# The Impact of the 2022-23 NBA Rule Change\*

Exploring the Relationship Between the New Transitional Foul Rule, Player Shooting Performance, and MVP Points

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Rule changes in professional basketball play a critical role in shaping both strategies and player performances. Specifically, the introduction of the transitional foul rule in the 2022-23 NBA season, which changes the way fouls are called during transition plays, could significantly affect players' shooting efficiency and their prospects for MVP awards. This paper examines the implications of this rule change, comparing player statistics between the 2022-23 and 2021-22 seasons. We find significant correlations between shooting efficiency and MVP points, highlighting the rule's influence on player recognition. This paper offers insights into the interplay between rule modifications, player performance, and individual accolades, underscoring the broader significance of adapting to regulatory shifts in professional sports.

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<sup>\*</sup>Code and data are available at: https://github.com/AmieLiu/the-NBA-Rule-Change.git.

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### 1 Introduction

In professional basketball, rule changes can reshape the game's landscape, altering strategies, player performances, and ultimately, the trajectory of teams' successes. The National Basketball Association (NBA), as the premier basketball league globally, often implements rule modifications to refine gameplay, increase scoring opportunities, and foster a more open and free-flowing style of basketball (Sorensen 2019). Under the new transitional foul rule in the 2022-23 NBA season, take fouls committed during transition scoring opportunities will result in one free throw and continued possession for the offended team (Reynolds 2022), encouraging defenders to make legitimate plays.

This paper analyzes the consequences of the 2022-23 NBA rule change, particularly its impact on player shooting performance and the dynamics of the Most Valuable Player (MVP) competition, comparing statistics with the previous season, 2021-22. Our estimand is to explore the relationship between player shooting performance and MVP points they won in these NBA seasons. We hypothesize that players who excel in transition offense may experience an increase in shooting efficiency as they capitalize on more fast-break opportunities without facing as many transition-take fouls, potentially leading to greater recognition in MVP voting due to the positive correlation between shooting efficiency and MVP consideration.

A clear gap exists in understanding the nuanced impacts of the new transitional foul rule on player performance and MVP considerations, especially when compared to the previous season. This paper seeks to fill this gap by thoroughly analyzing the statistical data from both the 2022-23 and 2021-22 NBA seasons to highlight the changes, continuities, and emerging

trends over these times. To address this gap, we used the R programming language (R Core Team 2023) to analyze shooting performance metrics, such as field goal percentage (FG%), three-point percentage (3P%), free throw percentage (FT%), as well as MVP voting points won from both seasons. Our findings reveal patterns in player shooting performance, MVP points accumulation, and the intertwined effects of the transitional foul rule... We discovered significant correlations between shooting efficiency and MVP points, with distinct variations observed between the two seasons.

Understanding the implications of the transitional foul rule in comparison to the previous season is crucial for stakeholders across the NBA landscape, from players and coaches to researchers and fans. By illuminating the intricate relationship between rule changes, player performance, and individual accolades, this paper contributes to a deeper understanding of the evolving dynamics within the league.

This paper is structured into three main sections: Data, Results, and Discussion. Section 2 Data explores sources and methodology. It includes various sources used for data collection, cleaning, and analyzing processes, drawing from the dataset. In Section 4 Results, we present our findings, focusing on data visualizations and analyses. Section 5 Discussion provides a comprehensive analysis of our research, highlights the implications for NBA basketball, the importance of limitations and considerations in our analysis, and identifies areas for future research. Section 6 Conclusion summarizes our main findings and the broader significance of our research.

#### 2 Data

Some of our data is of penguins (see Table 1 and Table 2).

