

# HOUSE PRICING

SELECTED TOPICS IN COM SCI II

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# MEET THE TEAM

## FIN NAK BID



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+ Create

Home

Competitions

Datasets

Models

Code

Discussions

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More

Your Work

VIEWED

House Prices - Adva...

Comprehensive dat...

View Active Events



KAGGLE · GETTING STARTED PREDICTION COMPETITION · ONGOING

Submit Prediction

...

# House Prices - Advanced Regression Techniques

Predict sales prices and practice feature engineering, RFs, and gradient boosting



Overview Data Code Models Discussion Leaderboard Rules Team Submissions

## Overview

∞ This competition runs indefinitely with a rolling leaderboard. [Learn more.](#)

Competition Host

Kaggle



Prizes & Awards

Knowledge

Does not award Points or Medals

Participation

4,145 Competitors

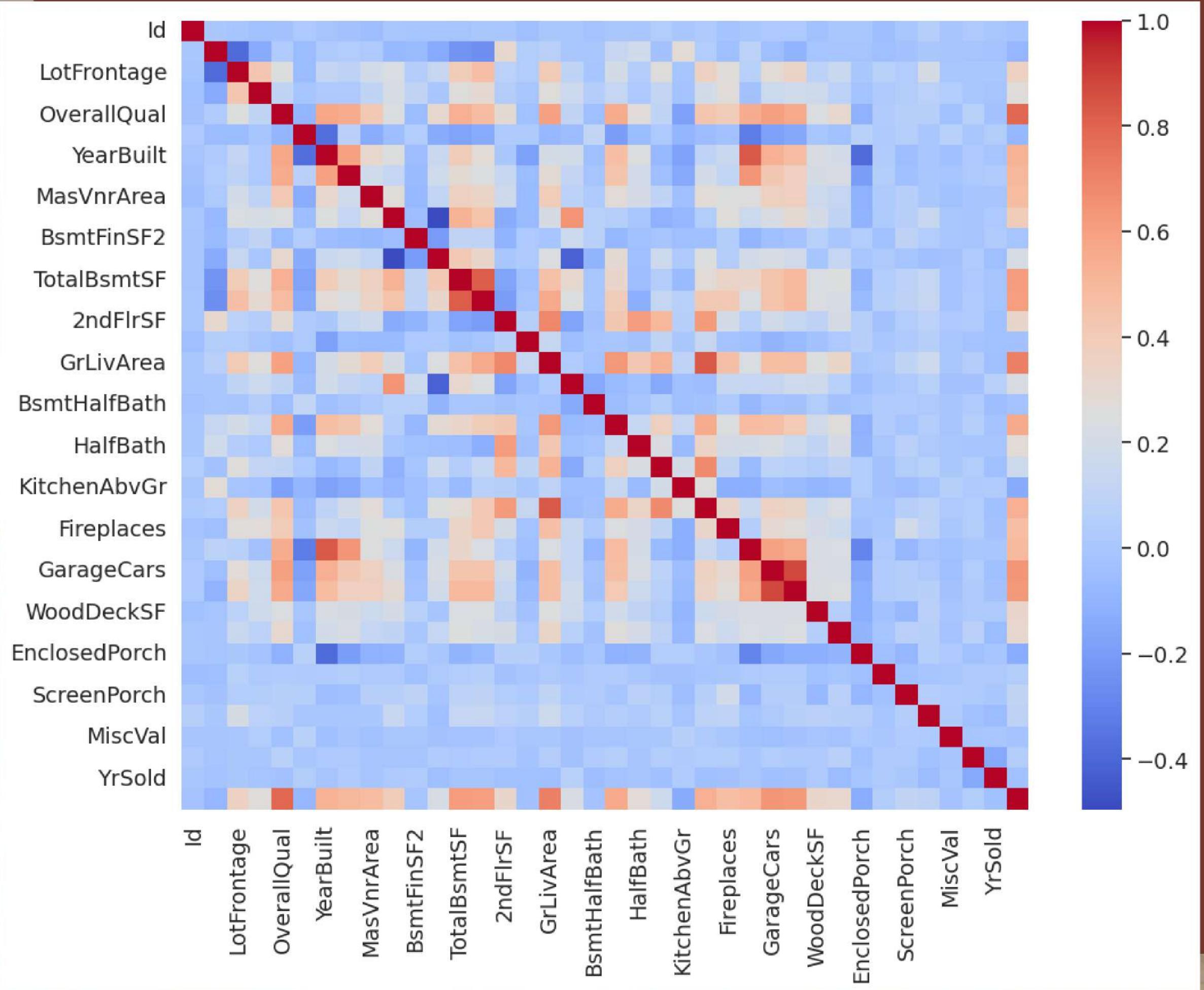
4,024 Teams

20,027 Entries

## Description

Start here if...

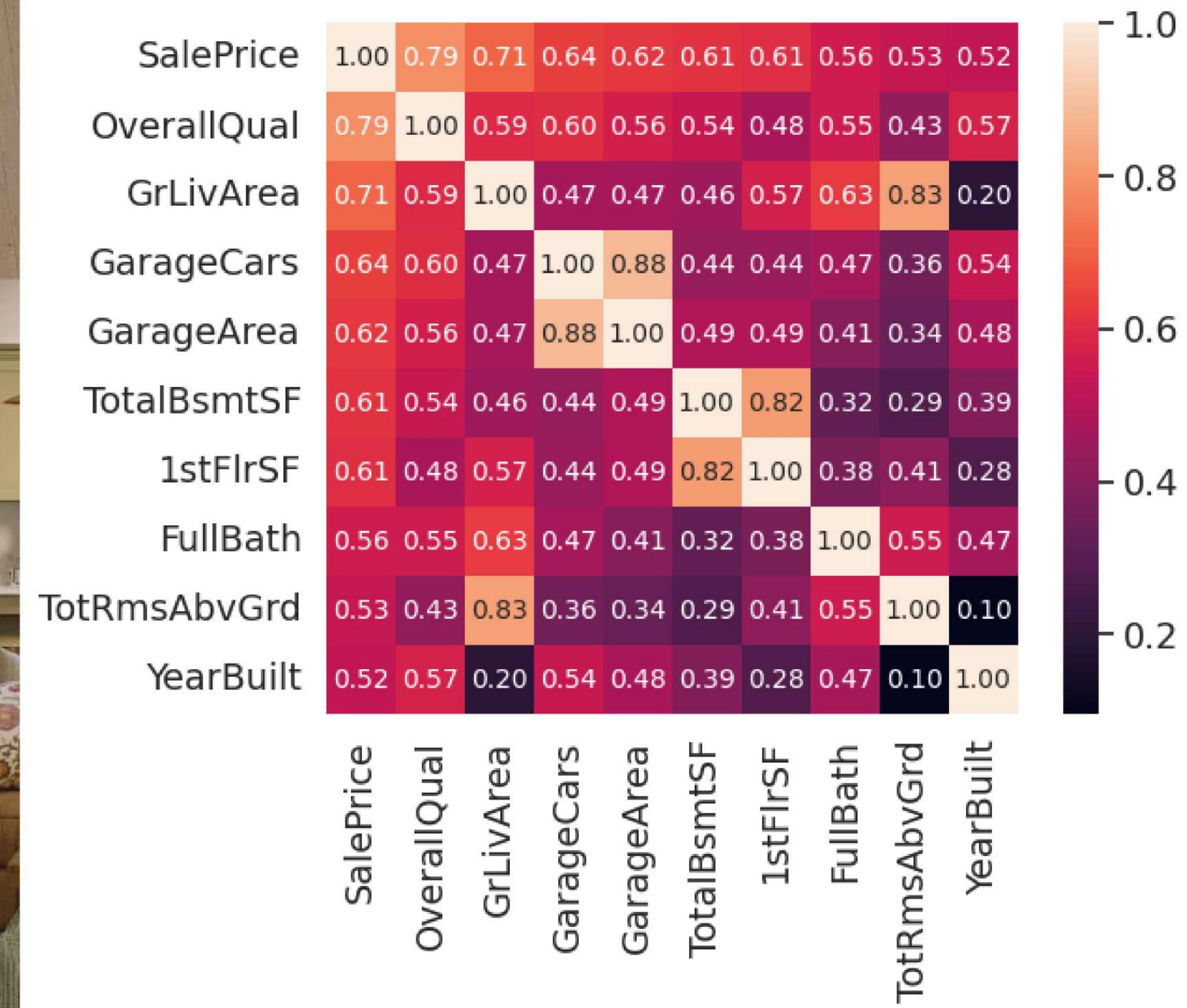
# FEATURE ENGINEERING



# FEATURE ENGINEERING

Top4

- OverallQual
- YearBuilt.
- TotalBsmtSF.
- GrLivArea.



# FEATURE ENGINEERING

Outlier  
StandardScaler

```
outer range (low) of the distribution:  
[[-1.83820775]  
 [-1.83303414]  
 [-1.80044422]  
 [-1.78282123]  
 [-1.77400974]  
 [-1.62295562]  
 [-1.6166617 ]  
 [-1.58519209]  
 [-1.58519209]  
 [-1.57269236]]
```

```
outer range (high) of the distribution:  
[[3.82758058]  
 [4.0395221 ]  
 [4.49473628]  
 [4.70872962]  
 [4.728631 ]  
 [5.06034585]  
 [5.42191907]  
 [5.58987866]  
 [7.10041987]  
 [7.22629831]]
```

# DATA PREPROCESSING



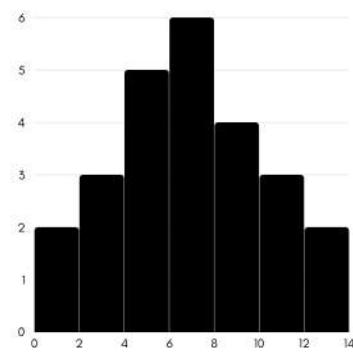
## DATA REDUCTION

Drop columns and a sample



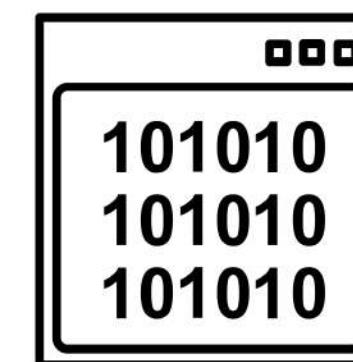
## HANDLING MISSING VALUE

Fill missing value with Mode and Median



## FEATURE SCALING

Reduce skewed curve



## ENCODING FEATURES

Ordinal encoding  
One-hot encoding



	Total	Percent
PoolQC	1453	0.995205
MiscFeature	1406	0.963014
Alley	1369	0.937671
Fence	1179	0.807534
FireplaceQu	690	0.472603
LotFrontage	259	0.177397
GarageYrBlt	81	0.055479
GarageCond	81	0.055479
GarageType	81	0.055479
GarageFinish	81	0.055479
GarageQual	81	0.055479
BsmtFinType2	38	0.026027
BsmtExposure	38	0.026027
BsmtQual	37	0.025342
BsmtCond	37	0.025342
BsmtFinType1	37	0.025342
MasVnrArea	8	0.005479
MasVnrType	8	0.005479
Electrical	1	0.000685
Id	0	0.000000

