



Fair or Forgotten? Forecasting and Evaluating 2024 Playing Time

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*2025 Cincinnati Reds
Baseball Analytics Hackathon*



Objective:

Predict Total Playing Time Players Will
Receive in 2024

Batters → Plate Appearances

Pitchers → Batters
Faced

Using 2021-2023 Baseball Savant and Lahman Data

Filtering Players



Filtering the Dataset



Remove Pitchers from PA & Batters from BF Datasets

Pitchers aren't signed to hit and batters aren't signed to pitch (minus Ohtani), helps reduce noise in the data

Pitchers from PA

Check players who appear in BF data and has <10 PAs in 2022 and 2023

Universal DH implemented in 2022

Batters from BF

Check players who appear in PA data and has <50 BFs across 2021-2023

This is AFTER pitchers are removed





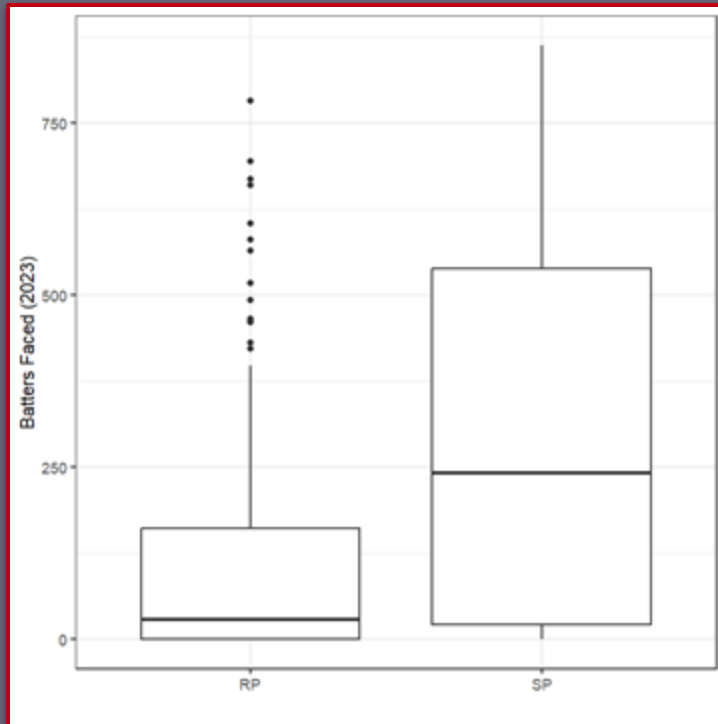
SPs Versus RPs



Differentiation

Number of Appearances
Based on Each Role

$$0.15 * (\# \text{ 2021 Appearances}) + \\ 0.85 * (\# \text{ 2022 Appearances})$$



Methodology



Position Players

Workload

- PA in Previous 2 Seasons
- Lineup + Batter Position
- Second Half PA
- Games Started After Debut
 - Rookies
- PA Per Game

Experience

- Debut Month (Rookies) and Year
- Age

Performance

- Avg and 90th % EV
- Total Bases, OBP, SLG
- Fielding/Framing RV
- Other Count/Rate Stats



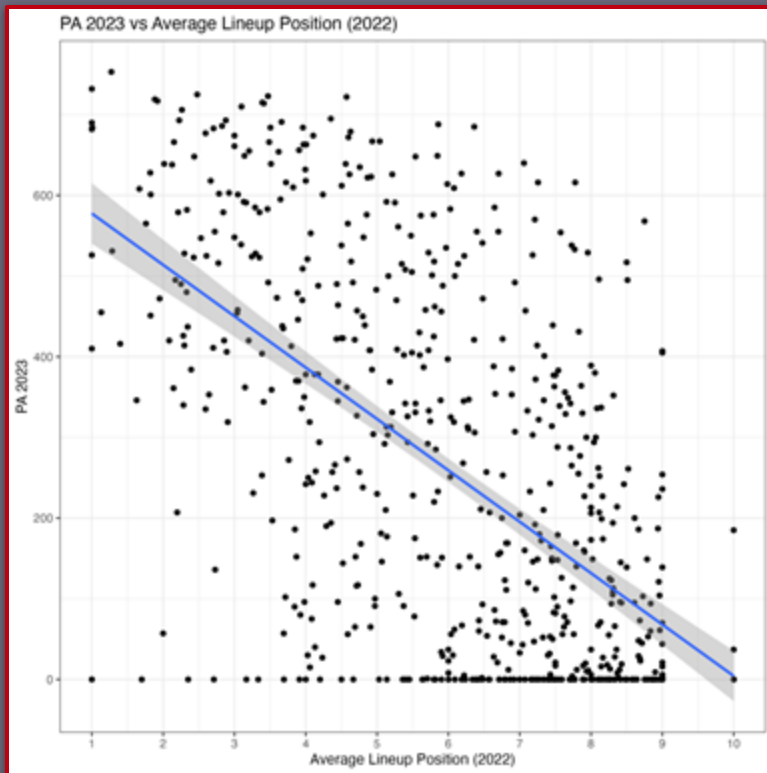
Lineup Position



Found the Average Batter Position in the Starting Lineup Across all 3 Years

Players who Didn't Start a Game in a Given Year Were Assigned 10 in the Lineup Position

