# Use Linear Regression to Predict House Price Based on Auto Selected Parameters from Known Dataset

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## **Project Statement**

#### **Problem:**

 When predicting house price, many parameters could influence the price. What parameters should we select?

#### **Project Statement:**

 The project utilized linear regression method to predict the house sale price by analyzing the existing information that provided house price and selected highly correlated factors with the price.

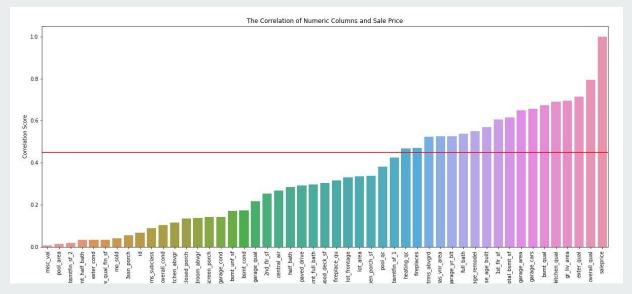
## **Basic Approach**

#### Data Source: Ames, Iowa Assessor's Office

- Two Dataset: Train and Test
  - Train dataset: Create linear model.
  - Test dataset: Predict house price with the linear model.
- Train dataset was splitted into 2 parts
  - Train (80%): Create linear model
  - Validation (20%): Evaluate the model accuracy
- Data Cleaning, Feature Engineering, Preprocessing
- Linear Regression
- Model Evaluation

### **Data Cleaning: Numeric Columns**

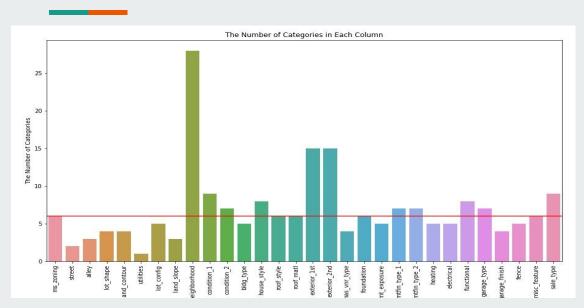
- Columns were splitted into numeric and object segments.
- Object columns only contained ratings were converted into numeric.
- The correlation score of each column versus sale price was calculated.



#### **Filtration Rule:**

- 1. Columns had correlation score over 0.45.
- 2. Columns had less than 5% of the missing values.

## **Data Cleaning: Object Columns**



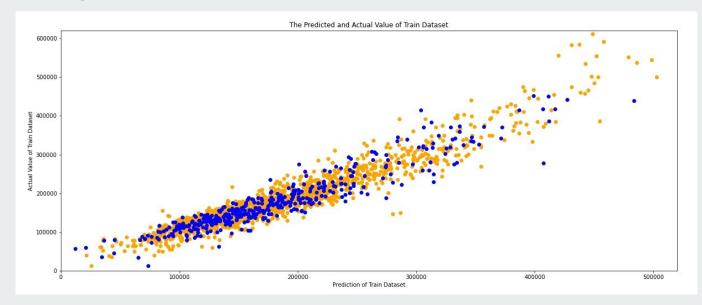
#### **Filtration Rule:**

- 1. Columns had no more than 6 categories.
- 2. Manually added the column of neighborhood

## Next Step:

- Combine numeric and object columns.
- Apply feature engineering and preprocessing, include Simple Imputer, OneHotEncoder and Standard Scaler

## **Linear Regression**



R^2 score:

Root Mean Squared Error:

Train (orange) 0.908

24,142

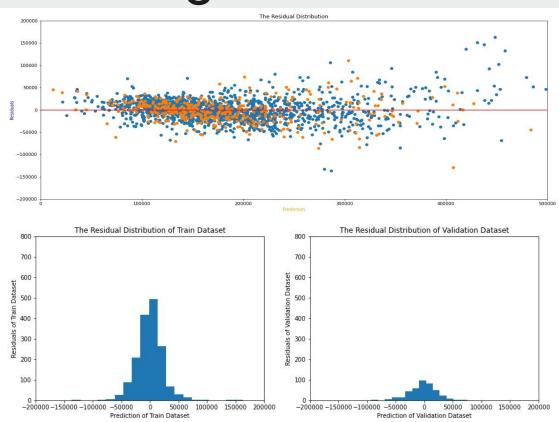
Validation (blue)

0.888

25,842

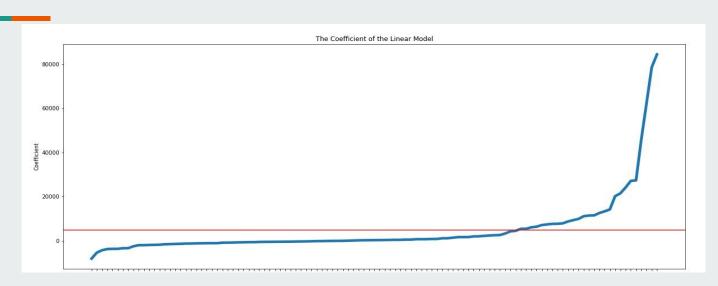
Low bias, Low Variance

## **Linear Regression**



- Low residuals between the range of \$100K to \$250K.
- The distribution of residuals was close to normal distribution.

## **Linear Regression**

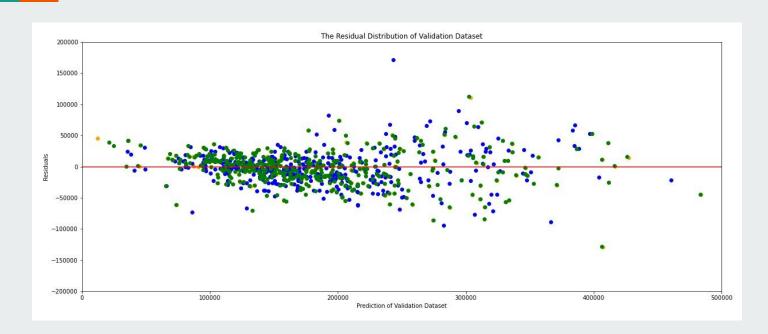


The coefficients showed the scales of impact on house sale prices.

Top Positive Features: Miscellaneous features not covered, Roof materials, Above ground living area

Top Negative Features: Building type - townhouse, House age, Garage - rough finish

## **Ridge and Lasso**



Original LR: orange, Ridge: blue, Lasso: green

#### **Conclusion and Recommendation**

- Linear model can predict!
- Prediction accuracy depends on the selected feature parameters.

#### **Suggestions:**

- For sale: roof material, masonry veneer area, kitchen, basement, heating system
- For investment: northridge heights, stone brook, green hills, northridge, crawford

## **Thank You for Listening**

**Questions?** 