# DATA REDUCTION

```
all_data = all_data.drop((missing_data[missing_data['Total'] > 1]).index,1)

df_train = df_train.dropna(subset=['Electrical'])
```

# HANDLING MISSING VALUE

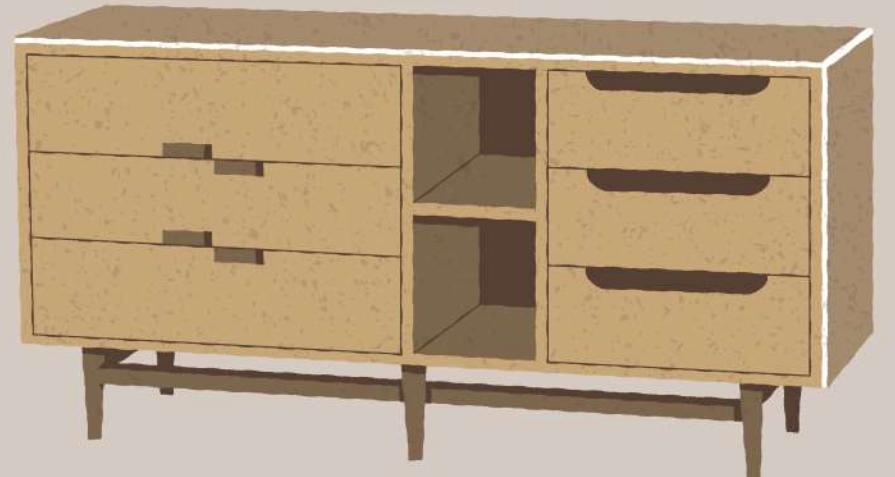
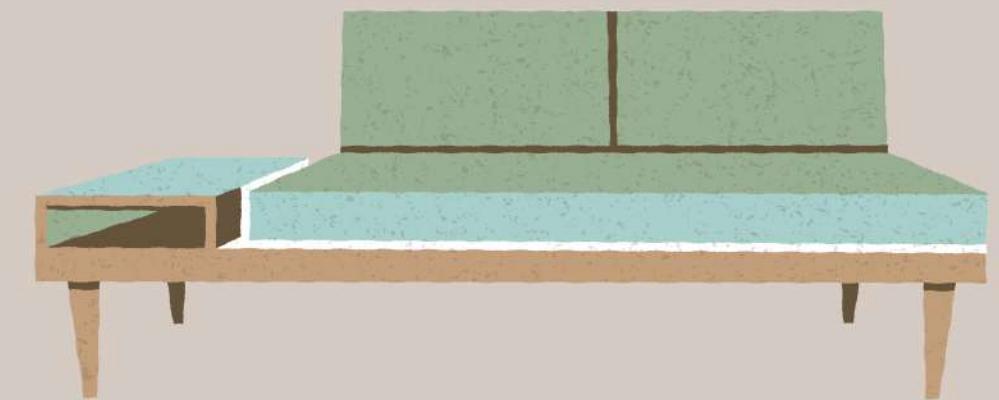
```
from sklearn.impute import SimpleImputer

original_dtypes = df_test.dtypes

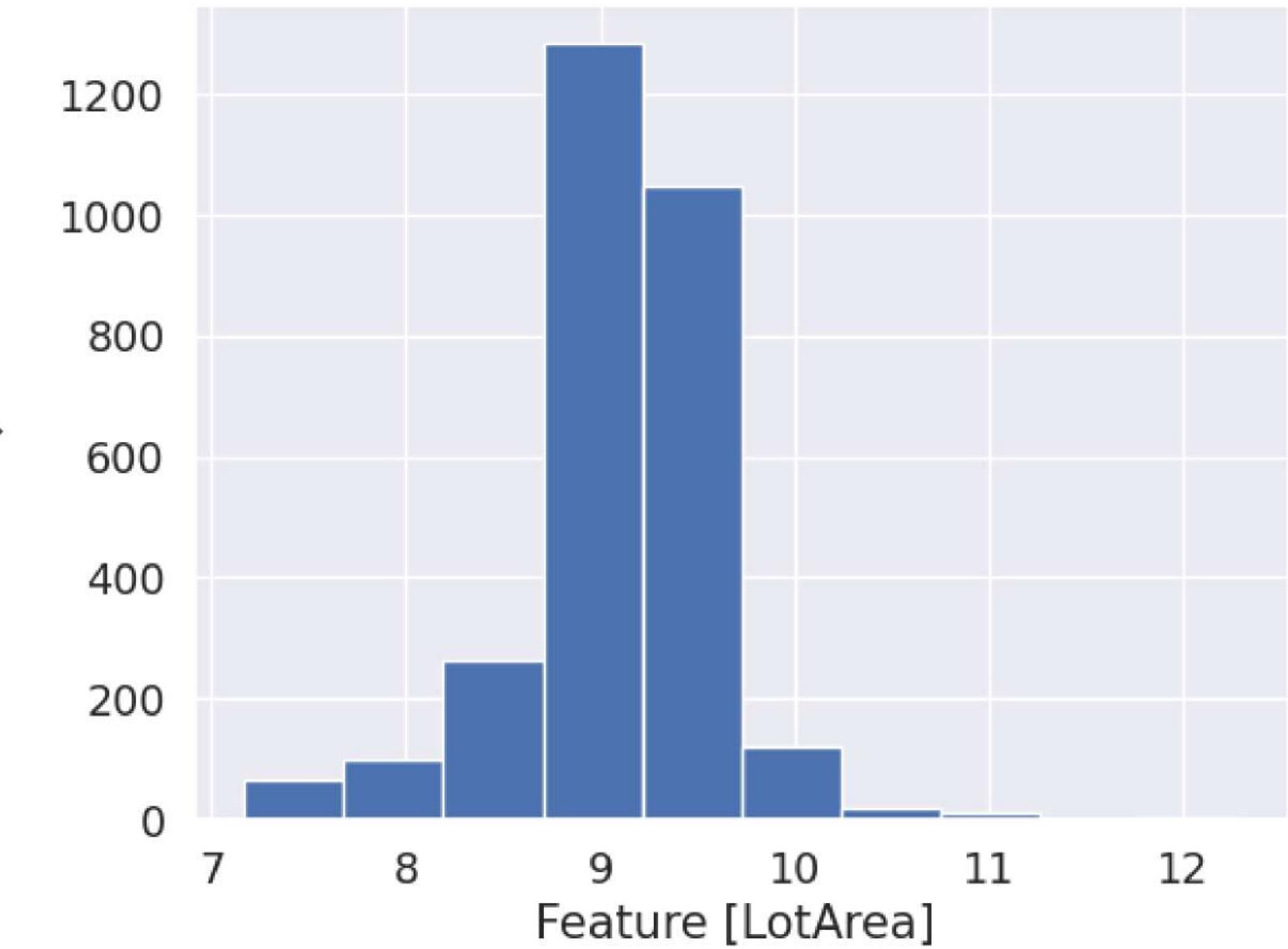
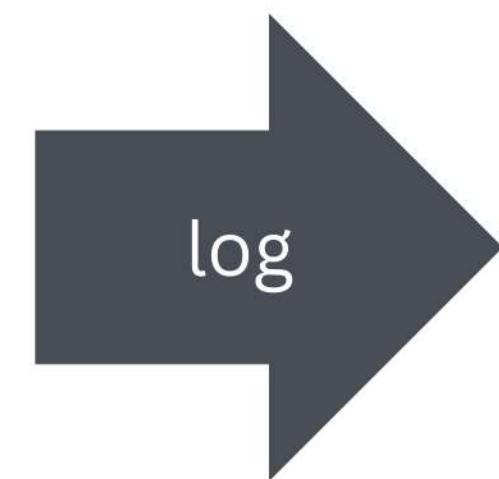
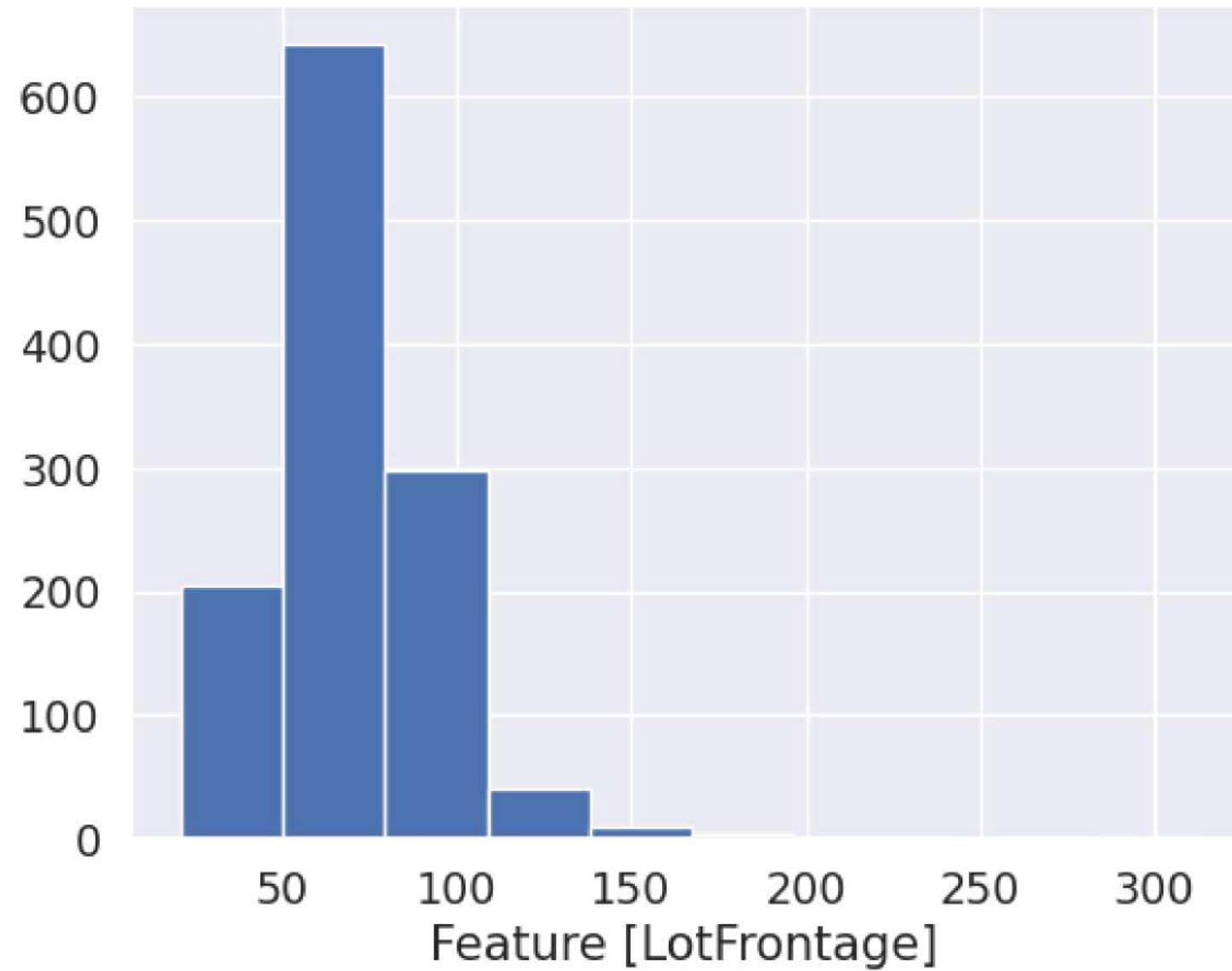
columns_to_impute = [ 'Functional', 'KitchenQual', 'BsmtHalfBath', 'BsmtFullBath', 'GarageCars', 'MSZoning', 'Utilities', 'SaleType', 'Exterior1st', 'Exterior2nd' ]

imputer = SimpleImputer(strategy='most_frequent')
x_train_imputed = imputer.fit_transform(df_train[columns_to_impute])
df_test[columns_to_impute] = imputer.transform(df_test[columns_to_impute])

columns_to_impute = [ 'BsmtFinSF1', 'TotalBsmtSF', 'BsmtUnfSF', 'BsmtFinSF2', 'GarageArea' ]
imputer = SimpleImputer(strategy='mean')
x_train_imputed = imputer.fit_transform(df_train[columns_to_impute])
df_test[columns_to_impute] = imputer.transform(df_test[columns_to_impute])
df_test = df_test.astype(original_dtypes)
```



