

# My title\*

My subtitle if needed

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First sentence. Second sentence. Third sentence. Fourth sentence.

## 1 Introduction

You can and should cross-reference sections and sub-sections. We use R Core Team (2023) and (rohan?).

The remainder of this paper is structured as follows. Section 2....

## 2 Data

The dataset is obtained from Basketball Reference Sports Reference LLC (2024), which contains NBA players' statistics in the 2023 to 2024 season. The dataset is directly downloaded from Basketball Reference using the given instruction. The dataset contains different statistics of NBA players, for instance players' position, age, assists, steals, etc. There are in total of 718 observations before any data cleanings.

Since this dataset contains 3-point field goals, 2-point field goals, free throws and field goals, points can be calculated solely using these variables. My goal is to predict a the numebr of scores a NBA player scores based on his performance, which means that these features are too closely related to the target variable (i.e. points). Thus, those variables are removed from the dataset using 'tidyverse' package in R Wickham et al. (2019). Moreover, there are 12 different positions in the dataset, I categorized them into a broader classification which has only three categories - Guards (G): SG, PG, SG-PG, PG-SG, Forwards (F): SF, PF, PF-SF, SF-PF, SF-SG, Centers (C): C, PF-C, C-PF. This is done so that I don't need to create overwhelming dummy variables when generating the model and this process of binning could simplify the model and potentially strengthen the signal by reducing noise and multicollinearity.

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\*Code and data are available at: <https://github.com/peachvegetable/NBA-player-points>

## 3 Model

Lasso regression is a variation of linear regressions that can actually perform feature selection by setting the coefficients of less important features to zero. It has an additional parameter  $\lambda$ , which is the regularization parameter that controls the strength of the penalty. The penalty term shrinks some of the coefficients (i.e. the less important variables) toward zero. As  $\lambda$  increases, more coefficients are set to zero, leading to a simpler model. The value of  $\lambda$  is determined through cross-validation.

### 3.1 Model set-up

$$y_i = \beta_0 + \beta_i \cdot X_i \quad (1)$$

In this equation,  $y_i$  is the number of points a player scores, which is the dependent variable I am trying to predict.  $\beta_0$  is the interception, and  $\beta_i$  is a matrix that contains the coefficients  $\beta_1, \beta_2, \dots, \beta_{18}$  for each predictor that the lasso regression will estimate.  $X_i$  is also a matrix contains the predictors: players position, age, games, game starts, minutes played, field goal percentage, 3-point field goal percentage, 2-point field goal percentage, effective field goal percentage, free throw percentage, offensive rebounds, defensive rebounds, total rebounds, assists, steals, blocks, turnovers, and personal fouls.

We run the model in R (R Core Team 2023) using the `tidymodels` package of Kuhn and Wickham (2020).

#### 3.1.1 Model justification

We expect a positive relationship between the size of the wings and time spent aloft. In particular...

We can use maths by including latex between dollar signs, for instance  $\theta$ .

```
# #| include: false
# #| warning: false
# #| message: false
#
# first_lasso_model <- readRDS("../models/first_lasso_model.rds")
# second_lasso_model <- readRDS("../models/second_lasso_model.rds")
# predictions <- predict(second_lasso_model, test_data)
# results <- bind_cols(test_data, predictions)
# # Calculate RMSE
# rmse_results_2 <- rmse(results, truth = pts, estimate = .pred)
```

```
# rsq_results_2 <- rsq(results, truth = pts, estimate = .pred)
```

## 4 Results

Our results are summarized in `?@tbl-modelresults`.

## 5 Discussion

### 5.1 First discussion point

If my paper were 10 pages, then should 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.

## 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...

#### B.2 Diagnostics

`?@fig-stanareyouokay-1` is a trace plot. It shows... This suggests...

`?@fig-stanareyouokay-2` is a Rhat plot. It shows... This suggests...

## References

- Kuhn, Max, and Hadley Wickham. 2020. *Tidymodels: A Collection of Packages for Modeling and Machine Learning Using Tidyverse Principles*. <https://www.tidymodels.org>.
- R Core Team. 2023. *R: A Language and Environment for Statistical Computing*. Vienna, Austria: R Foundation for Statistical Computing. <https://www.R-project.org/>.
- Sports Reference LLC. 2024. “2023-2024 NBA Player Stats: Totals.” Basketball-Reference.com. [https://www.basketball-reference.com/leagues/NBA\\_2024\\_totals.html](https://www.basketball-reference.com/leagues/NBA_2024_totals.html).
- Wickham, Hadley, Mara Averick, Jennifer Bryan, Winston Chang, Lucy D’Agostino McGowan, Romain François, Garrett Golemund, et al. 2019. “Welcome to the tidyverse.” *Journal of Open Source Software* 4 (43): 1686. <https://doi.org/10.21105/joss.01686>.