

NBA Player Career Dataset Analysis Class Project Final Report

Group 9

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Introduction

This project constitutes a comprehensive exploration into the intricacies of shot selection and performance, focusing on the eminent NBA player, Jimmy Butler. Through a synthesis of Exploratory Data Analysis (EDA) and advanced Machine Learning (ML) techniques, the overarching goal is to construct predictive models capable of unraveling the multifaceted decision-making processes that players, exemplified by Jimmy Butler, employ during critical shots. By delving into the granular details of Butler's shot data, this analysis seeks to uncover patterns and insights that contribute to a deeper understanding of his gameplay dynamics.

Meanwhile, this endeavor extends beyond individual player analysis to encompass broader implications for team management. Incorporating a salary analysis of players within each position, the project aims to provide valuable insights for basketball clubs in optimizing their recruitment strategies. Understanding the correlation between player performance and salary distributions within specific positions can empower clubs to make informed decisions, enhancing team dynamics and overall competitiveness.

This dual-pronged approach, combining individual pl ayer analytics with broader team management insights, positions the project at the intersection of player-centric performance evaluation and strategic decision-making for sports organizations. As the analysis unfolds, it aspires to offer a nuanced perspective on the intricate interplay between player performance, shot selection, and the financial considerations that underscore effective team management in the NBA.



Problem Statement

- 1. **Predictive Analytics:** The crux of our mission lies in foreseeing the upcoming season's performance of NBA teams, meticulously predicting win-loss records, points, and assists, all orchestrated through the lens of historical game data.
- Shooting Efficiency Analysis: Our second frontier involves a deep dive into the
 intricate tapestry of factors influencing a player's shooting success. This entails a
 meticulous dissection of various parameters, including action types, shot types,
 shot distances, and an array of other variables.
- 3. **Salary Prediction:** Unfurling the future financial landscape, we endeavor to predict NBA player salaries for an imminent season, weaving a narrative around player statistics, positions, team affiliations, and historical salary data, with a pronounced focus on the unfolding drama of the 2021 season.

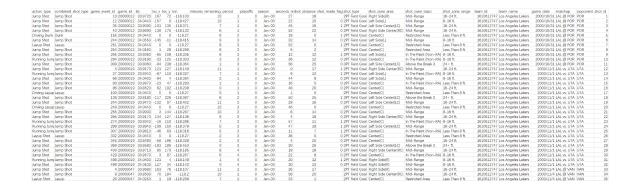
Related Works

- NBA Player Performance Analysis: Our quest involves an exhaustive exploration
 of existing research and studies, peering into the multifaceted world of NBA
 player performance analysis. This encompasses a detailed examination of player
 statistics, efficiency metrics, and their far-reaching impact on team performance.
- Predictive Modeling in Sports Analytics: With a discerning eye, we traverse
 through the annals of related works, unraveling the intricate tapestry of predictive
 modeling in sports analytics. Our focus spans player performance prediction,
 team win-loss prognosis, and the intricate factors that dance together to
 compose the symphony of success.
- 3. **Exploratory Data Analysis (EDA) in Sports Analytics:** Our journey into the heart of sports analytics literature includes a meticulous review of studies that have applied EDA techniques. We seek to glean insights into player performance nuances, team dynamics, and the capricious outcomes of riveting games.
- 4. Machine Learning Models for NBA Data: In our relentless pursuit of knowledge, we scrutinize research that has harnessed the power of machine learning models, notably the Random Forest algorithm, for the analysis of NBA data. Our gaze extends to the delicate art of feature engineering and the critical evaluation of model performance.



Datasets

 Career Dataset: This dataset comprises 30,697 records from Kobe Bryant's entire career, providing detailed information about individual shots taken by NBA players during games. It includes features like action type, shot type, shot distance, and game-related information.



30,697 pieces of data from Kobe Bryant's entire career from the '96 season to the 2016 season

Column	Meaning		
action_type	Action Type		
combined_shot_type	Combined Shot Type		
game_event_id	Game Event ID		
lat	Latitude		
loc_x	X Coordinate		
loc_y	Y Coordinate		
lon	Longitude		
minutes_remaining	Minutes Remaining		
period	Period		
playoffs	Playoffs		
season	Season		
seconds_remaining	Seconds Remaining		
shot_distance	Shot Distance		
shot_made_flag	Shot Made Flag		
shot_type	Shot Type		
shot_zone_area	Shot Zone Area		
shot_zone_basic	Shot Zone Basic		
shot_zone_range	Shot Zone Range		
team_id	Team ID		
team_name	Team Name		
game_date	Game Date		
matchup	Matchup		
opponent	Opponent		
game_id	Game ID		
shot_id	Shot ID		

Column name and meanings

2. NBA Player Information Dataset: This dataset contains information about 4,550 professional players, including career start and end years, team affiliations, and player statistics per season from 1950 to 2017.