Players Batting Earlier in the Lineup Tend to Receive More Playing Time





Batter Fielding Positions

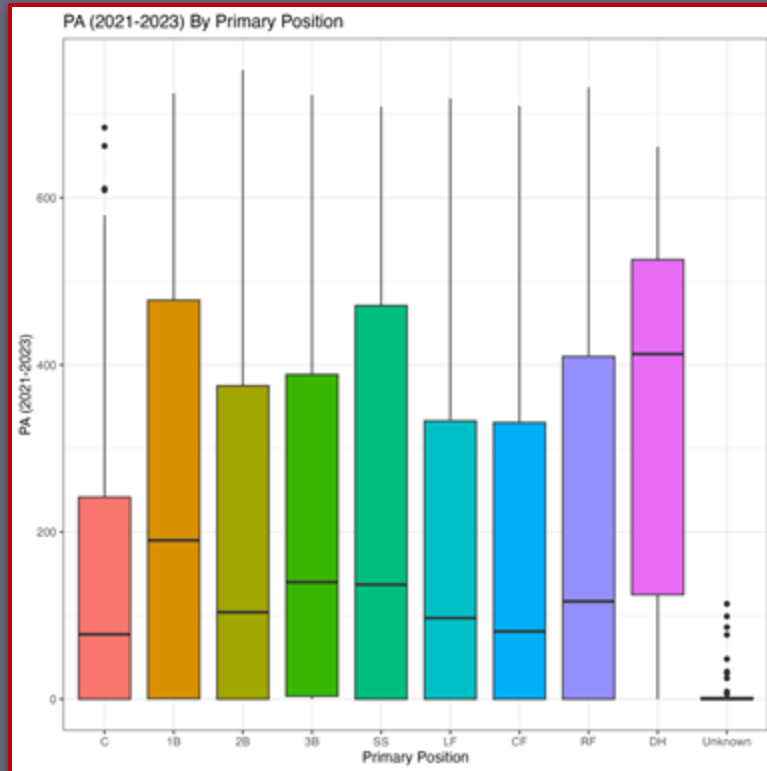


Identified Primary Position by Where They
Started the Most Over 2021-2023

Catchers Typically see Less Playing
Time

1B and DHs get a lot More Playing Time
Relative to Other Positions

“Unknown” Represents Players That Never
Started in the Field



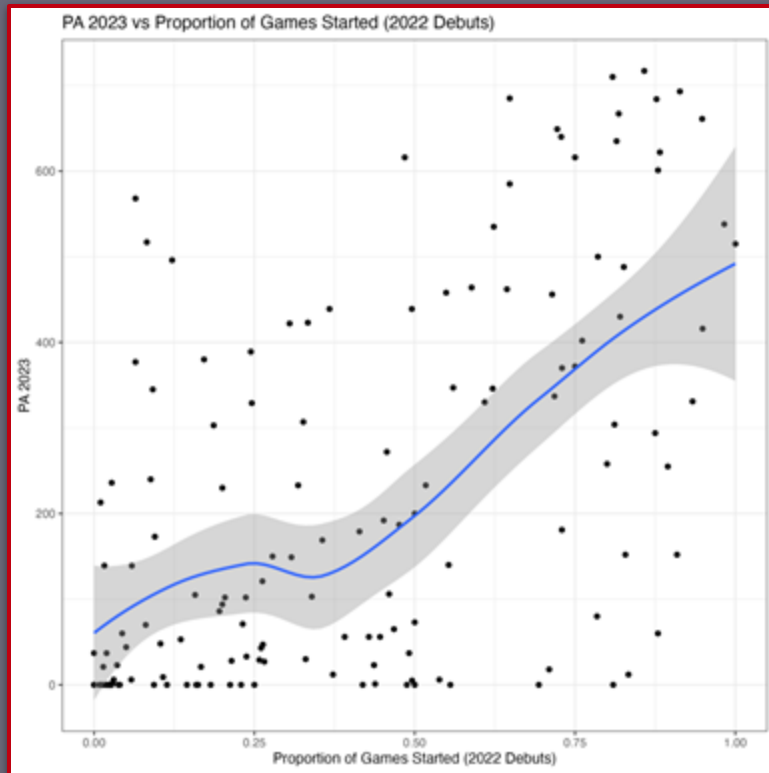


Proportion of Games Started (Debuts)



Identified Number of Games Remaining
on a Team's Schedule and Number of
Games a Player has Started Using Debut
Date

If a Rookie Starts Roughly 35% of
Games, Their Playing Time Should
Increase



Workload

- BF In Past 2 Seasons
- Pitches per Appearance
- BF per Appearance
- Avg. Times Through Order
- Pitches per PA
- BF in Second Half of Season
- BF Splits

Performance

- 90th Percentile Fastball Velocity
- Strikeouts and K%
- Other Count and Rate Statistics

SP: Number in Rotation on Opening Day
RP: Number of 9th Inning Appearances

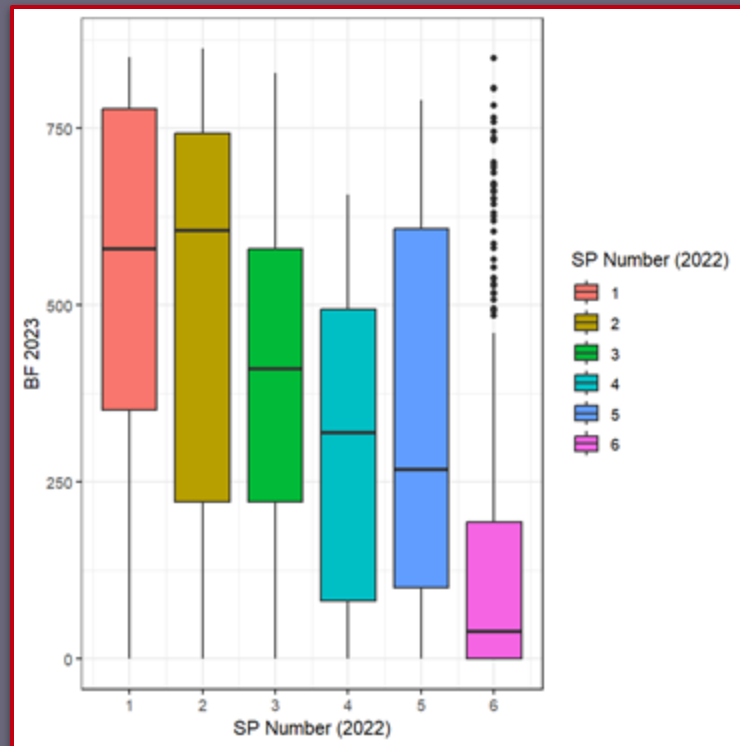


Number in Starting Rotation (Opening Day)



