



# Predicting Housing Category

Jhih-Rou Huang

# Goal

Build a classification model to predict the category of the houses

- Predicting the category of the house in the real estate company is basic or luxury
- Customers can find their ideal houses easily and efficiency based on their needs
- The sales of houses can increase

# Tools



Pandas

Numpy

Matplotlib

Seaborn

Sklearn

xgboost

# Data

ParisHousingClass from Kaggle  
10,000 houses with 18 features

- squareMeters
- numberOfRooms
- floors - number of floors
- cityPartRange- the higher the range, the more exclusive the neighborhood is
- numPrevOwners- number of previous owners
- Made- year
- hasStormProtector
- Basement- basement square meters
- Attic- attic square meters
- Garage- garage size
- hasStorageRoom
- hasGuestRoom- number of guest rooms
- Price- the price of a house

Size

Age

Price

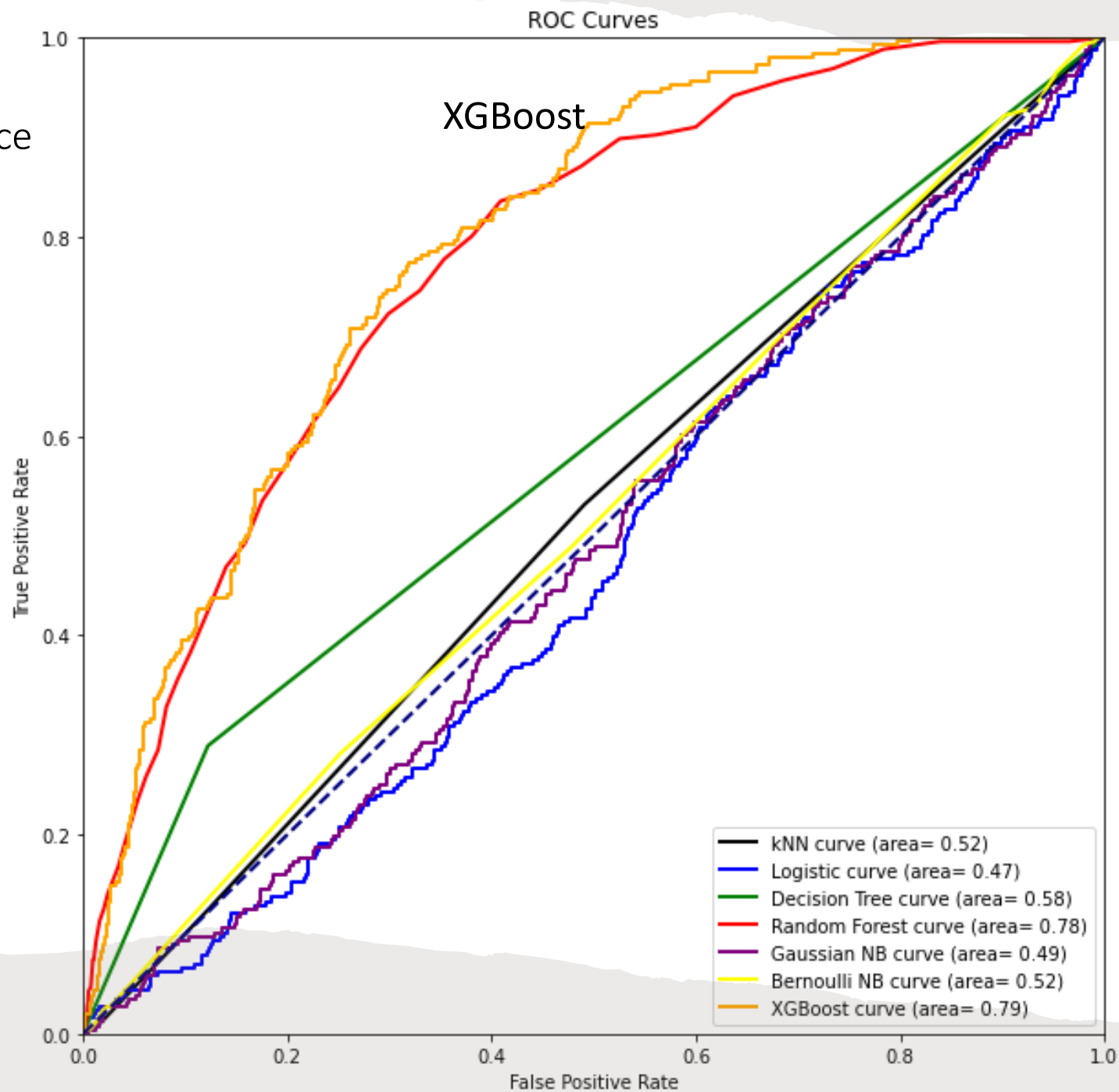
Target

Category: Luxury or Basic

Price per square meters=price/ square meters

Age =2021-made

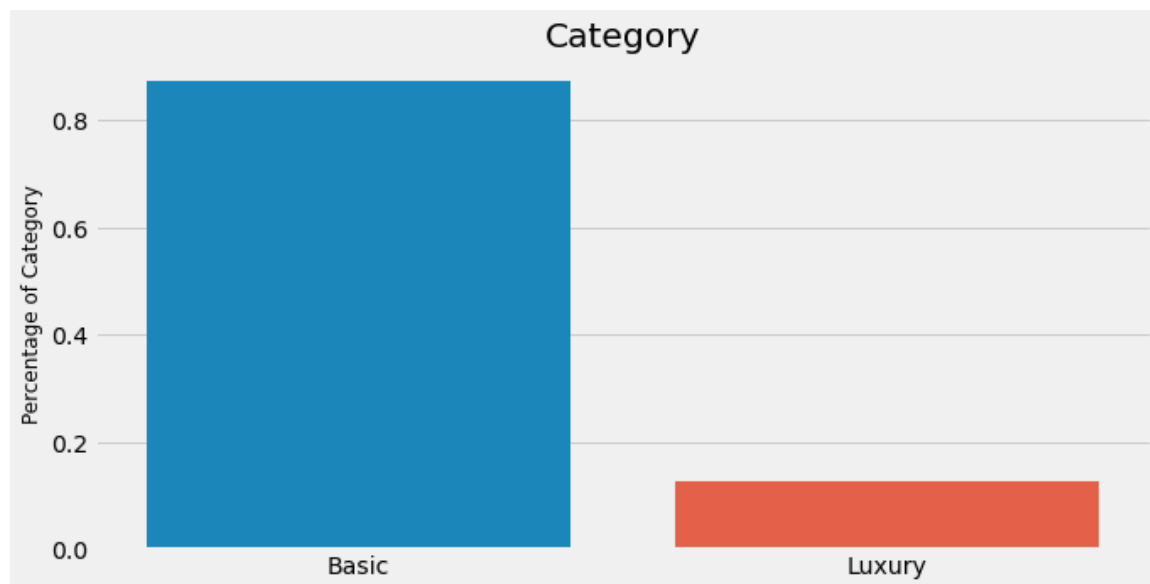
## Comparison of Model Performance



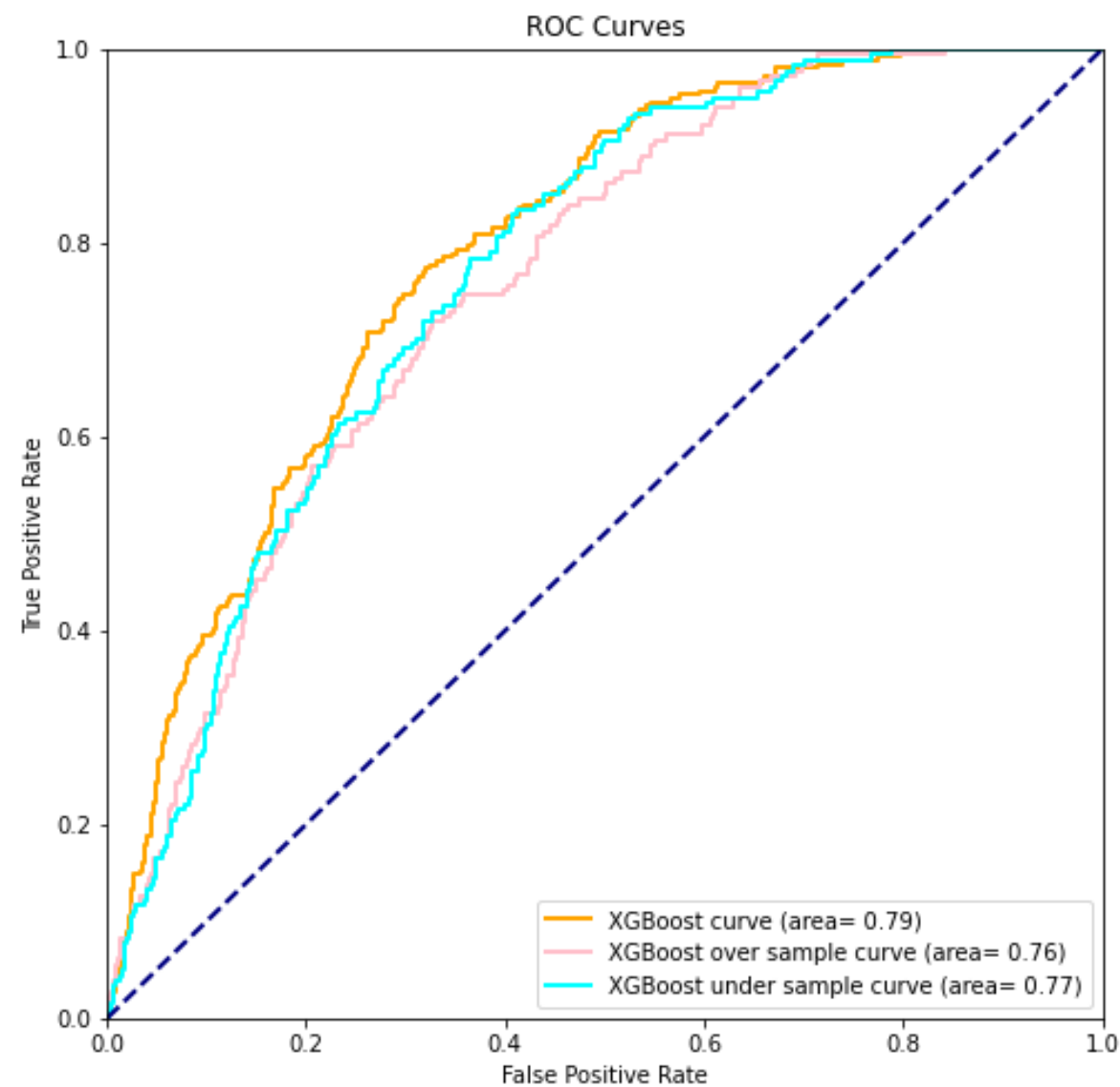
actual	basic	luxury
	1188	556
luxury	59	197
prediction		