# FEATURE SCALING



## Scaling Features:

LotArea BsmtFinSF1 BsmtFinSF2 TotalBsmtSF 1stFlrSF 2ndFlrSF LowQualFinSF  
WoodDeckSF OpenPorchSF EnclosedPorch 3SsnPorch PoolArea MiscVal

# ONE-HOT ENCODING



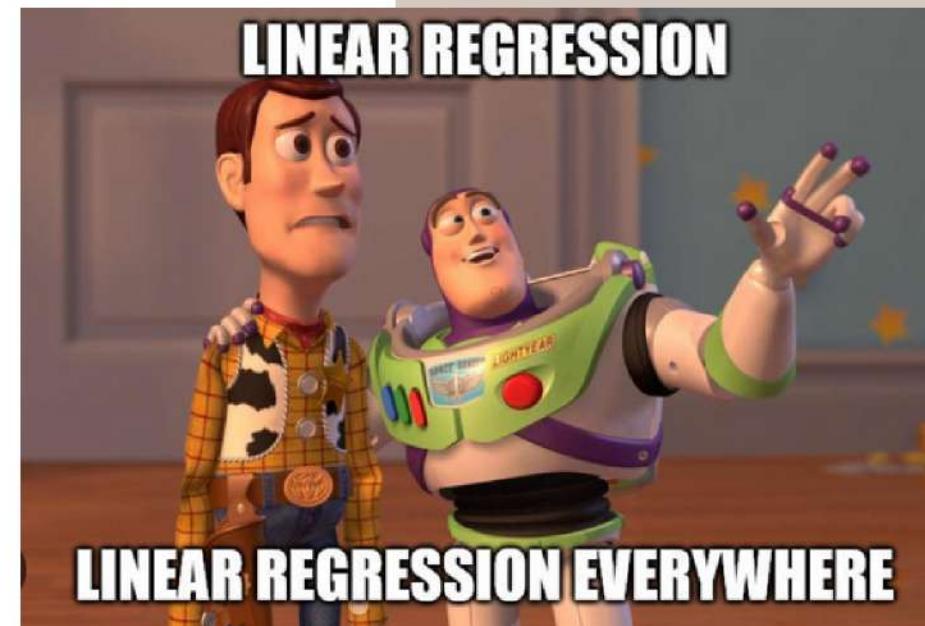
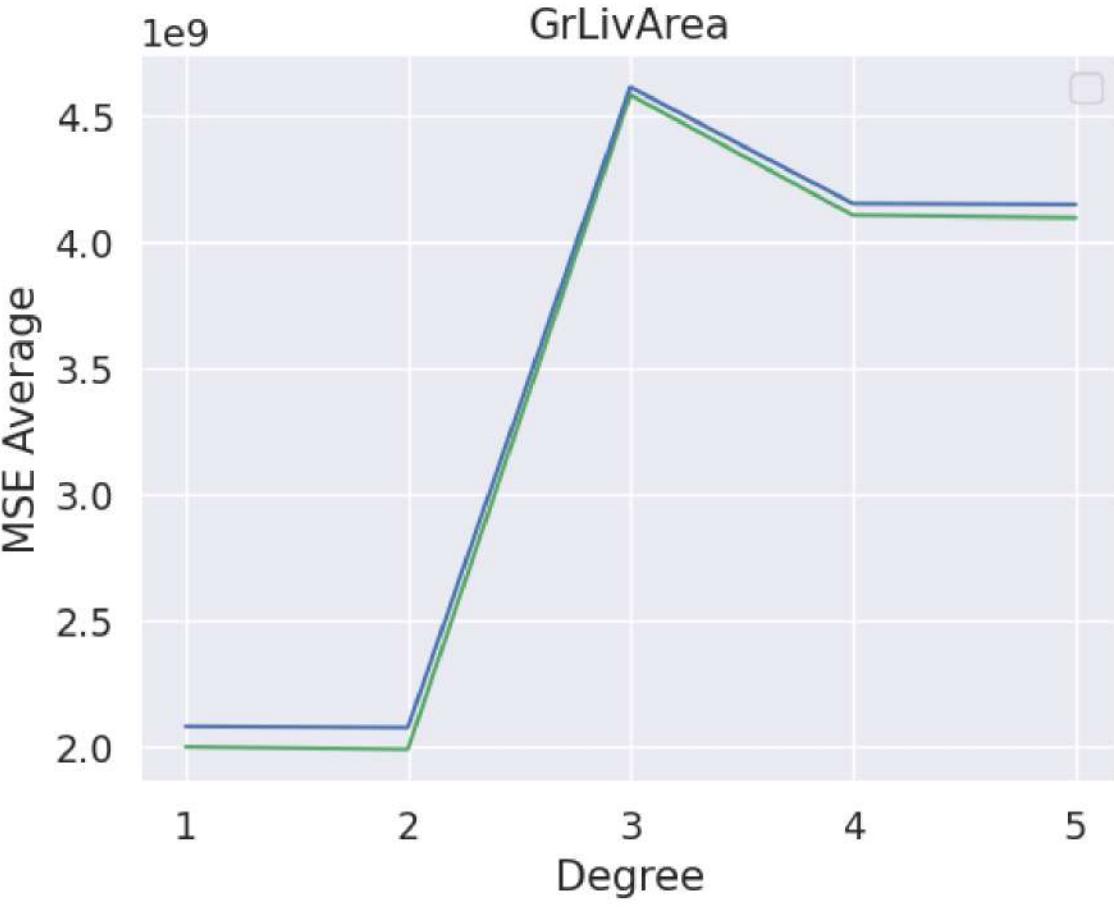
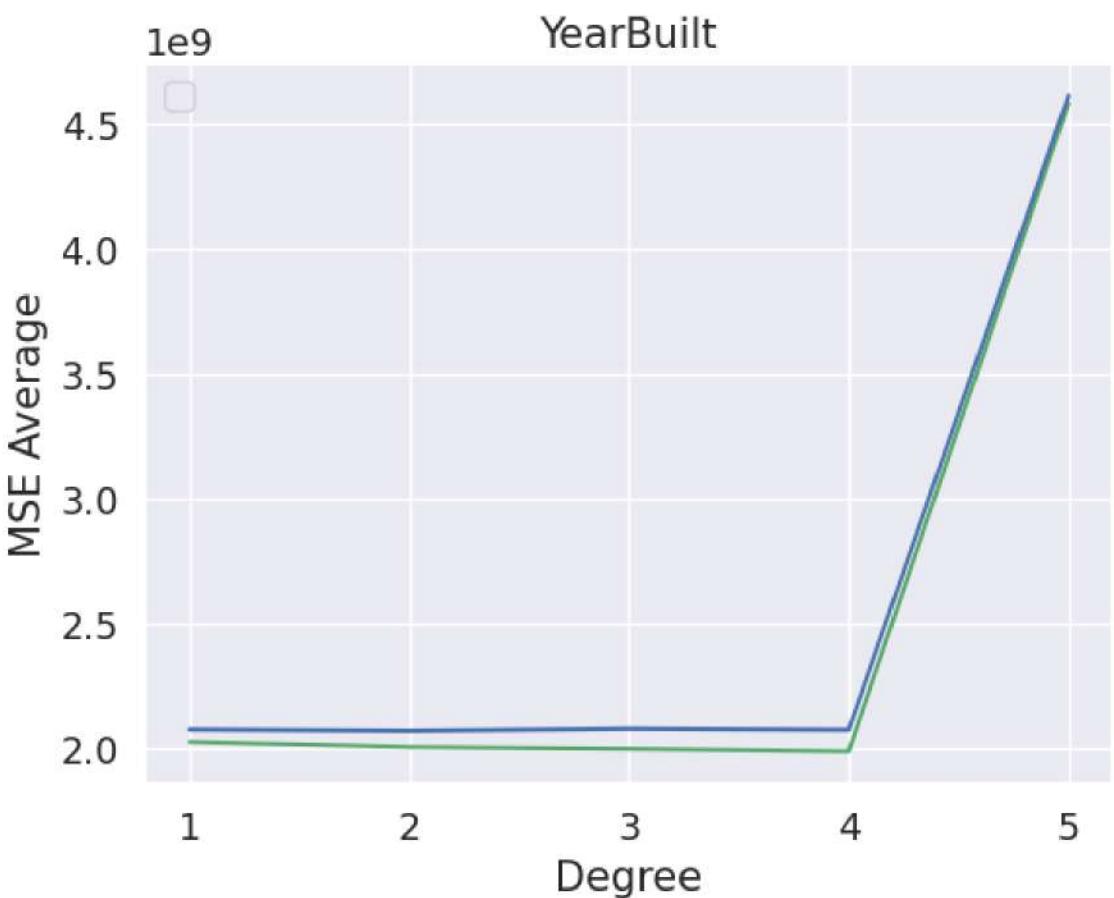
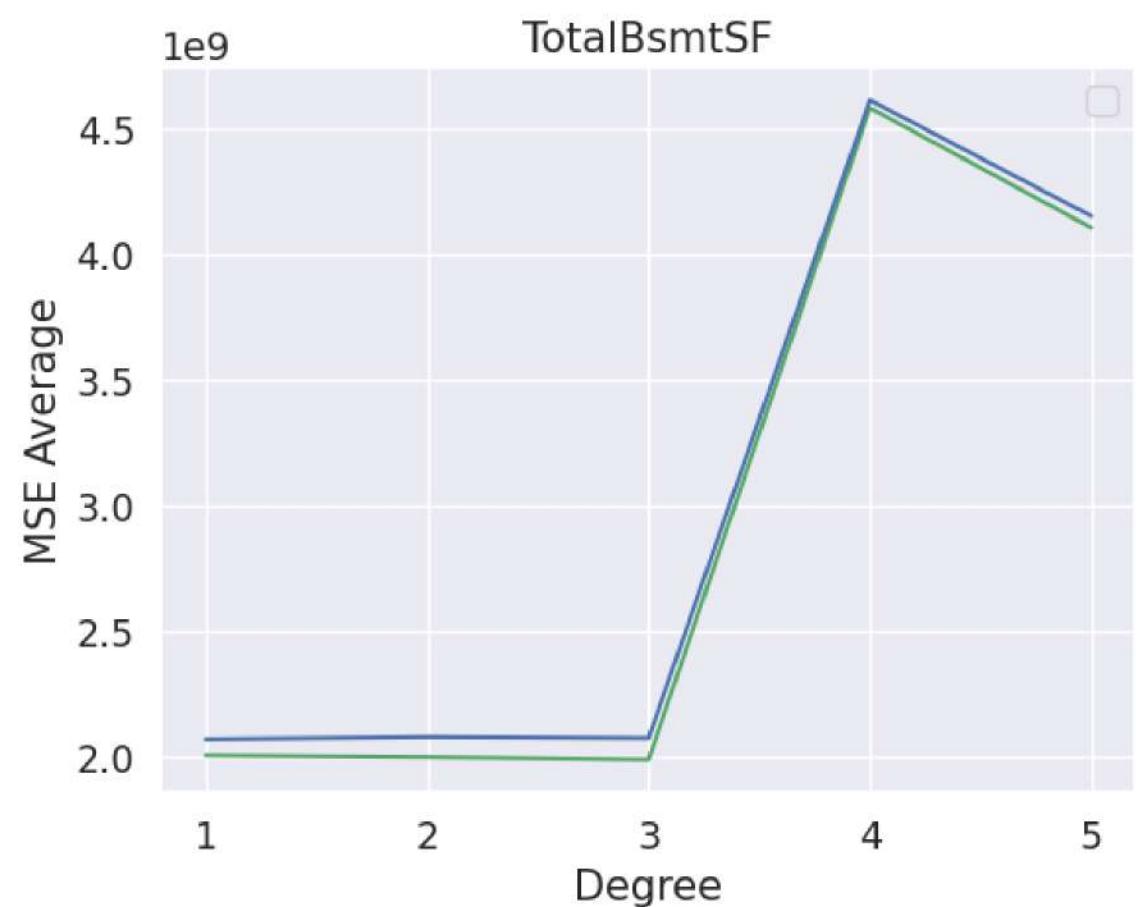
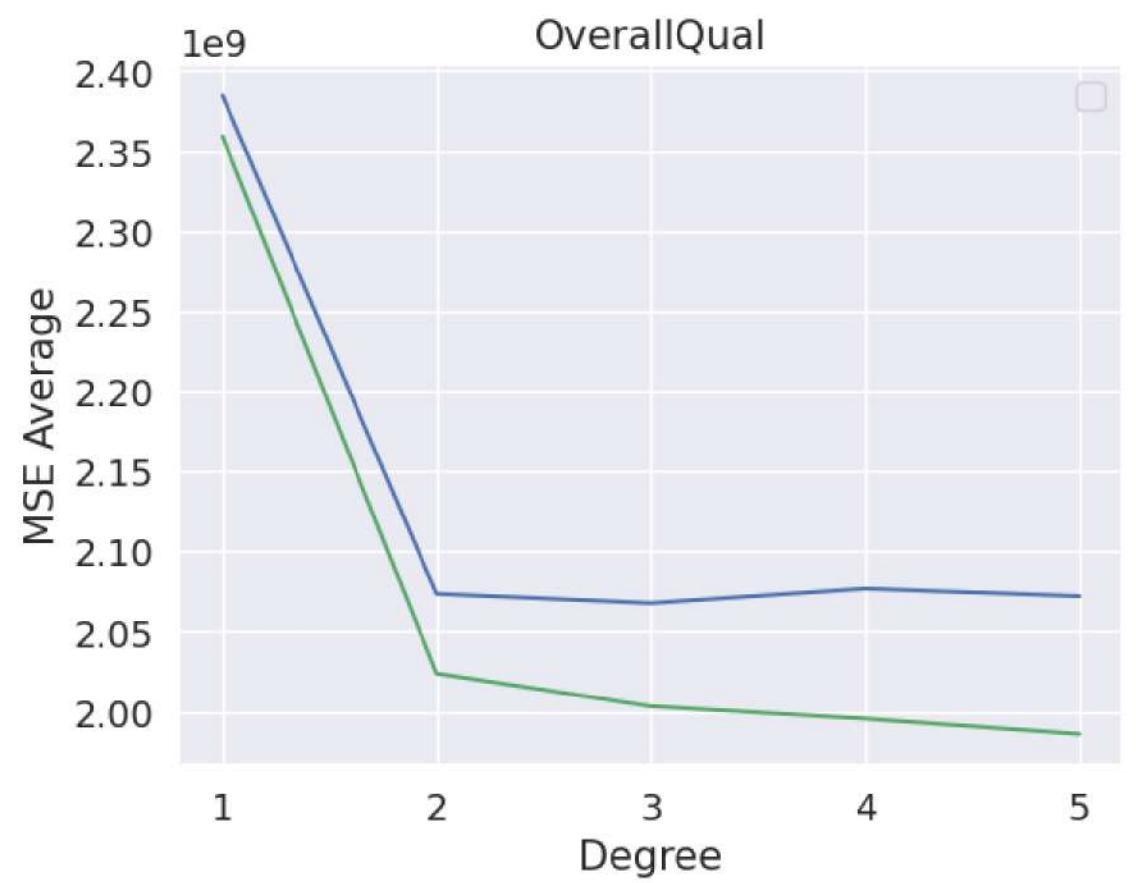
SaleCondition_Abnormal	SaleCondition_AdjLand	SaleCondition_Alloca	SaleCondition_Family	SaleCondition_Normal	SaleCondition_Partial
0	0	0	0	1	0
0	0	0	0	1	0
0	0	0	0	1	0
1	0	0	0	0	0
0	0	0	0	1	0
...	...	...	...	...	...
0	0	0	0	1	0
1	0	0	0	0	0
1	0	0	0	0	0
0	0	0	0	1	0
0	0	0	0	1	0