Opening Day Starters Face the Most Batters on Average Throughout the Season

Inverse Relationship Between Rotation Number and Workload



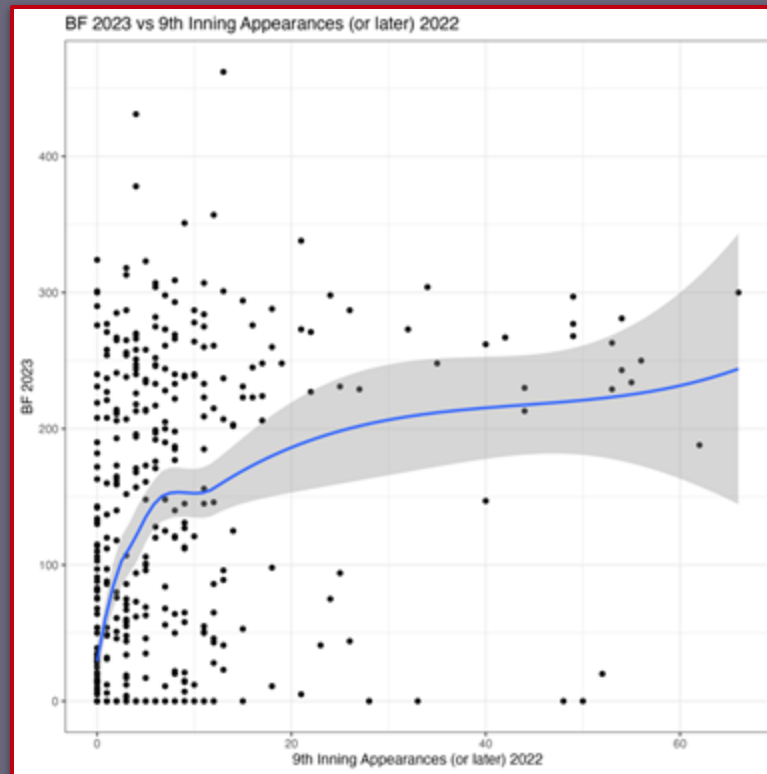


Number of 9th Inning Appearances



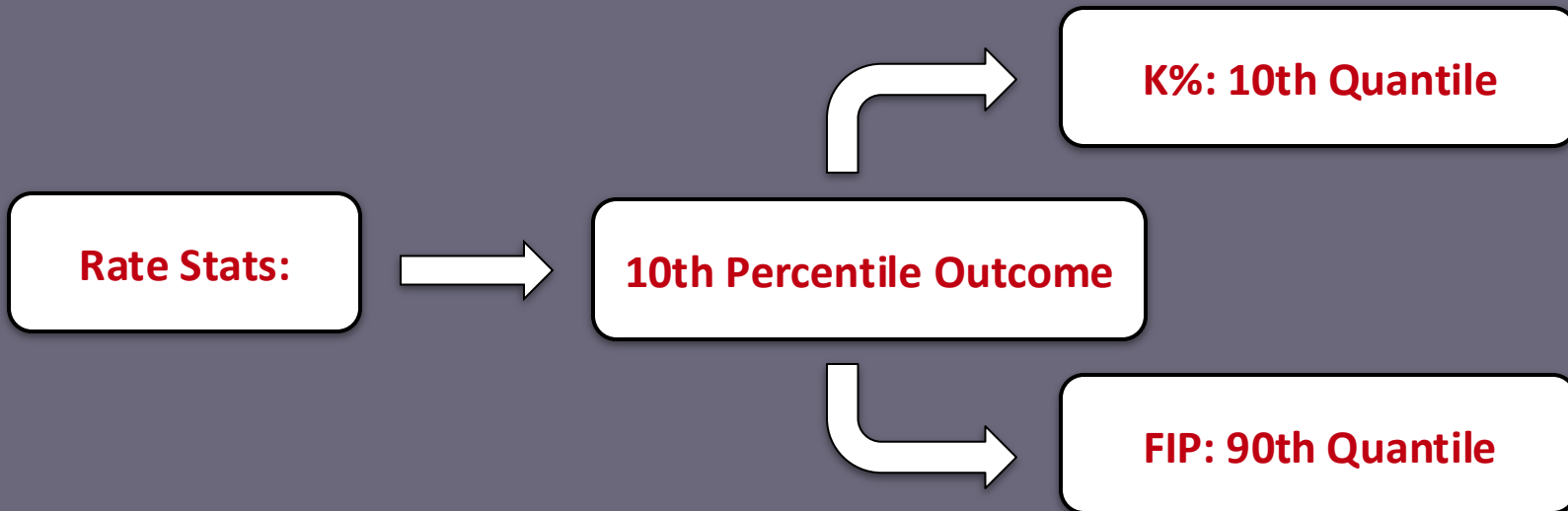
High Leverage Relievers are Going to Pitch in the 9th Often

Logarithmic Relationship Between 9th Inning Appearances and BF the Following Season





NA Imputation



Lack of PAs in a Season is Most Likely due to an Injury or Poor Ability

Clustering Analysis



Why This Approach?



