



AirBnb Analysis

AI for Business Final Project

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Agenda

1. **Background**
2. **Dataset Overview**
3. **Features Exploration**
4. **Occupancy Rate Prediction Model**
5. **Summary.**



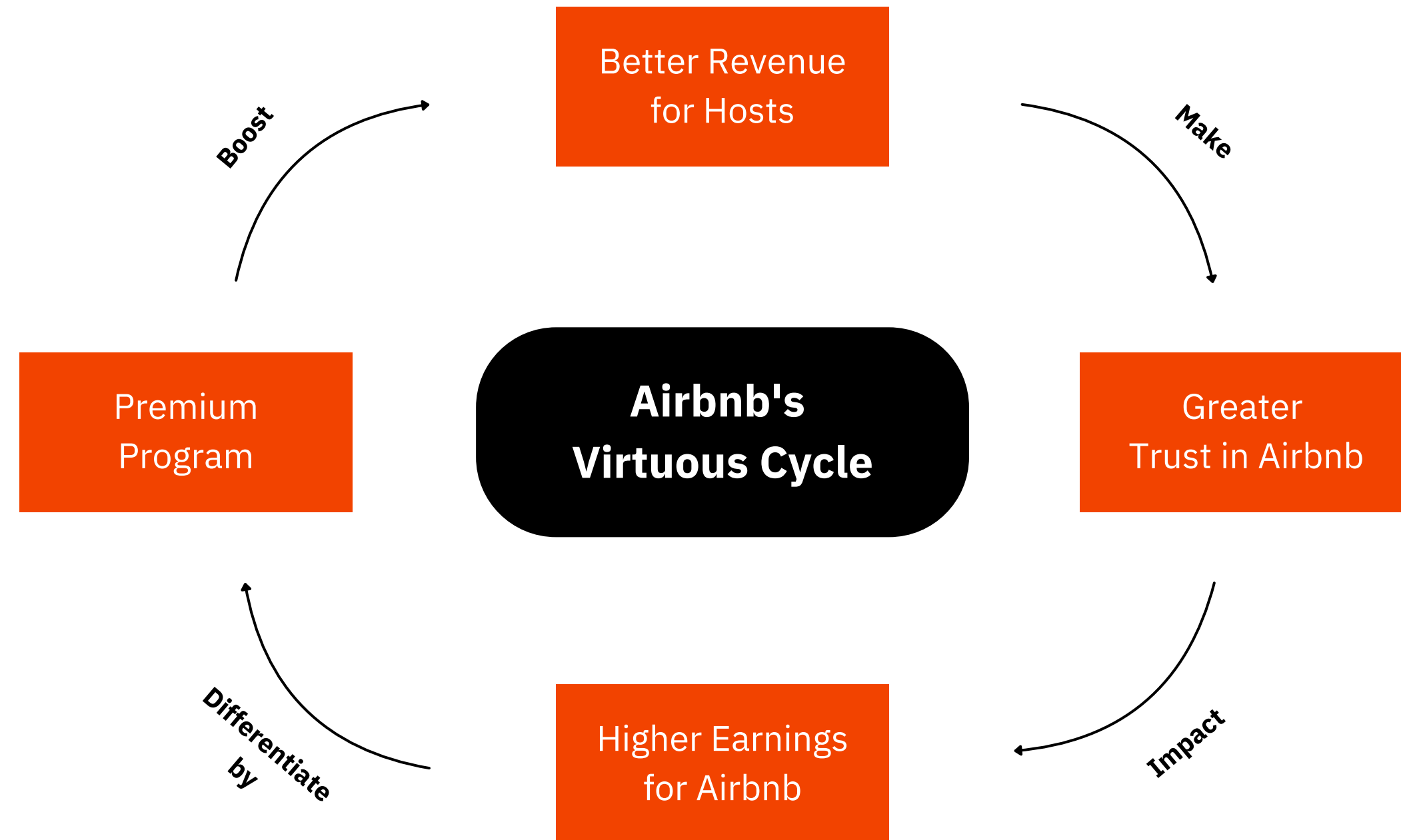
Background

Business Problem: Driving Revenue Growth for Airbnb and Clients

- Need to stand out in a competitive marketplace.
- Airbnb's Revenue Growth is tied to Clients' Success.
- Introducing a Premium Program

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Why We Increase Client Revenue:

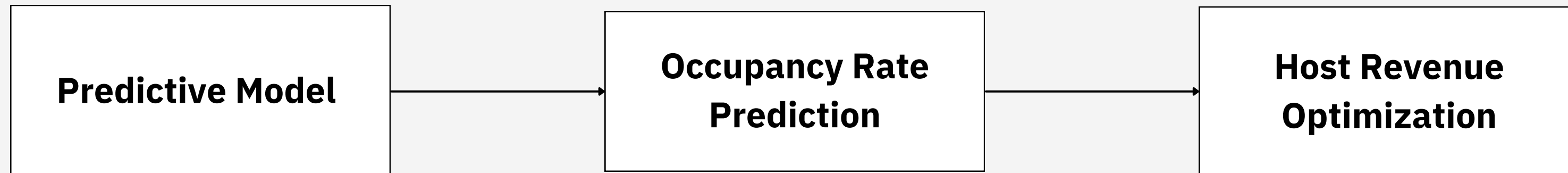


Background

Revenue Formula in AirBnb's:

$$\text{Revenue} \uparrow \downarrow = \text{Nightly Rate} \times \text{Occupancy Rate} \uparrow \downarrow$$

