

### IN SEATTLE

Analyzing prevailing market data trends and forecast rental prices for AirBNB listings in the Seattle Metro area.

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# OUR PROJECT GOALS



#### **PROJECT GOALS**

KEY
FEATURES
AFFECTING
PRICE PER
NIGHT

What are the **key features** and characteristics that make certain properties **highly sought after**, ultimately influencing their **per-night pricing**?

PERCEPTION
SHAPING
PRICING
DYNAMICS

How does the **perception** of property reliability, as influenced by **consumer feedback** and metrics, play a role in shaping these **pricing dynamics**?



#### **BACKGROUND**

You want to invest in AirBnB real estate or want the best deal possible. Our central question was to find the optimal AirBnB to suite the specific needs of the businessman or consumer.



#### **REASONING**

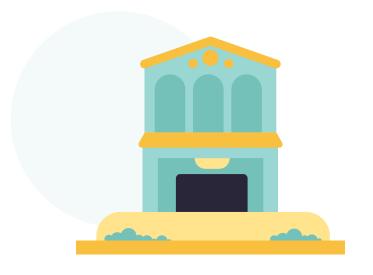
AirBnBs are a marketplace that are very hard to find a great deal on. Because of this, it is very useful to analyze AirBnB data to view many different details about them throughout their history.



#### **SIGNIFICANCE**

We were able to
predict the price of
AirBnBs, analyze
trends via linear
regression, and view
multivariable
relationships through
general exploratory
data analysis.

## **DATA SOURCES**& CLEANING



#### **DATA SOURCES & CLEANING**

Sourced **two** datasets from openly accessible **AirBNB** archives.

Combined dataset accentuating an array of features from over 3,000 properties in the Seattle region.

Filled in **missing data**, removed i**dentical duplicates**, and dealt with **unusual data** points.







 $\alpha = 0.05$ 

df = 3817

## AVERAGE PRICE

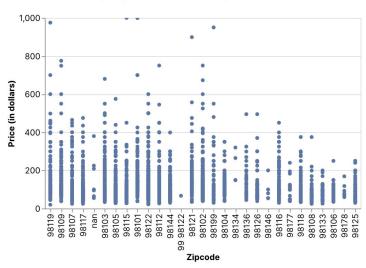
\$127.98 ±

\$176.92

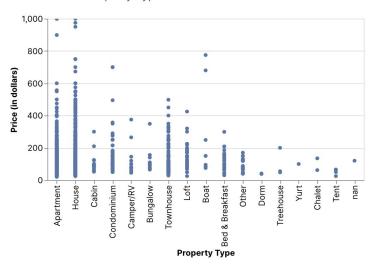
## STANDARD ERROR

\$90.24

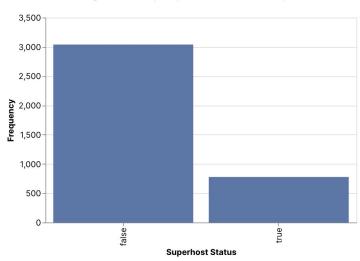
#### Distribution of Property Price Based on Zipcode



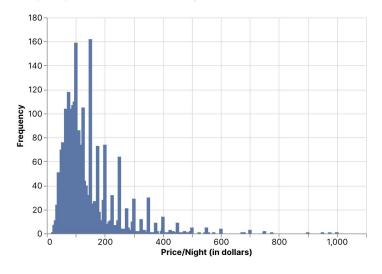
#### Distribution of Property Type vs. Price

