

A complex network graph composed of numerous small, semi-transparent gray dots connected by thin white lines, forming a dense web-like structure.

Capstone 3: Airbnb

By: Pedro Rodriguez

Data

Listing.csv

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 3817 entries, 0 to 3816
Data columns (total 15 columns):
 #   Column           Non-Null Count  Dtype  
--- 
 0   Unnamed: 0        3817 non-null    int64  
 1   id               3817 non-null    int64  
 2   latitude          3817 non-null    float64 
 3   longitude         3817 non-null    float64 
 4   zipcode           3817 non-null    int64  
 5   property_type     3817 non-null    object  
 6   room_type          3817 non-null    object  
 7   price              3817 non-null    float64 
 8   accommodates      3817 non-null    int64  
 9   cleaning_fee       3817 non-null    float64 
 10  availability_30    3817 non-null    int64  
 11  bathrooms          3817 non-null    float64 
 12  bedrooms           3817 non-null    float64 
 13  review_scores_rating 3817 non-null    float64 
 14  review_scores_cleanliness 3817 non-null    float64 
dtypes: float64(8), int64(5), object(2)
memory usage: 447.4+ KB
```

calendar.csv

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1393570 entries, 0 to 1393569
Data columns (total 5 columns):
 #   Column           Non-Null Count  Dtype  
--- 
 0   Unnamed: 0        1393570 non-null  int64  
 1   listing_id        1393570 non-null  int64  
 2   date              1393570 non-null  object  
 3   available          1393570 non-null  int64  
 4   price              1393570 non-null  float64 
dtypes: float64(1), int64(3), object(1)
memory usage: 53.2+ MB
```