How We Increase Client Revenue:

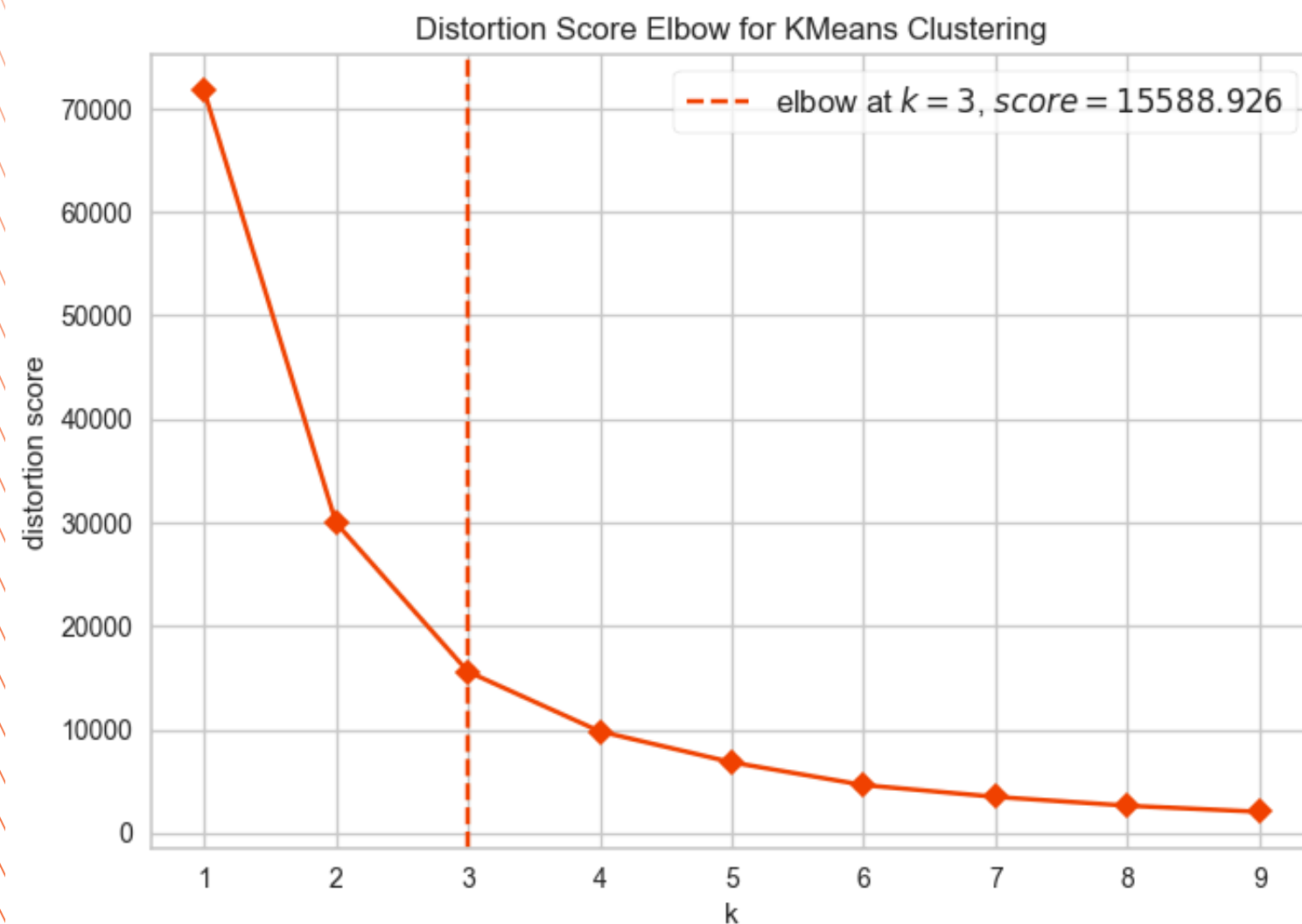




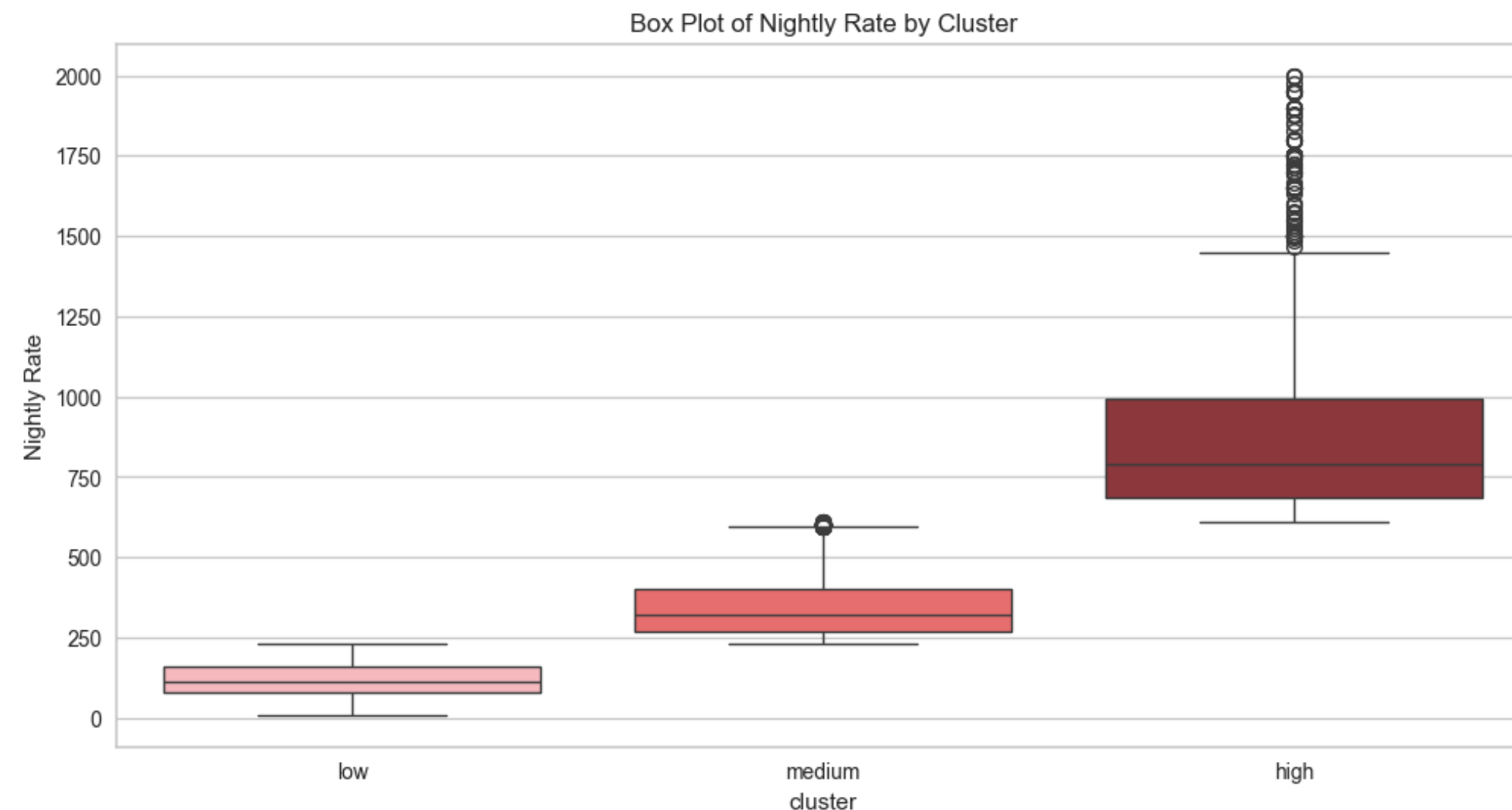
Dataset Overview

Clustering Airbnb Listings by Nightly Rates

Highlight the elbow point at $k=3$, showing the optimal cluster count.



Average nightly rate by three clusters, including low, medium, and high.



