678 Final Project

Paul Moon

Abstract

In the movie Moneyball, 'Brad Pitt' uses pure statistics to select which MLB players should play for the Athletics, disregarding what other teams and analysts say about their skill based on other parameters, such as "looks or personality." He used on-base percentage (OBP) and slugging percentage with the main goal of "getting on base" because "getting on base" meant "runs" and the more "runs" you score, the higher the chances of winning was.

In my final project, my goal is to do something similar but with the NBA. The NBA mainly consists of 3 statistics: points, rebounds, and assists (PRAs). First, I will find the average PRAs a playoff NBA team records throughout an entire 82-game season and compare it to an NBA team that does not make the playoffs. That will be the basis of our project. Find players who will get you the amount of PRAs you need to reach the playoffs.

In addition to PRAs, I will also be looking into Field Goals Made (FGM), Field Goals Attempted (FGA), and 3-Point Statistics. The reason for this comes from Wayne 'The Great One' Gretzky's quote "You miss 100% of the shots you don't take." I believe that the more shots a player attempts, the more likely they will score more points, resulting in more wins.

Finally, I will be looking into Games Started (GS) and Minutes Played (MIN) due to the fact that you want to draft/trade for players who know how to stay healthy and are less prone to injury.

The goal is to look at statistics of the past 3 years: the 2021-22, 2022-23, and 2023-24 NBA Seasons. This is due to the fact that the 2019-20 NBA Season was cut short due to Covid-19 and the 2020-21 NBA Season only had 72 instead of the regular 82 games due to the same reason. Most importantly, the new playoff format was updated in the 2020-21 NBA Season with the introduction of the play-in tournament.

Introduction

With the introduction of the play-in tournament, after the conclusion of the 82 game season, 6 teams per conference (total of 12 teams) will automatically make the playoffs. In addition

to that, 4 more teams per conference (total of 8 teams) will have the opportunity to still make it into the playoffs following this format:



1*

Data of both play-in tournament requirements and playoff requirements will be analyzed in this section; however, in this project, I will calculate the playoff scenario where play-in tournament teams have a 50% chance in making the playoffs. This makes sense since every year there are 8 play-in tournament teams and only 4 teams make it into the playoffs.

2021-22

Eastern Conference

| | w | L | PCT | <u>GB</u> ^ | HOME | AWAY | DIV | CONF | PPG | OPP PPG | DIFF | STRK | L10 |
|--------------------------|----|----|------|-------------|-------|-------|------|-------|-------|---------|------|------|------|
| z 💣 Miami Heat | 53 | 29 | .646 | - | 29-12 | 24-17 | 13-3 | 35-17 | 110.0 | 105.6 | +4.4 | L1 | 6-4 |
| y - Boston Celtics | 51 | 31 | .622 | 2 | 28-13 | 23-18 | 9-7 | 33-19 | 111.8 | 104.5 | +7.3 | W1 | 7-3 |
| y 😭 Milwaukee Bucks | 51 | 31 | .622 | 2 | 27-14 | 24-17 | 12-4 | 33-19 | 115.5 | 112.1 | +3.4 | L1 | 6-4 |
| x - 6 Philadelphia 76ers | 51 | 31 | .622 | 2 | 24-17 | 27-14 | 6-10 | 32-20 | 109.9 | 107.3 | +2.6 | W2 | 6-4 |
| x Toronto Raptors | 48 | 34 | .585 | 5 | 24-17 | 24-17 | 10-6 | 30-22 | 109.4 | 107.1 | +2.3 | L1 | 8-2 |
| x Chicago Bulls | 46 | 36 | .561 | 7 | 27-14 | 19-22 | 10-6 | 29-23 | 111.6 | 112.0 | -0.4 | W1 | 4-6 |
| xp B Brooklyn Nets | 44 | 38 | .537 | 9 | 20-21 | 24-17 | 10-6 | 31-21 | 112.9 | 112.1 | +0.8 | W4 | 6-4 |
| xp 🤰 Atlanta Hawks | 43 | 39 | .524 | 10 | 27-14 | 16-25 | 9-7 | 26-26 | 113.9 | 112.4 | +1.5 | W1 | 7-3 |
| pb C Cleveland Cavaliers | 44 | 38 | .537 | 9 | 25-16 | 19-22 | 10-6 | 27-25 | 107.8 | 105.7 | +2.1 | W1 | 3-7 |
| pb W Charlotte Hornets | 43 | 39 | .524 | 10 | 22-19 | 21-20 | 8-8 | 27-25 | 115.3 | 114.9 | +0.4 | W3 | 6-4 |
| e 🌄 New York Knicks | 37 | 45 | .451 | 16 | 17-24 | 20-21 | 5-11 | 22-30 | 106.5 | 106.6 | -0.1 | W2 | 7-3 |
| e 🕷 Washington Wizards | 35 | 47 | .427 | 18 | 21-20 | 14-27 | 7-9 | 24-28 | 108.6 | 112.0 | -3.4 | L3 | 5-5 |
| e 🔑 Indiana Pacers | 25 | 57 | .305 | 28 | 16-25 | 9-32 | 2-14 | 11-41 | 111.5 | 114.9 | -3.4 | L10 | 0-10 |
| e Detroit Pistons | 23 | 59 | .280 | 30 | 13-28 | 10-31 | 6-10 | 18-34 | 104.8 | 112.5 | -7.7 | L3 | 4-6 |
| e 🖜 Orlando Magic | 22 | 60 | .268 | 31 | 12-29 | 10-31 | 3-13 | 12-40 | 104.2 | 112.2 | -8.0 | W1 | 3-7 |

Western Conference

| | w | L | PCT | <u>GB</u> ^ | HOME | AWAY | DIV | CONF | PPG | OPP PPG | DIFF | STRK | L10 |
|----------------------------|----|----|------|-------------|-------|-------|------|-------|-------|---------|------|-------------|------|
| * 🥟 Phoenix Suns | 64 | 18 | .780 | - | 32-9 | 32-9 | 10-6 | 39-13 | 114.8 | 107.4 | +7.4 | L1 | 6-4 |
| y W Memphis Grizzlies | 56 | 26 | .683 | 8 | 30-11 | 26-15 | 11-5 | 36-16 | 115.6 | 109.9 | +5.7 | L1 | 7-3 |
| x Golden State Warriors | 53 | 29 | .646 | 11 | 31-10 | 22-19 | 12-4 | 33-19 | 111.0 | 105.5 | +5.5 | W5 | 6-4 |
| x Dallas Mavericks | 52 | 30 | .634 | 12 | 29-12 | 23-18 | 14-2 | 36-16 | 108.0 | 104.7 | +3.3 | W4 | 8-2 |
| y Utah Jazz | 49 | 33 | .598 | 15 | 29-12 | 20-21 | 15-1 | 33-19 | 113.6 | 107.6 | +6.0 | W1 | 4-6 |
| x Denver Nuggets | 48 | 34 | .585 | 16 | 23-18 | 25-16 | 6-10 | 29-23 | 112.7 | 110.4 | +2.3 | L1 | 6-4 |
| xp Minnesota Timberwolves | 46 | 36 | .561 | 18 | 26-15 | 20-21 | 12-4 | 32-20 | 116.0 | 113.3 | +2.7 | L1 | 4-6 |
| xp > New Orleans Pelicans | 36 | 46 | .439 | 28 | 19-22 | 17-24 | 6-10 | 25-27 | 109.3 | 110.3 | -1.0 | L2 | 6-4 |
| pb Clippers | 42 | 40 | .512 | 22 | 25-16 | 17-24 | 9-7 | 26-26 | 108.4 | 108.4 | 0.0 | W5 | 6-4 |
| pb 🥙 San Antonio Spurs | 34 | 48 | .415 | 30 | 16-25 | 18-23 | 6-10 | 24-28 | 113.2 | 113.0 | +0.2 | L3 | 6-4 |
| e 🐠 Los Angeles Lakers | 33 | 49 | .402 | 31 | 21-20 | 12-29 | 3-13 | 18-34 | 112.1 | 115.1 | -3.0 | W2 | 2-8 |
| e 🧱 Sacramento Kings | 30 | 52 | .366 | 34 | 16-25 | 14-27 | 6-10 | 20-32 | 110.3 | 115.8 | -5.5 | W1 | 5-5 |
| e Portland Trail Blazers | 27 | 55 | .329 | 37 | 17-24 | 10-31 | 1-15 | 11-41 | 106.2 | 115.1 | -8.9 | L11 | 0-10 |
| e 🐙 Oklahoma City Thunder | 24 | 58 | .293 | 40 | 12-29 | 12-29 | 6-10 | 17-35 | 103.7 | 111.8 | -8.1 | L3 | 4-6 |
| e Houston Rockets | 20 | 62 | .244 | 44 | 11-30 | 9-32 | 3-13 | 11-41 | 109.7 | 118.2 | -8.5 | L7 | 3-7 |

2*

In the 2021-2022 NBA Season, if you were in the Eastern Conference, a team needed 38 wins to qualify for the play-in tournament. To instantly qualify for the playoffs, a team needed 45 wins. However, in the Western Conference, a team needed only 34 wins to qualify for the play-in tournament; however, a team needed 47 wins to instantly qualify for the playoffs.

After the conclusion of the play-in tournament, playoff qualifying teams passed the average threshold of 44 wins in the Eastern Conference and 39 wins in the Western Conference (The average threshold was found by averaging the lowest playoff team to the highest non-playoff team).