Identify Playing Consistency Throughout the Season (Monthly Playing Time) and Evaluating its Impact on Future Playing Time

Insights to Whether a Player got Less PA/BF due to Being Benched or Injured

These Metrics are Only Valuable When Used Together, Rather Than Individually



K-Means Clustering



Batters and Relief Pitchers

**Monthly Playing
Time**

**Proportion of
Total/Season Playing
Time That Occurred
in Each Month**

**Whether the Player
was a Rookie**

**all for the prior 2 seasons*

Batters: 18 Clusters

RPs: 14 Clusters

Starting Pitches

Algorithm 1

**Monthly Playing Time
Throughout the Past 2
Seasons**

Algorithm 2

**Monthly Proportion of Total
Playing Time Throughout the
Past Season**

And Whether a Rookie in 2021 or 2022 (as separate variables)

Algorithm 1: 7 Clusters

Algorithm 2: 6 Clusters

Models



Finding the Right Fit



Model Construction

Removed players who debuted in 2023, didn't play in past or current season, or didn't exist in Lahman database

35 Fold CV

Limited Data, Model Was Efficient Despite Numerous Folds

BART

Applies Non-Parametric Approach to Bayesian Modeling

Less Prone to Overfitting, but Computationally Expensive and Limited Interpretability

XGBoost

Boosting Algorithm that Minimizes the Residuals

Efficient and High Performance, But Prone to Overfitting



Weighting Characteristics



Individualized Weights Were Created Based on Player Characteristics

Group 1

- 100+ PA/150+ BF 2 Seasons Ago
 - Less Than 30 PA/50 BF in Past Season
- (Weight 2)

Group 2

- 100+ PA/150+ BF in Past Season
 - Less Than 30 PA/50 BF 2 Seasons Ago
 - NOT a Rookie
- (Weight 3)

Group 3

Rookies Who Debuted Before May

Assigned 2022 Stats

Group 4

Rookies Who Debuted in May or June

Counting Stats Only

(Weight 4)

Group 5

Rookies Who Debuted in July or August

Counting Stats Only

(Weight 5)

Group 6

Rookies Who Debuted in September or October

Counting Stats Only

(Weight 6)

Group 7

All Other Players Who Don't Meet Groups 1-6

(Weight 1)

Results

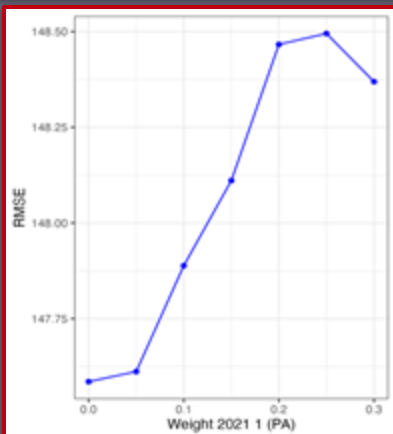


PA Weights



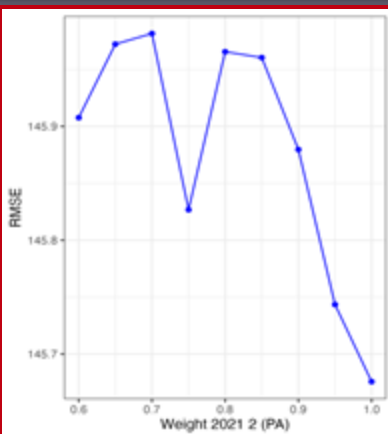
Weight 1

All Other Players Who Don't Meet Groups 1-6



Weight 2

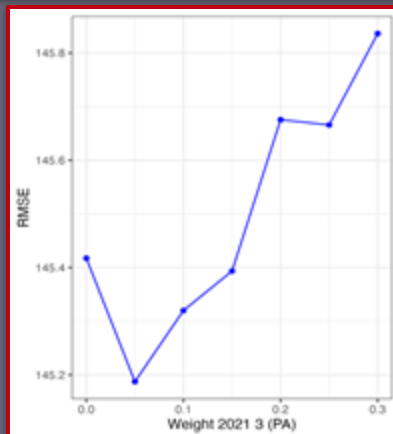
- 100+ PA/150+ BF 2 Seasons Ago
- Less Than 30 PA/50 BF in Past Season



Weight 4

Rookies Who Debuted in May or June

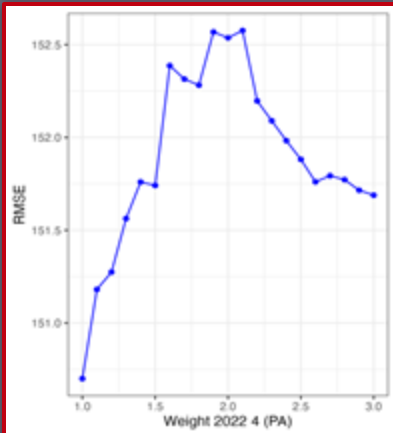
Counting Stats Only



Weight 5

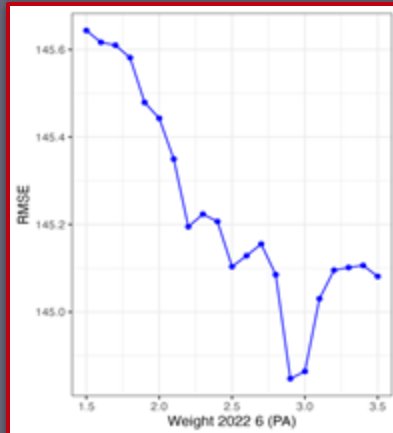
Rookies Who Debuted in July or August

Counting Stats Only



Weight 3

- 100+ PA/150+ BF in Past Season
- Less Than 30 PA/50 BF 2 Seasons Ago
- NOT a Rookie



