

Motivation



- Help the new host to list his property at FAIR price
- Help new hosts to improve occupancy rate by FAIR pricing
- Need a model to predict the price of the UNIQUE listing
 - Features
 - Specalities
 - Surge
 - Seasonality

Dataset

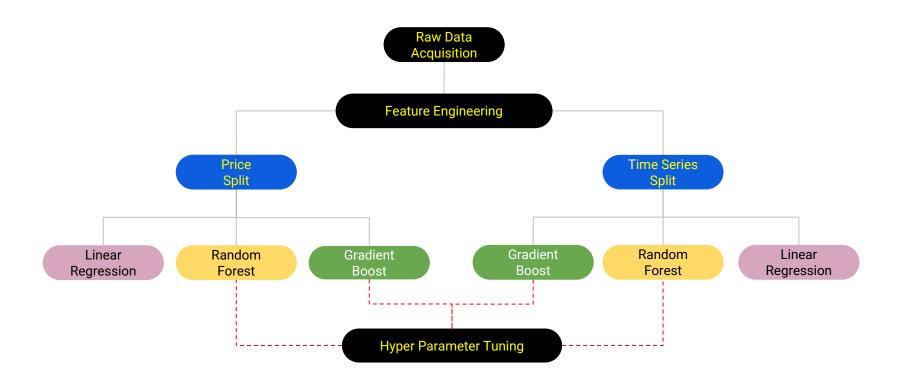
S.No.	Data	Size	Age	Data Size	
1	Calendar	1.8mil x 4		8 million	
2	Listings	5000 x 96		~ 500K	
3	Neighborhood	37 x 2			
4	Reviews	250K x 6	9 years	1.5 mil	
5	Geolocation			SF, Bayarea, LA County	

Rich Data with 96 features 5k listings = 500K data points/county



Flow





Feature Analysis

Data Cleanup

- Transform data
- Missing values
- Categorical

Relational Features

Category Condensation

Imputed sparse data sets

- Luxury
- Special Types



Feature Analysis

Data Cleanup

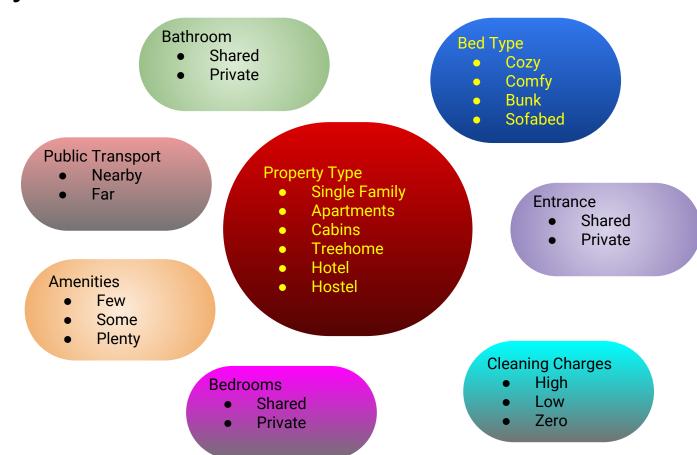
- Transform data
- Missing values
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Relational Features

Category Condensation

Imputed sparse data sets

- Luxury
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Model Ensemble



Models:

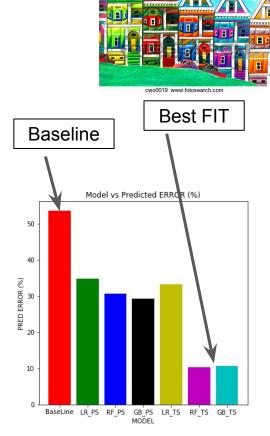
- Linear Regression
- Random Forest
- Gradient Boost

Data:

