



Generating Betting Insight on NBA Player Performance using Machine Learning



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Overview

We present an interactive dashboard to inform statistically optimal basketball betting strategy and describe trends in player-level data. We hope the use of interactive visualization around ML prediction should help to distill advanced data science concepts for everyone to receive data-driven advice to assist their intuition.

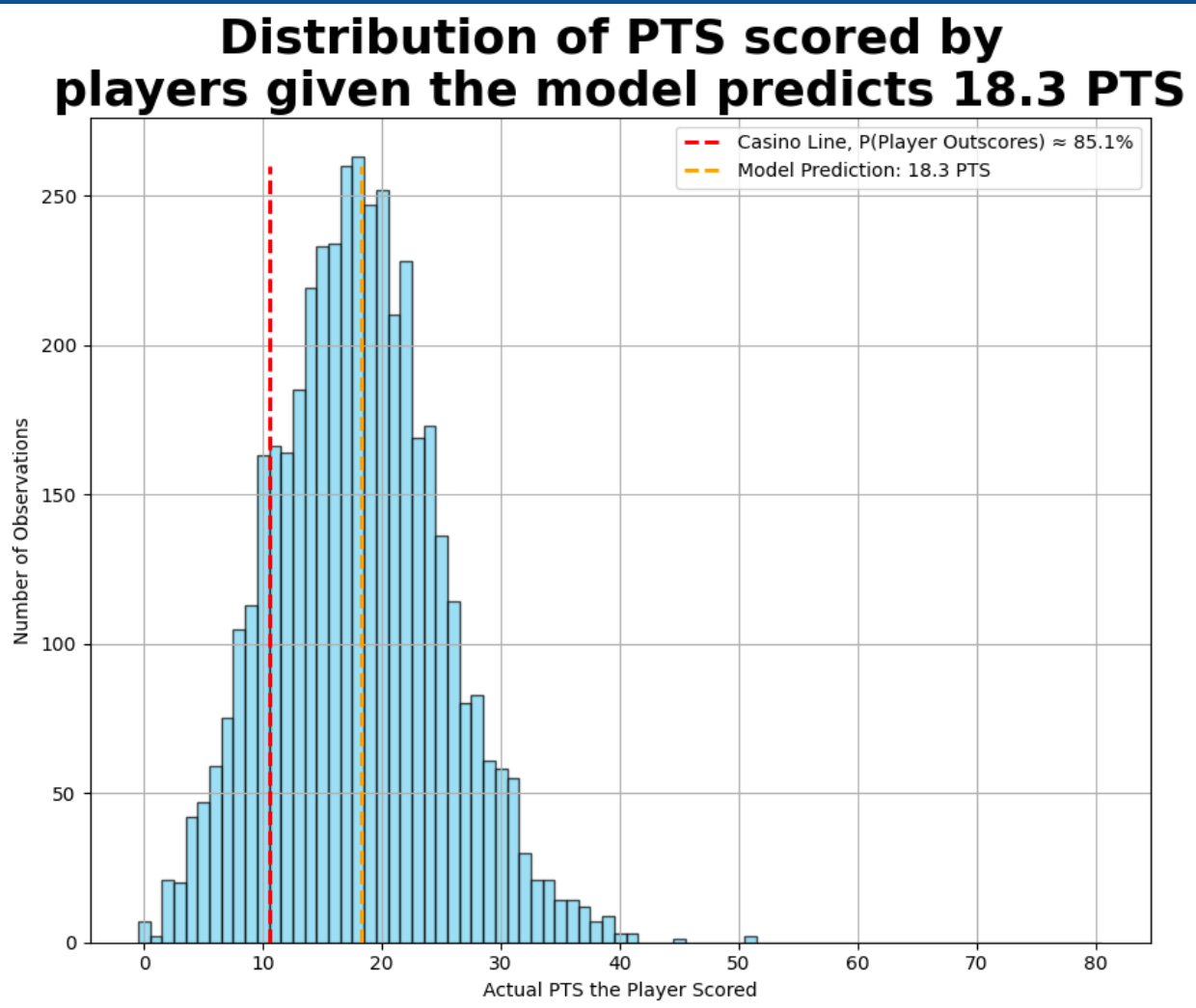
Modeling To identify profitable bets given historical professional basketball data, our team trained a model outputting a probability distribution of the five main player stats: Points (PTS), Assists (AST), Rebounds (REB), Blocks (BLK), Steals (STL).