# ORDINAL ENCODING

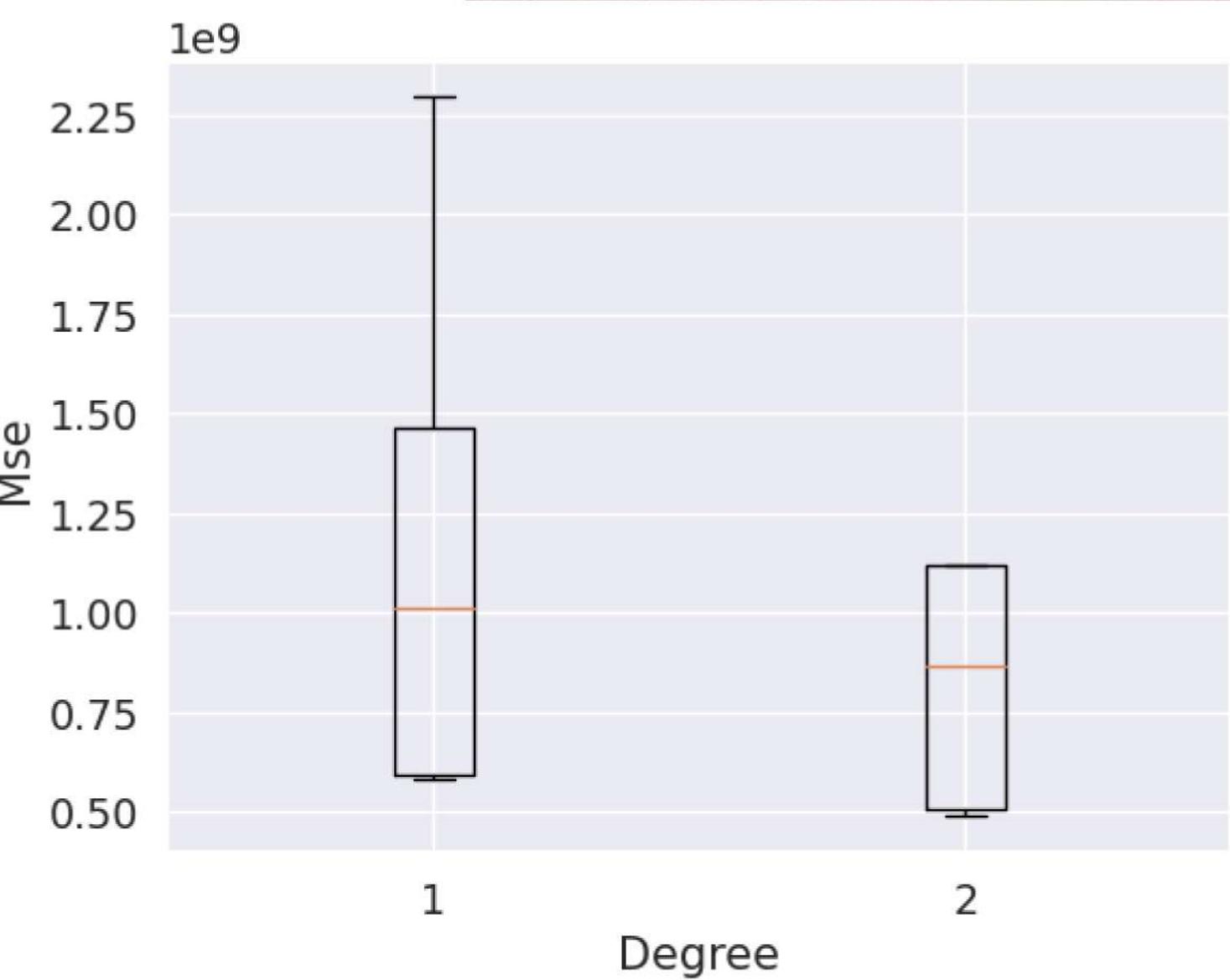
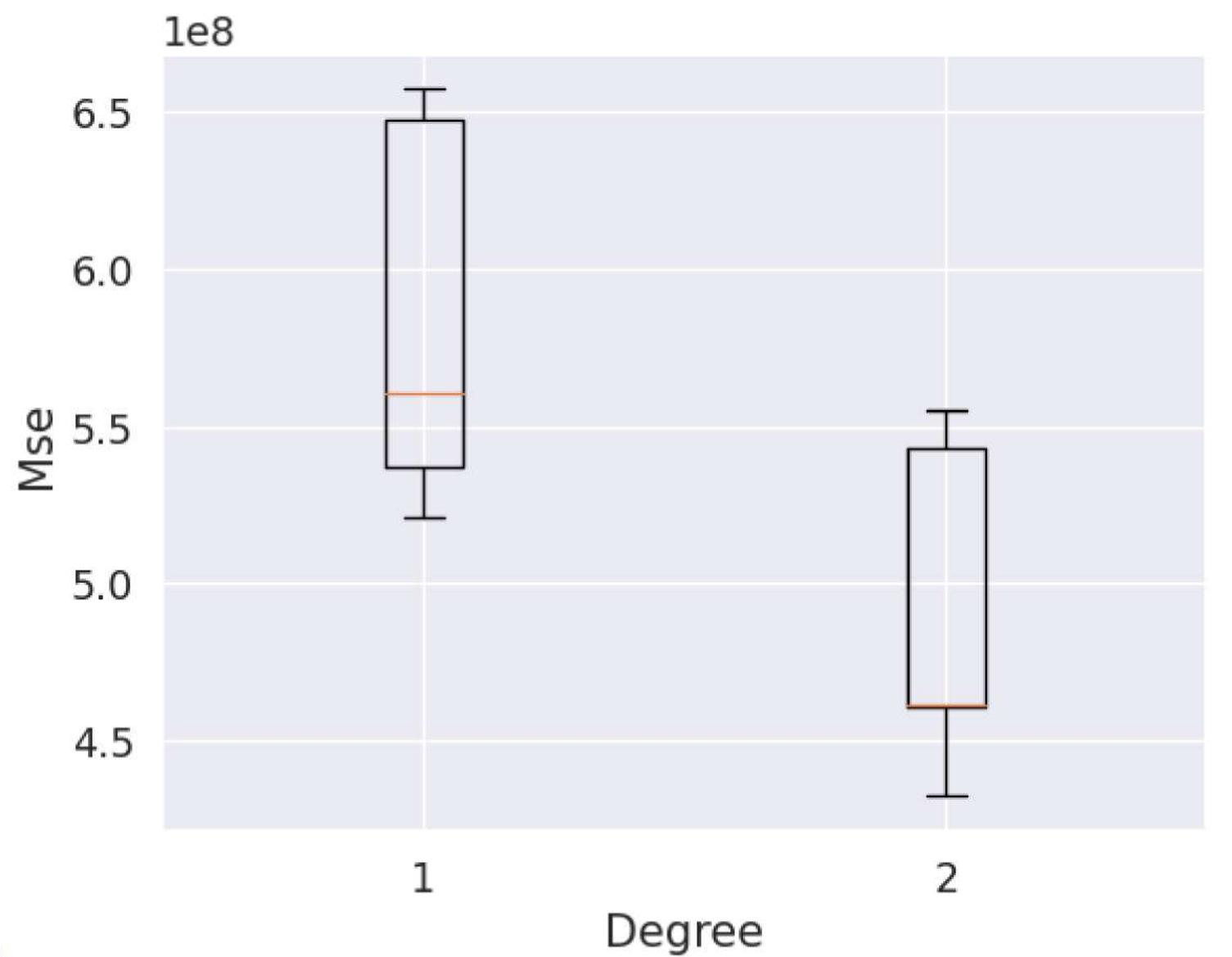