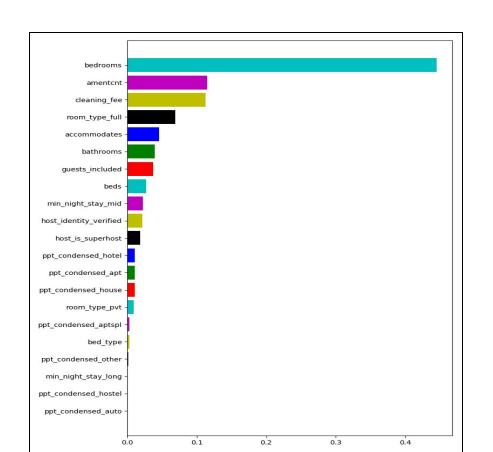
- Common Listings
- Time Series Split

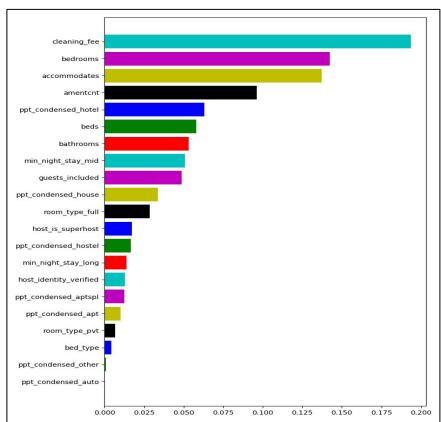
Result Analysis

Metric		Linear Regression	Random Forest	Gradient Boost	Linear Regression	Random Forest	Gradient Boost	
		Random Split			Time Series Split			
R ²		0.56	0.64	0.62	0.54	0.88	0.88	
MSE		4155	3366	3578	4618	1187	1190	
RMSE		64.46	58.02	59.82	67.96	34.46	34.50	
RMSLE		0.38	0.33	0.33	0.37	0.19	0.20	
PCT Error (MAPE)		34.86%	30.62%	29.20%	33.27%	10.37%	10.71%	
Baseline Error		\$77.42	\$77.42	\$77.42	\$79.72	\$79.72	\$79.72	
Predicted Error		\$49.56	\$43.98	\$43.46	\$50.37	\$15.44	\$15.90	



Feature Importance (RF vs GB)





Result Analysis

Prediction Metrics

• R^2 : 0.8817

• PCT ERROR : 10%

Baseline ERROR : \$ 79.00

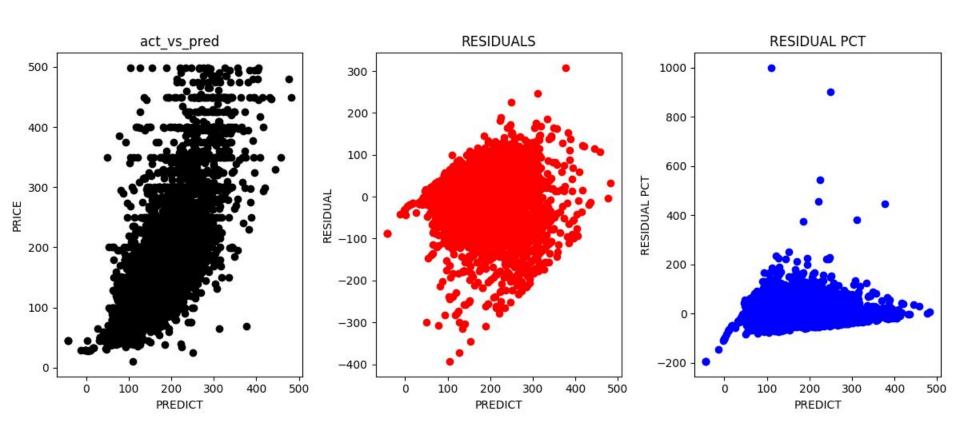
Predicted ERROR: \$ 15.90

Ensemble:

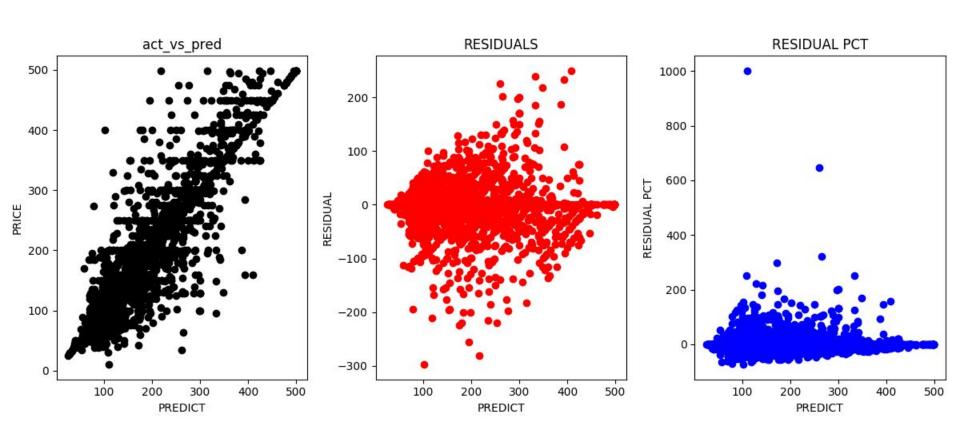
- Random Forest and Gradient Boost
- Time Series