Visualization These insights are displayed in a digestible and interactive formats allowing for user inputs to observe probability distributions for model

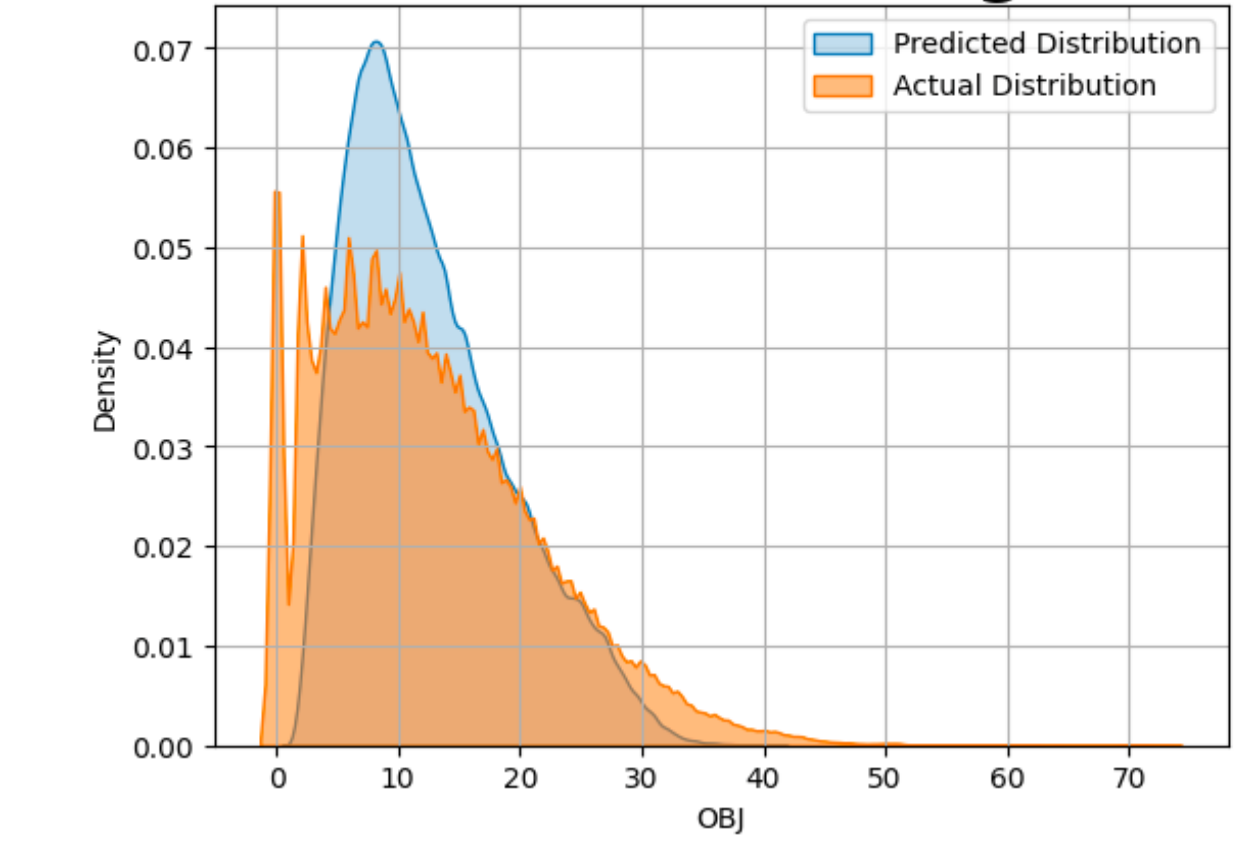
Visualization

Given user-input for a **player** and **statistic**, display two visuals:

- Time-series plot of the player's statistical output over their last ten games combined with the projection and sportsbook line for their upcoming game
- Distribution plot of historical performances amongst all players that had been previously projected the same statistical outcome by our model
 - Vertical lines included for the model projection and the sportsbook line
 - The proportion of the distribution falling above/below the sportsbook line represents the percent "confidence" of the model's recommendation for an over/under bet



Training vs. Predicted Distribution of PTS from Lasso Regression



Modeling

Data Summary:

- 142,320 Observations
- All Active NBA Players and Regular Season Games since 1995

Models Evaluated:

- Ridge/Lasso Regression
- Neural Networks
- K-Nearest Neighbors Regression
- AdaBoost and Random Forest Regression