	LotShape	LandSlope	ExterQual	ExterCond	HeatingQC	KitchenQual	Functional	
0	4	3	4	3	5	4	8	
1	4	3	3	3	5	3	8	
2	3	3	4	3	5	4	8	
3	3	3	3	3	4	4	8	
4	3	3	4	3	5	4	8	
...	...	...	...	...	...	...	...	
2913	4	3	3	3	4	3	8	
2914	4	3	3	3	3	3	8	
2915	4	3	3	3	5	3	8	
2916	4	3	3	3	3	3	8	
2917	4	2	3	3	5	3	8	

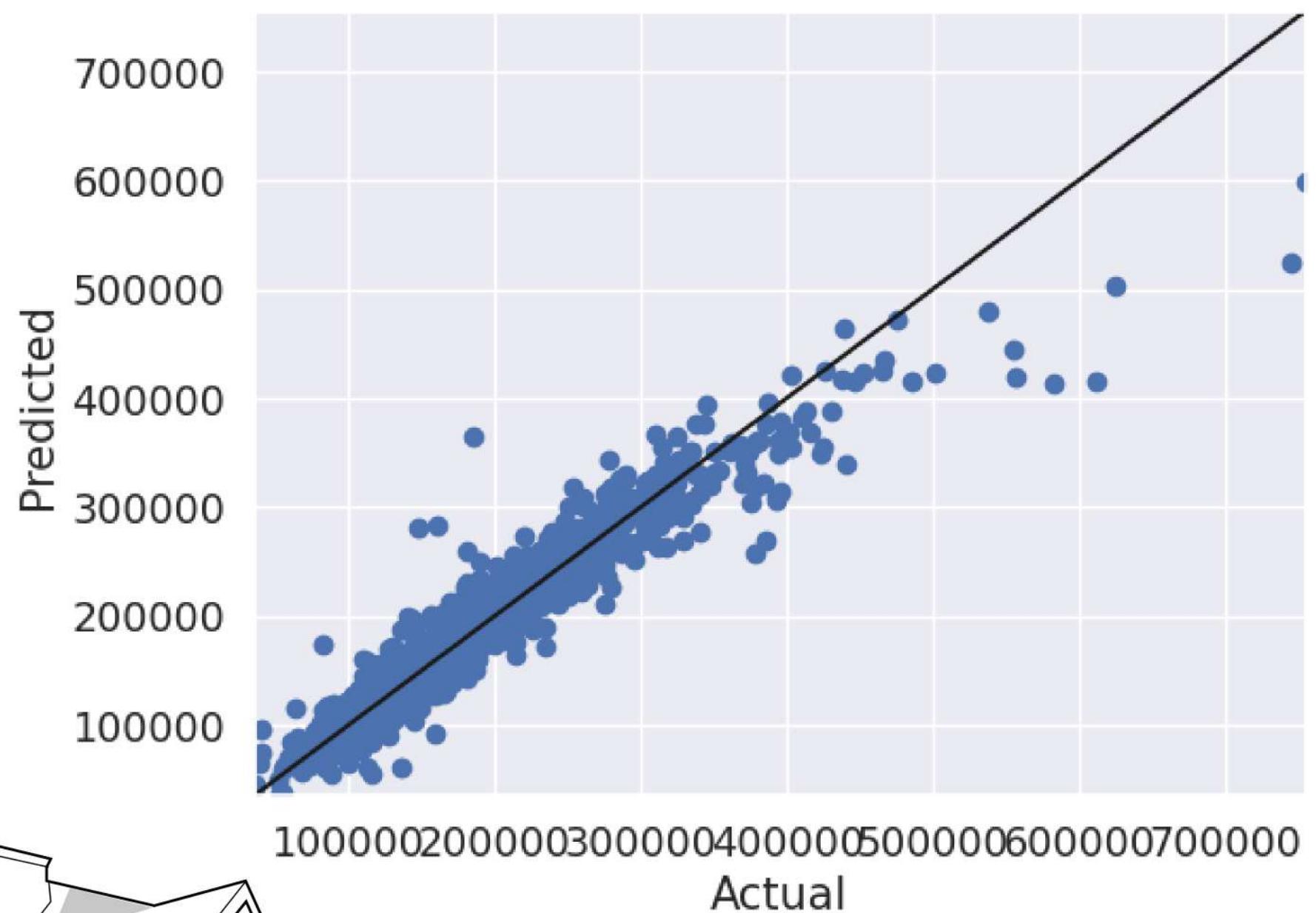
# LINEAR REGRESSION



# LINEAR REGRESSION



# LINEAR REGRESSION



```
model.score(x_poly, y_train)