2022-23
Eastern Conference

| | <u>w</u> | L | PCT | <u>GB</u> ^ | HOME | AWAY | DIV | CONF | PPG | OPP PPG | DIFF | STRK | L10 |
|--------------------------|----------|----|------|-------------|-------|-------|------|-------|-------|---------|------|-------------|-----|
| * 🦞 Milwaukee Bucks | 58 | 24 | .707 | - | 32-9 | 26-15 | 11-5 | 35-17 | 116.9 | 113.3 | +3.6 | L2 | 6-4 |
| y - Boston Celtics | 57 | 25 | .695 | 1 | 32-9 | 25-16 | 11-5 | 34-18 | 117.9 | 111.4 | +6.5 | W3 | 8-2 |
| x - 6 Philadelphia 76ers | 54 | 28 | .659 | 4 | 29-12 | 25-16 | 10-6 | 34-18 | 115.2 | 110.9 | +4.3 | W2 | 5-5 |
| x C Cleveland Cavaliers | 51 | 31 | .622 | 7 | 31-10 | 20-21 | 13-3 | 34-18 | 112.3 | 106.9 | +5.4 | L1 | 7-3 |
| x 🌄 New York Knicks | 47 | 35 | .573 | 11 | 23-18 | 24-17 | 8-8 | 32-20 | 116.0 | 113.1 | +2.9 | L2 | 5-5 |
| x B Brooklyn Nets | 45 | 37 | .549 | 13 | 23-18 | 22-19 | 7-9 | 30-22 | 113.4 | 112.5 | +0.9 | L1 | 6-4 |
| xp 🔵 Atlanta Hawks | 41 | 41 | .500 | 17 | 24-17 | 17-24 | 8-8 | 26-26 | 118.4 | 118.1 | +0.3 | L2 | 5-5 |
| y 🧨 Miami Heat | 44 | 38 | .537 | 14 | 27-14 | 17-24 | 10-6 | 24-28 | 109.5 | 109.8 | -0.3 | W1 | 6-4 |
| pb 😽 Chicago Bulls | 40 | 42 | .488 | 18 | 22-19 | 18-23 | 7-9 | 27-25 | 113.1 | 111.8 | +1.3 | W2 | 6-4 |
| pb 🥏 Toronto Raptors | 41 | 41 | .500 | 17 | 27-14 | 14-27 | 4-12 | 26-26 | 112.9 | 111.4 | +1.5 | W1 | 6-4 |
| e 🔑 Indiana Pacers | 35 | 47 | .427 | 23 | 20-21 | 15-26 | 7-9 | 24-28 | 116.3 | 119.5 | -3.2 | W1 | 3-7 |
| e 🖍 Washington Wizards | 35 | 47 | .427 | 23 | 19-22 | 16-25 | 8-8 | 21-31 | 113.2 | 114.4 | -1.2 | L1 | 3-7 |
| e 🖜 Orlando Magic | 34 | 48 | .415 | 24 | 20-21 | 14-27 | 7-9 | 20-32 | 111.4 | 114.0 | -2.6 | L4 | 5-5 |
| e 💝 Charlotte Hornets | 27 | 55 | .329 | 31 | 13-28 | 14-27 | 7-9 | 15-37 | 111.0 | 117.2 | -6.2 | W1 | 5-5 |
| e Detroit Pistons | 17 | 65 | .207 | 41 | 9-32 | 8-33 | 2-14 | 8-44 | 110.3 | 118.5 | -8.2 | L1 | 1-9 |

Western Conference

| | <u>w</u> | L | PCT | <u>GB</u> ^ | HOME | AWAY | DIV | CONF | PPG | OPP PPG | DIFF | STRK | L10 |
|----------------------------|----------|----|------|-------------|-------|-------|------|-------|-------|---------|-------|------|-----|
| z 🚷 Denver Nuggets | 53 | 29 | .646 | - | 34-7 | 19-22 | 10-6 | 34-18 | 115.8 | 112.5 | +3.3 | W1 | 5-5 |
| y Wemphis Grizzlies | 51 | 31 | .622 | 2 | 35-6 | 16-25 | 13-3 | 30-22 | 116.9 | 113.0 | +3.9 | L1 | 6-4 |
| y 🗱 Sacramento Kings | 48 | 34 | .585 | 5 | 23-18 | 25-16 | 9-7 | 32-20 | 120.7 | 118.1 | +2.6 | L3 | 5-5 |
| x 🥟 Phoenix Suns | 45 | 37 | .549 | 8 | 28-13 | 17-24 | 9-7 | 30-22 | 113.6 | 111.6 | +2.0 | L2 | 7-3 |
| x Clippers | 44 | 38 | .537 | 9 | 23-18 | 21-20 | 9-7 | 27-25 | 113.6 | 113.1 | +0.5 | W3 | 6-4 |
| x Golden State Warriors | 44 | 38 | .537 | 9 | 33-8 | 11-30 | 7-9 | 30-22 | 118.9 | 117.1 | +1.8 | W3 | 8-2 |
| xp 🔷 Los Angeles Lakers | 43 | 39 | .524 | 10 | 23-18 | 20-21 | 6-10 | 27-25 | 117.2 | 116.6 | +0.6 | W2 | 8-2 |
| xp Minnesota Timberwolves | 42 | 40 | .512 | 11 | 22-19 | 20-21 | 8-8 | 29-23 | 115.8 | 115.8 | 0.0 | W3 | 7-3 |
| pb 🐙 Oklahoma City Thunder | 40 | 42 | .488 | 13 | 24-17 | 16-25 | 9-7 | 25-27 | 117.5 | 116.4 | +1.1 | W2 | 4-6 |
| pb Plicans Pelicans | 42 | 40 | .512 | 11 | 27-14 | 15-26 | 11-5 | 29-23 | 114.4 | 112.5 | +1.9 | L1 | 7-3 |
| e Dallas Mavericks | 38 | 44 | .463 | 15 | 23-18 | 15-26 | 9-7 | 28-24 | 114.2 | 114.1 | +0.1 | L2 | 2-8 |
| e d Utah Jazz | 37 | 45 | .451 | 16 | 23-18 | 14-27 | 6-10 | 24-28 | 117.1 | 118.0 | -0.9 | L1 | 2-8 |
| e Portland Trail Blazers | 33 | 49 | .402 | 20 | 17-24 | 16-25 | 7-9 | 23-29 | 113.4 | 117.4 | -4.0 | L4 | 1-9 |
| e Houston Rockets | 22 | 60 | .268 | 31 | 14-27 | 8-33 | 4-12 | 12-40 | 110.7 | 118.6 | -7.9 | W3 | 4-6 |
| e 🥊 San Antonio Spurs | 22 | 60 | .268 | 31 | 14-27 | 8-33 | 3-13 | 10-42 | 113.0 | 123.1 | -10.1 | W1 | 3-7 |
| | | | | | | | | | | | | | |

3*

In the 2022-2023 NBA Season, if you were in the Eastern Conference, a team needed 36 wins to qualify for the play-in tournament. To instantly qualify for the playoffs, a team needed 45 wins. In the Western Conference, a team needed 39 wins to qualify for the play-in tournament and 44 wins to instantly qualify for the playoffs.

After the conclusion of the play-in tournament, playoff qualifying teams passed the average threshold of 41 wins in the Eastern Conference and 42 wins in the Western Conference.

2023-24

Eastern Conference

| <u>DIFF</u> +11.4 +4.6 | <u>STRK</u> L1 W2 7- W5 6- |
|------------------------------|---|
| +4.6 | |
| | W5 6- |
| [| |
| +2.6 | L2 3- |
| +2.4 | L1 4- |
| +2.0 | W1 5- |
| +3.1 | W1 7- |
| +3.1 | W8 8- |
| +1.7 | W2 7- |
| -1.4 | L1 5- |
| -2.2 | L6 3- |
| -2.9 | L2 5- |
| -6.4 | L4 2- |
| -10.2 | W1 3- |
| -9.3 | L6 1- |
| -9.1 | L1 2- |
| | +2.0 +3.1 +3.1 +1.7 -1.4 -2.2 -2.9 -6.4 -10.2 -9.3 |

