but generate fewer reviews.



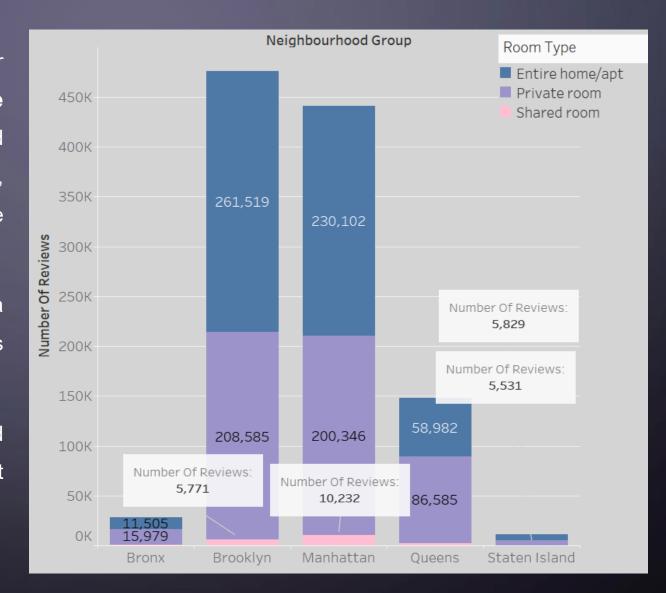
TOP 10 TRENDING NEIGHBORHOODS FOR AIRBNB LISTINGS

- While neighborhoods like Bedford-Stuyvesant and Williamsburg excel in review counts, others like Hell's Kitchen exhibit a high number of listings with lower review counts.
- Harlem has the most engagement among Manhattan neighborhoods, suggesting strong interest area.
- Neighborhoods like East Village, Crown Heights, East Harlem, Upper West Side with fewer listings but high reviews, indicate strong demand.

Bedford-Stuyvesant Brooklyn Total Reviews:107,881 Num_Of_Listings:9,687	Harlem Manhattan Total Reviews:72,803 Num_Of_Listings:5,814	East Village Manhattan Total Reviews:42,494	Crown Heights Brooklyn Total Reviews:35,909
Williamsburg Brooklyn Total Reviews:82,851 Num_Of_Listings:5,596	Bushwick Brooklyn Total Reviews:51,189 Num_Of_Listings:6,377	Manhattan Total	Upper West Side Manhattan Total Reviews:35,017
	Hell's Kitchen Manhattan Total Reviews:48,683 Num_Of_Listings:17,943		
		Upper East Side Manhattan Total Reviews:31,170 Num_Of_Listings:12,000	

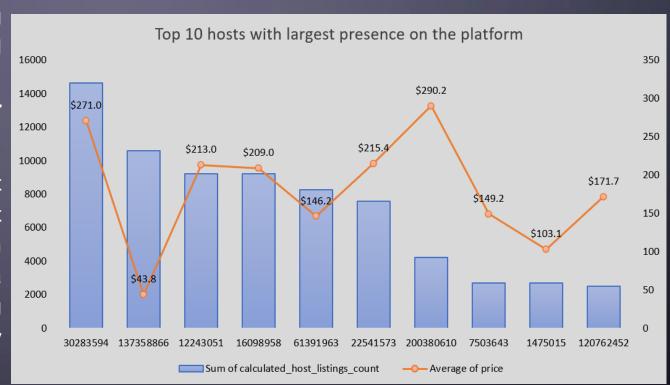
ANALYSIS OF THE MOST PREFERRED ROOM TYPES

- With over 200K reviews each for
 Private rooms and Entire
 homes/apt , Brooklyn and
 Manhattan continue to dominate,
 showcasing the popularity of these
 room types in high-demand areas.
- Both Queens and the Bronx show a unique trend, private rooms dominate over entire home/apt.
- Shared rooms in Manhattan and Brooklyn, though fewer still attract more reviews than in other locations.