```

```
0.9150407700354773
```

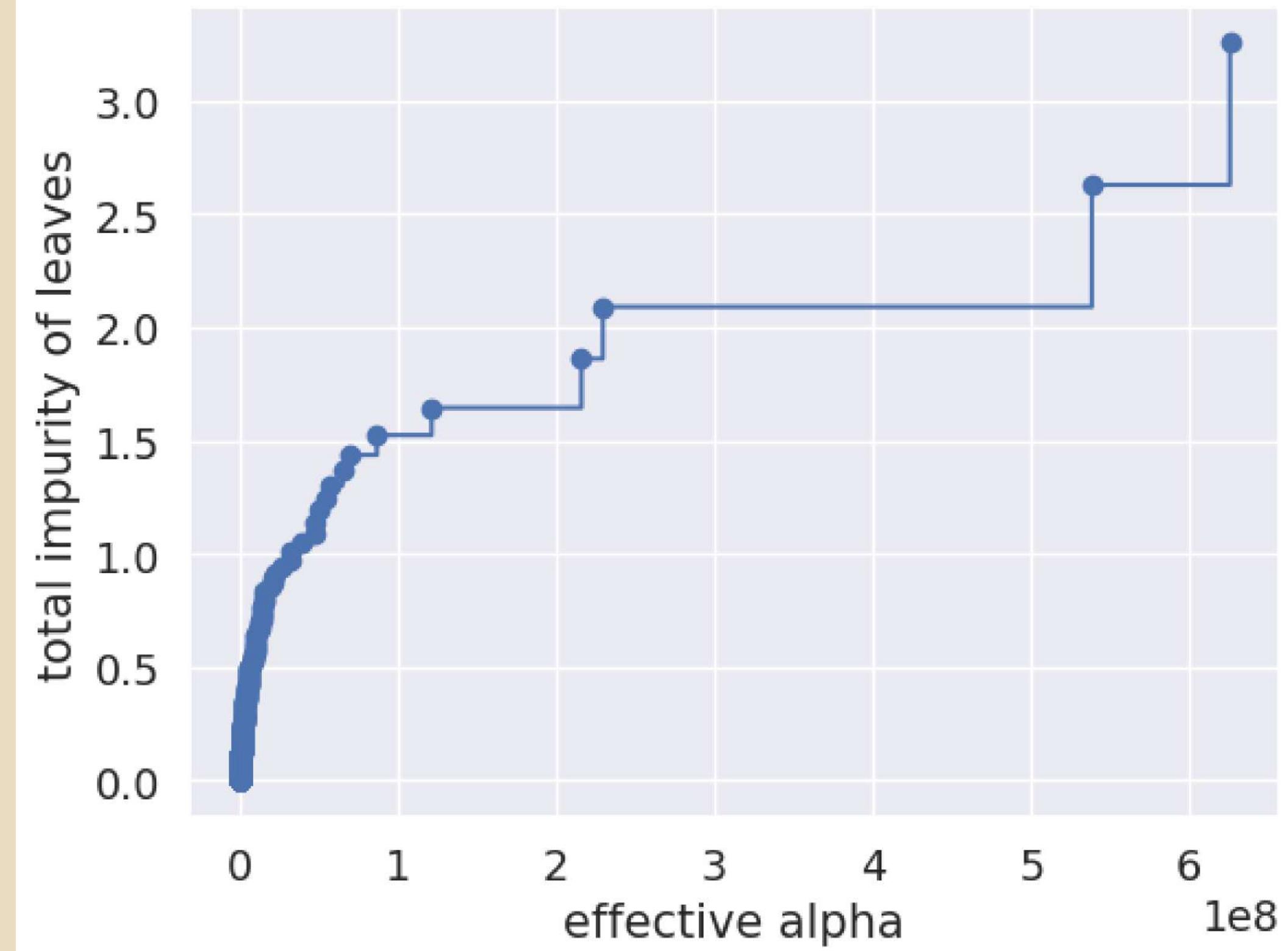
predicted\_results.csv

Complete · Jump Hopper · 7h ago · Ridge Regression

0.34343

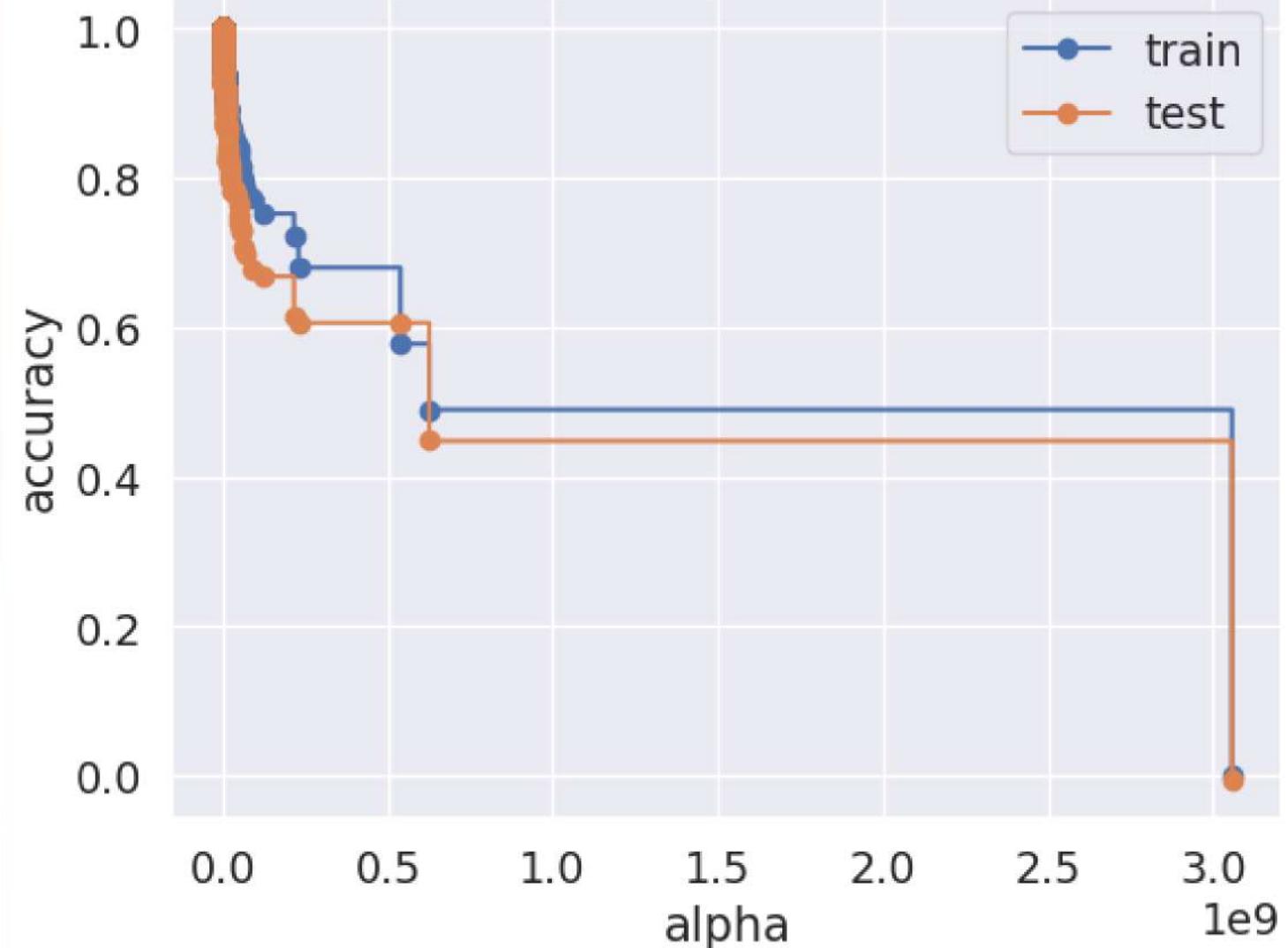
# DECISION TREE

Total Impurity vs effective alpha for training set  
1e9

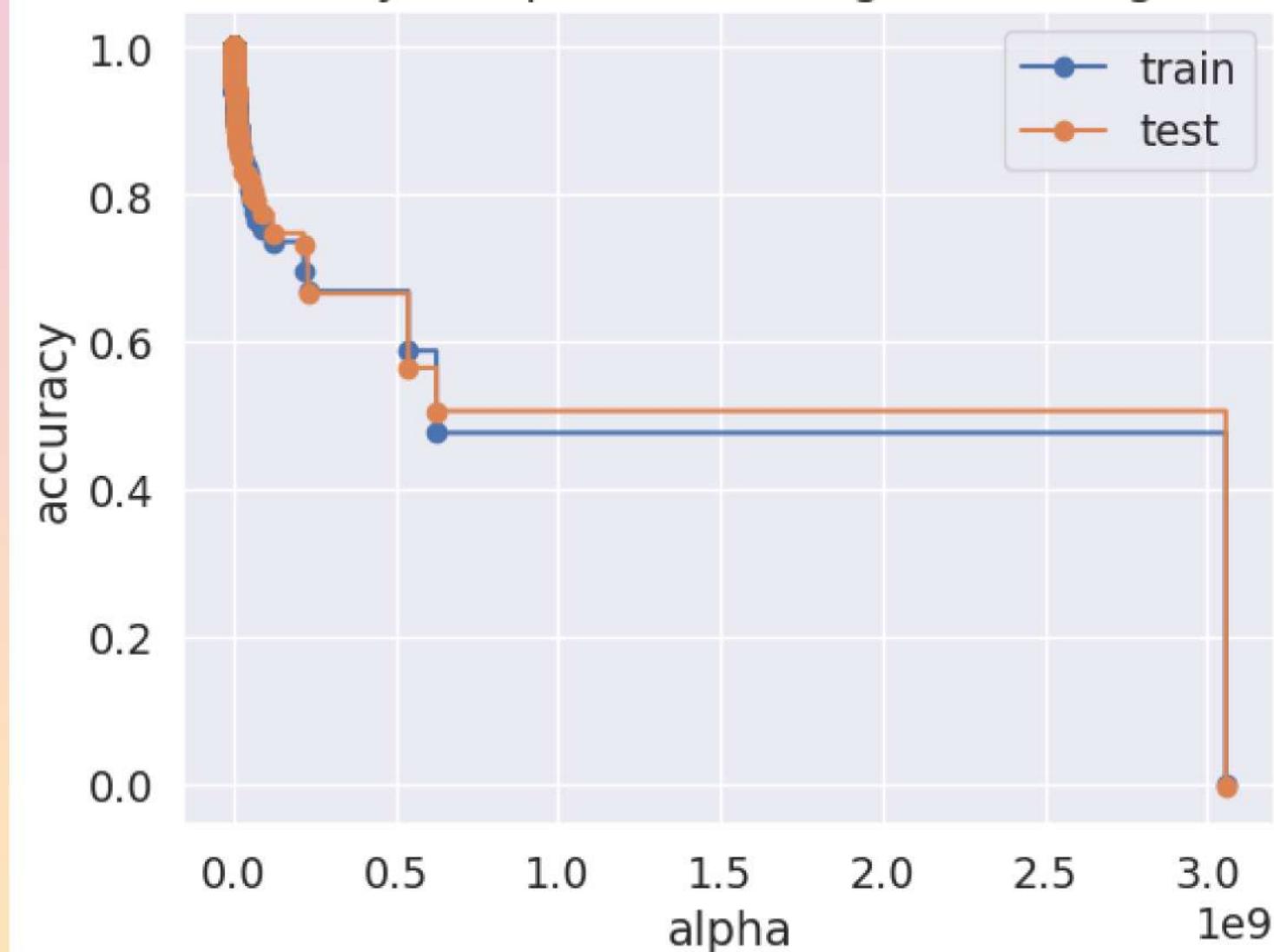


# DECISION TREE

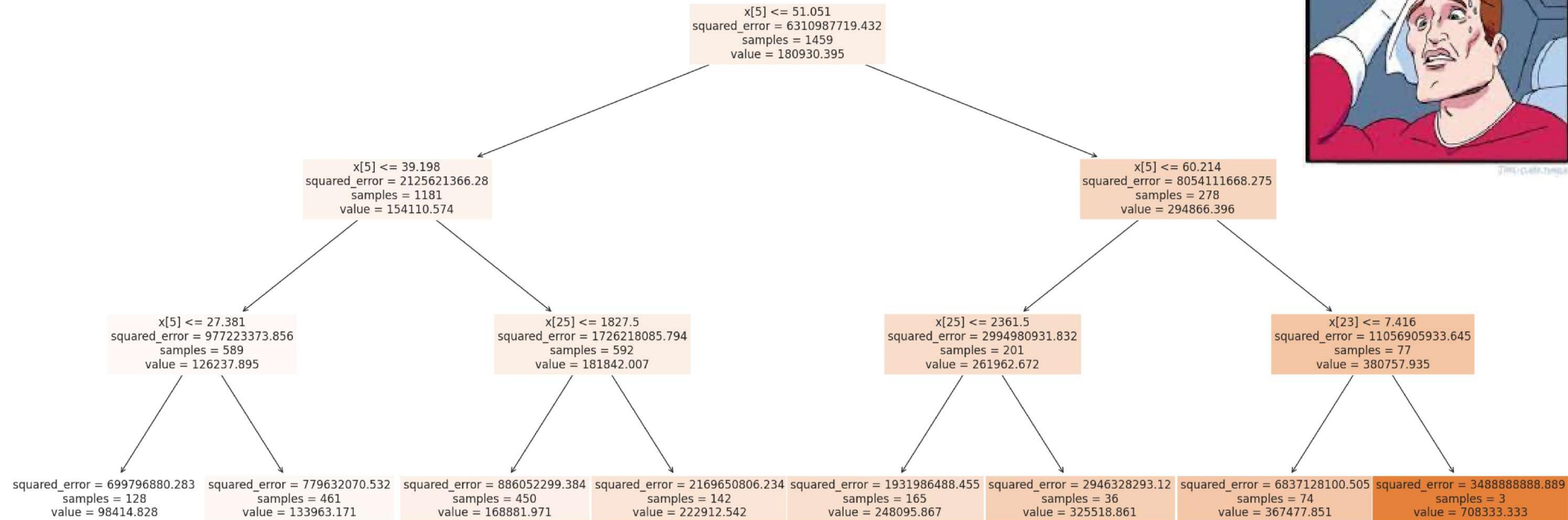
Accuracy vs alpha for training and testing sets