Test MAE By Model, Statistic

PTS REB AST STL BLK

Ridge Regression

Lasso Regression

4.813

Neural Network

KNN Regression

AdaBoost

Random Forest

Final Model: Lasso Regression ($\alpha=9e-3$)

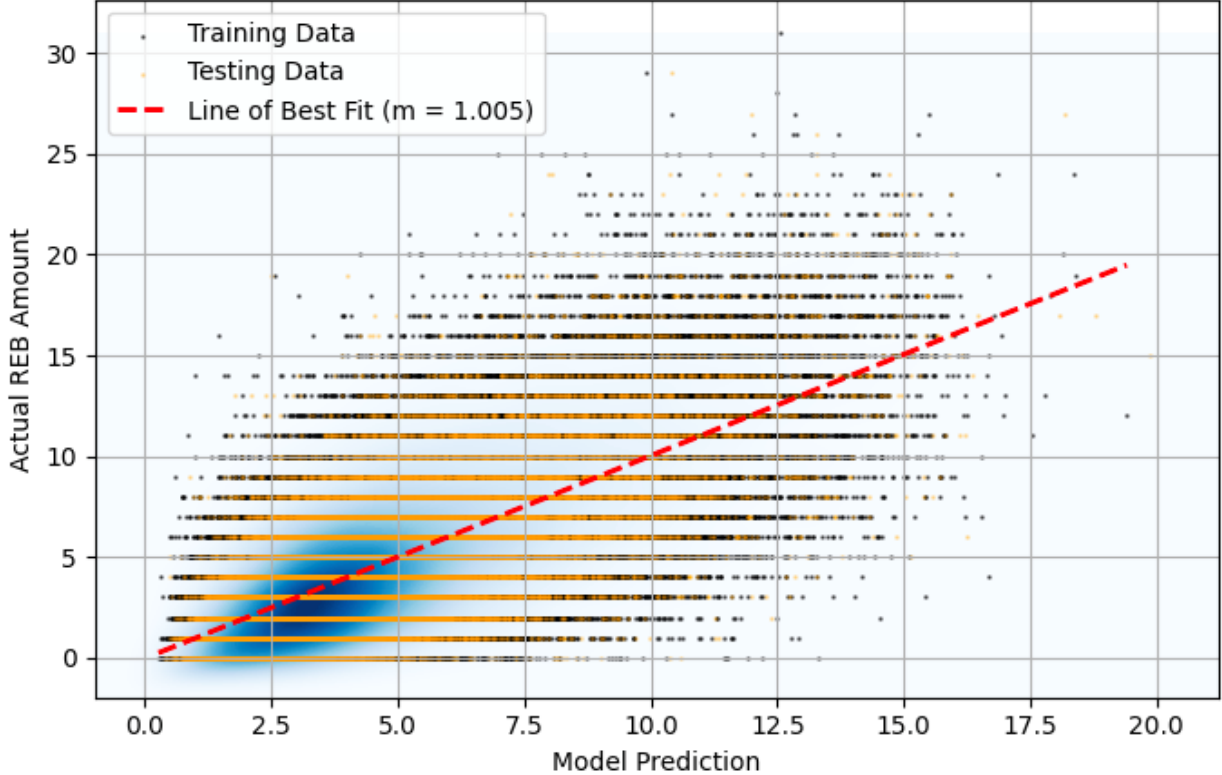
- Chosen for its accuracy, interpretability, and efficiency both in training and evaluation. Notably avoids overfitting.

Features include:

