

AirBnb Ratings Classification

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Overview

AirBnb success is often determined by customer experience. I hope to gain valuable insights to form strategy and demonstrate success.

In this case, I will be a company analyst looking to help AirBnb hosts enhance overall customer satisfaction by classifying listings into three categories: Subpar(0), Good(1) and Best(2); further, to ascertain which aspects of properties can be improved.



Understanding the Data

Source:

Inside AirBnb data set from insideairbnb.com

Specifically, the Los Angeles location using the listings..csv.gz file.

Details:

The data set contains around 31,000 rows (properties) and 74 columns.

After cleaning, around 29,000 rows and 36 columns

Target:

3 classes based on rating. These classes were named Subpar(0), Good(1) and Best(2).

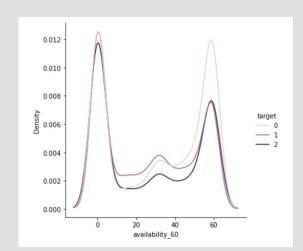
```
Out[130]: 0 0.370931
1 0.334184
2 0.294884
Name: target, dtype: float64
```

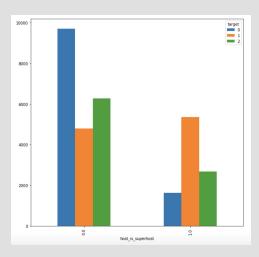
```
Out[55]: 0 11662
1 10479
2 9324
```

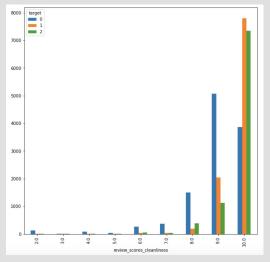
Name: target, dtype: int64

EDA

- Visualized and explored descriptive statistics
- Examined spreads of classes across each predictor
- Researched features, and processed accordingly
- Created dummy variables for relevant categoricals







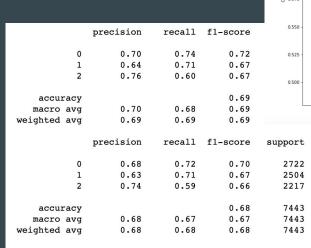
Intermediate Model: Decision Tree Classifier

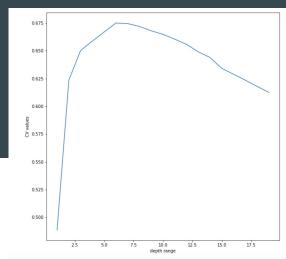
Vanilla Model: Perfect training metrics but accuracy of .60 for testing.

Max_depth and GridSearch CV optimal parameters:

Max_depth = 6, criterion = 'gini', min_samples_split = 10

Improved model: .68 accuracy





Baseline Model: Random Forest Classifier

Without any tuning and running a vanilla model

Again, perfect training set metrics with an accuracy of 1.00

However, test set has an accuracy of .68

What I tuned: n_estimators, max_depth, min_samples_leaf, min_samples_split, criterion and max_features

	precision	recall	f1-score	support
0 1 2	1.00 1.00 1.00	1.00 1.00 1.00	1.00 1.00 1.00	8321 7445 6562
accuracy macro avg weighted avg	1.00	1.00	1.00 1.00 1.00	22328 22328 22328
	precision	recall	f1-score	support
0 1 2	0.70 0.68 0.66	0.71 0.67 0.65	0.70 0.67 0.66	2722 2504 2217
accuracy macro avg weighted avg	0.68 0.68	0.68 0.68	0.68 0.68 0.68	7443 7443 7443

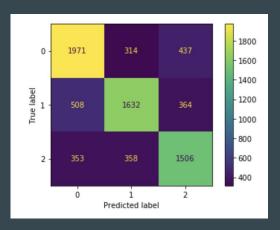
Final Model: Random Forest Classifier

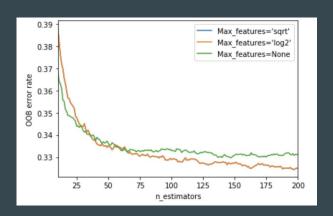
After GridSearchCV and Out-Of-Bag comparison, final model had an accuracy and averages of .70.

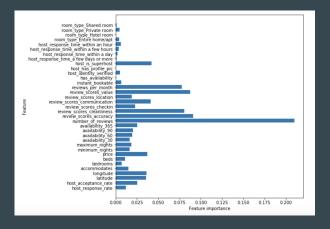
Confusion Matrix: Majority predicted correctly and less than half predicted incorrectly.

Illustration of features importances

Similar to decision trees, but the levels of importance changed as a result of ensemble of multiple decision trees considering different features







Conclusions



Effectively classify AirBnb listings according to each class Subpar(0), Good(1), Best(2)

Especially useful for predicting listings with not a lot of reviews

Recommendations:

- Properties should be cleaned thoroughly
- Consider being an Airbnb superhost
- Improve communication

Future Work

 Explore different types of property and how they affects a customer's visit

 Analysis of different cities provided by the Inside AirBnb website.



Thank You

GitHub Repository https://github.com/mzcode98/airbnb-rating-classification mattzhang989@gmail.com Email LinkedIn https://www.linkedin.com/in/matthew-zhang-167859183/ Acknowledgements Flatiron School, Instructor Yish Lim, Classmates