Accuracy vs alpha for training and testing sets



# DECISION TREE



# DECISION TREE

```
[1]: from sklearn.model_selection import cross_val_score
      from sklearn.tree import DecisionTreeRegressor
      regressor = DecisionTreeRegressor(random_state= RAND_STATE)
      regressor.fit(x_poly, y_train)
      print(np.mean(cross_val_score(regressor, x_poly, y_train, cv=10)))
```

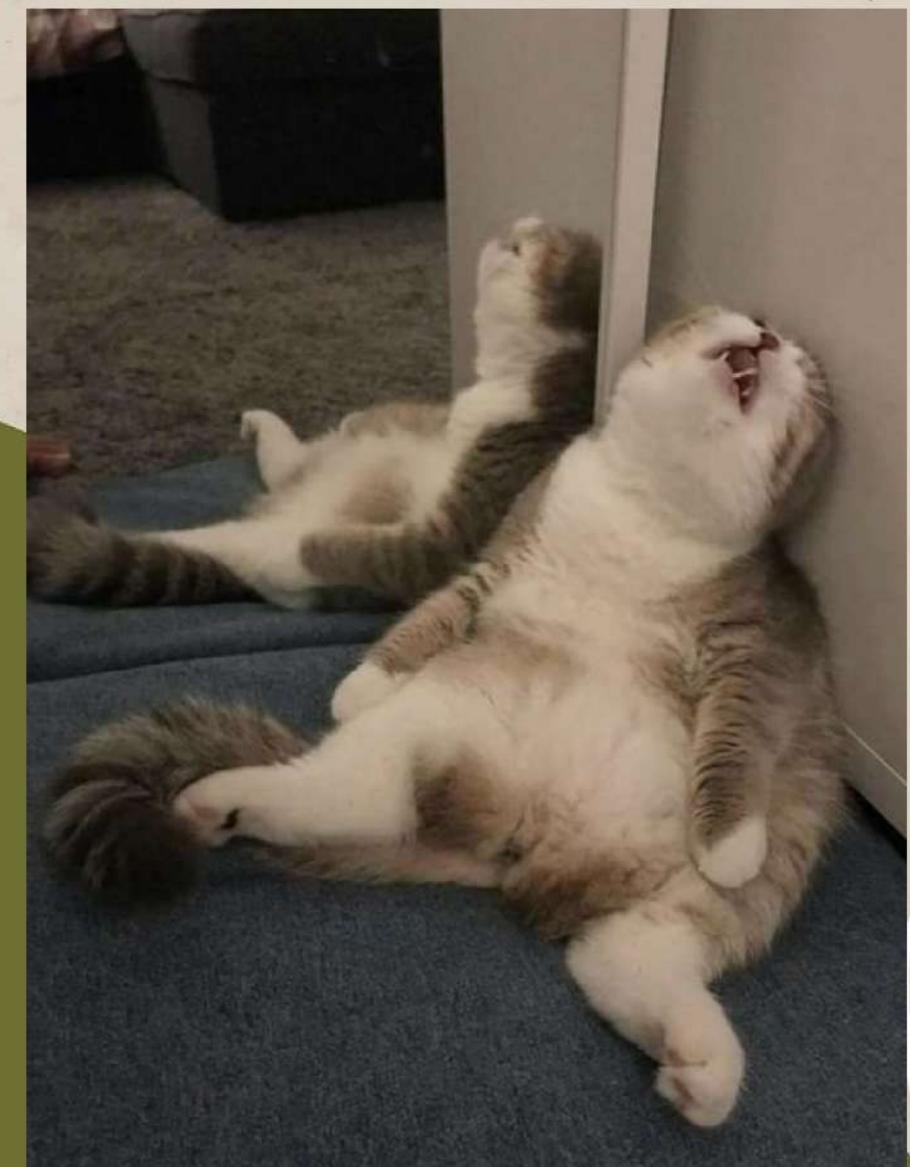
0.7526313923410217

DT\_results.csv

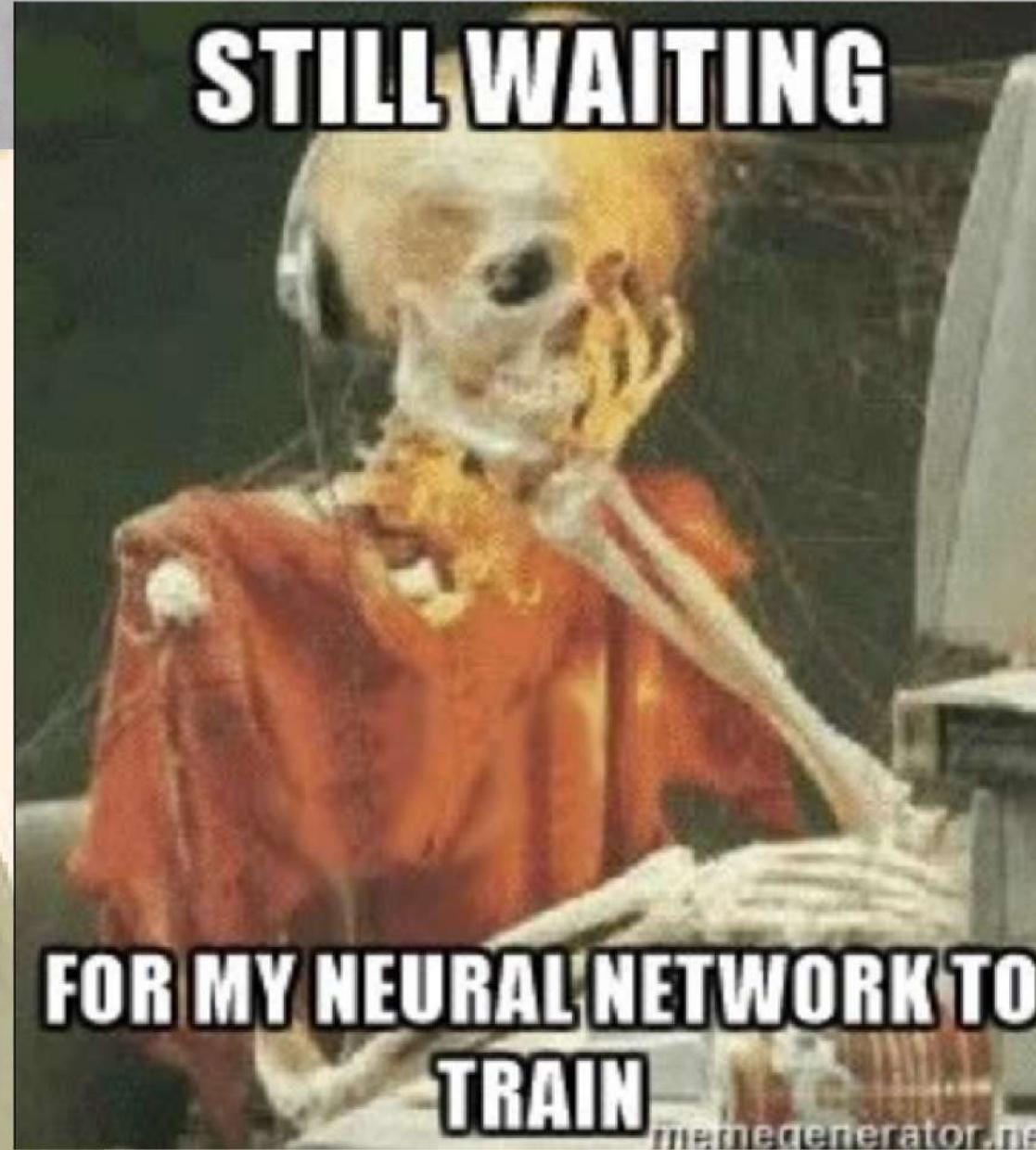


Complete · Jump Hopper · 6h ago · Decision Tree

0.20827



# NEURAL NETWORK



```
from sklearn.model_selection import GridSearchCV

param_grid = {
    'alpha':[0.00005, 0.00008, 0.0001],
    'hidden_layer_sizes': [
        (100,), (150,), (180,)
    ]
}

grid_search = GridSearchCV(nn, param_grid, cv=3,
                           scoring='r2', refit=True)
grid_search.fit(x_poly, y_train)

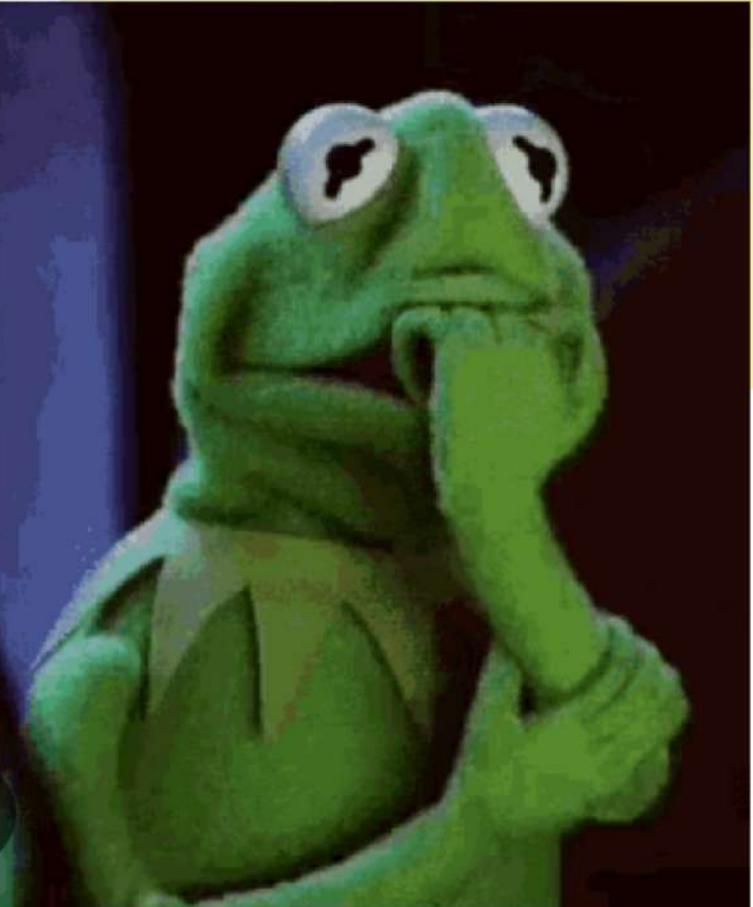
print("Best parameters set found on development set:")
print(grid_search.best_params_)

Best parameters set found on development set:
{'alpha': 8e-05, 'hidden_layer_sizes': (150,)}
```

# NEURAL NETWORK

```
from sklearn.neural_network import MLPRegressor  
nn = MLPRegressor(hidden_layer_sizes=150, alpha=8e-05, random_state= RAND_STATE, max_iter=1000).fit(x_poly, y_train)  
nn.score(x_poly, y_train)
```