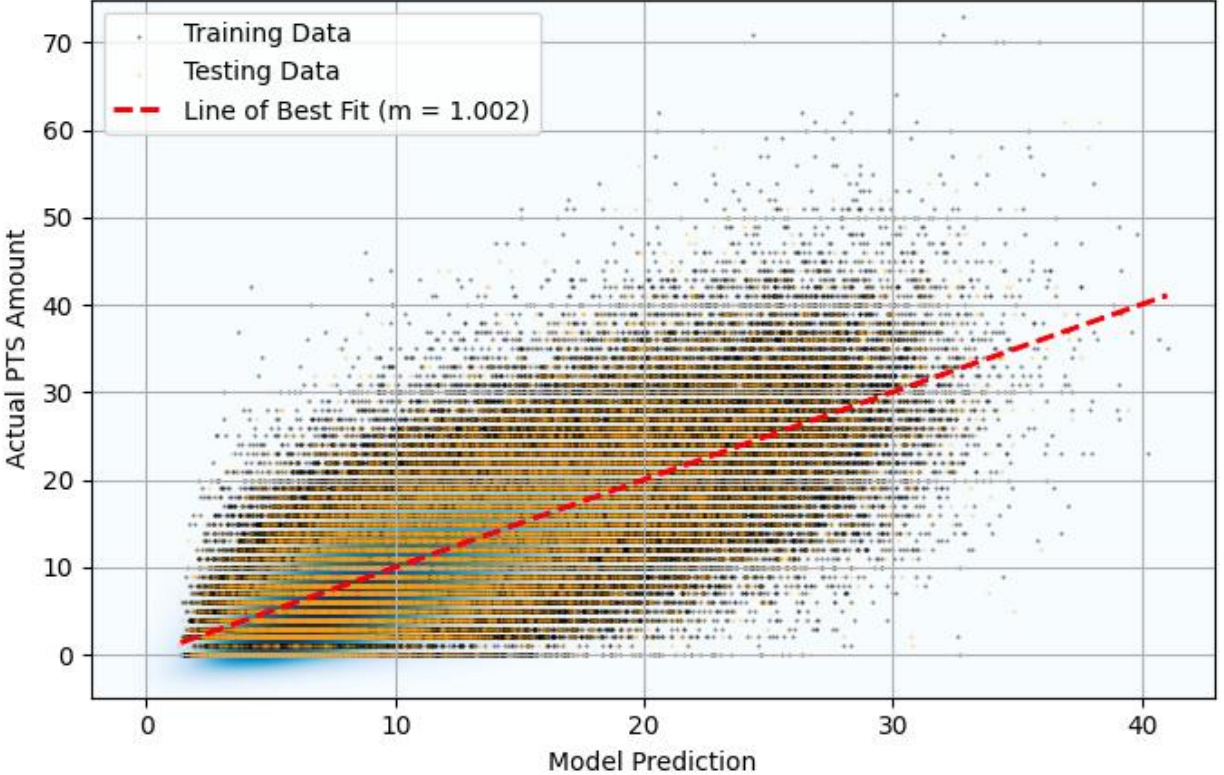
- Seasonal Running Average of W/L, MIN, FGM, FGA, FG_PCT, FG3M, FG3A, FG3_PCT, FTM, FTA, FT_PCT, OREB, DREB, REB, AST, STL, BLK, TOV, PF, PTS, PLUS_MINUS
- Whether or not the player is on the Home Team
- 5-Game Lookback of Stat of Interest
- Seasonal Running Average of Team Box Score Statistics

Results

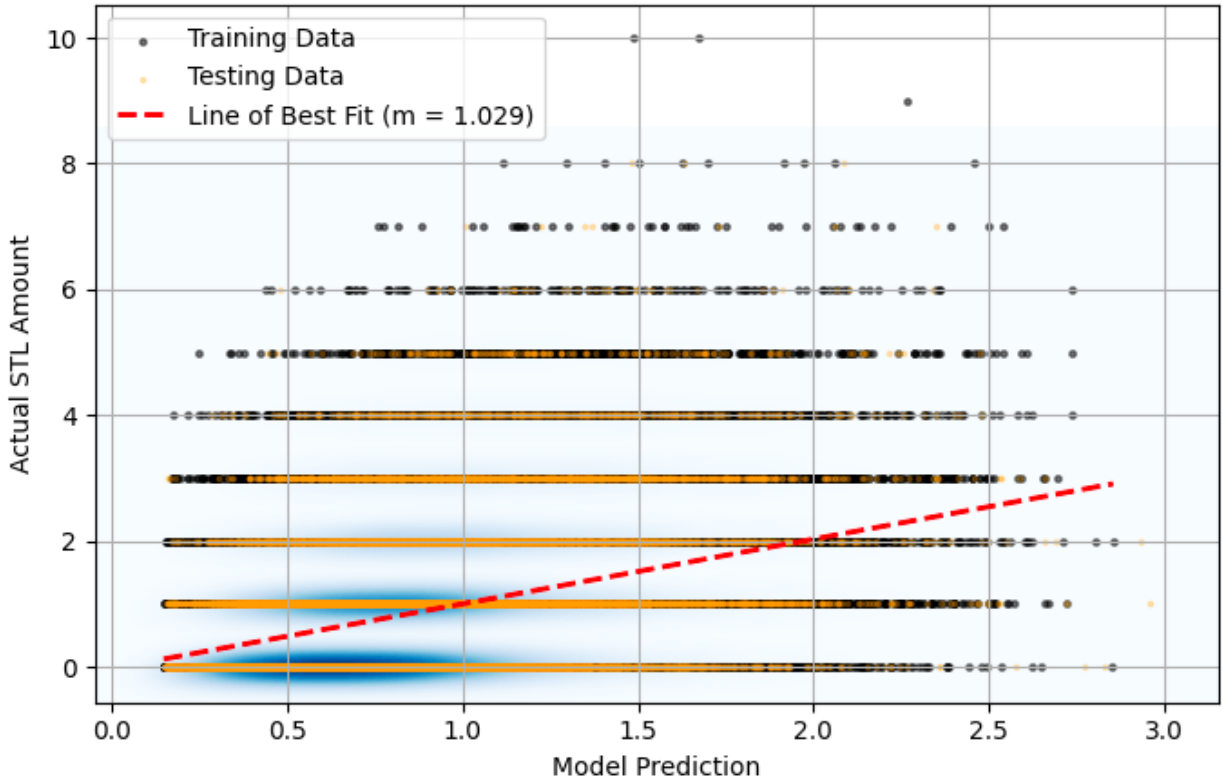
Actual REB Scored vs. Predicted REB Scored