Color scale: 0 to 1000 (dark blue to light blue)

# Resample



Model	accuracy	precision	recall	f1	AUC
KNN	0.858	0.088	0.012	0.021	0.518
Logistic	0.872	0.000	0.000	0.000	0.473
Decision Tree	0.803	0.258	0.289	0.273	0.583
Random Forest	0.872	0.500	0.004	0.008	0.778
Gaussian	0.872	0.000	0.000	0.000	0.487
Bernoulli	0.872	0.000	0.000	0.000	0.516
<b>XGBoost</b>	<b>0.863</b>	<b>0.402</b>	<b>0.152</b>	<b>0.221</b>	<b>0.792</b>
<b>XGBoost over sample</b>	<b>0.839</b>	<b>0.289</b>	<b>0.287</b>	<b>0.288</b>	<b>0.758</b>
<b>XGBoost under sample</b>	<b>0.684</b>	<b>0.223</b>	<b>0.718</b>	<b>0.340</b>	<b>0.771</b>

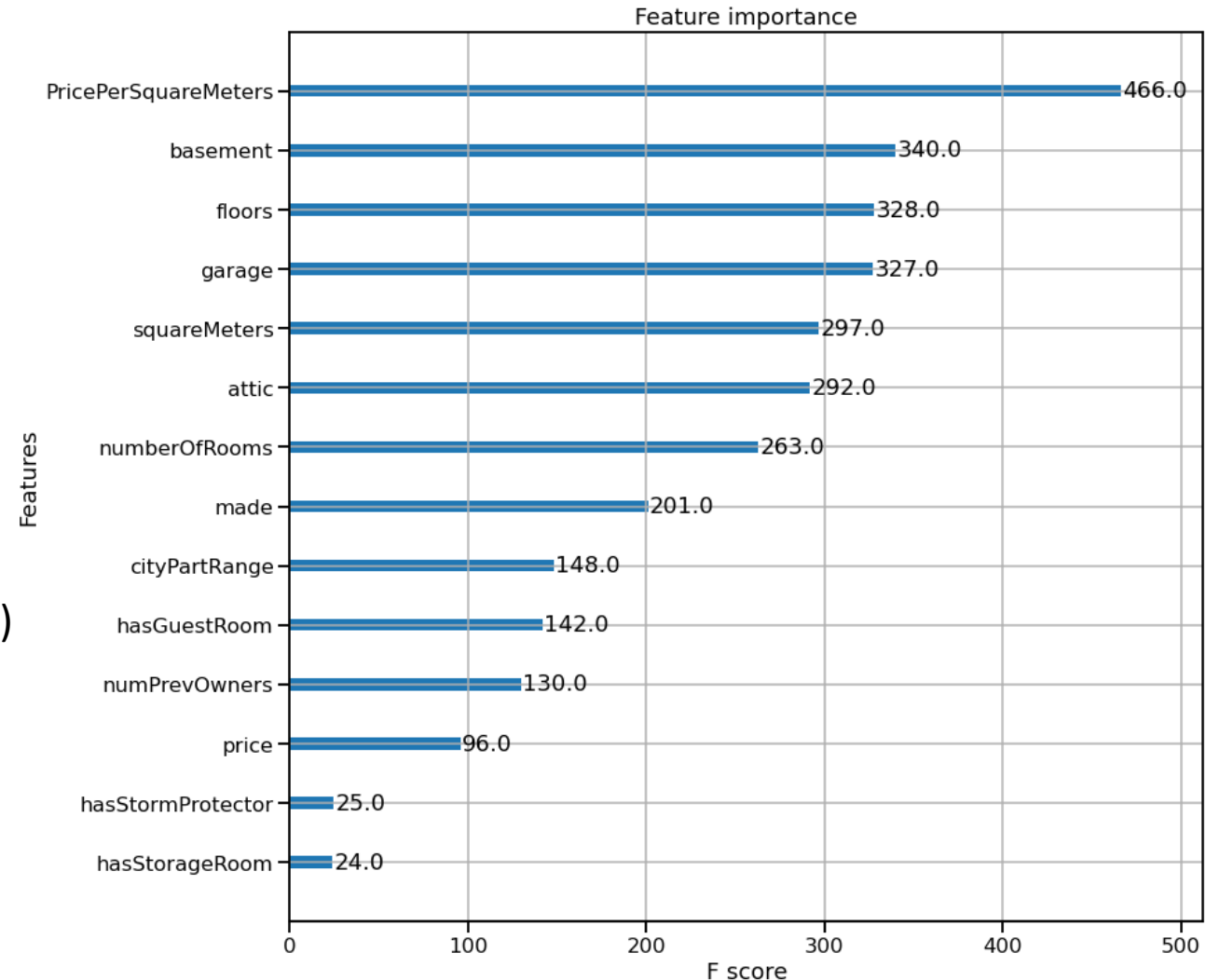


# Insights and Recommendations

- Price per square meters is the most important feature
- The size of the house also plays an important role

## Future Improvements

Build more robust model (more data and features)



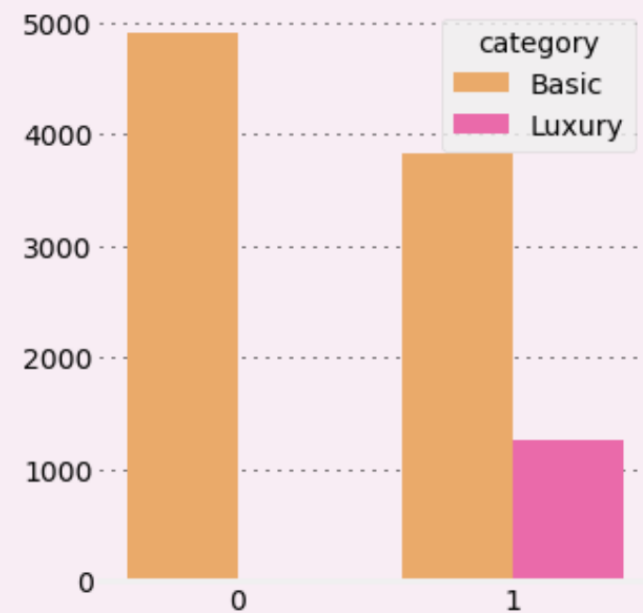
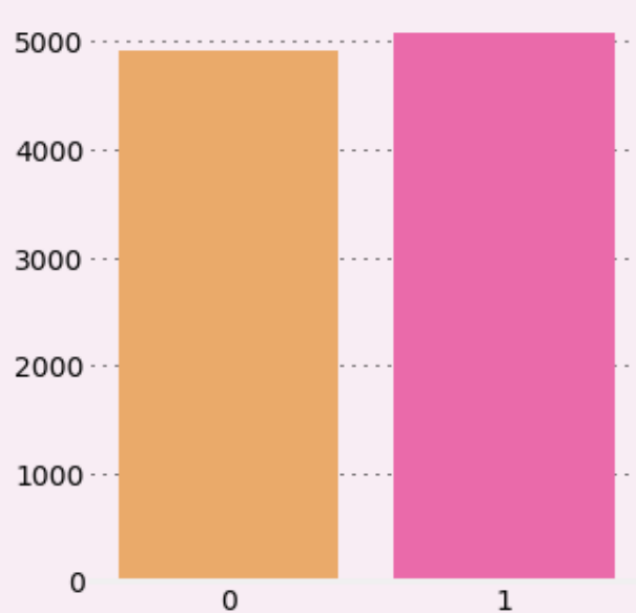
# Appendix



# 'has Yard'

hasYard	0	1	All
category			
Basic	4913	3822	8735
Luxury	0	1265	1265
All	4913	5087	10000

Countplot of 'hasYard'



# 'has Pool'

hasPool	0	1	All
category			
Basic	5032	3703	8735
Luxury	0	1265	1265
All	5032	4968	10000

Countplot of 'hasPool'