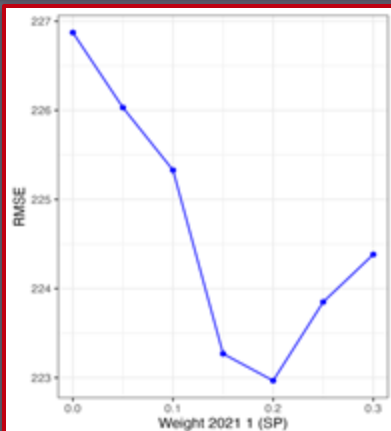


SP Weights



Weight 1

All Other Players Who Don't Meet Groups 1-6



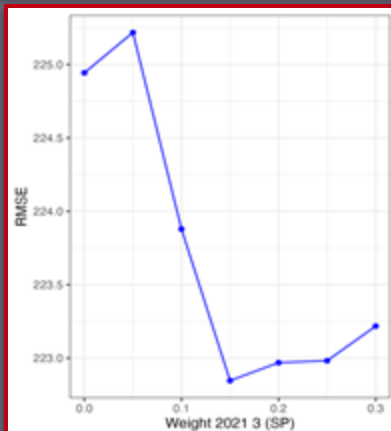
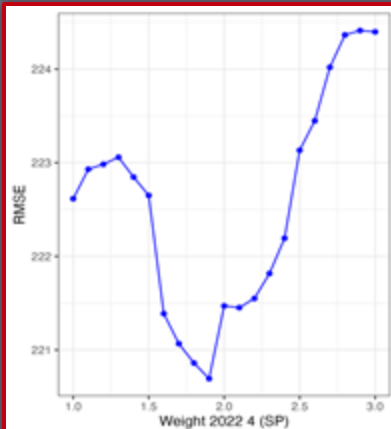
Weight 2

- 100+ PA/150+ BF 2 Seasons Ago
- Less Than 30 PA/50 BF in Past Season



Weight 3

- 100+ PA/150+ BF in Past Season
- Less Than 30 PA/50 BF 2 Seasons Ago
- NOT a Rookie



Weight 4

Rookies Who Debuted in May or June

Counting Stats Only

Weight 5

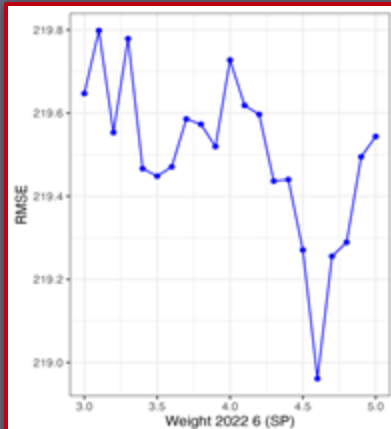
Rookies Who Debuted in July or August

Counting Stats Only

Weight 6

Rookies Who Debuted in September or October

Counting Stats Only





RP Weights



Weight 1

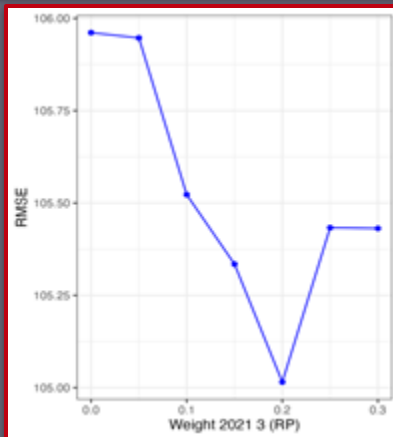
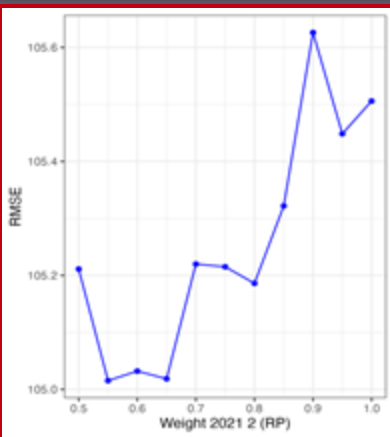
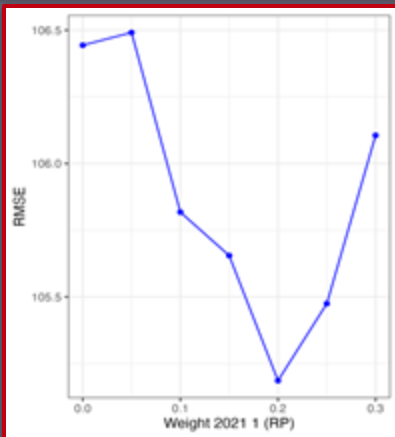
All Other Players Who Don't Meet Groups 1-6

Weight 2

- 100+ PA/150+ BF 2 Seasons Ago
- Less Than 30 PA/50 BF in Past Season

Weight 3

- 100+ PA/150+ BF in Past Season
- Less Than 30 PA/50 BF 2 Seasons Ago
- NOT a Rookie



Weight 4

Rookies Who Debuted in May or June

Counting Stats Only

Weight 5

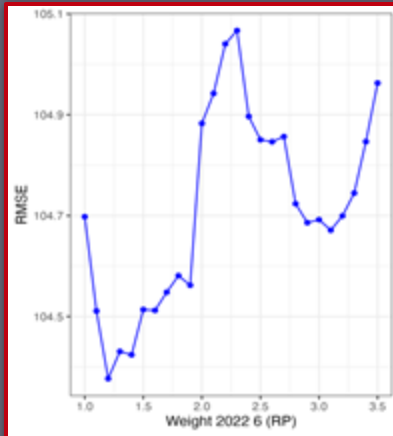
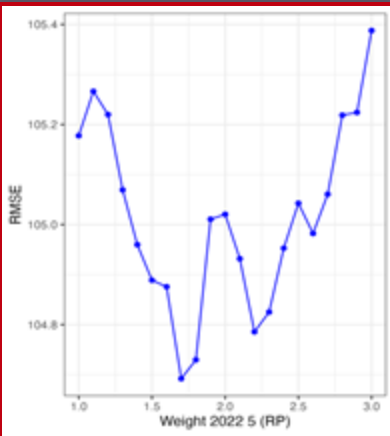
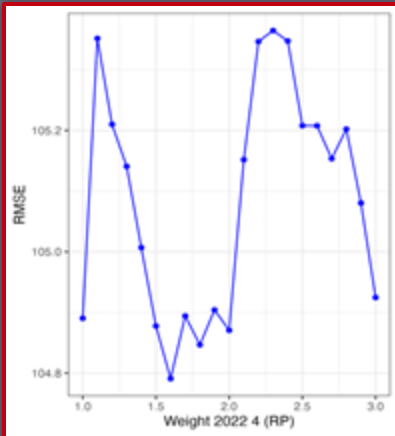
Rookies Who Debuted in July or August

