

Introduction

Background

About NBA:

- NBA (National Basketball Association)
- Prestigious Professional Basketball leagues

About Career Trajectory:

- Player Performance Statistics
- Game Records and Historical Data



Introduction

We want to predict the future career development of NBA players, which is a complex task that involves analyzing a variety of factors, both quantitative and qualitative. There are some different aspects to measure their development.

- More suitable position in a team
- Performance in subsequent games
- Expected retirement age
- Future influence in the league





Player's performance data in contests

• Basic Statistics: Points, rebounds, assists, steals, blocks, field goal percentage, three-

point percentage, free throw percentage, etc.

SEASON LEADERS		See All P		II Player St	
POINTS PER GAME		REBOUNDS PER GAME		ASSISTS PER GAME	
1. Shai Gilgeous-Alexander OKC	32.6	1. Domantas Sabonis SAC	14.1	1. Trae Young ATL	11.
2. Giannis Antetokounmpo MIL	30.8	2. Karl-Anthony Towns NYK	13.4	 Nikola Jokić DEN 	10.
3. Nikola Jokić DEN	28.9	28.9 3. Nikola Jokić DEN 12.7 3. (3. Cade Cunningham DET	9.
4. Anthony Edwards MIN	27.4	4. Ivica Zubac LAC	12.5	4. Tyrese Haliburton IND	8.
5. Kevin Durant PHX	26.9	5. Walker Kessler UTA	12.1	5. James Harden LAC	8.
BLOCKS PER GAME		STEALS PER GAME		FIELD GOAL PERCENTAGE	
1. Victor Wembanyama SAS	3.8	1. Dyson Daniels ATL	3.0	1. Walker Kessler UTA	72.
 Walker Kessler UTA 	2.3	2. Shai Gilgeous-Alexander OKC	1.8	2. Jarrett Allen CLE	71.
3. Brook Lopez MIL 4. Daniel Gafford DAL	1.9 3. Kris Dunn LAC		1.8	3. Jalen Duren DET	70.3
	1.9	1.9 4. Nikola Jokić DEN		4. Daniel Gafford DAL	69.
5. Myles Turner IND	1.9	5. Cason Wallace OKC	1.8	5. Rudy Gobert MIN	65.
THREE POINTERS MADE		THREE POINT PERCENTAGE		FANTASY POINTS PER GAME	
1. Malik Beasley DET	247	1. Luke Kennard MEM	45.6	Nikola Jokić DEN	63.
2. Anthony Edwards MIN	246	2. Taurean Prince MIL	45.3	Giannis Antetokounmpo MIL	56.
Stephen Curry GSW	235	 Nikola Jokić DEN 	44.3	Victor Wembanyama SAS	54.
4. Tyler Herro MIA	208	4. Ty Jerome CLE	44.1	Shai Gilgeous-Alexander okc	54.
5. Jayson Tatum BOS	206	5. Keon Ellis SAC	43.8	Jayson Tatum BOS	48.



Player's performance data in contests

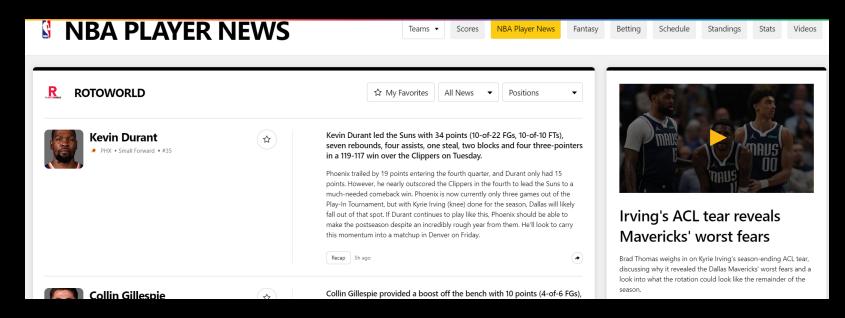
 Advanced Statistics: True Shooting Percentage (TS%), Player Efficiency Rating (PER), Win Shares, Box Plus-Minus (BPM), Value Over Replacement Player (VORP), etc.

TRUE SHOOTING PERCENTAGE		USAGE PERCENTAGE		OFFENSIVE REBOUND %	
Jarrett Allen CLE	72.8	Giannis Antetokounmpo MIL	35.4	Zach Edey MEM	15.0
Jalen Duren DET	71.6	Shai Gilgeous-Alexander окс	33.3	Donovan Clingan POR	14.9
Daniel Gafford DAL	71.6	Cade Cunningham DET	31.8	Kevon Looney GSW	14.8
Walker Kessler UTA	71.4	Anthony Edwards MIN	31.2	Walker Kessler UTA	14.2
Luke Kornet BOS	68.7	Victor Wembanyama SAS	30.0	Jalen Duren DET	13.9



Player's Injury Data

- Injury History: Past injuries, recovery times, and games missed.
- Injury Risk: Playing style, workload, and medical history.





Team and Contextual Data

- Team Performance: Win-loss records, offensive and defensive ratings, and playoff success.
- Role and Usage: Player usage rate, time on court, and role within the team.

				e Stats Lineups Tool Media Cer	
SEASON LEADERS					See All Stat
POINTS PER GAME		REBOUNDS PER GAME		ASSISTS PER GAME	
1. Cleveland Cavaliers	123.3	23.3 2. Memphis Grizzlies 47.5 2. Atlanta Hawks 21.1 3. Denver Nuggets 46.0 3. Indiana Pacers		1. Denver Nuggets	31.1
1. Memphis Grizzlies	123.3			2. Atlanta Hawks	29.4
3. Denver Nuggets	121.1			3. Indiana Pacers	29.2
4. Oklahoma City Thunder	119.6			3. Memphis Grizzlies	29.2
5. New York Knicks	117.4	5. Charlotte Hornets	45.6	3. San Antonio Spurs	29.2
BLOCKS PER GAME		STEALS PER GAME		FIELD GOAL PERCENTAGE	
1. San Antonio Spurs	6.3	1. Oklahoma City Thunder	10.8	1. Denver Nuggets	50.8
2. Orlando Magic	6.1	2. Atlanta Hawks	10.0	2. Cleveland Cavaliers	49.7
3. Dallas Mavericks	5.9	3. LA Clippers	9.8	3. Indiana Pacers	49.2
3. Memphis Grizzlies	5.9	4. Philadelphia 76ers	9.4	3. New York Knicks	49.2
5. Oklahoma City Thunder	5.7	5. New Orleans Pelicans	9.1	5. Memphis Grizzlies	48.4
THREE POINTERS MADE		THREE POINT PERCENTAGE		FREE THROW PERCENTAGE	
1. Boston Celtics	1084	1. Cleveland Cavaliers	39.4	1. Oklahoma City Thunder	82.4
2. Cleveland Cavaliers	991	2. Milwaukee Bucks	38.2	2. Sacramento Kings	81.0
3. Chicago Bulls	979	3. Denver Nuggets	38.1	3. Phoenix Suns	80.6
4. Golden State Warriors	948	4. Phoenix Suns	37.8	4. Chicago Bulls	80.4



Methodology

- Objective:
 - Predict NBA players' career trajectories based on historical performance data
- Approach:
 - Combine time-series forecasting, regression models, and clustering techniques to analyze career progression.
- Key Components:
 - Data Collection & Preprocessing
 - Feature Engineering
 - Predictive Modeling
 - Model Evaluation & Interpretation



Data Collection & Preprocessing

- