

The 2019-20 NBA Season: What Could Have Been

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Abstract

The COVID-19 pandemic has affected our society in various ways and has changed numerous events on schedule for 2020. One such event that we were looking forward to was the end of the 2019-20 NBA season as well as the 2020 NBA Playoffs. Since a large portion of the 2019-20 NBA regular season games have already been played, we utilized the data collected from these games to calculate Elo ratings for each team in order to predict the standings at the end of the regular season as well as the matchups and results of the Playoffs. From our analysis, we found that ... **[NEED TO ADD MORE HERE]**.

Introduction

Due to the widespread impact of the COVID-19 pandemic throughout the world, almost every company, organization, and public event has canceled or suspended any activities that involve interpersonal contact for the foreseeable future. Many of these activities are moving to a virtual format if possible, but several others have been forced to shut down.

As avid sports fans, the absence of the major sporting events during this time has hit us and many others around the world especially hard [1]. Some of the events that we particularly were looking forward to include the NBA, NCAA March Madness tournament, MLB, and the 2020 Summer Olympics.

In our curiosity, we decided to utilize this opportunity to exercise our data science and modeling skills in order to predict what could have been. Specifically, we focused on the NBA and the NBA Playoffs. Since the 2019-20 NBA season was suspended approximately one month prior to the end of the regular season (and the beginning of the playoffs), we used the 2019-20 season data accumulated from the games played before the suspension to predict how the season and the playoffs would have ended had everything gone according to schedule.

In this analysis, we will examine data from the 2019-20 NBA season as well as some data from previous NBA seasons in order to draw some meaningful conclusions about the remainder of the 2019-20 NBA season.

Methods

Predicting the 2019-20 NBA Season

Since we missed one of the most exciting times of the year (the NBA Playoffs & Finals), we made some predictions on how the rest of the season might have played out using a popular methodology referred to as the Elo ratings system [2]. This tool, created by Hungarian-American physics professor Arpad Elo, was originally designed to rate chess players, but is now used for all sorts of competitions ranging anywhere from sports to video games. This is a methodology that FiveThirtyEight and many other popular sports analysts take advantage of due to its simplicity and effectiveness [3].

These ratings depend only on the final score of each game as well as where it was played (home-court advantage). In other words, this system is built on a Win/Loss basis. We will be analyzing the 2018-19 NBA Season in its entirety to validate its performance, then we will apply it to the 2019-20 regular season in order to predict the matchups for the Playoffs and the Finals and ultimately the NBA Champions. For this

project, we retrieved several types of data sources including game-by-game scores and schedules for several seasons from Basketball-Reference.com [4].

How does Elo work?

The long-run average for an Elo score in the NBA sits around 1500. An Elo of 1500 means that the teams performance would be normally distributed around an average of 1500 with the chance of performing better or worse. The formula for Elo below shows how the probability of one team beating another is calculated using the ratings.

When Player A competes in a match against Player B , Player A has an expected outcome (probability or score) for Team A ($E[A]$) where R_A is the rating for Team A and R_B is the rating for Team B . The expected outcome for Team A ($E[A]$) can be calculated by the formula below:

$$E[A] = \frac{1}{1 + 10^{\frac{(R_B - R_A)}{400}}}$$

The same calculation ($E[B]$) has to be done for Player B , but with R_A (current rating A) and R_B (current rating B) swapped so that $E[A] + E[B] = 1$. Once the match is played and S_A (actual outcome or score for Team A) and S_B (actual outcome or score for Team B) are determined, R'_A (the new rating for A) and R'_B (the new rating for B) are calculated with the formula below:

$$R'_A = R_A + K(S_A - E[A])$$

The S value in our case would either be 1 for a win, or 0 for a loss. This is because there are no ties in the NBA.

In this equation, K is an optimization constant that usually takes different values according the sport and the amount of games available. In other words, this value is the maximum amount by which a score can change in one match. If K is set too high, the ratings will jump around too much; if K is set too low, Elo will take too long to recognize important changes in team quality. Determining the right value of K is an entirely different and more complicated topic, so for this experiment we will be using $K = 20$, the optimal K for the NBA determined by FiveThirtyEight [3]. This is higher than most other sports and can likely be attributed to the fact that the NBA plays more games (81 games per team) and is subject to relatively little randomness.

Home-court advantage is set as equivalent to 100 Elo rating points. One hundred Elo points is equivalent to about 3.5 NBA points, so it can also be interpreted as the home team being favored by 3 to 4 points if the teams were otherwise evenly matched (obviously this value fluctuates from season to season). Since every team plays about half of their games at home and the other half away, a change in the home-court advantage value does not produce a significant difference in the ratings, but is still an important factor to consider.

Elo strikes a nice balance between ratings systems that account for margin of victory and those that do not. While teams always gain Elo points after wins and lose Elo points after losses, they also gain or lose more with larger margins of victory.