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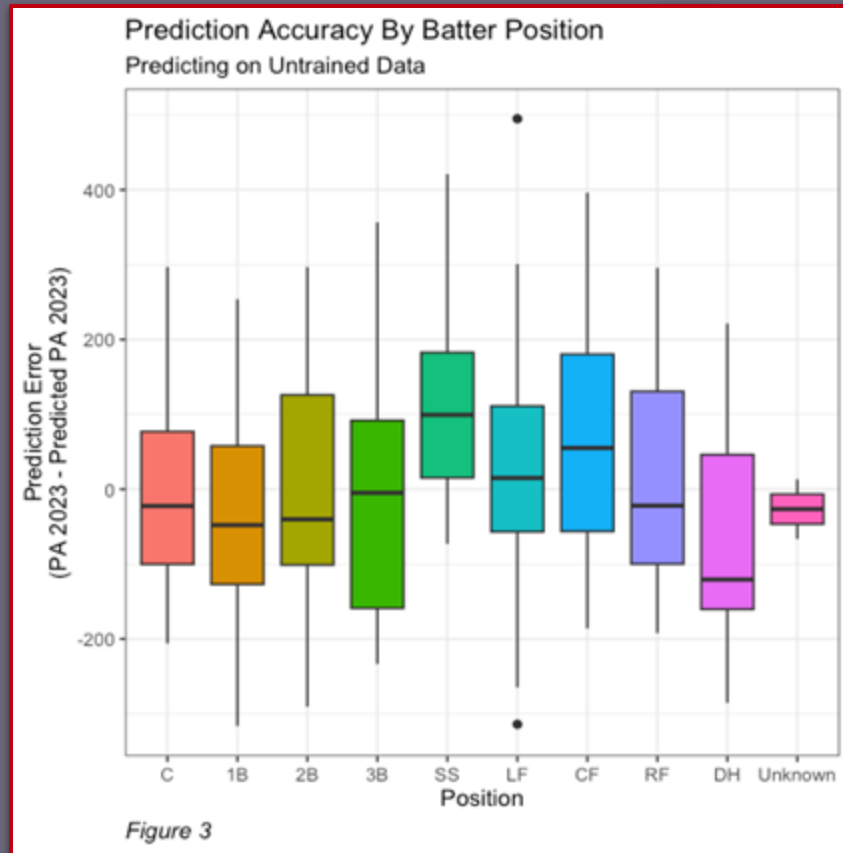
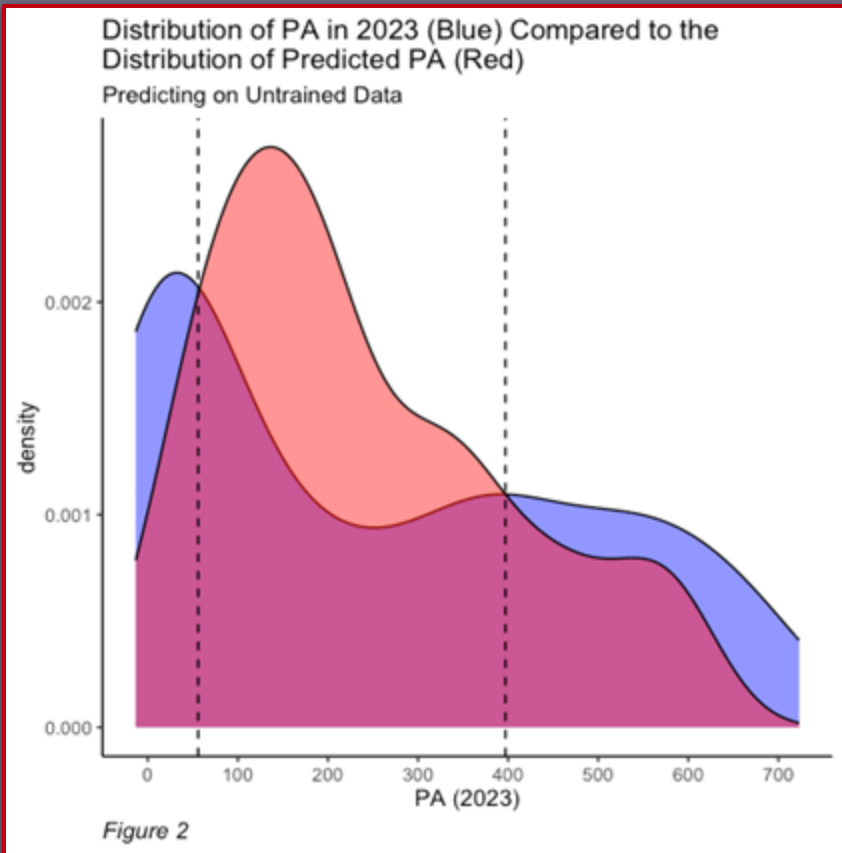
Rookies Who Debuted in September or October