LOW: \$119

MEDIUM: \$342

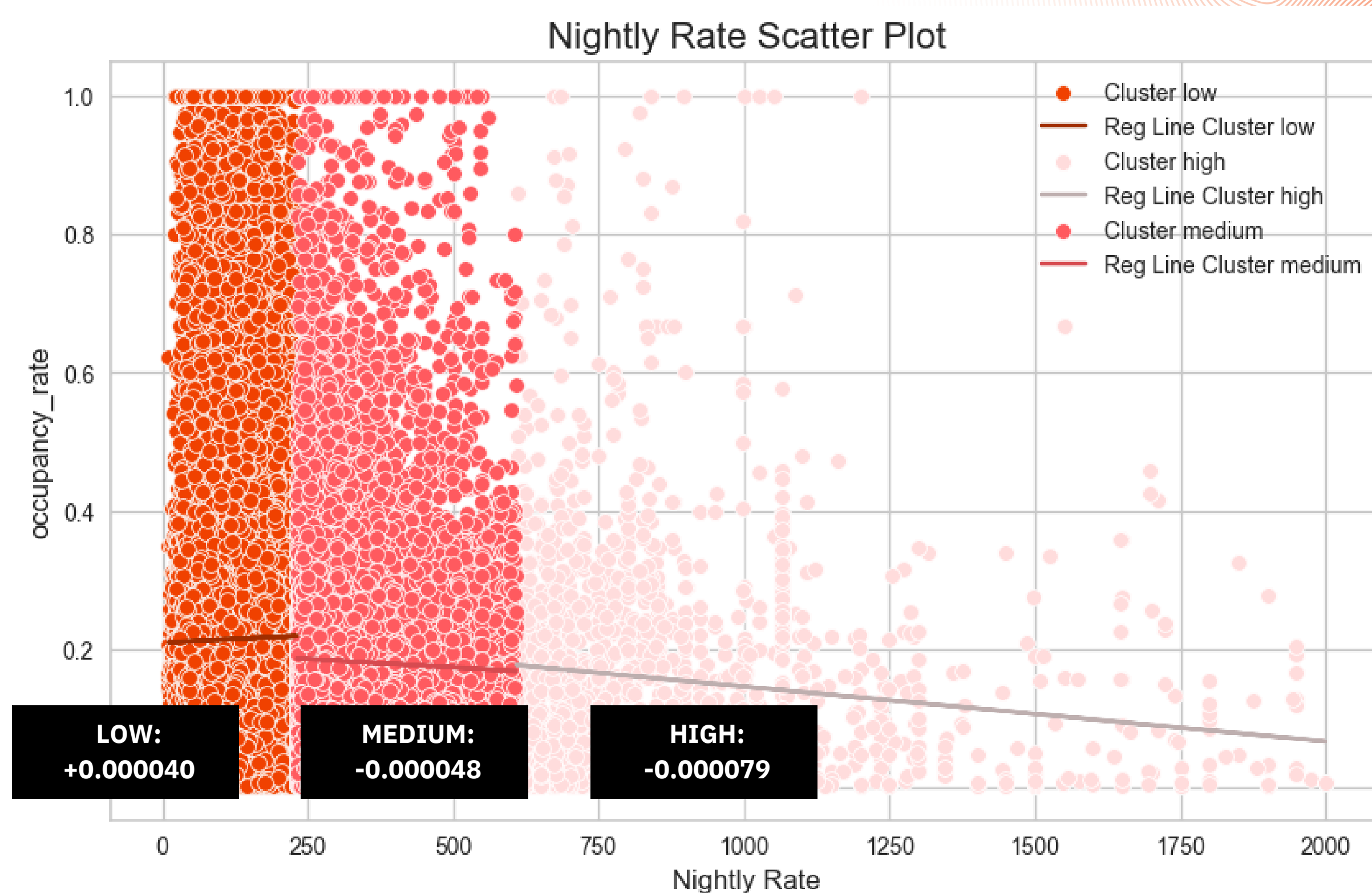
HIGH: \$873



Dataset Overview

Understanding Price Sensitivity in Clusters

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Features Exploration



Significant Variables to Occupancy Rate

- Use linear regression to identify which variables significantly impact the occupancy rate

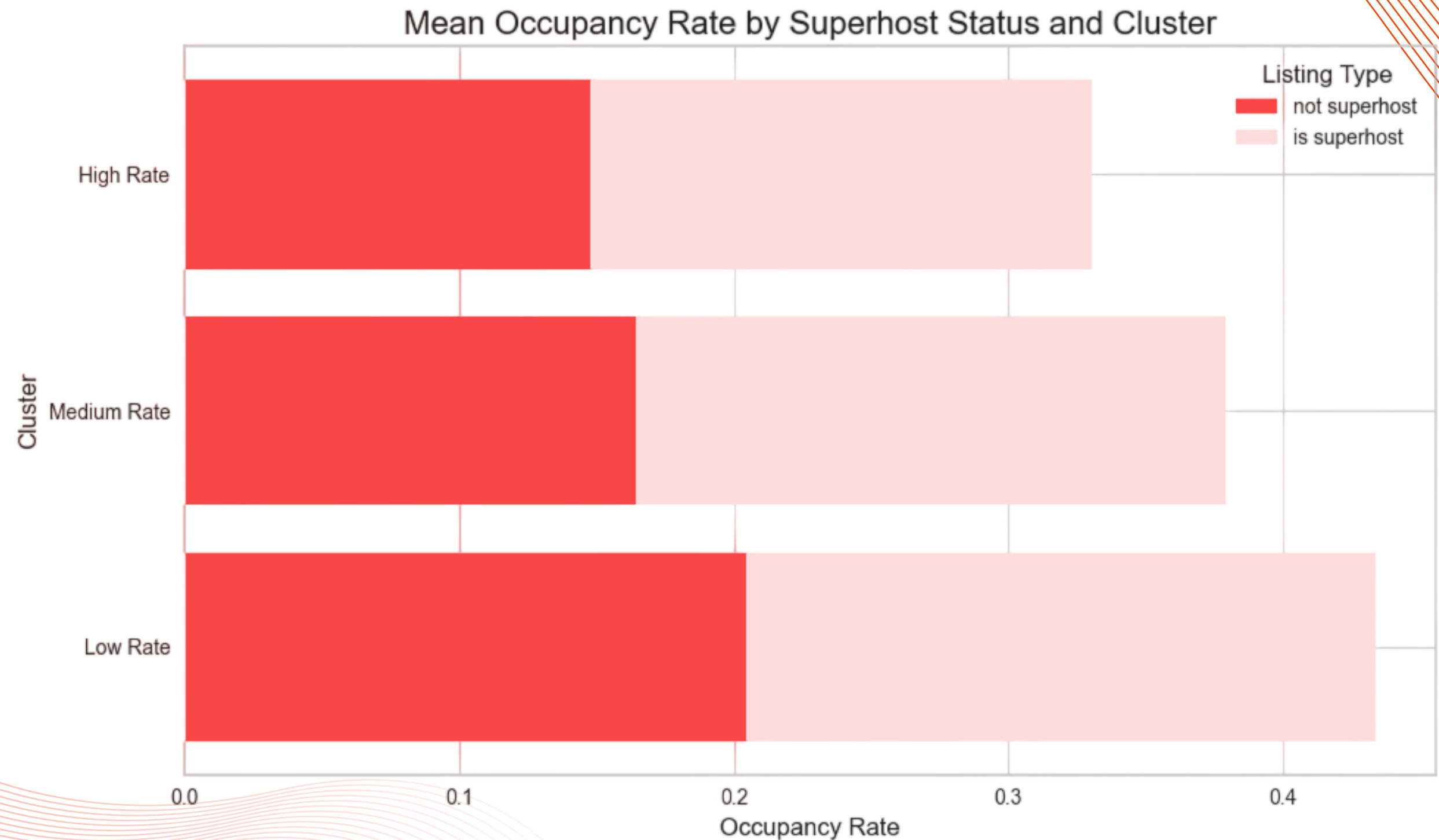
Rate Cluster	Low	Medium	High
Positive features	<ul style="list-style-type: none">booked_days 0.00681rating_ave_pastYear 0.00526Max Guests 0.00135rev_available_days 0.00037Number of Reviews 0.00019Rating Overall 0.00013	<ul style="list-style-type: none">tract_booking_share 2.10435tract_superhosts_ratio 0.08094booked_days 0.00602zip_black_nothispanic_percent 0.00117tract_prev_superhosts 0.00083hostResponseNumber 0.00011	<ul style="list-style-type: none">Pets Allowed 0.01554booked_days 0.00655numCancel_pastYear 0.00141prev_numReserv_pastYear 0.00014numReservedDays_pastYear 0.00004
Negative features	<ul style="list-style-type: none">Pets Allowed -0.00594available_days -0.00193Number of Photos -0.00026Number of Reviews -0.00021Cleaning Fee (USD) -0.00011booked_days_avePrice -0.00007	<ul style="list-style-type: none">Pets Allowed -0.00808available_days -0.00163tract_superhosts -0.00074Cleaning Fee (USD) -0.00007booked_days_period_tract -0.00002numReserv_pastYear -0.00001	<ul style="list-style-type: none">available_days -0.00155tract_unique_prices -0.00097prev_time_to_date_mean -0.00027numReserv_pastYear -0.00011numReservedDays_pastYear -0.00005revenue -0.000001