Table 1: 2022-2023 season most valuable player table

•	MVP Points	Field Goal	Three Points	Free Throw
Rank	Won	Percentage	Percentage	Percentage
1	915	0.548	0.330	0.857
2	674	0.632	0.383	0.822
3	606	0.553	0.275	0.645
4	280	0.466	0.350	0.854
5	46	0.510	0.345	0.905
6	30	0.484	0.386	0.867
7	27	0.615	0.373	0.742
8	10	0.496	0.342	0.742
9	5	0.493	0.427	0.915
10	3	0.539	0.350	0.850

Table 2: 2021-2022 season most valuable player table

Darah	MVP Points Won	Field Goal	Three Points	Free Throw
Rank	VVOII	Percentage	Percentage	Percentage
1	875	0.583	0.337	0.810
2	706	0.499	0.371	0.814
3	595	0.553	0.293	0.722
4	216	0.466	0.383	0.868
5	146	0.457	0.353	0.744
6	43	0.453	0.353	0.853
7	10	0.493	0.344	0.761
8	4	0.437	0.380	0.923
9	2	0.493	0.317	0.837
10	1	0.504	0.352	0.877

Talk more about it.

And also planes (?@fig-planes). (You can change the height and width, but don't worry about doing that until you have finished every other aspect of the paper - Quarto will try to make it look nice and the defaults usually work well once you have enough text.)

Talk way more about it.

#### 2.1 Data Visualization

# 2.1.1 The distribution of MVP points won by top 10 players in the 2021-22 and 2022-23 seasons

# 2.1.2 Different shooting performances by top 10 players in the 2021-22 and 2022-23 seasons

#### 3 Model

To explore the impact of the 2022-23 NBA transitional foul rule on player shooting performance and its subsequent effect on MVP voting, we constructed a linear regression model. The objective is to quantify the relationship between shooting performance metrics and MVP voting outcomes.

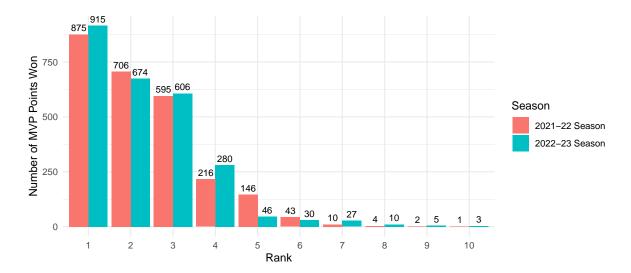


Figure 1: Number of MVP points won by each player from the top 10 rankings in the 2021-22 and 2022-23 seasons

## 3.1 Model set-up

The dependent variable is MVP Voting Points and the independent variables are Field Goal Percentage (FG%), Three-Point Field Goal Percentage (3P%), and Free Throw Percentage (FT%). Each independent variable was chosen based on its potential influence on a player's performance under the conditions created by the new foul rule. We run the model in R (R Core Team 2023) utilizing the rstanarm package (Brilleman et al. 2018). Our model is constructed as follows:

$$Y=\beta_0+\beta_1X_1+\beta_2X_2+\beta_3X_3+\epsilon$$

where:

- Y is the MVP Voting Points. It is the dependent variable that represents the points a player receives in the MVP voting process.
- $\beta_0$  is the intercept of the regression model that represents the expected value of Total MVP Voting Points when all predictor variables (FG%, 3P%, FT%) are at zero.
- $\beta_1$  is the coefficient for FG%. It indicates how changes in a player's overall shooting efficiency from the field affect their MVP voting points.
- $\beta_2$  is the the coefficient for 3P%. This reflects the impact of three-point shooting efficiency on MVP voting.
- $\beta_3$  is the coefficient for FT%. This measures the influence of free throw shooting efficiency on MVP voting points.
- $X_1$  measures the FG% made out of attempts.