Counting Stats Only

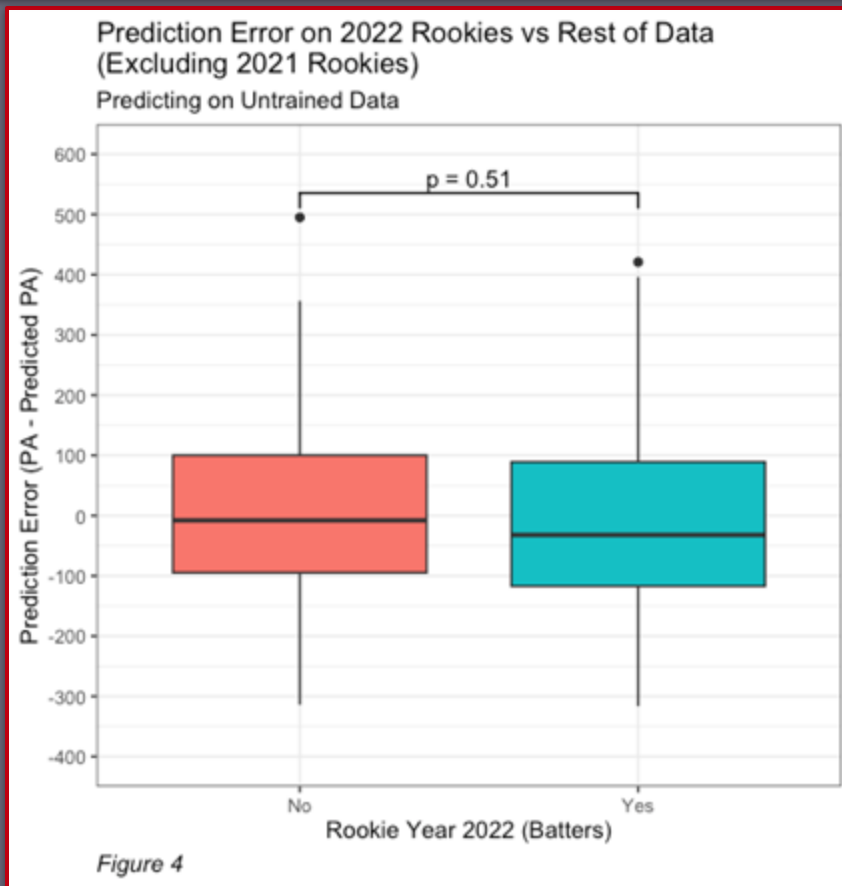




Batter Distribution → RMSE: 144.3



Batter Distribution





SP Distribution → RMSE: 220.4



Distribution of BF in 2023 (Blue) Compared to the Distribution of Predicted BF (Red) for SP

Predicting on Untrained Data

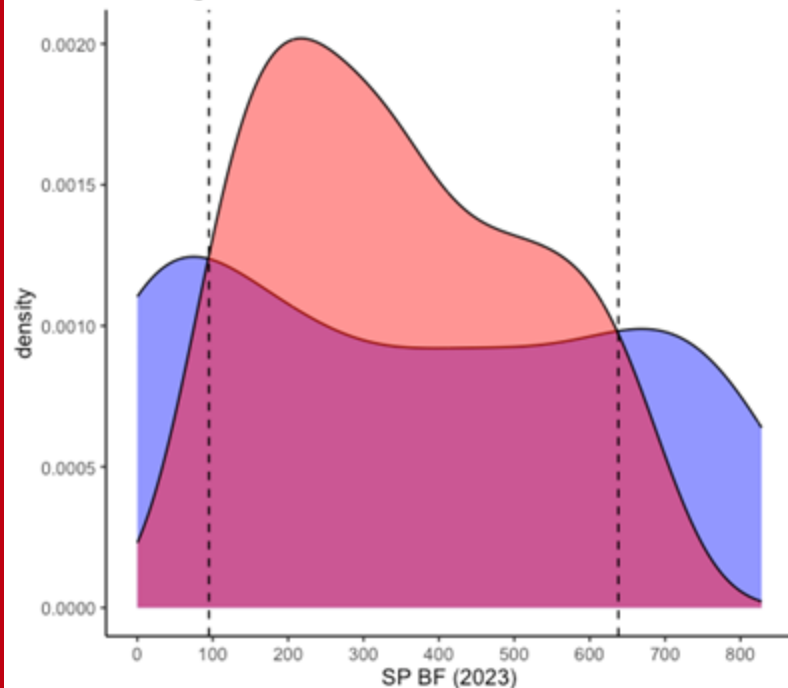


Figure 5

Prediction Error on 2022 Rookies vs Rest of Data (Excluding 2021 Rookies)