0.717196792965644



NN\_results.csv

Complete · Jump Hopper · 3h ago · NN

0.18884

# RANDOM FOREST



 **I mean, trees? Everywhere trees?!  
What the hell is this place?**



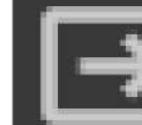
**submission (1).csv**

**0.14151**

Complete · bkkpup · 3h ago



**rf.score(x\_poly, y\_train)**



**0.9816694712337685**

```
[ ] from sklearn.ensemble import RandomForestRegressor
```

```
rf = RandomForestRegressor(n_estimators=500, min_samples_split=2, min_samples_leaf=1, max_depth=None, random_state=42)
```

# XGBOOST

```
[237] import xgboost as xgb

xgboost = xgb.XGBRegressor(learning_rate=0.01,n_estimators=3460,
                           max_depth=3, min_child_weight=0,
                           gamma=0, subsample=0.7,
                           colsample_bytree=0.7,
                           objective='reg:linear', nthread=-1,
                           scale_pos_weight=1, seed=27,
                           reg_alpha=0.00006)
```

```
# make predictions
xgboost.fit(x_poly, y_train, verbose=False)
predictions = xgboost.predict(x_test_poly)
```

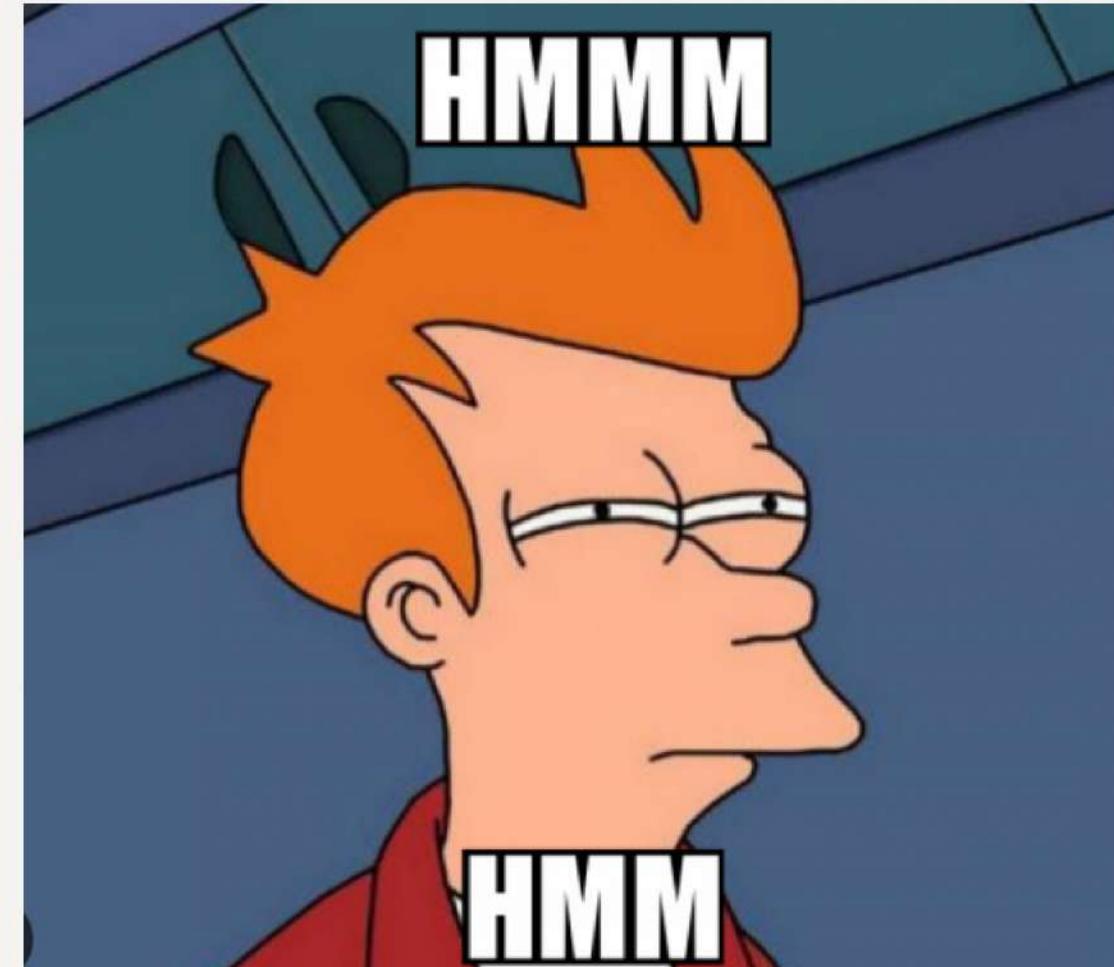


# XGBOOST

 **submission.csv** 0.12906  
Complete · Jump Hopper · 5h ago

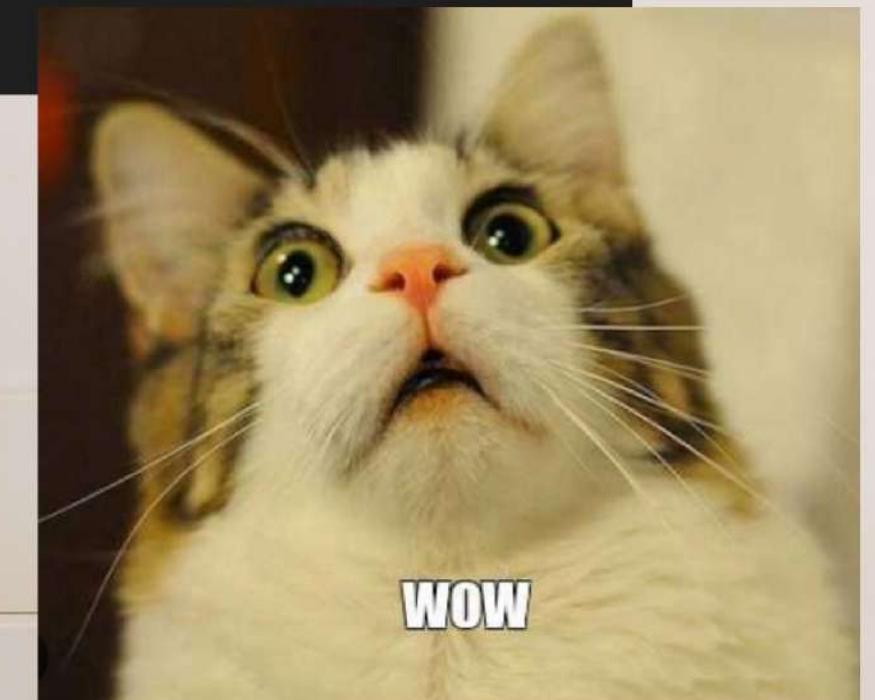
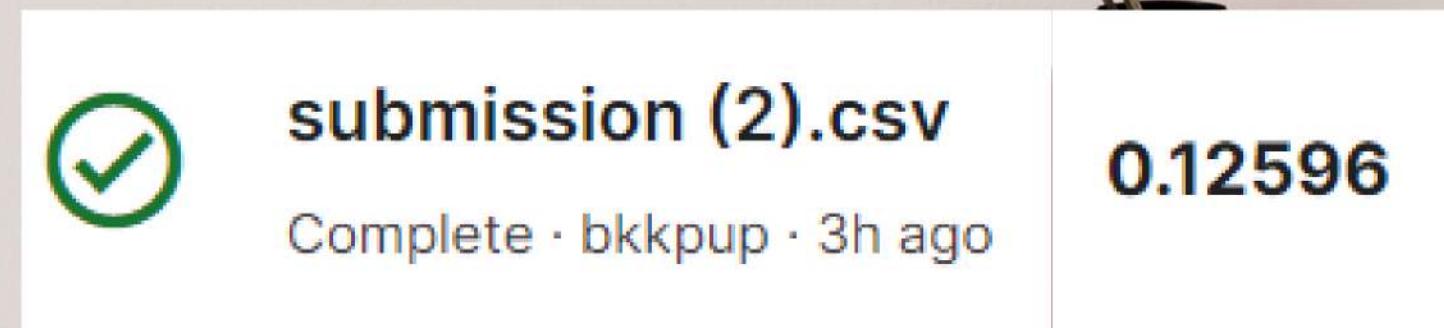
```
np.mean(cross_val_score(xgboost, x_poly, y_train, cv=5, scoring='r2'))
```

```
0.8973252875403605
```



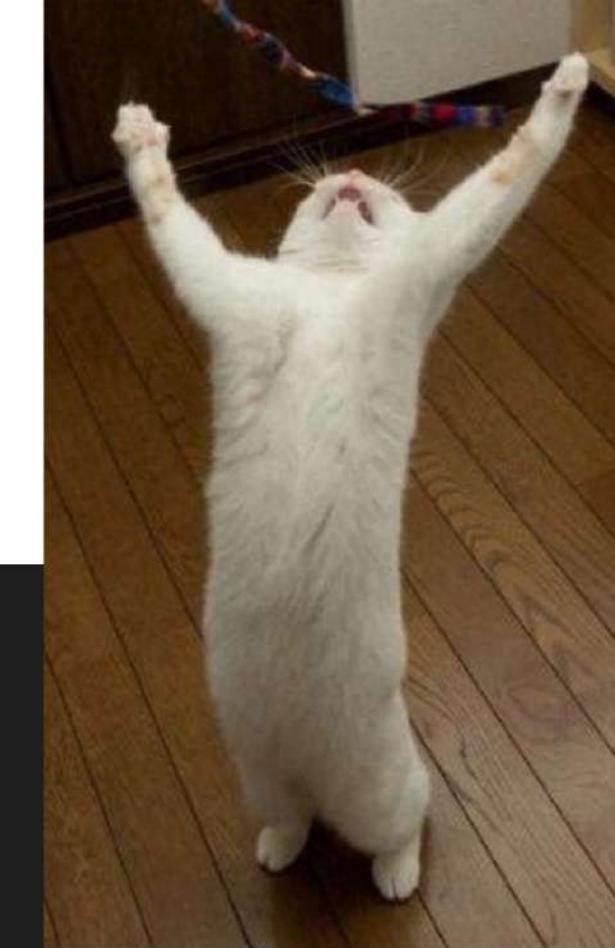
# STACKING MODEL

```
from mlxtend.regressor import StackingCVRegressor  
  
stack_gen = StackingCVRegressor(regressors=(xgboost, model, regressor, nn, rf),  
                                 meta_regressor=xgboost,  
                                 use_features_in_secondary=True)  
  
stack_gen_model = stack_gen.fit(np.array(x_poly), np.array(y_train))  
predictions = stack_gen_model.predict(x_test_poly)
```



# BLENDED PREDICTION

```
def blended_predictions(X):
    return ((0.3 * xgboost.predict(X)) + \
            (0.05 * model.predict(X)) + \
            (0.05 * regressor.predict(X)) + \
            (0.05 * nn.predict(X)) + \
            (0.15 * rf.predict(X)) + \
            (0.4 * stack_gen_model.predict(np.array(X))))
```



	blended.csv	0.12479
	Complete · 18s ago	

# OUR RANK

553	Kongphob Laoarun		0.12479	1	1m
	Your First Entry! Welcome to the leaderboard!				



**THANK YOU**

