

Group 3: Predicting WAR

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Overview

- ▶ Goal: build a model to predict WAR (Wins Above Replacement)
- ▶ WAR measures a player's total contribution compared to a replacement-level player
- ▶ Interpreted as the number of additional wins a player adds to a team



Figure 1: War Ratings Chart

Data used

- we love baseball

Models used

- ▶ OLS

- ▶ Simplest model used

- ▶ LASSO

- ▶ Feature selection

- ▶ Boosting

- ▶ Tree based method

OLS Models

- ▶ OLS identifies and measures the relationship between a response variable and predictor variables.
- ▶ Finds a best-fitting line through a set of data points
- ▶ Pros: Convenient, accurate regression results for linearly related data
- ▶ Cons: May be too simplistic for real world examples, assumptions of Linear Regression

OLS Metrics Plot

LASSO Models

- ▶ LASSO models perform regularization (L_1), which shrinks some coefficients to exactly zero
 - ▶ Essentially feature selection
- ▶ Pros: Produces a more interpretative model, prevents over fitting
- ▶ Cons: LASSO performs poorly when predictors are highly correlated

LASSO metrics plot

- ▶ Metrics ran on split training data
- ▶ Good performance?

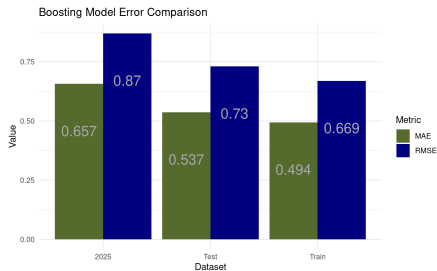


Boosting Models

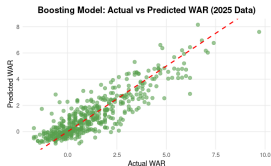
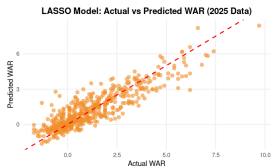
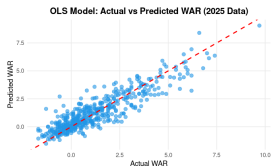
- ▶ Boosting grows trees sequentially using information from previously grown trees
 - ▶ Each tree fit on a modified version of the original data set
- ▶ Pros: High predictive accuracy and captures complex, nonlinear relationships automatically.
- ▶ Cons: Prone to overfitting and requires careful tuning of hyperparameters to perform well.

Boosting metrics plot

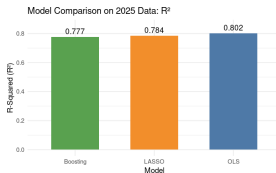
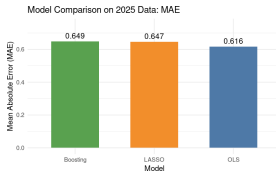
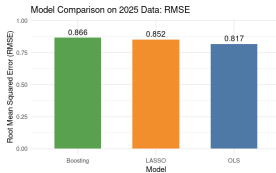
- Metrics ran on split training data



Model Comparison: OLS, LASSO, and Boosting



Model Comparison: RMSE, MAE, and R²



Player Examples

Player	Prediction	Actual
Aaron Judge	9.051733	9.7
Hunter Goodman	3.174746	3.7
Michael Toglia	-1.032315	-1.7
Bobby Witt Jr.	6.125421	7.1