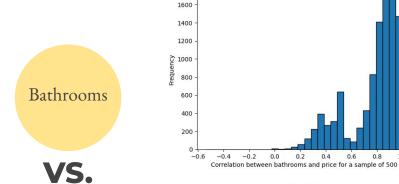


#### Number of Listings Posted by Superhosts vs. Non-Superhosts

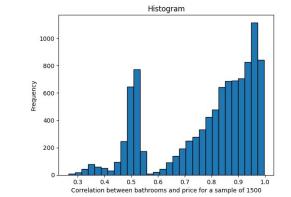


#### Frequency Distribution of Price/Night Values for Airbnbs in Seattle

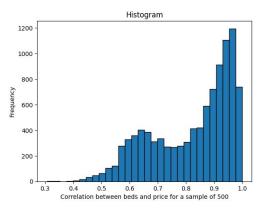


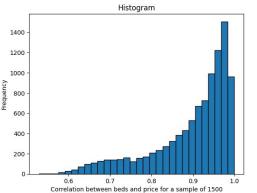


Price



Histogram



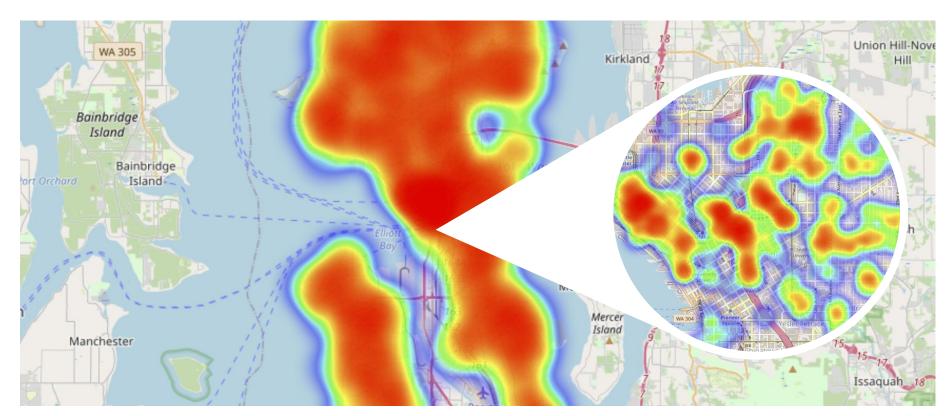






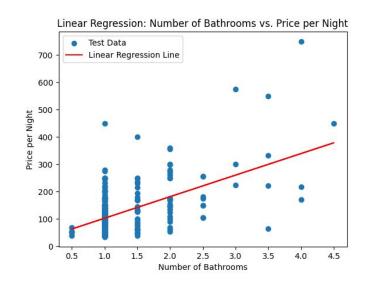


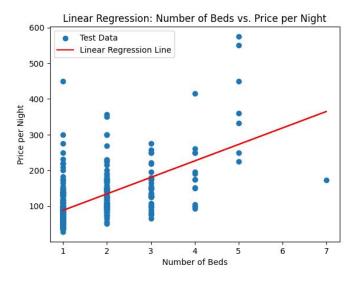




# LINEAR PRICE PREDICTION MODELS









## RANDOM FOREST REGRESSOR ALGORITHM



#### **RANDOM FOREST REGRESSOR ALGORITHM**

#### **ML ALGORITHM**

The Random Forest Regressor is a Machine learning algorithm that enhances regression predictions.

#### **DECISION TREES**

It builds multiple decision trees through **bootstrapping**, utilizing random subsets of the data and features.

#### **PREDICTIONS**

Decision trees are **used to make predictions** by
averaging outputs,
reducing overfitting, and
improving accuracy.







#### **RANDOM FOREST REGRESSOR ALGORITHM**

```
Host_listings_count: 5
           Accomodation capacity: 10
                         Bathrooms: 5
                         Bedrooms: 4
                              Beds: 7
                  Security_deposit: 45
                     Cleaning fee: 50
                   Guests included: 7
Price_per_head_over_max_capacity: 25
                  Minimum_nights: 1
                 Maximum_nights: 10
                 Availability 365: 156
             Number of reviews: 500
              Review scores rating: 8
    Calculated_host_listings_count: 20
              Reviews_per_month: 10
```

```
Ready
                                                             # Render the graph as a PNG image
graph.render("decision_tree")
# Initialize an empty dictionary to store user inputs
user_inputs = {}
# Prompt the user to input values for each feature
for feature in features:
    user_input = input(f"Enter the value for {feature}: ")
    user_inputs[feature] = float(user_input) # Assuming the input is numeric
# Create a DataFrame with the user's inputs
user_data = pd.DataFrame([user_inputs])
# Make a price prediction
predicted_price = model.predict(user_data)
# Display the predicted price
print(f"Predicted Price: ${predicted_price[0]:.2f}/night")
Predicted Price: $473.28/night
```

## FINAL TAKEAWAYS



#### **FINAL TAKEAWAYS**