2016

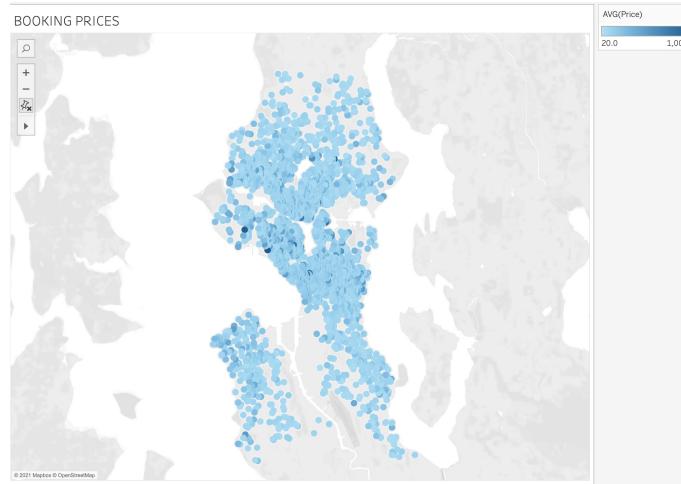


Weekly Availability

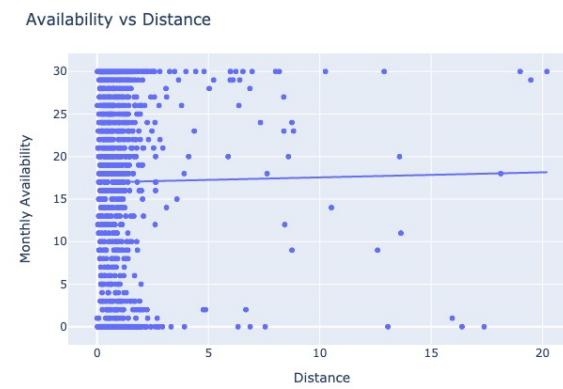


Average booking price in 2016



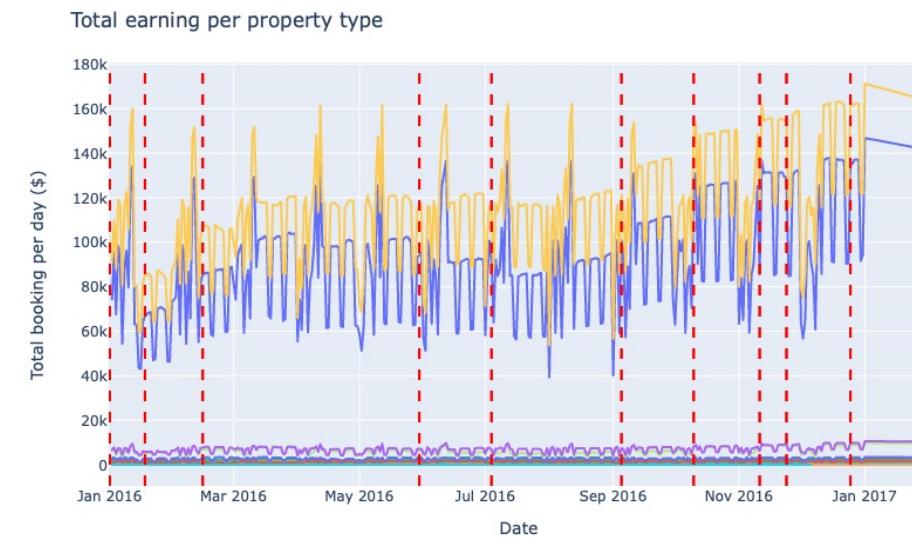
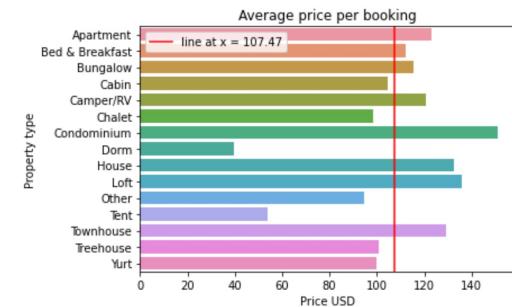
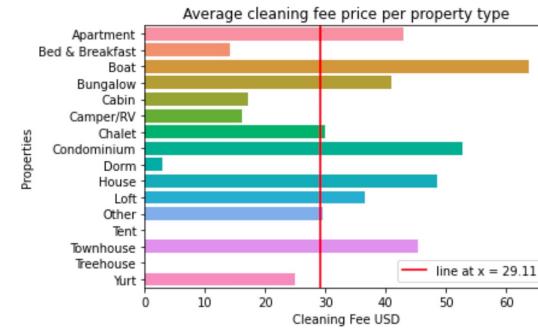


Locations



Property Type

Property	Average Price	Average Cleaning fee
Condominium	\$151.11	\$54.51
Loft	\$135.70	\$37.09
House	\$132.35	\$42.24
Apartment	\$122.93	\$38.58

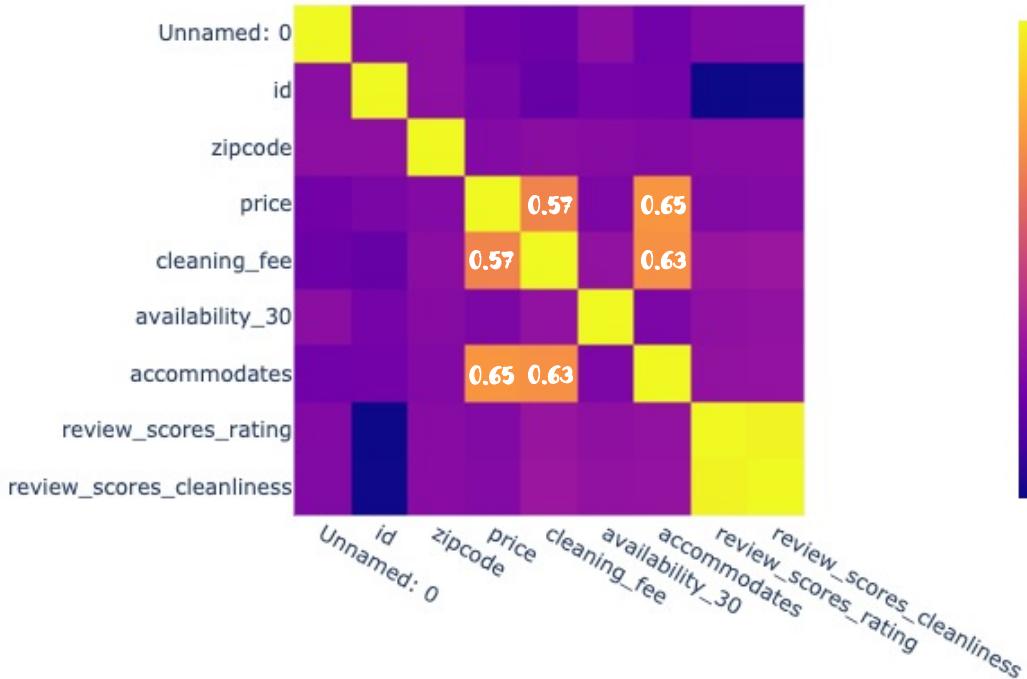


property_type

- Apartment
- Bed & Breakfast
- Boat
- Bungalow
- Cabin
- Camper/RV
- Chalet
- Condominium
- Dorm
- House
- Loft
- Other
- Tent
- Townhouse
- Treehouse
- Yurt