Western Conference

| | <u>w</u> | L | PCT | <u>GB</u> ^ | HOME | AWAY | DIV | CONF | PPG | OPP PPG | DIFF | STRK | L10 |
|----------------------------|----------|----|------|-------------|-------|-------|------|-------|-------|---------|------|------|-----|
| z 🐙 Oklahoma City Thunder | 57 | 25 | .695 | - | 33-8 | 24-17 | 12-4 | 36-16 | 120.1 | 112.7 | +7.4 | W5 | 7-3 |
| x Oenver Nuggets | 57 | 25 | .695 | - | 33-8 | 24-17 | 10-6 | 33-19 | 114.9 | 109.6 | +5.3 | W1 | 6-4 |
| x - Minnesota Timberwolves | 56 | 26 | .683 | 1 | 30-11 | 26-15 | 12-4 | 37-15 | 113.0 | 106.5 | +6.5 | L1 | 6-4 |
| y Clippers | 51 | 31 | .622 | 6 | 25-16 | 26-15 | 9-7 | 30-22 | 115.6 | 112.3 | +3.3 | L3 | 6-4 |
| y 🚺 Dallas Mavericks | 50 | 32 | .610 | 7 | 25-16 | 25-16 | 11-5 | 31-21 | 117.9 | 115.6 | +2.3 | L2 | 7-3 |
| x 🥖 Phoenix Suns | 49 | 33 | .598 | 8 | 25-16 | 24-17 | 9-9 | 29-23 | 116.2 | 113.2 | +3.0 | W3 | 7-3 |
| xp 🐣 Los Angeles Lakers | 47 | 35 | .573 | 10 | 28-14 | 19-21 | 7-10 | 27-25 | 118.0 | 117.4 | +0.6 | W2 | 7-3 |
| xp > New Orleans Pelicans | 49 | 33 | .598 | 8 | 21-19 | 28-14 | 9-7 | 30-22 | 115.1 | 110.7 | +4.4 | L1 | 5-5 |
| pb 🥮 Sacramento Kings | 46 | 36 | .561 | 11 | 24-17 | 22-19 | 10-7 | 30-22 | 116.6 | 114.8 | +1.8 | W1 | 4-6 |
| pb Golden State Warriors | 46 | 36 | .561 | 11 | 21-20 | 25-16 | 7-9 | 26-26 | 117.8 | 115.2 | +2.6 | W1 | 8-2 |
| e 👫 Houston Rockets | 41 | 41 | .500 | 16 | 27-14 | 14-27 | 9-7 | 28-24 | 114.3 | 113.2 | +1.1 | W2 | 4-6 |
| e 🗲 Utah Jazz | 31 | 51 | .378 | 26 | 21-20 | 10-31 | 5-11 | 16-36 | 115.7 | 120.5 | -4.8 | L1 | 2-8 |
| e W Memphis Grizzlies | 27 | 55 | .329 | 30 | 9-32 | 18-23 | 8-8 | 14-37 | 105.8 | 112.8 | -7.0 | L5 | 3-7 |
| e 🥊 San Antonio Spurs | 22 | 60 | .268 | 35 | 12-29 | 10-31 | 3-13 | 14-37 | 112.1 | 118.6 | -6.5 | W2 | 6-4 |
| e M Portland Trail Blazers | 21 | 61 | .256 | 36 | 11-30 | 10-31 | 1-15 | 8-44 | 106.4 | 115.4 | -9.0 | L5 | 2-8 |

4*

In the 2022-2023 NBA Season, if you were in the Eastern Conference, a team needed 33 wins to qualify for the play-in tournament. To instantly qualify for the playoffs, a team needed 47 wins. In the Western Conference, a team needed 42 wins to qualify for the play-in tournament and 48 wins to instantly qualify for the playoffs.

After the conclusion of the play-in tournament, playoff qualifying teams passed the average threshold of 40 wins in the Eastern Conference and 47 wins in the Western Conference.

To summarize, Eastern Conference teams needed 38, 36, 33 wins to make the play-in tournament and the Western Conference teams needed 34, 39, and 42 wins to make the play-in tournament in 2022, 2023, and 2024 respectively.

According to NBA play-in tournament Wikipedia (5*), the average number of wins required to secure the 10th seed typically falls between 38 and 42 wins. For a more detailed breakdown, The Eastern Conference is historically less competitive, requiring around 38-40 wins whereas the more competitive Western Conference requires around 40-42 wins.

Combining this information with the data that we have, let's set the weight of each of these 50/50 (csv data and online data), the "magic" number that NBA teams are trying to reach to secure a spot in the play-in tournament league-wide will be set to 39.

In order to secure a NBA playoff spot without having to go into the play-in tournament, in previous years, Eastern Conference teams needed 45, 45, 47 wins and the Western Conference teams needed 47, 44, 48 wins in 2022, 2023, 2024 respectively.

According to the NBA's official website (6*), the Eastern Conference requires around 42-45 wins and the Western Conference requires around 44-47 wins to achieve this feat, with the league average being around 45 wins.

Again, combining our csv data and online data, the "magic" number that NBA teams are trying to reach to secure a top 6 spot to go straight into the playoffs will be set to 46.

Combining the 50% chance that teams have in making the playoffs after qualifying for the play-in tournament, this project: the EDA and Modeling will revolve around achieving one single number:

43 wins

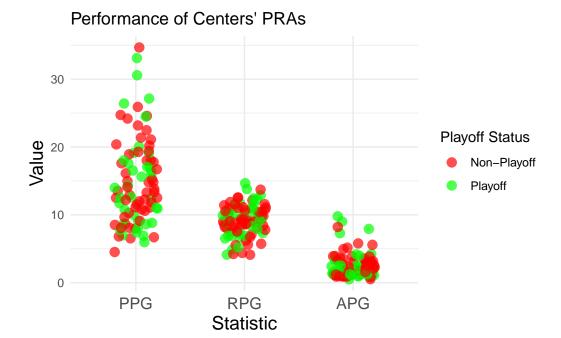
EDA

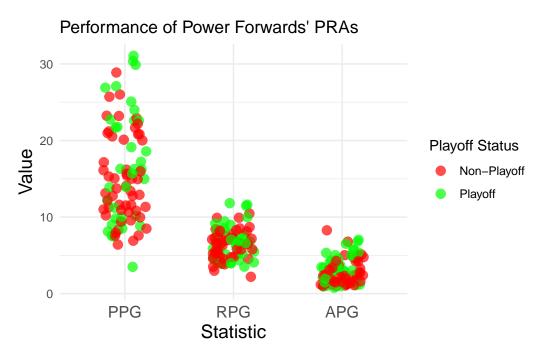
In our EDA, we will visualize the data to see if anything catches our eyes before we dive into the Modeling section. We will be combining and looking at all 3 years of data.

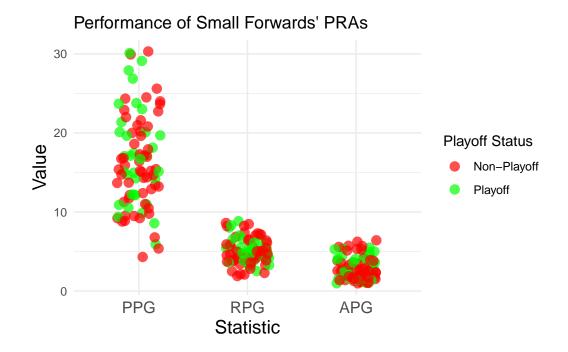
PRAs

Playoff vs Non-Playoffs

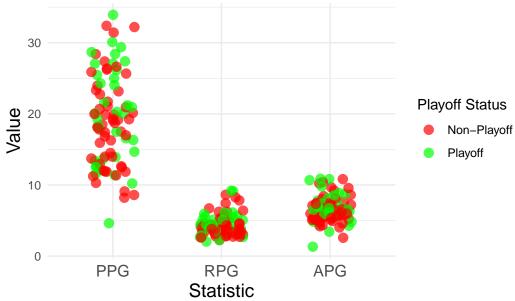
We will first look to see if PRA averages between playoff clinching players vs players who did not clinch the playoffs differ.

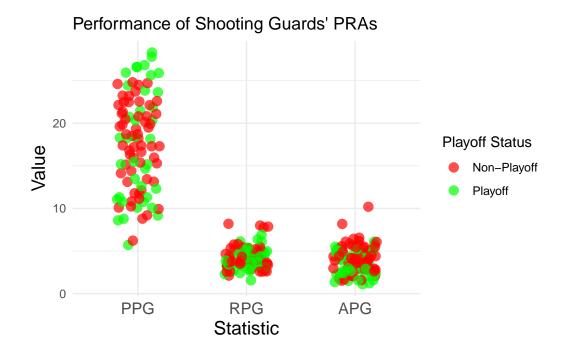






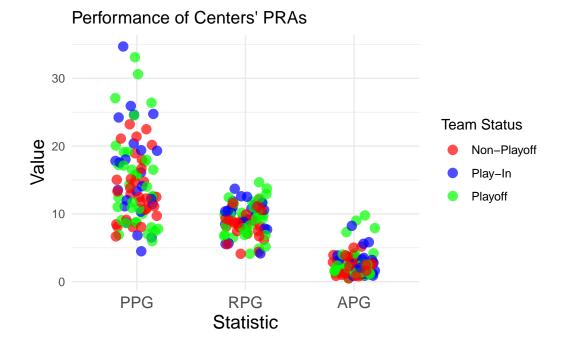
Performance of Point Guards' PRAs

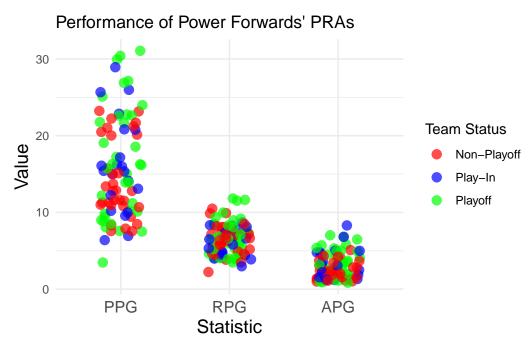


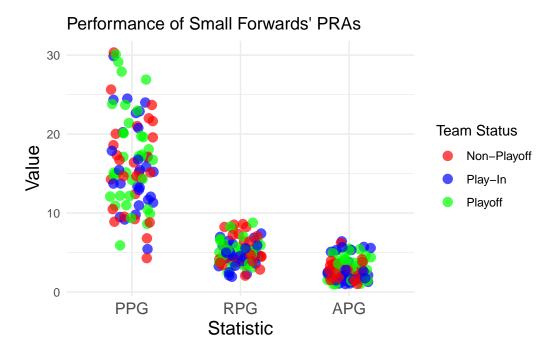


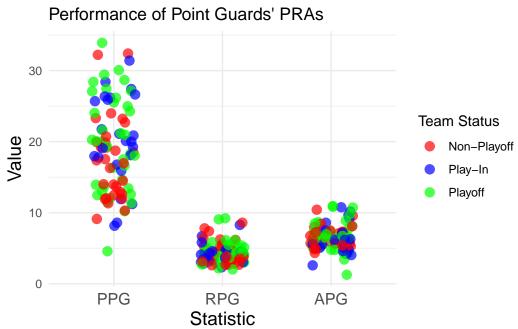
Looking at the data, the green and red dots seem to be scattered "randomly" with only some of the apex dots (players) being green. If playoff teams were "better" than non-playoff teams, then we would expect to see more green dots at the top and the red dots at the bottom. Maybe players on teams that are close to making the playoffs are the red dots at the top; therefore, let's do the same plots again but with play-in tournament players shown in blue.