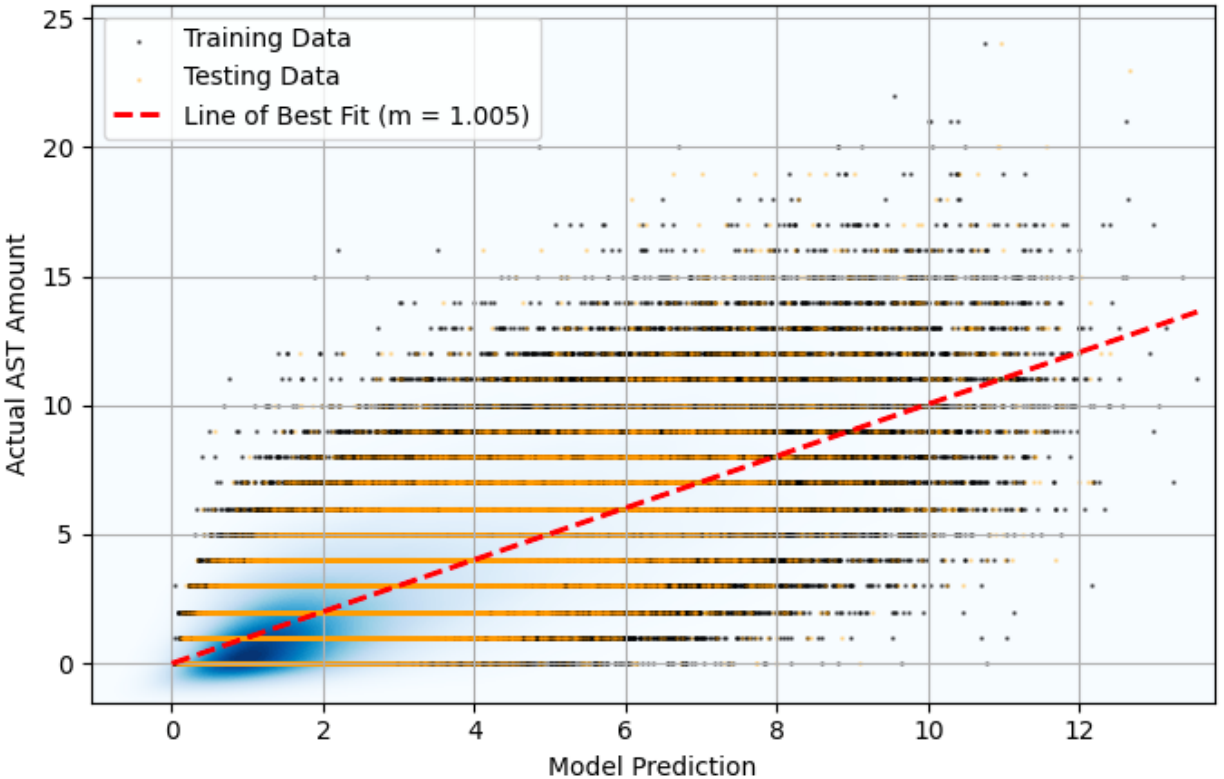
Actual PTS Scored vs. Predicted PTS Scored



Actual STL Scored vs. Predicted STL Scored



Actual AST Scored vs. Predicted AST Scored



Conclusion