Relationships

Heatmap



Cleaning Fee vs Price

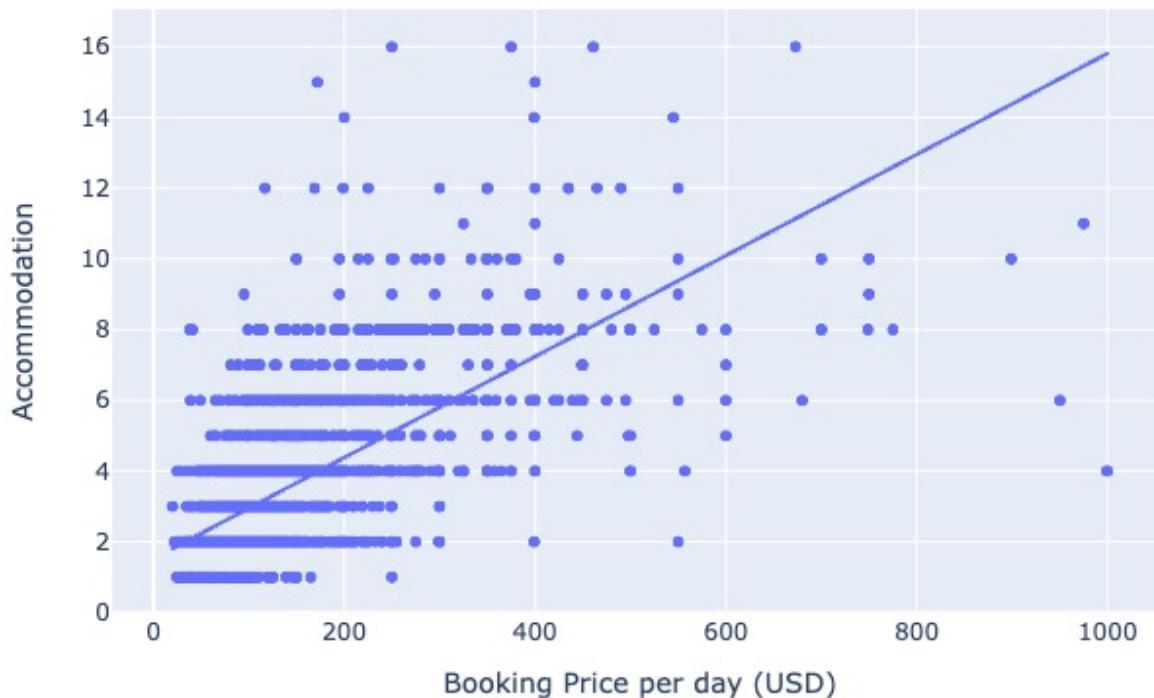


Cleaning Fee vs Availability

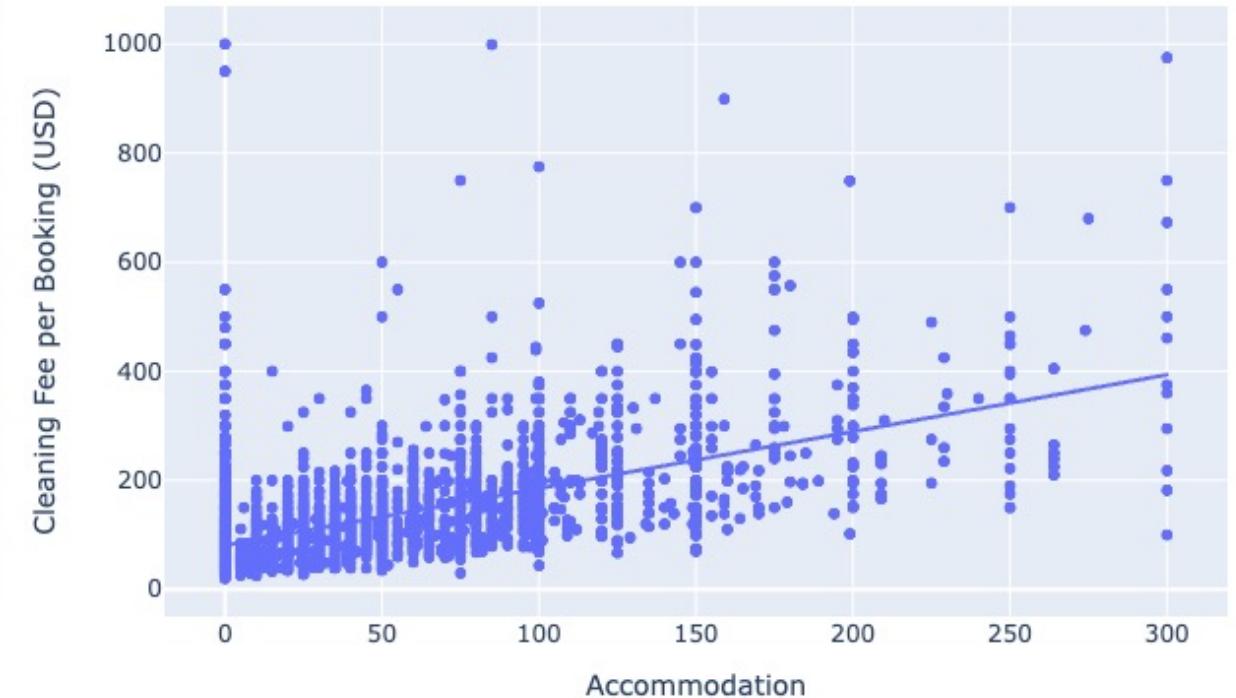


Relationships

Price vs Accommodation



Cleaning fee vs Accommodation



Prediction Model Construction

- Linear Regression
- Random Forest
- Gradient Boost
- Extreme Gradient Boost

```
from sklearn.model_selection import train_test_split

X = house_scaled[['distance', 'accommodates', 'availability_30', 'cleaning_fee', 'bedrooms',
                   'bathrooms', 'property_type_Apartment', 'property_type_House']].values
y = house_scaled['price'].values

X_train, X_test, y_train, y_test = train_test_split(X, y, test_size = 0.2, random_state = 0)
```

Features	Importance
Bedrooms	0.449162
cleaning_fee	0.152625
distance	0.129531
availability_30	0.084503
bathrooms	0.079779
accommodates	0.077149
property_type_House	0.013932
property_type_Apartment	0.013319

Models Result