Features **Exploration**

Variables That Impact Super Host



- Showing the average impact on occupancy rates of host with or without a superhost

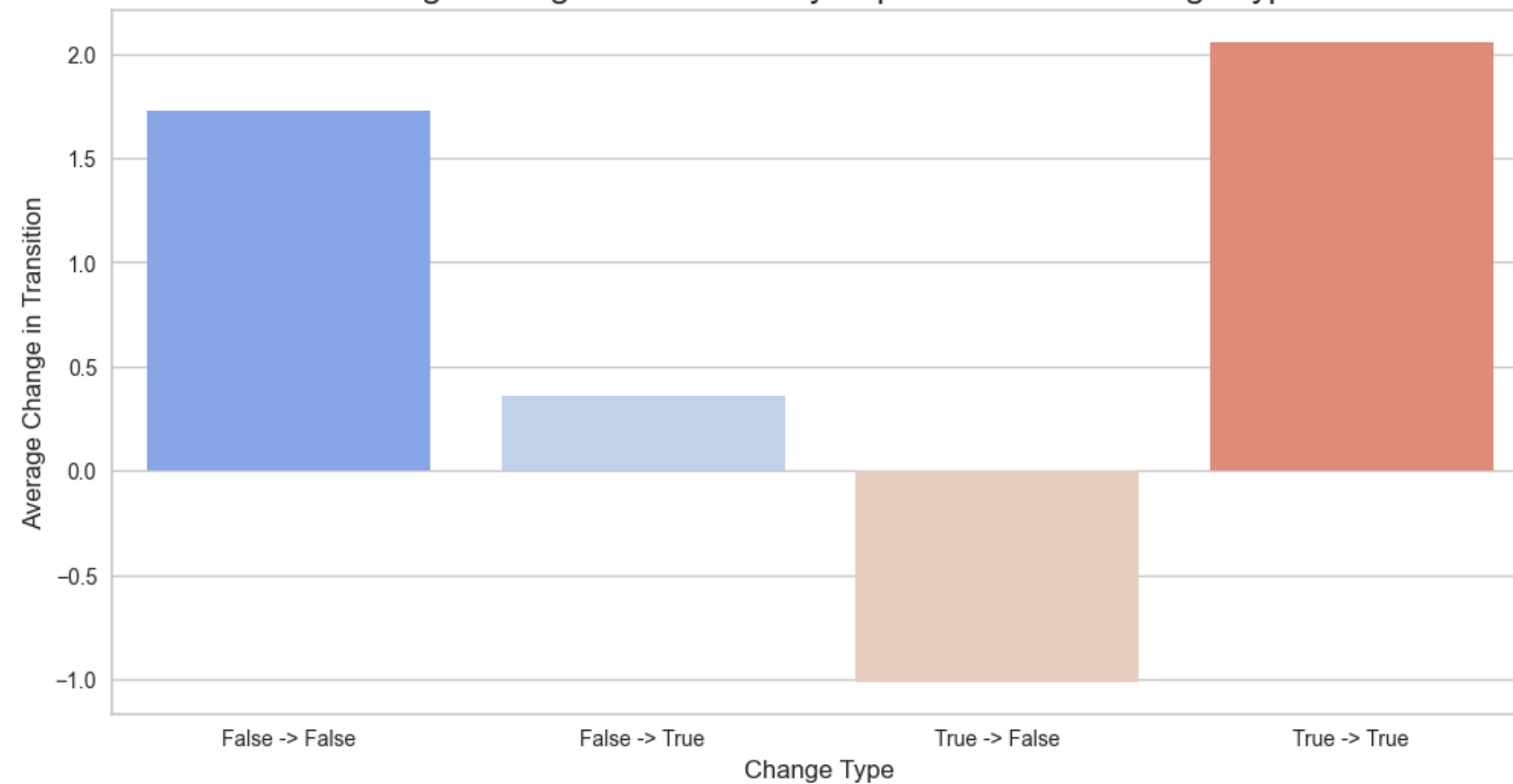


Features Exploration

Superhost Transition



Average Change in Transition by Superhost Status Change Type



Examine how superhost status impacts occupancy rate

- Label
 - X: Superhost status changes
 - Y: Mean of changes in occupancy rate
- Insight
 - Getting superhost status (False → True) leads to increased occupancy.
 - Losing superhost status (True → False) negatively impacts occupancy.

Features Exploration



Listing type

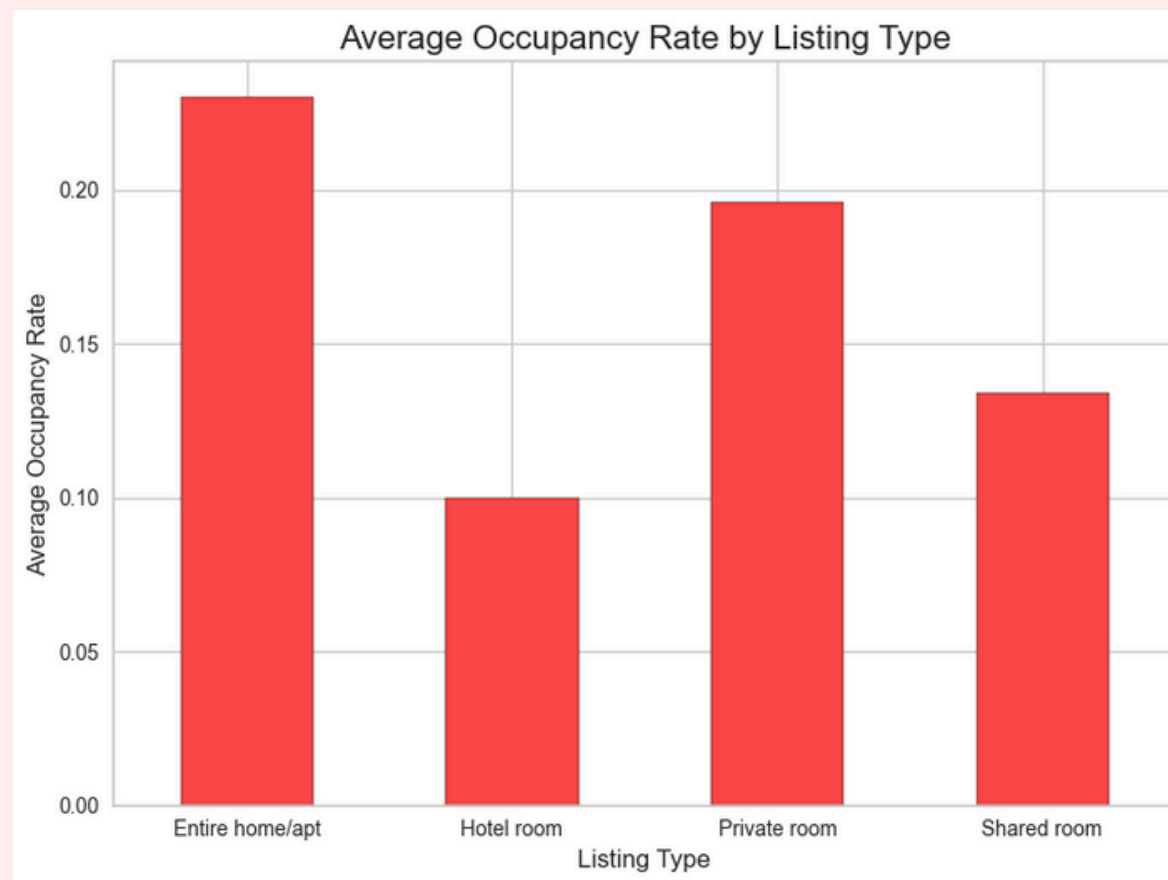
Low

a. Entire home/apt

Performs best and can be promoted more.

b. Hotel room, Shared room

Strategies need to be optimized to attract guests.



