House Prices Advanced Regression Techniques

W207 - Applied Machine Learning - Final Project Yatin Majmudar, Anu Sankar, Eugene Tang Dec 17, 2018

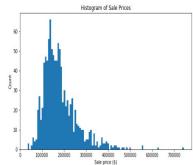


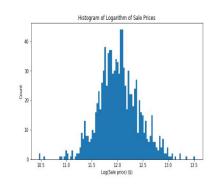
Introduction

- The goal of this project is to use the features of a residential home to predict the sales price for each house ("SalePrice" variable) in the test dataset scored on the rmse(root mean square error).
- Kaggle Competition Dataset <u>lowa House Prices</u>
- The dataset contains home sales made between 2006 and 2010.
- 79 explanatory variables.
- Training dataset 1460 rows
- Testing dataset 1459 rows.
- This data was originally compiled by Dean De Cock as <u>Ames Housing dataset</u> for use in data science education.

Approach

- Split the given Training dataset into train (70%) and dev (30%)
- Exploratory Data Analysis
 - Missing Data
 - Explore the outcome variable SalePrice
 - SalePrice is skewed to the right
 - Log(SalePrice) is normally distributed
 - Additional Variables
 - Age of the house (BuiltAge)
 - Years since the last remodeling (RemodelAge)
 - Correlation Matrix
 - Numeric and Ordinal Predictor variables with SalePrice and Log(SalePrice)
 - Univariate and Bivariate Analysis
 - Of numeric and ordinal predictor variables that are highly correlated with SalePrice and Log(SalePrice)
 - Apply Transformation to some predictor variables

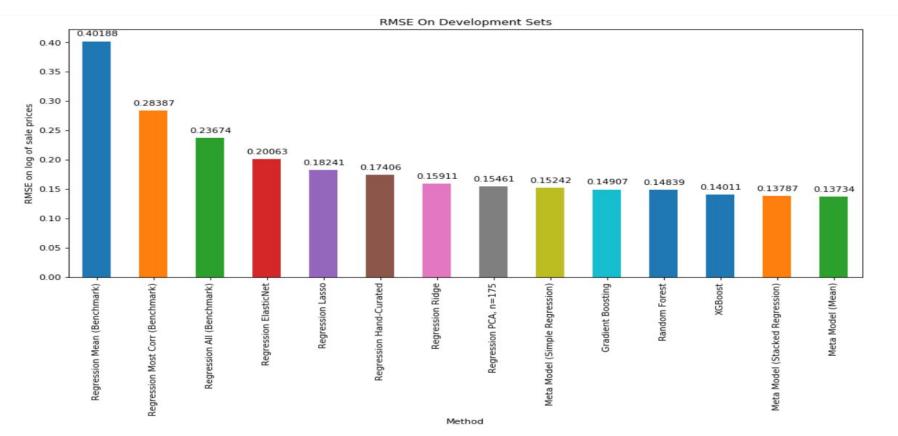




PreProcessing

- New Outcome Variable
 - Log(SalePrice)
- Transformations / Feature Engineering
 - Log(SalePrice)
 - BuiltAge Age of the house
 - RemodelAge Years since the last remodeling
 - Log(GrLivArea)
 - Log(LotArea)
 - Log(1stFlrSF)
- Dummy Variables
 - Created dummy variables for categorical and ordinal variable columns

Model Performance



Next Steps

- Reducing Feature Set (292 predictors, including dummies)
 - Remove features that are highly correlated to each other
 - Removing some of these provides a noticeably small improvement, but nothing major
 - LotFrontage, LotShape, GarageYrBlt, GarageCars, PoolArea, PoolQC, TotRmsAbvGrd
 - Removing features are sparsely populated
 - PoolQC, PoolArea, Alley, MiscFeatures
- Feature Engineering
 - Combine total square footage of the house (finished and unfinished)
 - Separate owned vs. rental properties
- Additional Models
 - Different models for different neighborhoods or different years (w/macroeconomic factors)
 - Neural Networks
 - o SVM

Backup: After removing strongly correlated predictors

