Foundation Of Data Science Project Report: Ames House Price Estimation

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Score: 0.11619

• Language: Python

• Data tidying:

- 1. Removed rows with > 60% Nan (Alley, PoolQC, Fence, MiscFeature)
- 2. For categorical put None Instead of Nan's, and for numeral's put mode, median or '0' based on the values.
- 3. Removed outliers Based on scatter plot of every 'feature value' and 'target value'.
- 4. Normalization: for features with skewness >60%, we do log + 1 transform.

• Feature engineering:

- Delete Features with correlation value > 80%. (GarageArea, 1stFlrSF, TotRmsAbvGrd, GarageYrBlt, KitchenAbvGr)
- 2. Create 'Dummy Variables' for categorical ones.
- 3. Add Poly (poly = 2) Feature for 12 most important Features. (Based on lasso)
- 4. Add feature 'NewHouse' (1 for new house and 0 for others) based on 'YrSold' and 'YearBuilt' features subtraction.
- 5. Add feature 'OverallSF' based on sum of '2ndFlrSF' and 'TotalBsmtSF' features.

• Base Models:

- 1. One LassoCV
- 2. One LassoLarsCV
- 3. One Elastic Net
- 4. One Linear Regression
- 5. Two Random Forest Regressor
- 6. Two Gradient Boosting Regressor

Training

- 1. Run every base model with 10-fold KFOLD validation and do prediction for test data.
- 2. Then use average of base modes to create the model.
- 3. After creating final model we use GridSearchCV from sklearn as stacking model with Ridge as estimator and find best hyper parameter for this estimator.
- 4. We do prediction based on GridSearchCV and achieve 0.1012 in python test and 0.11619 in Kaggle website.