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**Team Name:** 1873995 & 1851049  
**Score:** 0.11358

**Language:** Python

**Data Tidying:**

1. Removed outliers with data GrLivArea > 4000 & SalePrice < 300000
2. Dropped columns which have Multicollinearity > 0.8.
3. Normalize the Target feature i.e. Sale Price
4. Apply Log-Transformation if the Skewness > 0.6.
5. Make dummy features for categorical data.
6. Replacing missing numerical values with mean of the feature.

**Feature Engineering:**

1. Create TotalHousePorchSF which contains all features about porch.
2. Create TotalSF which is the total square feet of floors.
3. Created OverallQual\_2 which is squared OverallQual because it is mostly correlated.

**Modelization:**

1. Linear Regression

**Training:**

We have Stacked the below models to improve our prediction on Sale Price.

1. Lasso
2. Elastic Net Regression (ENet)
3. Kernel Ridge Regression (KRR)
4. Support Vector Regression (SVM)
5. Gradient Boosting Regression (GBoost)
6. Light Gradient Boosting (LGB)
7. Extreme Gradient Boosting (XGB)

After we evaluate the weight for each model, we apply the model below.

$$ENet * 0.6 + GBoost * 0.3 + LGB * 0.1$$