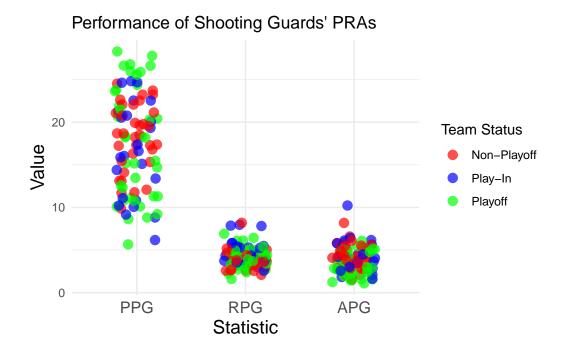
Playoffs vs Play-Ins vs Non-Playoffs











Based on these dot plots, we can see that there are fewer red dots near the top. It is mostly filled with green and blue dots, which is what we are aiming for.

Conclusion

It seems as if RPGs, more specifically APGs, are not as important as PPGs due to the low amount players record in general. However, it is important to notice that certain positions will be recording higher counts of either PPGs, RPGs, or APGs than other positions simply due to the play-style of the game. Therefore, this is something that we will be looking at in the Modeling section.

Shooting

Playoff vs Non-Playoffs

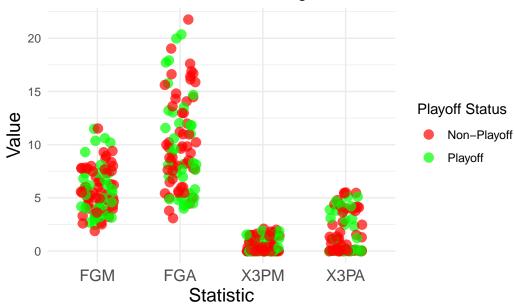
Now, let's see if we see differences in shooting within playoffs vs non-playoffs teams.

I have two predictions:

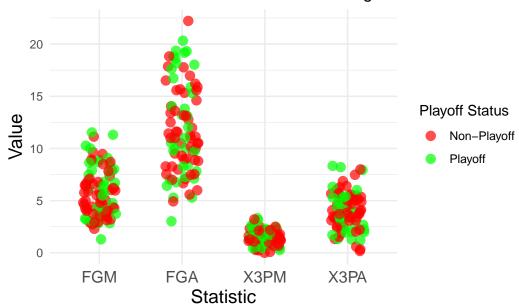
• "Better" teams will have players with higher shooting percentages. In this case, it will be more green dots near the top of FGM and X3PM but lower in FGA and X3PA

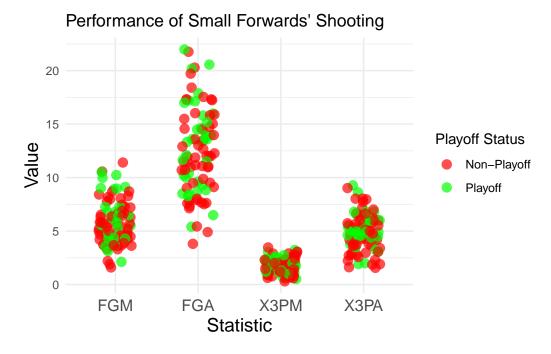
• "Better" teams will have players with higher FGM, FGA, X3PM, and X3PA due to the fact that these players are creating more shooting opportunities and thus with normal shooting percentages, are scoring more.

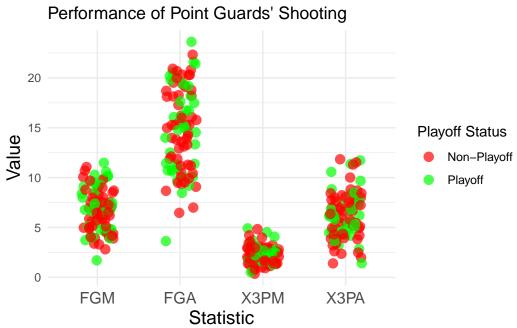


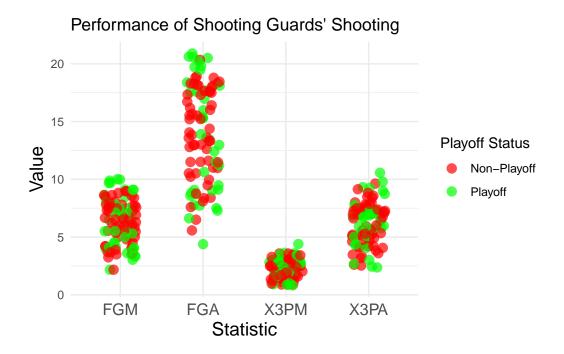


Performance of Power Forwards' Shooting





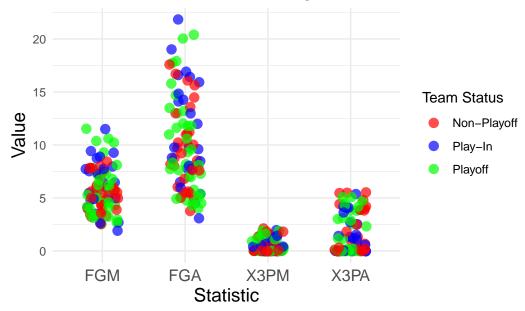




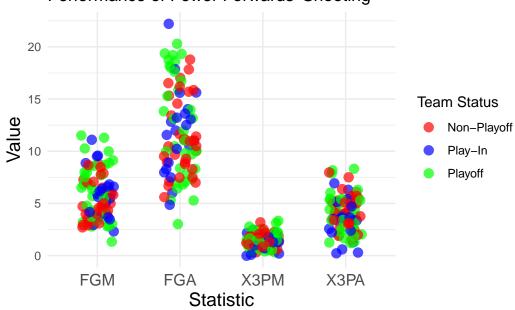
Again, for a clearer visual, let's try adding in the play-in teams to see if the higher red dots in the dot plots are from teams who are close to making the playoffs.

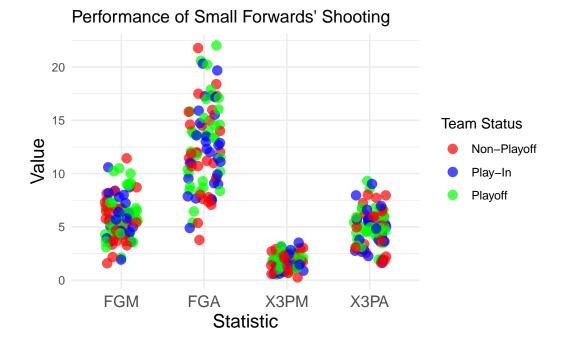
Playoffs vs Play-Ins vs Non-Playoffs

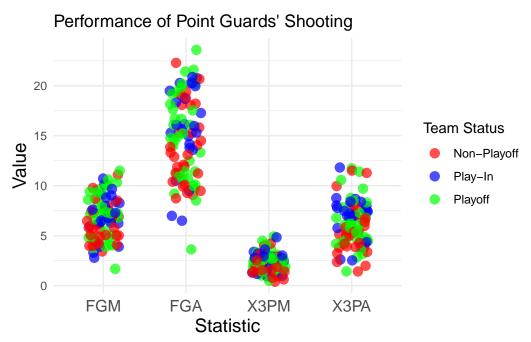
Performance of Centers' Shooting

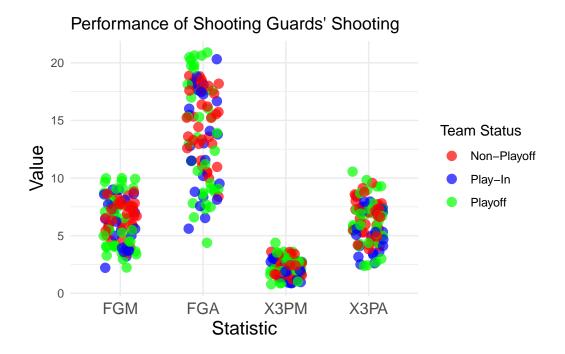


Performance of Power Forwards' Shooting









Conclusion

It seems like my second prediction was more accurate. Based on the dot plot, we can see that in both sections (FGM and FGA), there are more greens and blues at the top. However, when it comes to three pointers, there was not really a big difference between playoffs and non-playoffs teams. This might be due to the fact that some teams have players who are just better long range shooters or are known for their long range shot, like Stephen Curry, and some teams aren't.

