

NCAA Basketball Point Spread Prediction Model

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Model Architecture

We developed an ensemble model combining Ridge Regression (30%) and LightGBM (70%) to predict point spreads:

$$\text{Predicted Spread} = 0.3 \times \text{Ridge}_{\text{pred}} + 0.7 \times \text{LightGBM}_{\text{pred}}.$$

Ridge provides a stable, interpretable linear baseline, while LightGBM captures complex non-linear patterns. The ensemble balances model transparency with predictive accuracy.

Data Sources & Features

Our model uses **11 features** derived from two primary data sources:

- **Barttorvik efficiency ratings**¹: Adjusted Offensive/Defensive Efficiency (AdjOE/AdjDE), measured as points per 100 possessions and adjusted for opponent strength. Efficiency Margin (AdjEM) = AdjOE - AdjDE.
- **Elo ratings**²: FiveThirtyEight-style Elo with K-factor = 38, home-court advantage = 4.0 points, and 64% season carryover to conference average.

Complete feature set (11):

- **Team efficiencies (6)**: home/away AdjOE, AdjDE, AdjEM.
- **Elo features (3)**: home/away Elo, Elo differential.
- **Derived features (2)**: efficiency differential, Elo-based spread.

Training Data

Trained on **18,024 NCAA games** from 2020–2025 (6 seasons), representing 53.4% of games after filtering for complete Barttorvik data. Games processed chronologically for accurate Elo histories and to avoid data leakage.

Model Evaluation

We used **5-fold time-series cross-validation** to respect temporal ordering, evaluating on future games to mimic real-world prediction.

Performance metrics:

Model	MAE (points)	RMSE (points)
Ridge	5.975 ± 0.160	7.85
LightGBM	4.65 ± 0.29	6.22
Ensemble	5.171 ± 0.248	6.89

The ensemble achieves MAE of **5.2 points** with lower variance, demonstrating improved robustness.

Key Decisions

- **Feature selection**: Retained 11 baseline features (efficiency + Elo). Advanced features decreased performance. AdjOE/AdjDE and Elo differential were strongest predictors.
- **Data quality**: Comprehensive validation ensures correct team-stat alignment.
- **Ensemble weighting**: LightGBM 70%, Ridge 30%. Optimized via validation.
- **Hyperparameters**: Ridge $\alpha = 1.0$; LightGBM: 100 trees, depth = 8, rate = 0.1.

Code: github.com/calebyhan/triangle-sports-analytics-2

¹barttorvik.com

²fivethirtyeight.com/features/how-we-calculate-nba-elo-ratings/