name	year_start	year_end	position	height	weight	birth_date	college
Alaa Abdelnaby	1991	1995	F-C	6月10日	240	24-Jun-68	Duke University
Zaid Abdul-Aziz	1969	1978	C-F	6月9日	235	7-Apr-46	Iowa State University
Kareem Abdul-Jabbar	1970	1989	С	7月2日	225	16-Apr-47	University of California, Los Ange
Mahmoud Abdul-Rauf	1991	2001	G	6月1日	162	9-Mar-69	Louisiana State University
Tariq Abdul-Wahad	1998	2003	F	6月6日	223	3-Nov-74	San Jose State University
Shareef Abdur-Rahim	1997	2008	F	6月9日	225	11-Dec-76	University of California
Tom Abernethy	1977	1981	F	6月7日	220	6-May-54	Indiana University
Forest Able	1957	1957	G	6月3日	180	27-Jul-32	Western Kentucky University
John Abramovic	1947	1948	F	6月3日	195	9-Feb-19	Salem International University
Alex Abrines	2017	2018	G-F	6月6日	190	1-Aug-93	
Alex Acker	2006	2009	G	6月5日	185	21-Jan-83	Pepperdine University

NBA Player Information Dataset

3. NBA Player Salaries Dataset: This Dataset provides a comprehensive overview of player compensation in the National Basketball Association (NBA) over a span of two decades, from 2000 to 2020. With 9,456 records, the dataset includes valuable information such as player rank based on salary, player names, positions (e.g., Center, Power Forward, Small Forward, Point Guard, Shooting Guard), and affiliated NBA teams. This dataset offers insights into the financial landscape of professional basketball, showcasing the evolution of player salaries over time. Analysts and enthusiasts alike can utilize this rich dataset to explore trends, compare player earnings, and gain a deeper understanding of the economic dynamics within the NBA during the specified period.



NBA Player Salaries Dataset

Methodology

Exploratory Data Analysis (EDA):



- Objectives: Our EDA serves as a compass, guiding us through the dense thicket
 of data intricacies. The mission is expansive, ranging from unraveling the
 underlying data structure to detecting outliers and anomalies, testing
 fundamental assumptions, and crystallizing optimal factor settings.
- Graphical Techniques in EDA: Armed with a formidable arsenal of graphical techniques, our EDA journey involves the orchestration of data traces, histograms, bivariate histograms, probability plots, lag plots, block plots, mean plots, standard deviation plots, box plots, main effects plots, probability plot correlation coefficient plots, univariate and multivariate control charts, 4-plots, block plots, scatter plots, and a myriad of visual tools tailored to the unique contours of our data.
- Applications in Different EDA Problems: Tailoring our approach to the specific
 nature of the EDA problem at hand, we deftly wield tools like probability plots,
 probability plot correlation coefficient plots, univariate and multivariate control
 charts, 4-plots, block plots, scatter plots, and more. Our arsenal is diverse,
 ensuring that we unravel the data mysteries through a kaleidoscope of analytical
 lenses.

Pre-processing:

In the prelude to the grand symphony of machine learning, we navigate the
preliminary steps of data preprocessing. Our voyage involves the delicate art of
handling missing values, encoding categorical variables, and orchestrating the
harmonious scaling or normalization of features to create a symphony of
meaningful data for machine learning.

Machine Learning Training/Testing:

 Random Forest Prediction: The marquee event of our methodological extravaganza unfolds with the deployment of the Random Forest algorithm, an ensemble learning method that begets a "forest" of decision trees. The training process, infused with a judicious dose of randomness, births a diverse array of decision trees, enhancing model robustness and orchestrating a ballet that mitigates the specter of overfitting.

EDA Highlights:

• Shot Type Distribution: Our visual symphony takes a dramatic turn, depicting the distribution of Butler's shot types. This visual narrative unveils a predilection for



- certain shot types, such as jump shots and layups, offering a tantalizing glimpse into the intricate dance of his offensive strategy.
- **3-Point vs. Mid-Range Shots:** Another act in our visual spectacle showcases the distribution between 3-point and mid-range shots, unfurling the canvas of Butler's proficiency in different shooting categories.
- Shooting Range Analysis: The canvas expands further with histograms and boxplots, offering a panoramic view of Butler's shooting range. This visual panorama underscores the critical importance of understanding shot distances in the rich tapestry of player performance.

Machine Learning Predictive Model:

 Random Forest Classifier: The pièce de résistance in our methodological saga, the Random Forest Classifier, takes center stage. Its performance, akin to a virtuoso performance, yields promising results in predicting Butler's shot outcomes. Through meticulous feature importance analysis, we identify the instrumental elements that orchestrate successful shots.

Conclusions and future works

In the grand finale of our project, the curtain falls, and we draw the threads of our analysis into a harmonious conclusion. The developed predictive models, crafted with precision and insight, offer a robust framework for the evaluation and forecasting of player performance. These insights, akin to a compass, guide teams, enabling them to tailor strategies steeped in the rich tapestry of historical performance data.

As we cast our gaze towards the future, the melody of future works beckons. The symphony of refinement calls for the incorporation of additional features, the exploration of advanced machine learning techniques, and an expansion of our analysis to include datasets from other players. This expansive vision seeks to create a more comprehensive understanding of the captivating world of basketball analytics, where each player's narrative enriches the collective saga.