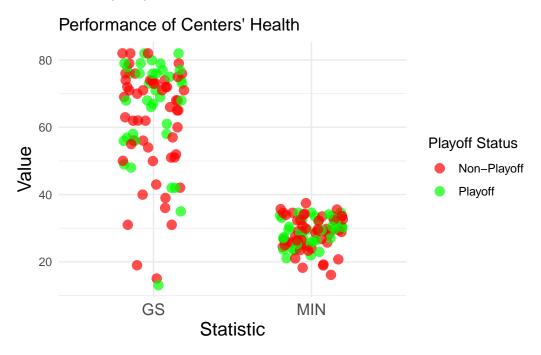
Health

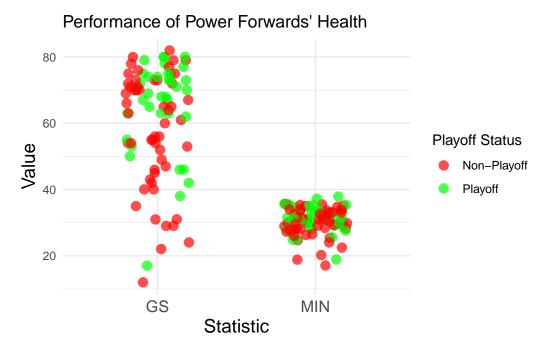
Now, when drafting or trading for a player, it is very important to assess their health. Are they injury prone? Have they missed numerous games due to illnesses or other personal issues? Usually, when a player is good enough to be a starter, they will stay a starter whenever they are in good condition. When players are considered "active" for the game, but start on the bench, it means that there is something wrong with the player. Therefore, we will look at players' games started (GS) and minutes played (MIN) in order to see how much they will be able to contribute to the team during the season.

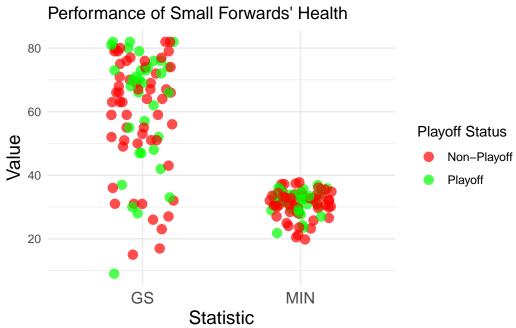
Playoff vs Non-Playoffs

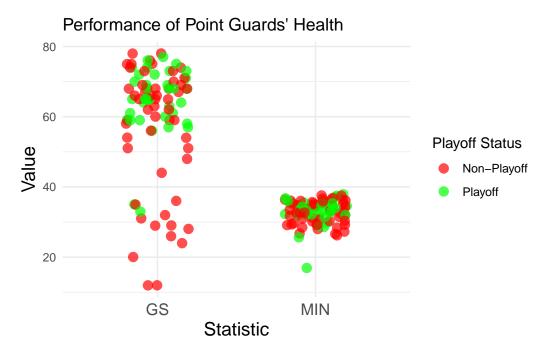
First, we will look at the just playoff teams vs non-playoff teams with play-in qualifying teams being considered non-playoff teams like before.

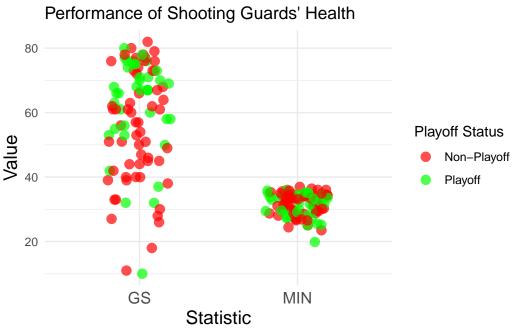
My prediction is that "better" teams will have players who are not injury prone (GS) and have better stamina (MIN).





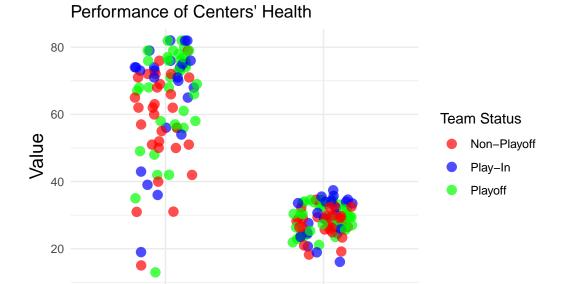






Playoffs vs Play-Ins vs Non-Playoffs

Now, let us look at when play-in teams are differentiated.

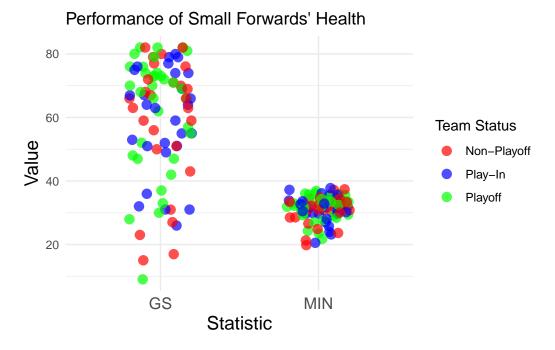


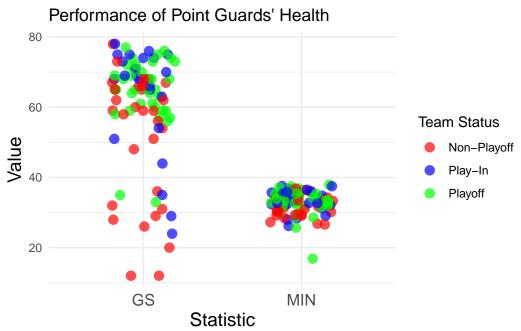
Statistic

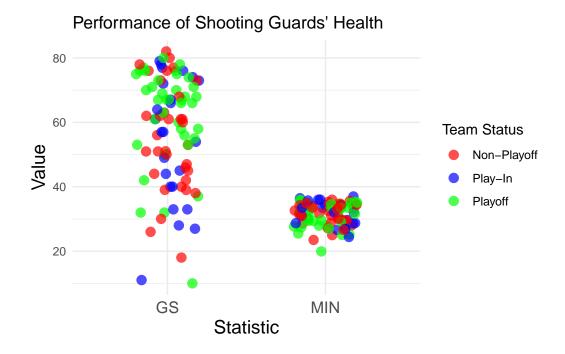
MIN

GS

Performance of Power Forwards' Health Team Status Non-Playoff Play-In Playoff Statistic







Conclusion

It's clear in both dot plots that in the amount of games started, playoff teams had players who started most, if not all 82, games of the season. Other than some outliers, my initial prediction was correct. However, for the number of minutes a player plays during the game, it seems as if there was no correlation between playoff teams and non-playoff teams. The reason for this could be the fact that players on both "good" and "bad" teams are still playing all 48 minutes of the game.

Modeling

We now have an idea of what teams have to do in order to secure a playoff spot. However, just eyeballing our data is not good enough for scouts to use and pick out specific players to draft or trade, especially with all the scattered range in our dot plots. Therefore, we will do three different models: the Null Model, Linear Model, and Multilinear Model. Within each of these models, we will do: Complete Pooling, No Pooling, and Partial Pooling.

As for the predictors from the EDA, we have decided to only test: GS, PPG, RPG, and APG. We already established that MIN, 3PM, and 3PA are not really important and for FGM and FGA, we decided not to use this since more shots and more field goals will directly translate to PPG, RPG, and APG.

The Null Model

Games Started

Complete Pooling

Playoff Teams

```
Call:
```

lm(formula = GS ~ 1, data = playoff_data)

Residuals:

Min 1Q Median 3Q Max -54.911 -5.911 4.089 10.089 18.089

Coefficients:

Estimate Std. Error t value Pr(>|t|)
(Intercept) 63.911 1.093 58.48 <2e-16 ***

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 14.66 on 179 degrees of freedom

Play-In Teams

Call:

lm(formula = GS ~ 1, data = playin_data)

Residuals:

Min 1Q Median 3Q Max -49.099 -9.099 5.901 13.901 21.901

Coefficients:

Estimate Std. Error t value Pr(>|t|)
(Intercept) 60.10 1.55 38.77 <2e-16 ***

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 17.05 on 120 degrees of freedom

Non-Playoff Teams

Call:

lm(formula = GS ~ 1, data = non_playoff_data)

Residuals:

Min 1Q Median 3Q Max -43.967 -10.967 4.033 13.283 26.033

Coefficients:

Estimate Std. Error t value Pr(>|t|)
(Intercept) 55.967 1.438 38.93 <2e-16 ***

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 17.72 on 151 degrees of freedom

No Pooling

Playoff Teams

A tibble: 5 x 5

| | Position | ${\tt mean_gs}$ | std_err | t_value | p_value |
|---|-------------|------------------|-------------|-------------|-------------|
| | <chr></chr> | <dbl></dbl> | <dbl></dbl> | <dbl></dbl> | <dbl></dbl> |
| 1 | C | 65.7 | 2.56 | 25.7 | 2.78e-24 |
| 2 | PF | 65.2 | 2.38 | 27.3 | 9.97e-25 |
| 3 | PG | 64.5 | 1.56 | 41.5 | 5.50e-32 |
| 4 | SF | 61.9 | 3.18 | 19.5 | 1.13e-19 |
| 5 | SG | 62.3 | 2.46 | 25.3 | 5.57e-25 |

Play-In Teams

A tibble: 5 x 5

| | Position | mean_gs | std_err | t_value | p_value |
|---|-------------|-------------|-------------|-------------|-------------|
| | <chr></chr> | <dbl></dbl> | <dbl></dbl> | <dbl></dbl> | <dbl></dbl> |
| 1 | C | 66.1 | 3.37 | 19.6 | 7.24e-16 |
| 2 | PF | 57.7 | 3.61 | 16.0 | 7.35e-13 |
| 3 | PG | 63.1 | 3.15 | 20.0 | 4.80e-16 |
| 4 | SF | 59.1 | 3.08 | 19.2 | 2.85e-17 |
| 5 | SG | 54.4 | 3.93 | 13.8 | 1.20e-12 |

Non-Playoff Teams

```
# A tibble: 5 x 5
 Position mean_gs std_err t_value p_value
 <chr>
           <dbl>
                 <dbl> <dbl>
                                 <dbl>
1 C
            57.6
                   2.78
                          20.8 1.54e-18
2 PF
            57.4 3.24 17.7 8.85e-18
                 3.45 14.8 4.47e-15
3 PG
            51.2
4 SF
            58.6 3.87 15.1 2.11e-14
5 SG
            55.3
                   2.81 19.7 8.15e-20
```

Partial Pooling

Playoff Teams

Linear mixed model fit by REML ['lmerMod']
Formula: GS ~ 1 + (1 | Position)
 Data: playoff_data

REML criterion at convergence: 1474.5

Scaled residuals:

Min 1Q Median 3Q Max -3.7452 -0.4032 0.2789 0.6881 1.2338

Random effects:

Groups Name Variance Std.Dev.