Predicting on Untrained Data

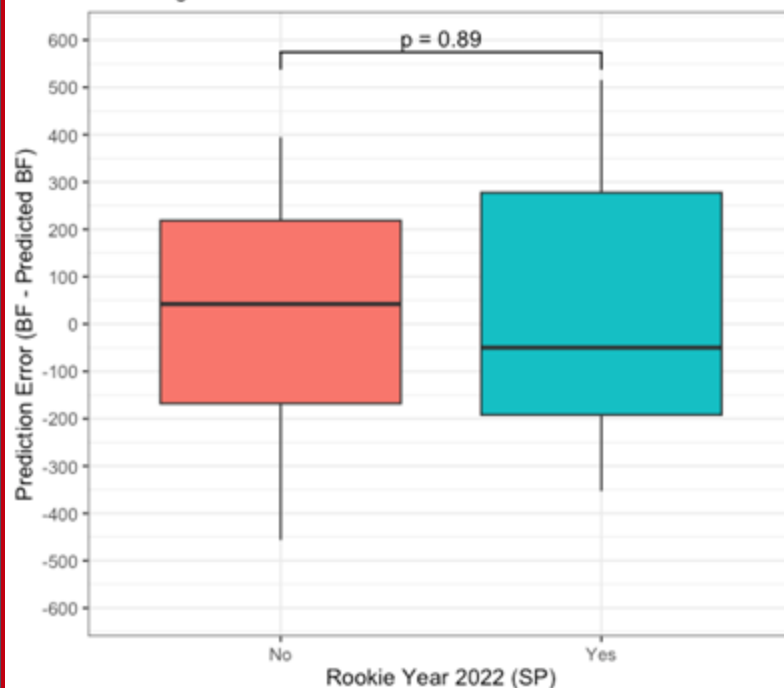


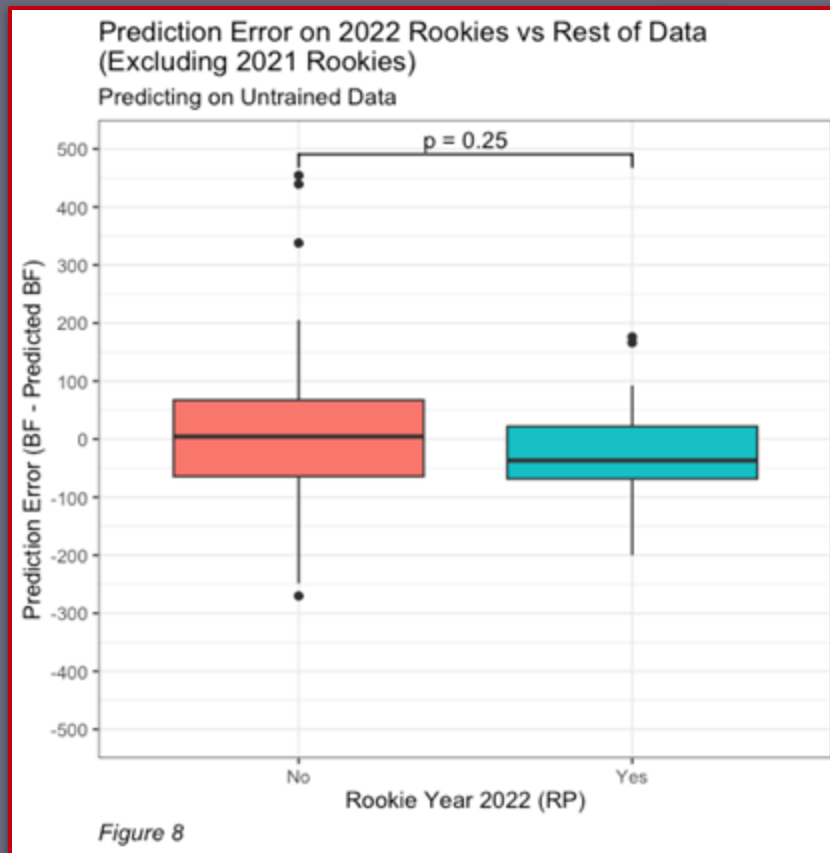
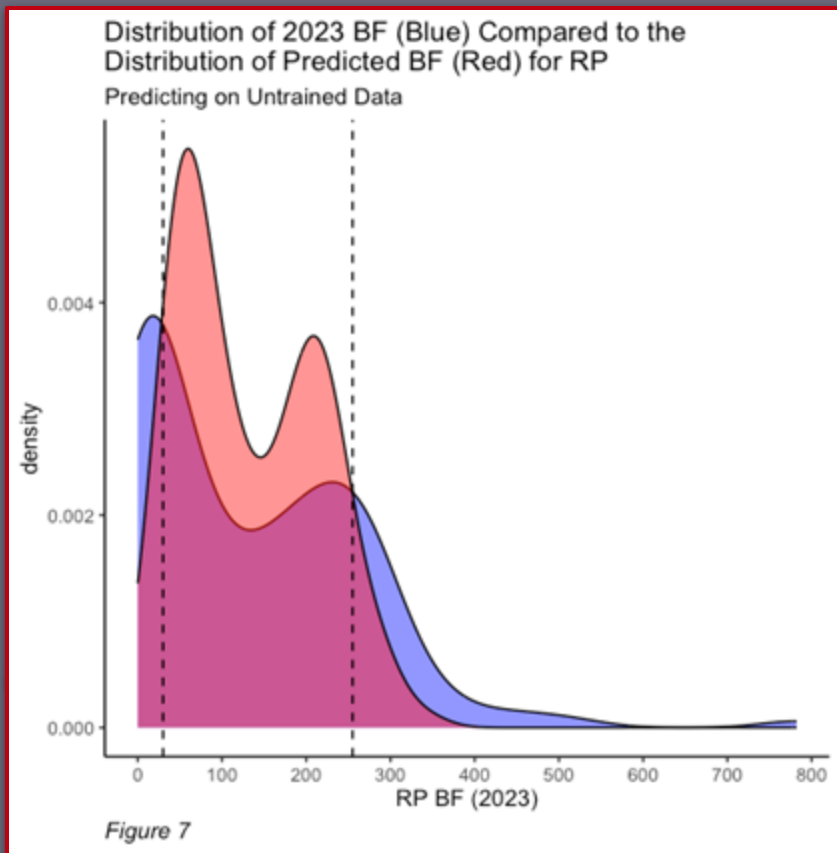
Figure 6

Batters Faced Outliers (2023)

Pitcher	2021 BF	2022 BF	2023 BF
Pitcher 1	586	574	0
Pitcher 2	779	608	7
Pitcher 3	538	646	0
Pitcher 4	774	780	19



RP Distribution → RMSE: 103.9



**RHETT LOWDER** (274)

RELIEF PITCHERS:

SCOTT BARLOW [189]

PREDICTED PLAYING TIME:



Preseason and
Offseason Injuries

Lack of Information on Contracts
and Service Time

Doesn't Account for
Team Quality

Don't Know Time Spent
on 26-Man Roster



Questions?