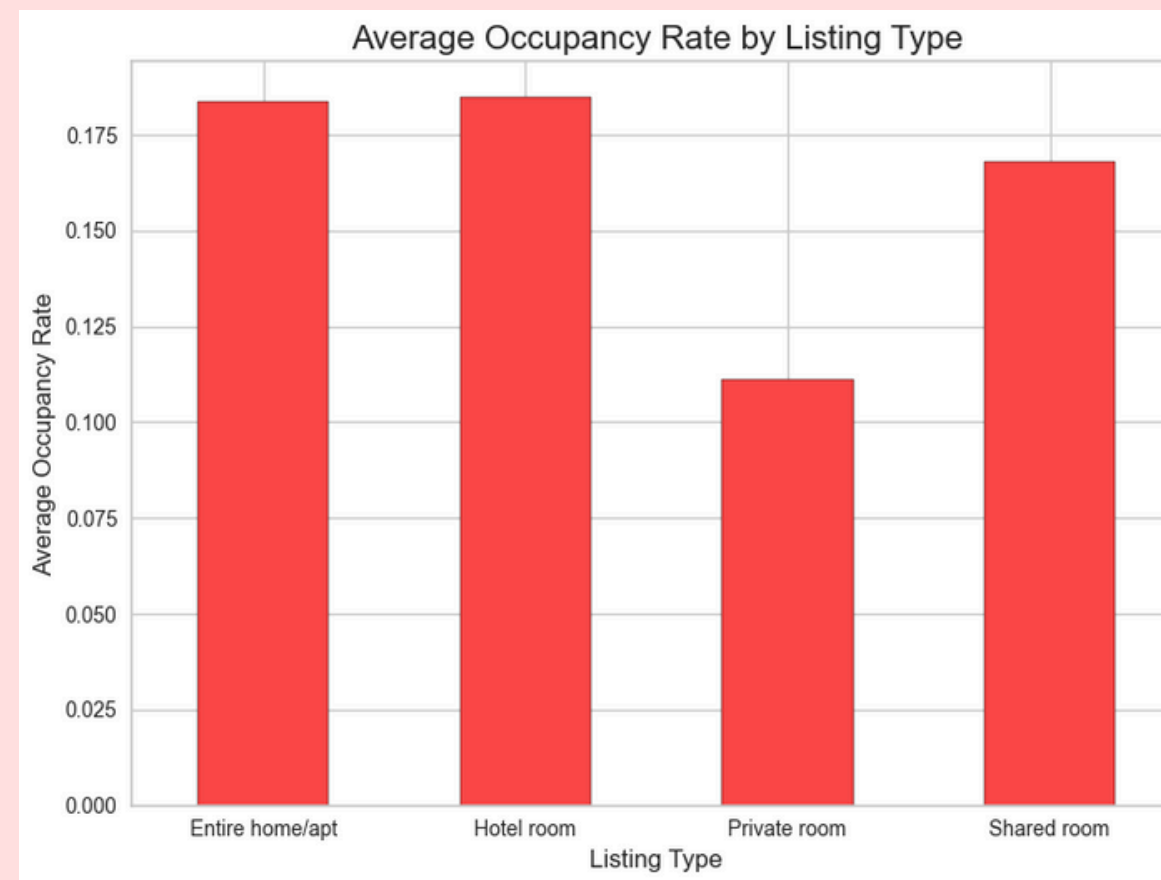
Medium

a. Hotel room

Best performing, first choice for mid-priced guests.

b. Private room

The weakest performance, needs to improve its attractiveness.