- The prediction model doesn't perform well with Apartment and House prices.
- The prediction model predicts better the house's prices.

For House and Apartment's prices

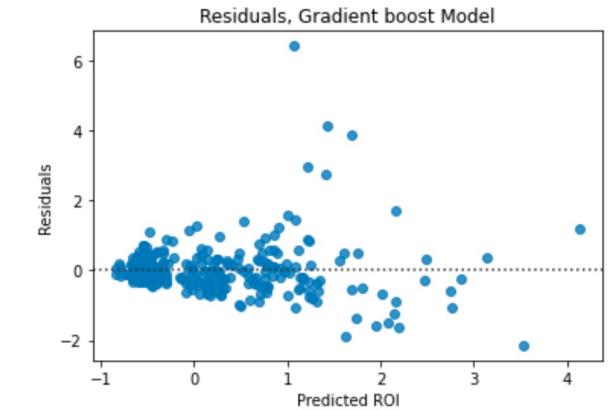
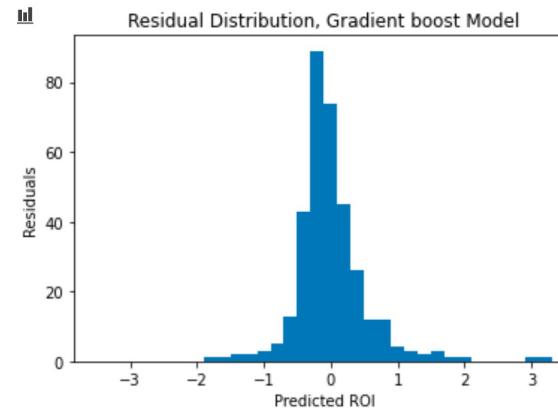
Model	Test Score	MAE	RMSE
Base Linear Regression	0.5228	0.4232	0.6294
Base Random Forest	0.4894	0.4288	0.6511
Base Gradient Boost	0.5194	0.3939	0.6317
Base XGB	0.488	0.4156	0.652

For House's prices

Model	Test Score	MAE	RMSE
Base Linear Regression	0.604	0.4029	0.7332
Base Random Forest	0.5837	0.41	0.7518
Base Gradient Boost	0.5971	0.3998	0.7396
Base XGB	0.5664	0.4363	0.7673

Models tuning

Model	Test Score	MAE	RMSE
Base Linear Regression	0.604	0.4029	0.7332
Base Random Forest	0.5837	0.41	0.7518
Base Gradient Boost	0.5971	0.3998	0.7396
Base XGB	0.5664	0.4363	0.7673
Tune Model	Test Score	MAE	RMSE
Tune Linear Regression	0.604	0.4029	0.7332
Tune Random Forest	0.6141	0.4083	0.701
Tune Gradient Boost	0.6322	0.3861	0.7067
Tune XGB	0.6286	0.3949	0.6876





Q & A