Position (Intercept) 0 0.00

Residual 215 14.66

Number of obs: 180, groups: Position, 5

Fixed effects:

Estimate Std. Error t value (Intercept) 63.911 1.093 58.48 optimizer (nloptwrap) convergence code: 0 (OK) boundary (singular) fit: see help('isSingular')

Play-In Teams

Linear mixed model fit by REML ['lmerMod']
Formula: GS ~ 1 + (1 | Position)

Data: playin_data

REML criterion at convergence: 1025.3

Scaled residuals:

Min 1Q Median 3Q Max -2.7684 -0.5158 0.3791 0.7491 1.3577

Random effects:

Groups Name Variance Std.Dev.
Position (Intercept) 9.023 3.004
Residual 283.539 16.839
Number of obs: 121, groups: Position, 5

Fixed effects:

Estimate Std. Error t value (Intercept) 60.091 2.039 29.47

Non-Playoff Teams

Linear mixed model fit by REML ['lmerMod']
Formula: GS ~ 1 + (1 | Position)

Data: non_playoff_data

REML criterion at convergence: 1301.8

Scaled residuals:

Min 1Q Median 3Q Max -2.4806 -0.6188 0.2275 0.7494 1.4687

Random effects:

Groups Name Variance Std.Dev.
Position (Intercept) 0.0 0.00
Residual 314.2 17.72
Number of obs: 152, groups: Position, 5

Fixed effects:

Estimate Std. Error t value (Intercept) 55.967 1.438 38.93 optimizer (nloptwrap) convergence code: 0 (OK) boundary (singular) fit: see help('isSingular')

Points Per Game

Complete Pooling

Non-Playoff Teams

Playoff Teams

```
Call:
lm(formula = PPG ~ 1, data = playoff_data)
Residuals:
    Min
              1Q
                   Median
                                ЗQ
                                        Max
                            5.3889 16.6889
-13.7111 -5.9111 -0.7111
Coefficients:
           Estimate Std. Error t value Pr(>|t|)
(Intercept) 17.2111 0.5177 33.24 <2e-16 ***
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
Residual standard error: 6.946 on 179 degrees of freedom
Play-In Teams
Call:
lm(formula = PPG ~ 1, data = playin_data)
Residuals:
    Min
              1Q
                   Median
                                ЗQ
                                        Max
-12.5157 -5.0157 -0.8157
                            4.0843 17.6843
Coefficients:
           Estimate Std. Error t value Pr(>|t|)
(Intercept) 17.0157 0.5759
                                 29.55
                                        <2e-16 ***
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
Residual standard error: 6.335 on 120 degrees of freedom
```

Call:

lm(formula = PPG ~ 1, data = non_playoff_data)

Residuals:

Min 1Q Median 3Q Max -11.6618 -4.1618 -0.7618 4.0382 16.4382

Coefficients:

Estimate Std. Error t value Pr(>|t|)
(Intercept) 15.9618 0.4338 36.8 <2e-16 ***

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 5.348 on 151 degrees of freedom

No Pooling

Playoff Teams

A tibble: 5 x 5

| | Position | mean_ppg | std_err | t_value | p_value |
|---|-------------|-------------|-------------|-------------|-------------|
| | <chr></chr> | <dbl></dbl> | <dbl></dbl> | <dbl></dbl> | <dbl></dbl> |
| 1 | C | 14.8 | 1.15 | 12.9 | 7.15e-15 |
| 2 | PF | 16.7 | 1.28 | 13.1 | 7.84e-15 |
| 3 | PG | 19.8 | 1.13 | 17.5 | 3.75e-19 |
| 4 | SF | 17.1 | 1.05 | 16.3 | 2.56e-17 |
| 5 | SG | 17.6 | 1.08 | 16.3 | 1.70e-18 |

Play-In Teams

A tibble: 5 x 5

| | ${\tt Position}$ | mean_ppg | $\mathtt{std}_\mathtt{err}$ | t_{value} | p_value |
|---|------------------|-------------|------------------------------|-------------|-------------|
| | <chr></chr> | <dbl></dbl> | <dbl></dbl> | <dbl></dbl> | <dbl></dbl> |
| 1 | C | 16.1 | 1.47 | 10.9 | 1.36e-10 |
| 2 | PF | 16.1 | 1.35 | 11.9 | 1.53e-10 |
| 3 | PG | 20.3 | 1.28 | 15.9 | 7.01e-14 |
| 4 | SF | 16.1 | 1.09 | 14.8 | 1.83e-14 |
| 5 | SG | 16.5 | 1.15 | 14.4 | 5.73e-13 |

Non-Playoff Teams

A tibble: 5 x 5

| | Position | ${\tt mean_ppg}$ | std_err | t_value | p_value |
|---|-------------|-------------------|-------------|-------------|-------------|
| | <chr></chr> | <dbl></dbl> | <dbl></dbl> | <dbl></dbl> | <dbl></dbl> |
| 1 | С | 13.8 | 0.879 | 15.8 | 1.87e-15 |
| 2 | PF | 14.2 | 0.896 | 15.9 | 2.00e-16 |
| 3 | PG | 17.4 | 1.06 | 16.4 | 3.06e-16 |
| 4 | SF | 15.9 | 1.17 | 13.6 | 2.38e-13 |
| 5 | SG | 18.2 | 0.671 | 27.1 | 3.86e-24 |

Partial Pooling

Playoff Teams

Linear mixed model fit by REML ['lmerMod']

Formula: PPG \sim 1 + (1 | Position)

Data: playoff_data

REML criterion at convergence: 1204.8

Scaled residuals:

Min 1Q Median 3Q Max -2.06971 -0.89628 -0.07599 0.78056 2.53686

Random effects:

Groups Name Variance Std.Dev.
Position (Intercept) 1.925 1.388
Residual 46.686 6.833
Number of obs: 180, groups: Position, 5

Fixed effects:

Estimate Std. Error t value (Intercept) 17.1970 0.8029 21.42

Play-In Teams

Linear mixed model fit by REML ['lmerMod']

Formula: PPG ~ 1 + (1 | Position)

Data: playin_data

REML criterion at convergence: 787.1

Scaled residuals:

Min 1Q Median 3Q Max -1.93830 -0.73335 -0.09513 0.67603 2.91363

Random effects:

Groups Name Variance Std.Dev.
Position (Intercept) 1.731 1.316
Residual 38.742 6.224
Number of obs: 121, groups: Position, 5

Fixed effects:

Estimate Std. Error t value (Intercept) 17.0226 0.8172 20.83

Non-Playoff Teams

Linear mixed model fit by REML ['lmerMod']
Formula: PPG ~ 1 + (1 | Position)

Data: non_playoff_data

REML criterion at convergence: 932.7

Scaled residuals:

Min 1Q Median 3Q Max -2.2627 -0.7224 -0.1032 0.7092 2.9922

Random effects:

Groups Name Variance Std.Dev.
Position (Intercept) 2.868 1.694
Residual 26.231 5.122
Number of obs: 152, groups: Position, 5

Fixed effects:

Estimate Std. Error t value (Intercept) 15.9213 0.8643 18.42

Rebounds Per Game

Complete Pooling

Playoff Teams

```
Call:
```

lm(formula = RPG ~ 1, data = playoff_data)

Residuals:

Min 1Q Median 3Q Max -4.3633 -1.8633 -0.7633 1.2367 8.7367

Coefficients:

Estimate Std. Error t value Pr(>|t|)
(Intercept) 5.9633 0.1997 29.85 <2e-16 ***

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 2.68 on 179 degrees of freedom

Play-In Teams

Call:

lm(formula = RPG ~ 1, data = playin_data)

Residuals:

Min 1Q Median 3Q Max -3.9413 -1.9413 -0.8413 1.8587 7.8587

Coefficients:

Estimate Std. Error t value Pr(>|t|)
(Intercept) 5.8413 0.2437 23.97 <2e-16 ***

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 2.681 on 120 degrees of freedom

Non-Playoff Teams

Call:

lm(formula = RPG ~ 1, data = non_playoff_data)

Residuals:

1Q Median ЗQ Max -3.5776 -1.8776 -0.6776 1.8224 6.4224

Coefficients:

Estimate Std. Error t value Pr(>|t|) (Intercept) 5.6776 0.1897 29.93 <2e-16 ***

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 2.339 on 151 degrees of freedom