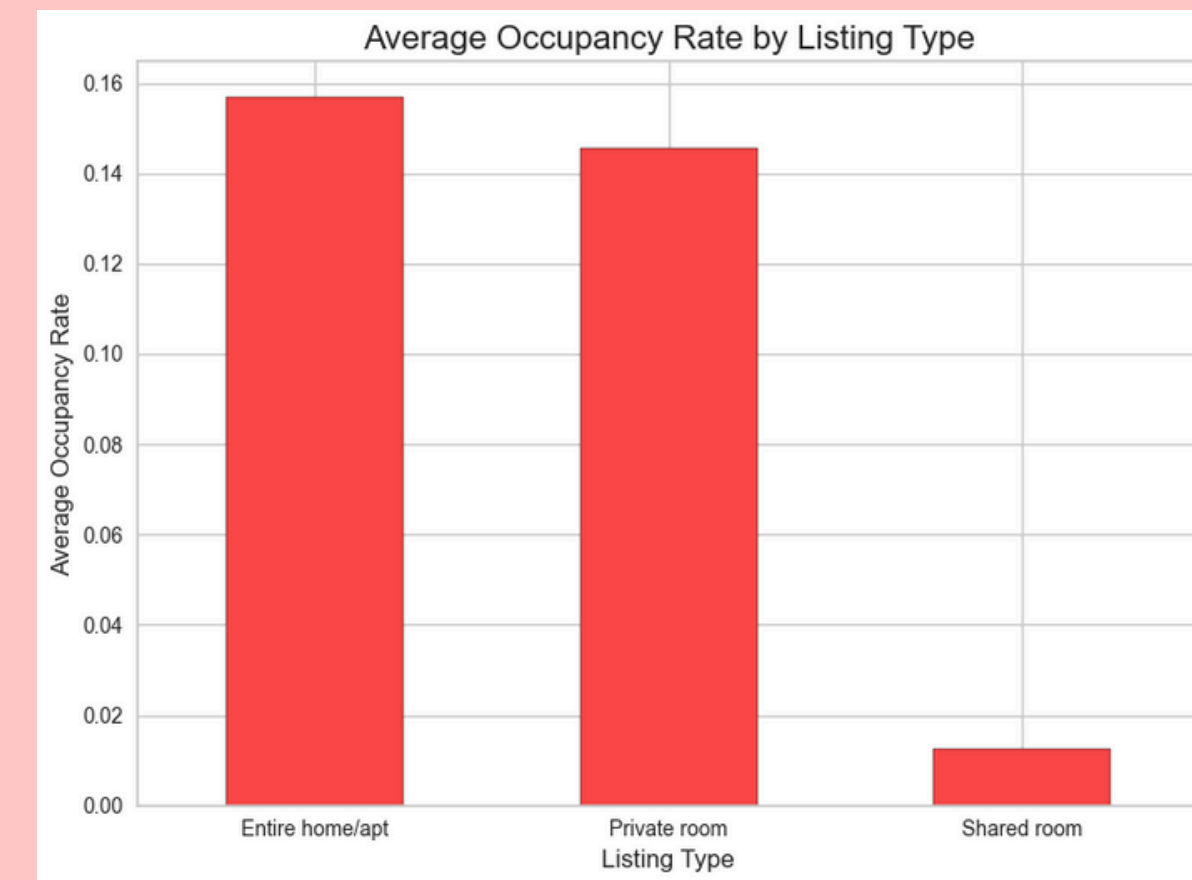
High

a. Entire home/apt, Private room

Performance is similar.

b. Shared room

almost no need and needs to be repositioned.



Features **Exploration**



Variable Selection

- **Dummy variables** that significantly influence the occupancy rate

Low

- host_is_superhost_in_period
- prev_host_is_superhost
- superhost_change
- Pets Allowed
- Instantbook Enabled

Medium

- host_is_superhost_in_period
- prev_host_is_superhost
- superhost_change

High

- host_is_superhost_in_period
- Pets Allowed
- Instantbook Enabled

- Identify significant **interaction terms**:
 - a. Create interaction features based on correlations exceeding 0.8
 - b. Feature selection using Lasso



Occupancy Rate Prediction Model

Gradient Boosting Tree



High predictive power

It provides better predictive accuracy.



Flexibility

It can be adapted for regression, classification, and ranking tasks.



Suitability for moderate dataset size

It performs well on datasets of moderate size.

Model Performance

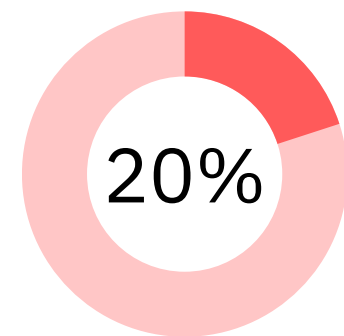
Gradient Boosting Tree



Input

All significant variables

- Low: 54
- Medium: 43
- High: 32



test dataset

Test Result

Root Mean Squared Error

- low: 0.0091
- Medium: 0.0093
- High: 0.0305

R-squared

- Low: 0.9977
- Medium: 0.9968
- High: 0.9591

Premium Program

Benefits

Driving Revenue Growth
for Customers



Occupancy Rate Prediction

Maximize hosts' profits, regardless of their price segment.



Customized Suggestions

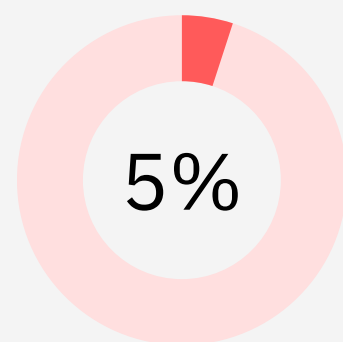
Friendly for new hosts while also benefiting existing hosts by enhancing their service quality.



Premium Program

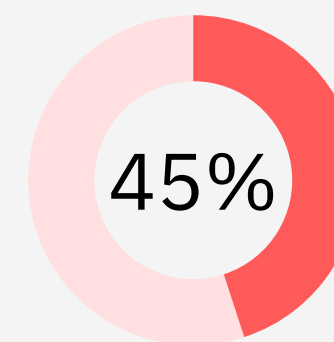
Customized suggestions

Assume you were a low-performing host, Eric:



Before Adjustment

Superhost = 0
Pets Allowed = False
available_days = 245
booked_days = 11



After Adjustment

Superhost = 1
Pets Allowed = True
available_days = 300
booked_days = 100