

House Price Prediction



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Machine Learning Project

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1.Introduction

1.1HousePricePrediction

Dataset link(.xlsx)

<https://docs.google.com/spreadsheets/d/1EsFosB6JxawN201VJqBmBL3Av3ruMbBe/edit#gid=1199153200>

1.2 Description

This dataset concerns the housing prices in the housing city of Boston. The dataset provided has instances with 13 features.

The Description of the dataset is shown below:

1	Id	To count the records.
2	MSSubClass	Identifies the type of dwelling involved in the sale.
3	MSZoning	Identifies the general zoning classification of the sale.
4	LotArea	Lot size in square feet.
5	LotConfig	Configuration of the lot
6	BldgType	Type of dwelling
7	OverallCond	Rates the overall condition of the house
8	YearBuilt	Original construction year

9	YearRemodAdd	Remodel date (same as construction date if no remodeling or additions).
10	Exterior1st	Exterior covering on house
11	BsmtFinSF2	Type 2 finished square feet.
12	TotalBsmtSF	Total square feet of basement area
13	SalePrice	To be predicted

2. Libraries Used

- **Pandas** A powerful data manipulation and analysis library in Python.
- **Scikit-learn** A widely-used machine learning library for building and analyzing models.
- **XGBoost** An optimized gradient boosting library for classification and regression tasks.
- **Matplotlib** A comprehensive plotting library for creating visualizations in Python.
- **Seaborn** A data visualization library that provides an enhanced interface to create aesthetic and informative statistical graphics.

3.Algorithm(s) Used

Random Forest Regression-

Random Forest is an ensemble technique that uses multiple of decision trees and can be used for both regression and classification tasks.

Linear Regression-

Linear Regression predicts the final output-dependent value based on the given independent features. Like, here we have to predict SalePrice depending on features like MSSubClass, YearBuilt, BldgType, Exterior1st etc. To read more about Linear Regression refer this.

Regression Algorithm-

XGBoost Regressor is used in our project. XGBoost is an efficient implementation of gradient boosting that can be used for regression predictive modeling.

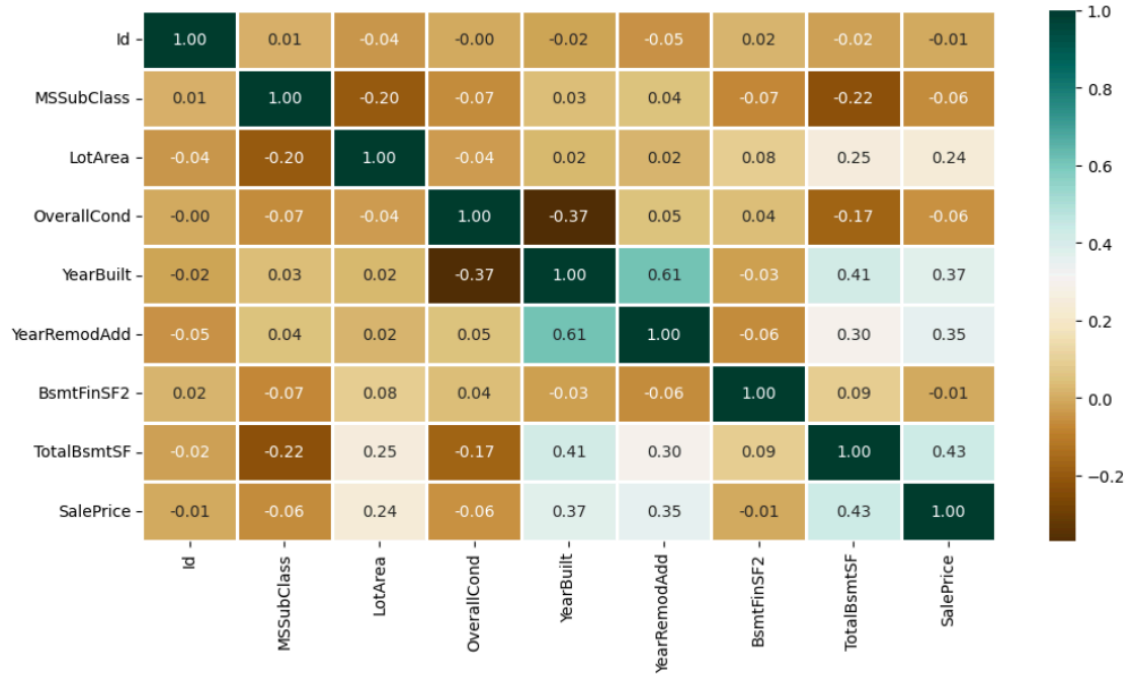
4.Code and Screenshots

1. Exploratory Data Analysis

EDA refers to the deep analysis of data so as to discover different patterns and spot anomalies. Before making inferences from data it is essential to examine all your variables.

```
plt.figure(figsize=(12, 6))
sns.heatmap(dataset.corr(),
             cmap = 'BrBG',
             fmt = '.2f',
             linewidths = 2,
             annot = True)
```

SALES. /



2.XGBoost Regression

XGBoost stands for “Extreme Gradient Boosting”. XGBoost is an optimized distributed gradient boosting library designed to be highly efficient, flexible, and portable. It implements Machine Learning algorithms under the Gradient Boosting framework. It provides a parallel tree boosting to solve many data science problems in a fast and accurate way.


```
import xgboost as xgb
from sklearn.metrics import r2_score


xgb_model = xgb.XGBRegressor()
xgb_model.fit(X_train, Y_train)
preds = xgb_model.predict(X_valid)

xgb_r2_score = r2_score(Y_valid, preds)
xgb_r2_score
```

▼ XGBoost Regression

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 0s



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0.28458200472226935