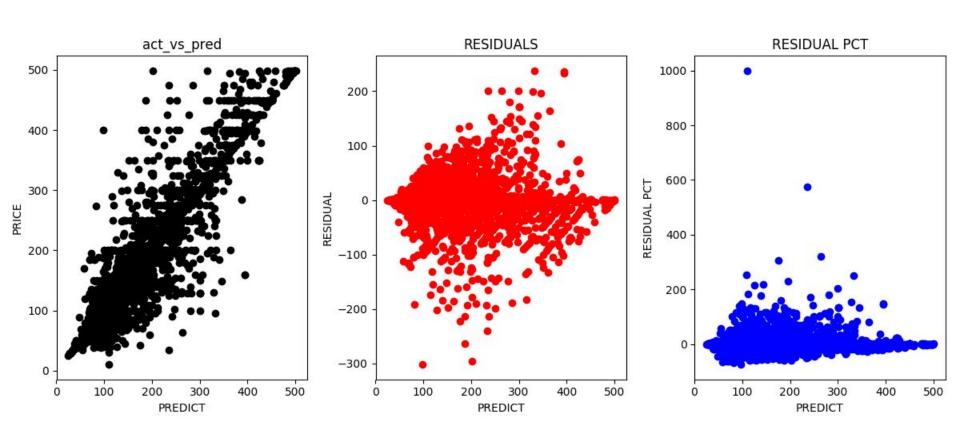
Model Metrics: Linear Regression



Model Metrics: Random Forest

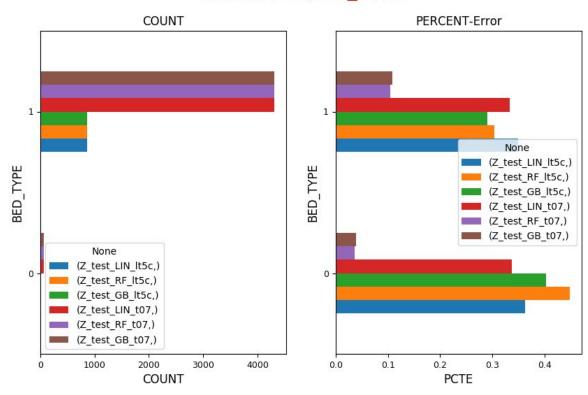


Model Metrics: Gradient Boost



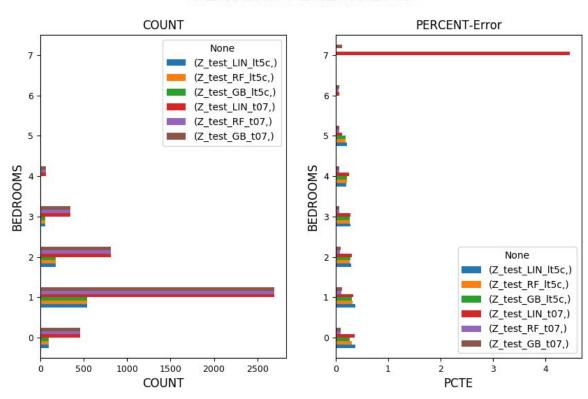
Model Performance Vs Bedtype





Model Performance Vs Bedrooms

FEATURE: BEDROOMS



Model Performance Vs Location



Future Improvements

- Add complete time series to include seasonal signal
- Analyze real time demand predict a surge for Host
- Analyze the reviews to provide a feature sentiment analysis to Host
- Update the model to support special cases

Thank you

Questions?