This works by assigning a multiplier to each game based on the final score and dividing it by a team's projected margin of victory conditional upon having won the game. For instance, the Golden State Warriors' 4-point margin over the Houston Rockets in Game 1 of the 2018-19 Western Conference finals was lower than Elo would expect for a Warriors win. So the Warriors gain Elo points, but not as many as if they'd won by a larger margin. The formula accounts for diminishing returns; going from a 5-point win to a 10-point win matters more than going from a 25-point win to a 30-point win. For the exact formula, see the footnotes.

Instead of resetting each team's rating when a new season begins, Elo carries over a portion of a team's rating from one season to the next. This is to account for any momentum that a team may build from season-to-season (i.e. sports dynasties). In NBA ratings, three-quarters of the previous score are kept. The high fraction reflects the fact that NBA teams are more consistent from year to year. For example, the Miami Heat ended the 2012-13 NBA season with an Elo rating of 1754. The team's Elo rating for the start of the 2013-14 season is calculated as follows:

$$(0.75 * 1754) + (0.25 * 1500) = 1692$$

Since this is a consistent method, we will also initialize the Elo scores for the 2019-20 NBA Season using the Elo scores from the 2018-19 season.

After incorporating a constant for home court advantage, our formula is as follows with $A = 100$ points (the value we previously determined represents a home-court advantage):

$$P(\text{Home team wins}) = \frac{1}{1 + 10^{-\frac{(H-R+A)}{400}}}$$

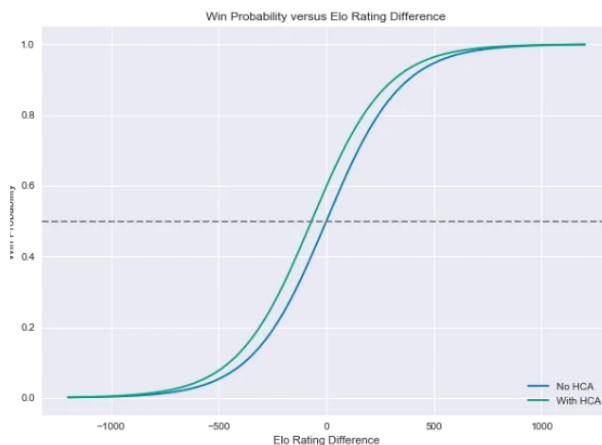


Figure 1: Logistic Function of Win Probability by Elo Rating Difference

In Figure 1, we see an example of a logistic function for win probability by Elo rating difference.

The 2018-19 NBA Season

Results

[Tables, figures, and text that illustrate your findings. Keep the focus on the numbers here. You will interpret your results in the next section.]

First, we calculated the Elo ratings for each team for the games played so far in the 2019-20 season. As explained earlier, we incorporated 25% of the previous seasons Elo ratings with 75% of this season's current Elo ratings. Below are the Elo ratings we derived for each team.

Table 1: Elo Ratings for Every NBA Team (Descending)

Team	Elo Rating
Milwaukee Bucks	1628.4919
Los Angeles Lakers	1627.8301
Toronto Raptors	1601.2836
Los Angeles Clippers	1588.4583
Oklahoma City Thunder	1582.6036
Boston Celtics	1569.6068
Denver Nuggets	1560.4563
Utah Jazz	1559.3314
Houston Rockets	1546.1499
Indiana Pacers	1544.9614
Philadelphia 76ers	1536.7509
Miami Heat	1531.0399
Dallas Mavericks	1530.4395

Team	Elo Rating
Memphis Grizzlies	1512.1054
New Orleans Pelicans	1507.3374
Sacramento Kings	1496.7678
Brooklyn Nets	1491.0335
Orlando Magic	1490.3970
Portland Trail Blazers	1478.5000
San Antonio Spurs	1478.2600
Phoenix Suns	1453.5151
Washington Wizards	1446.2512
Charlotte Hornets	1435.6679
New York Knicks	1430.9475
Atlanta Hawks	1424.2177
Cleveland Cavaliers	1412.5029
Chicago Bulls	1410.1883
Minnesota Timberwolves	1391.4969
Golden State Warriors	1387.3738
Detroit Pistons	1383.5339

Conclusion

[Interpret what you found. What are the main lessons we should take away from your report?]

References

- [1] List of all sporting events canceled around the world during the coronavirus pandemic (https://www.espn.com/olympics/story/_/id/28824781/list-sporting-events-canceled-coronavirus)
- [2] Elo ratings system (https://en.wikipedia.org/wiki/Elo_rating_system)
- [3] FiveThirtyEight NBA Elo Ratings (<https://fivethirtyeight.com/features/how-we-calculate-nba-elo-ratings/>)
- [4] Compilation of in-depth NBA statistics (<https://www.basketball-reference.com/>)