No Pooling

Playoff Teams

A tibble: 5 x 5

| | Position | ${\tt mean_rpg}$ | std_err | t_value | p_value |
|---|-------------|-------------------|-------------|-------------|-------------|
| | <chr></chr> | <dbl></dbl> | <dbl></dbl> | <dbl></dbl> | <dbl></dbl> |
| 1 | C | 9.38 | 0.414 | 22.6 | 1.74e-22 |
| 2 | PF | 6.81 | 0.386 | 17.6 | 1.12e-18 |
| 3 | PG | 4.48 | 0.257 | 17.4 | 4.18e-19 |
| 4 | SF | 5.17 | 0.253 | 20.4 | 2.77e-20 |
| 5 | SG | 4.10 | 0.192 | 21.4 | 1.91e-22 |

Play-In Teams

A tibble: 5 x 5

| | ${\tt Position}$ | mean_rpg | $\mathtt{std}_\mathtt{err}$ | t_{value} | p_value |
|---|------------------|-------------|------------------------------|-------------|-------------|
| | <chr></chr> | <dbl></dbl> | <dbl></dbl> | <dbl></dbl> | <dbl></dbl> |
| 1 | C | 9.82 | 0.476 | 20.6 | 2.42e-16 |
| 2 | PF | 6.1 | 0.390 | 15.6 | 1.11e-12 |
| 3 | PG | 4.11 | 0.272 | 15.1 | 1.97e-13 |
| 4 | SF | 4.57 | 0.279 | 16.4 | 1.43e-15 |
| 5 | SG | 4.85 | 0.304 | 16.0 | 6.24e-14 |

Non-Playoff Teams

A tibble: 5 x 5

Position mean_rpg std_err t_value p_value <chr> <dbl> <dbl> <dbl> <dbl> 1 C 8.46 0.356 23.8 4.23e-20

| 2 | PF | 6.55 | 0.335 | 19.5 | 5.70e-19 |
|---|----|------|-------|------|----------|
| 3 | PG | 4.48 | 0.290 | 15.5 | 1.54e-15 |
| 4 | SF | 5.27 | 0.356 | 14.8 | 3.49e-14 |
| 5 | SG | 3.86 | 0.191 | 20.2 | 3.62e-20 |

Partial Pooling

Playoff Teams

Linear mixed model fit by REML ['lmerMod']

Formula: RPG ~ 1 + (1 | Position)

Data: playoff_data

REML criterion at convergence: 751

Scaled residuals:

Min 1Q Median 3Q Max -2.79908 -0.58552 -0.03962 0.47916 2.89797

Random effects:

Groups Name Variance Std.Dev.
Position (Intercept) 4.576 2.139
Residual 3.462 1.861
Number of obs: 180, groups: Position, 5

Fixed effects:

Estimate Std. Error t value (Intercept) 5.9877 0.9667 6.194

Play-In Teams

Linear mixed model fit by REML ['lmerMod']

Formula: RPG ~ 1 + (1 | Position)

Data: playin_data

REML criterion at convergence: 489.3

Scaled residuals:

Min 1Q Median 3Q Max -3.2370 -0.6716 0.0017 0.5441 2.4274

Random effects:

Groups Name Variance Std.Dev.
Position (Intercept) 5.264 2.294
Residual 2.925 1.710
Number of obs: 121, groups: Position, 5

Fixed effects:

Estimate Std. Error t value (Intercept) 5.889 1.038 5.674

Non-Playoff Teams

Linear mixed model fit by REML ['lmerMod']
Formula: RPG ~ 1 + (1 | Position)

Data: non_playoff_data

REML criterion at convergence: 605.5

Scaled residuals:

Min 1Q Median 3Q Max -2.5660 -0.6152 -0.1167 0.5313 2.5444

Random effects:

Groups Name Variance Std.Dev.
Position (Intercept) 3.254 1.804
Residual 2.841 1.686
Number of obs: 152, groups: Position, 5

Fixed effects:

Estimate Std. Error t value (Intercept) 5.7237 0.8183 6.994

Assists Per Game

Complete Pooling

Playoff Teams

Call:

lm(formula = APG ~ 1, data = playoff_data)

```
Residuals:
```

Min 1Q Median 3Q Max -3.3772 -1.9022 -0.3772 1.4228 7.0228

Coefficients:

Estimate Std. Error t value Pr(>|t|)

(Intercept) 3.8772 0.1724 22.49 <2e-16 ***

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Residual standard error: 2.313 on 179 degrees of freedom

Play-In Teams

Call

lm(formula = APG ~ 1, data = playin_data)

Residuals:

Min 1Q Median 3Q Max -3.0008 -2.0008 -0.4008 1.5992 6.8992

Coefficients:

Estimate Std. Error t value Pr(>|t|)

(Intercept) 3.9008 0.2106 18.52 <2e-16 ***

Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1

Residual standard error: 2.317 on 120 degrees of freedom

Non-Playoff Teams

۱۱ء۲

lm(formula = APG ~ 1, data = non_playoff_data)

Residuals:

Min 1Q Median 3Q Max -3.1092 -1.6342 -0.5092 1.5908 6.7908

Coefficients:

```
Estimate Std. Error t value Pr(>|t|)
(Intercept) 3.6092 0.1663 21.7 <2e-16 ***
---
Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
```

Residual standard error: 2.051 on 151 degrees of freedom

No Pooling

Playoff Teams

A tibble: 5 x 5 Position mean_apg std_err t_value p_value <dbl> <dbl> <dbl> <dbl> 1 C 2.73 0.382 7.15 2.45e- 8 2 PF 3.45 0.306 11.3 4.83e-13 3 PG 6.68 0.329 20.3 2.53e-21 4 SF 3.04 0.246 12.3 6.58e-14 5 SG 3.38 0.228 14.8 3.72e-17

Play-In Teams

| # | A tibble: 5 x 5 | | | | | | |
|---|------------------|-------------|------------------------------|-------------|-------------|--|--|
| | ${\tt Position}$ | mean_apg | $\mathtt{std}_\mathtt{err}$ | t_value | p_value | | |
| | <chr></chr> | <dbl></dbl> | <dbl></dbl> | <dbl></dbl> | <dbl></dbl> | | |
| 1 | C | 2.78 | 0.364 | 7.63 | 9.49e- 8 | | |
| 2 | PF | 3.23 | 0.437 | 7.39 | 3.86e- 7 | | |
| 3 | PG | 6.45 | 0.424 | 15.2 | 1.76e-13 | | |
| 4 | SF | 2.88 | 0.328 | 8.76 | 2.25e- 9 | | |
| 5 | SG | 4.27 | 0.386 | 11.1 | 1.10e-10 | | |

Non-Playoff Teams

A tibble: 5 x 5 Position mean_apg std_err t_value p_value <dbl> <dbl> <dbl> <dbl> 1 C 2.29 0.240 9.54 2.69e-10 2 PF 2.31 0.200 11.6 8.73e-13 3 PG 6.45 0.259 24.9 3.90e-21 4 SF 2.69 12.6 1.39e-12 0.213 5 SG 4.18 0.262 15.9 4.59e-17

Partial Pooling

Playoff Teams

Linear mixed model fit by REML ['lmerMod']

Formula: APG ~ 1 + (1 | Position)

Data: playoff_data

REML criterion at convergence: 741

Scaled residuals:

Min 1Q Median 3Q Max -2.9020 -0.7527 -0.2077 0.5791 3.8599

Random effects:

Groups Name Variance Std.Dev.
Position (Intercept) 2.488 1.577
Residual 3.314 1.820
Number of obs: 180, groups: Position, 5

Fixed effects:

Estimate Std. Error t value (Intercept) 3.8563 0.7184 5.368

Play-In Teams

Linear mixed model fit by REML ['lmerMod']

Formula: APG ~ 1 + (1 | Position)

Data: playin_data

REML criterion at convergence: 509.6

Scaled residuals:

Min 1Q Median 3Q Max -1.9459 -0.7564 -0.2347 0.6416 3.1449

Random effects:

Groups Name Variance Std.Dev.
Position (Intercept) 2.199 1.483
Residual 3.586 1.894
Number of obs: 121, groups: Position, 5

Fixed effects:

Estimate Std. Error t value (Intercept) 3.9175 0.6853 5.717

Non-Playoff Teams

Linear mixed model fit by REML ['lmerMod']

Formula: APG ~ 1 + (1 | Position)

Data: non_playoff_data

REML criterion at convergence: 532

Scaled residuals:

Min 1Q Median 3Q Max -2.03495 -0.78500 -0.06839 0.62289 3.06616

Random effects:

Groups Name Variance Std.Dev.
Position (Intercept) 3.117 1.766
Residual 1.725 1.313
Number of obs: 152, groups: Position, 5

Fixed effects:

Estimate Std. Error t value (Intercept) 3.5845 0.7968 4.499

The Linear Model

Complete Pooling

Playoff Teams

Call:

glm(formula = Playoff_Prob ~ GS + PPG + RPG + APG, family = binomial,
 data = nba_data)

Coefficients:

Estimate Std. Error z value Pr(>|z|)
(Intercept) 0.597662 0.487366 1.226 0.2201
GS 0.014001 0.006780 2.065 0.0389 *

```
PPG 0.002854 0.023055 0.124 0.9015

RPG -0.021522 0.046566 -0.462 0.6440

APG -0.016067 0.063483 -0.253 0.8002
```

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

(Dispersion parameter for binomial family taken to be 1)

Null deviance: 367.78 on 452 degrees of freedom Residual deviance: 363.43 on 448 degrees of freedom