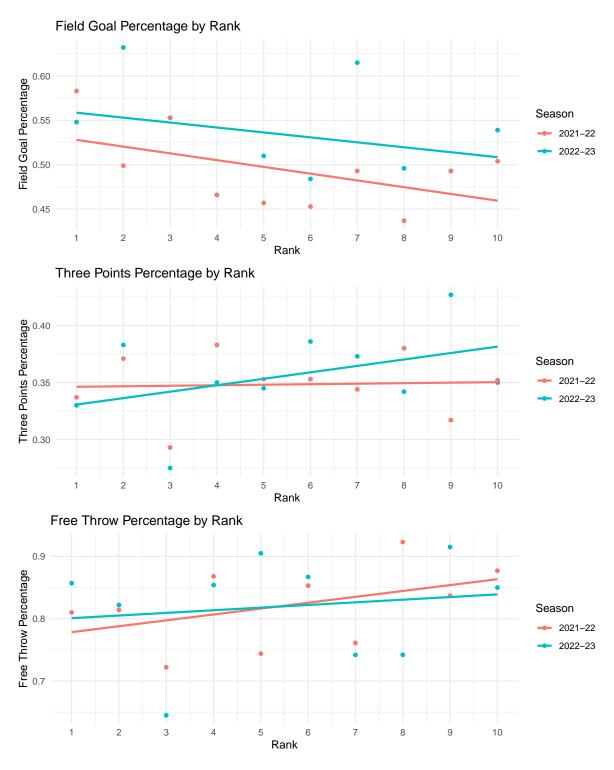


Figure 2: Different shooting percentage by each player from the top 10 rankings in the 2021-22 and 2022-23 seasons

- $X_2$  measures the 3P% made out of attempts.
- $X_3$  measures the FT% made out of attempts.
- $\epsilon$  is the error term, representing the residual element not explained by the model. This captures the variability in MVP voting points not explained by shooting percentages alone.

#### 3.1.1 Model justification

The selection of linear regression for our analysis suggests linear relationships between the shooting metrics and MVP voting points. We hypothesized that improvements in these shooting metrics, reflecting enhanced performance in transition plays free from take fouls, would correlate positively with higher MVP voting scores. The use of FG%, 3P%, and FT% as predictors is justified by their direct impact on game performance, which is critical in MVP considerations. Moreover, these metrics are commonly used in basketball analytics to assess player efficiency and effectiveness, making them relevant for evaluating the consequences of rule changes.

Statistical significance of the model coefficients will be tested using t-tests, where a p-value less than 0.05 will indicate significant influence of the predictor on MVP voting points. The overall fit of the model will be assessed by the R-squared value, which explains the proportion of variance in MVP points accounted for by the predictors.

#### 4 Results

Our results are summarized in ?@tbl-modelresults. (see Figure 1 and Figure 3)

Table 3: Summary of regression model coefficients

Term	Estimate	Std. Error	Statistic	P-value
	Listinate	Did. Ellor	Diadistic	
(Intercept)	-920.9892	1833.862	-0.5022128	0.6334126
Field Goal Percentage	3432.8779	2139.048	1.6048624	0.1596463
Three Points Percentage	-6037.2306	3553.444	-1.6989801	0.1402370
Free Throw Percentage	1827.8611	1828.513	0.9996432	0.3560769

Table 4: Summary of additional regression model metrics

Metric	Value
Residual Standard Error	314.3
Multiple R-squared	0.4446

Table 4: Summary of additional regression model metrics

Metric	Value
Adjusted R-squared	0.1669
F-statistic p-value	1.601 $0.2852$

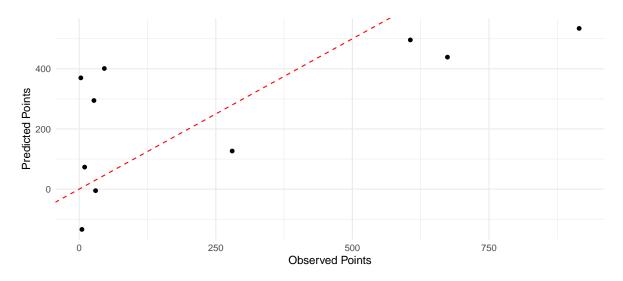


Figure 3: Observed vs. predicted values of MVP points won

## 5 Discussion

## 5.1 First discussion point

If my paper were 10 pages, then should be be at least 2.5 pages. The discussion is a chance to show off what you know and what you learnt from all this.

## 5.2 Second discussion point

#### 5.3 Third discussion point

## 5.4 Weaknesses and next steps

Weaknesses and next steps should also be included.

## 6 Conclusion

## **Appendix**

## A Additional data details

## **B** Model details

## **B.1** Posterior predictive check

In **?@fig-ppcheckandposteriorvsprior-1** we implement a posterior predictive check. This shows...

In ?@fig-ppcheckandposteriorvsprior-2 we compare the posterior with the prior. This shows...

## References

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