ANALYZING TOP HOSTS' PRICING STRATEGY

- Host ID 30283594 dominates the market, while most top hosts managing over 8,000 listings have maintained their pricing in the \$140-\$200 range, indicating a trend towards mid-tier pricing among high-volume hosts.
- Host ID 137358866 offers the lowest average price among top hosts at \$43.8, contrasting significantly with Host ID 200380610, who has set a premium rate of \$290.2, indicating potential differences in property quality.
- Hosts such as 1475015 and 7503643, with consistent average prices around \$100–\$150, demonstrate stability.



PRECOMMENDATIONS

- Prioritize acquiring properties in high-demand, high-price neighborhoods like Tribeca and Brooklyn Heights to maximize revenue.
- Develop a marketing strategy for budget-friendly neighborhoods to boost bookings and increase overall occupancy.
- Focus on acquiring more properties in Staten Island to balance demand and reduce prices, attracting more bookings. Highlight and promote Staten Island properties to increase visibility and bookings, leveraging year-round availability.
- Given the high demand for private rooms in Queens and the Bronx, hosts should focus on improving and marketing these listings to attract more bookings.
- Enhance the appeal of shared rooms by upgrading amenities and ensuring clean, safe and comfortable conditions to attract more customers.
- Improve property listings in East Village to take advantage of the strong demand shown by reviews, and consider adding special features or experiences to attract more customers.
- Implement data-driven pricing strategies for hosts with many listings but low prices to increase occupancy and align supply with the right customer segment for each neighborhood.
- Study the strategies of high-value hosts to understand their strategies, pricing models.
- Given the concentration of high reviews in neighborhoods like Hell's Kitchen (Manhattan) and East Elmhurst (Queens), Airbnb should consider acquiring hosts in these areas.
- Focus marketing efforts on neighborhoods with higher average prices but fewer listings, such as Fort Wadsworth or Neponsit.

APPENDIX: DATA METHODOLOGY

- Data description:
- Data wrangling:
 - Checking duplicate entries

```
# check for duplicates
airbnb.duplicated().sum()
0
```

• Handling the missing values by imputing and removing rows with less than 5% of missing values.

```
# filling the missing values for reviews_per_month with 0
airbnb['reviews_per_month'].fillna(0, inplace = True)
# drop the 'last review' column
airbnb.drop('last_review', axis=1, inplace=True)
# drop rows with missing values in the 'name' column
airbnb = airbnb[~airbnb['name'].isnull()]
# check missing values
round((airbnb.isnull().sum()/len(airbnb)*100).sort_values(ascending = False), 2)
name
host_id
                                   0.0
neighbourhood_group
neighbourhood
latitude
longitude
room_type
price
minimum_nights
number of reviews
reviews per month
calculated_host_listings_count
availability_365
dtype: float64
```

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	availability_365 number of days when listing is available for booking	

APPENDIX: DATA METHODOLOGY

- Identifying and handing outlier: Capping values above the 99th percentile
- Creating Calculated Fields:
 - To calculate average price across the neighbourhood groups to see aggregate

t capping the ouliers at 99th percentile

for col in num cols:

'availability_365']

percentile 99 = airbnb[col].quantile(0.99)

- Binning minimum nights
- Supporting Tools:
 - Python
 - Excel
 - Tableau

num cols = ['price', 'minimum nights', 'number of reviews', 'reviews per month', 'calculated host listings count',

airbnb[col] = np.where(airbnb[col] > percentile_99, percentile_99, airbnb[col])

```
Custom Bins for Mini

IF [Minimum Nights] <= 5 THEN "1-5"

ELSEIF [Minimum Nights] <= 10 THEN "6-10"

ELSEIF [Minimum Nights] <= 15 THEN "11-15"

ELSEIF [Minimum Nights] <= 20 THEN "16-20"

ELSEIF [Minimum Nights] <= 25 THEN "21-25"

ELSEIF [Minimum Nights] <= 30 THEN "26-30"

ELSEIF [Minimum Nights] <= 35 THEN "31-35"

ELSE "35+"

END
```

APPENDIX: DATA ASSUMPTIONS

- We assumed that a higher number of reviews reflects the popularity and demand for a listing, suggesting higher occupancy rates.
- We assumed that hosts' pricing and availability strategies are consistent over time.
- We assumed that higher pricing generally correlates with higher quality.
- We assumed that historical data on reviews and occupancy would be indicative of future trends.