Github: https://github.com/sgjcyp/capstone_price_is_right

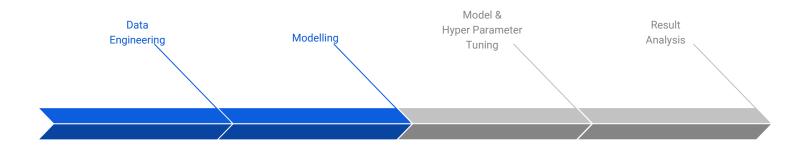
Motivation



- Airbnb is a portal for homeowners to host the Property
- Typically Hosts review the current listings to decide the price
- Need a model to predict the price of the listing
 - Features
 - Uniqueness
 - Surge
 - Seasonality

Process





Feature Analysis

Identify Prime Features

Create Relational Features

Regression Models

Linear Regression

Random Forest

Gradient Boost

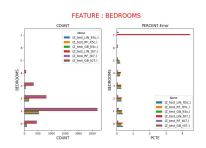
Measure Performance

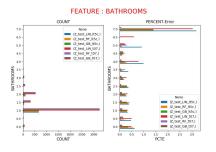
Same Metrics to measure multiple models and datasets

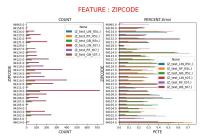
Results vs Model

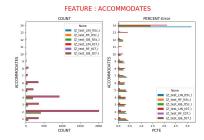
Model Performance in Various Categories

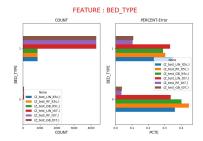
Model Performance Vs Features

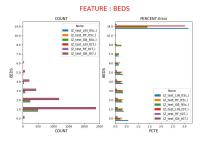






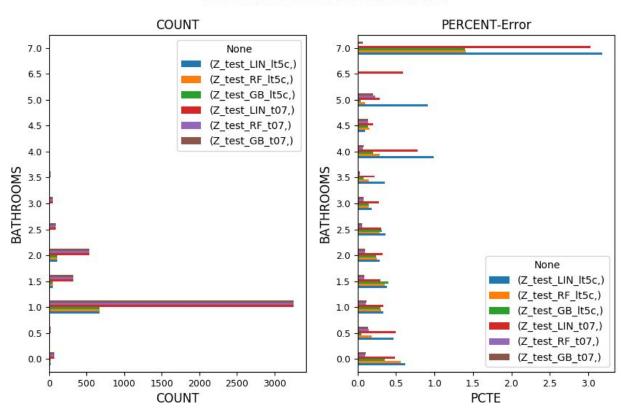






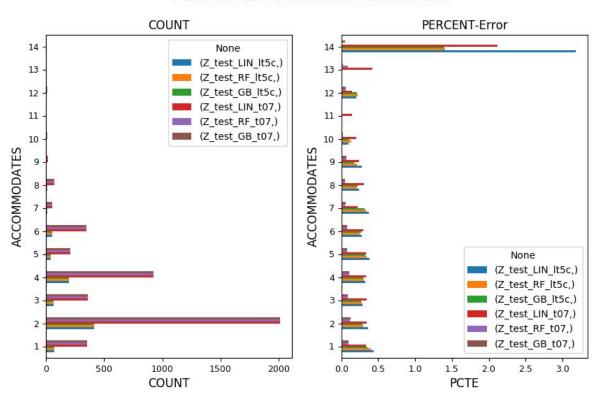
Model Performance Vs Bathrooms

FEATURE: BATHROOMS



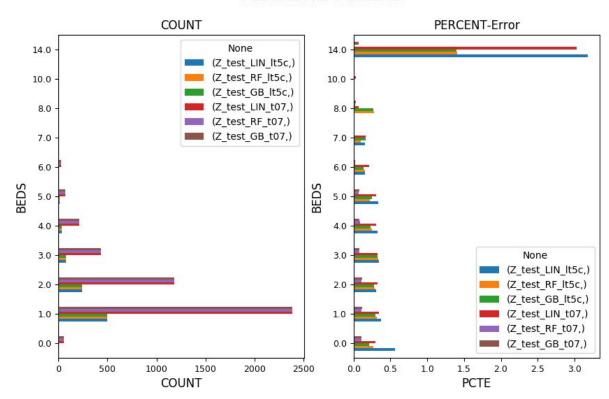
Model Performance Vs Accommodates

FEATURE: ACCOMMODATES



Model Performance Vs Beds

FEATURE: BEDS



Model Performance Vs Zipcode

FEATURE : ZIPCODE