AIC: 576.92

Number of Fisher Scoring iterations: 4

No Pooling

Playoff Teams

A tibble: 5 x 6

| | ${\tt Position}$ | Intercept | GS_coef | ${\tt PPG_coef}$ | ${\tt RPG_coef}$ | ${\tt APG_coef}$ |
|---|------------------|-------------|-------------|-------------------|-------------------|-------------------|
| | <chr></chr> | <dbl></dbl> | <dbl></dbl> | <dbl></dbl> | <dbl></dbl> | <dbl></dbl> |
| 1 | C | 1.07 | 0.0234 | -0.00243 | -0.154 | 0.0891 |
| 2 | PF | 1.17 | 0.00898 | -0.101 | -0.0106 | 0.471 |
| 3 | PG | -0.215 | 0.0334 | 0.0234 | -0.0181 | -0.127 |
| 4 | SF | 1.12 | -0.00392 | 0.0928 | -0.233 | 0.0213 |
| 5 | SG | 0.224 | 0.0104 | -0.0578 | 0.866 | -0.480 |

Partial Pooling

Playoff Teams

```
Generalized linear mixed model fit by maximum likelihood (Laplace
```

Approximation) [glmerMod]
Family: binomial (logit)

Formula: Playoff_Prob ~ GS + PPG + RPG + APG + (1 | Position)

Data: nba_data

AIC BIC logLik deviance df.resid 561.0 585.7 -274.5 549.0 447

Scaled residuals:

Min 1Q Median 3Q Max

```
-1.8529 -0.3877 0.6020 0.6662 0.8940
Random effects:
Groups
        Name
                    Variance Std.Dev.
Position (Intercept) 0
                            0
Number of obs: 453, groups: Position, 5
Fixed effects:
           Estimate Std. Error z value Pr(>|z|)
(Intercept) 0.314758 0.448338 0.702 0.4826
GS
           PPG
           0.009169 0.020667 0.444 0.6573
          RPG
APG
          -0.044555 0.056590 -0.787 0.4311
___
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
Correlation of Fixed Effects:
   (Intr) GS
                PPG
                      RPG
GS -0.620
PPG -0.263 -0.131
RPG -0.342 -0.149 -0.218
APG -0.106 -0.036 -0.580 0.180
optimizer (Nelder_Mead) convergence code: 0 (OK)
boundary (singular) fit: see help('isSingular')
The Multilinear Model
Complete Pooling
Call:
lm(formula = Playoff_Prob ~ GS + PPG + RPG + APG + PPG:RPG, data = nba_data)
Residuals:
   Min
           1Q Median
                          3Q
                                Max
-0.8697 -0.2552 0.1777 0.2319 0.4192
Coefficients:
            Estimate Std. Error t value Pr(>|t|)
```

(Intercept) 0.9330929 0.1236742 7.545 2.56e-13 ***

0.0027561 0.0010317 2.671 0.00783 **

GS

```
PPG -0.0170693 0.0070862 -2.409 0.01641 *

RPG -0.0513466 0.0182734 -2.810 0.00517 **

APG -0.0023775 0.0092597 -0.257 0.79748

PPG:RPG 0.0027778 0.0009876 2.813 0.00513 **
```

Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1

Residual standard error: 0.3541 on 447 degrees of freedom Multiple R-squared: 0.03037, Adjusted R-squared: 0.01953

F-statistic: 2.8 on 5 and 447 DF, p-value: 0.0167

No Pooling

A tibble: 5 x 8

| | Position I | ntercept | GS_coef | PPG_coef | RPG_coef | APG_coef | PPG_RPG_coef | R_squared |
|---|-------------|-------------|-------------|-------------|-------------|-------------|--------------|-------------|
| | <chr></chr> | <dbl></dbl> | <dbl></dbl> | <dbl></dbl> | <dbl></dbl> | <dbl></dbl> | <dbl></dbl> | <dbl></dbl> |
| 1 | C | 1.29 | 0.00521 | -0.0460 | -0.0933 | -0.00215 | 0.00491 | 0.0689 |
| 2 | PF | 1.47 | 0.00236 | -0.0599 | -0.118 | 0.0526 | 0.00670 | 0.122 |
| 3 | PG | 0.690 | 0.00644 | -0.00794 | -0.0594 | -0.0174 | 0.00255 | 0.0924 |
| 4 | SF | 1.04 | -0.000190 | -0.00468 | -0.0989 | 0.00385 | 0.00354 | 0.0776 |
| 5 | SG | 1.48 | 0.00115 | -0.0553 | -0.0959 | -0.0704 | 0.0117 | 0.176 |

Partial Pooling

Linear mixed model fit by REML ['lmerMod']

Formula: Playoff_Prob ~ GS + PPG + RPG + APG + PPG:RPG + (1 | Position)

Data: nba_data

REML criterion at convergence: 395

Scaled residuals:

Min 1Q Median 3Q Max -2.4562 -0.7206 0.5020 0.6549 1.1838

Random effects:

Groups Name Variance Std.Dev.
Position (Intercept) 0.0000 0.0000
Residual 0.1254 0.3541
Number of obs: 453, groups: Position, 5

Fixed effects:

```
Estimate Std. Error t value
(Intercept)
             0.9330929 0.1236742
                                    7.545
GS
             0.0027561 0.0010317
                                    2.671
PPG
            -0.0170693 0.0070862
                                   -2.409
RPG
            -0.0513466 0.0182734
                                   -2.810
APG
            -0.0023775
                       0.0092597
                                    -0.257
PPG:RPG
             0.0027778
                       0.0009876
                                    2.813
Correlation of Fixed Effects:
        (Intr) GS
                      PPG
                             RPG
                                    APG
GS
        -0.316
PPG
        -0.774 -0.143
RPG
        -0.814 -0.136
                       0.781
APG
        -0.058 -0.022 -0.282
                              0.057
PPG:RPG 0.796 0.093 -0.881 -0.929 0.009
optimizer (nloptwrap) convergence code: 0 (OK)
boundary (singular) fit: see help('isSingular')
```

Conclusion

Now, since we created and took a look at all these different models, we can analyze our results to reach our "magic" number of 43 wins.

From the null model, we were able to determine the overall average GS, PPG, RPG, and APG of players and the average of players based on their position in the NBA. With the p-value being statistically significant, we were able to use weighted average calculations to determine the average threshold players needs to create a successful playoff team.

Since only 50% of play-in teams make it to the playoffs, we said half of their performance contributes to the playoff picture; therefore, using this formula:

Weighted Statistic for Playoffs = (Playoff Statistic) + 0.5 * (Play-In Statistic - Playoff Statistic)

We will be using the No Pooling Null Model to determine the average for each position and will be testing both Linear and Multilinear Model in order to calculate the chances the average playoff qualifying player had to make the playoffs.

Center

Null Model

• Games Played: 65.88

• Points Per Game: 15.47

• Rebounds Per Game: 9.60

• Assists Per Game: 2.75

Linear Model

With this model, the team with this center will have a 79.1% chance to make the playoffs.

Multilinear Model

With this model, the team with this center will have a 68.0% chance to make the playoffs.

Power Forward

Null Model

• Games Played: 59.44

• Points Per Game: 16.41

• Rebounds Per Game: 6.48

• Assists Per Game: 3.34

Linear Model

With this model, the team with this center will have a 82.5% chance to make the playoffs.

Multilinear Model

With this model, the team with this power forward will have a 68.0% chance to make the playoffs.

Small Forward

Null Model

• Games Played: 60.49

• Points Per Game: 17.56

• Rebounds Per Game: 5.161

• Assists Per Game: 2.96

Linear Model

With this model, the team with this center will have a 79.8% chance to make the playoffs.

Multilinear Model

With this model, the team with this power forward will have a 68.3% chance to make the playoffs.

Point Guard

Null Model

• Games Played: 64.81

• Points Per Game: 20.0

• Rebounds Per Game: 4.29

• Assists Per Game: 6.56

Linear Model

With this model, the team with this center will have a 81.8% chance to make the playoffs.

Multilinear Model

With this model, the team with this center will have a 69.0% chance to make the playoffs.

Shooting Guard

Null Model

• Games Played: 59.33

• Points Per Game: 17.07

• Rebounds Per Game: 4.47

• Assists Per Game: 4.84

Linear Model

With this model, the team with this center will have a 80.3% chance to make the playoffs.

Multilinear Model

With this model, the team with this center will have a 67.4% chance to make the playoffs.

Statement

With this, I believe that scouts will be able to see NBA prospects in a different way. To use sheer numbers in order to draft and trade players. I think that with the right combination of players with the right numbers, even the worst NBA teams can make the playoffs.

Error Analysis

We can see that there is a huge difference in playoff percentage between the linear and multilinear model, even with the same statistics. Let's dive into why this might be the case.

When viewing the p-values of the linear model, we can see that all predictors except for GS were statistically insignificant whereas in the multilinear model, all predictors except APG were statistically significant; therefore, it will be better to accept the multilinear model over the linear model. Although, there were a lot of large p-values, we still decided to run the tests to see whether or not the percentages were something out of the ordinary or if they made sense.

References

https://www.basketball-reference.com/teams/

```
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2* - https://www.espn.com/nba/standings/_/season/2022
3* - https://www.espn.com/nba/standings/_/season/2023
4* - https://www.espn.com/nba/standings/_/season/2024
5* - https://en.wikipedia.org/wiki/NBA_play-in_tournament?utm_source=chatgpt.com
6* - https://www.nba.com/news/playoff-primer-seeding-matchup-postseason-history-all-16-teams
The websites I used to create my csv files are:
https://www.nbastuffer.com/nba-stats/player/
https://www.statmuse.com/
```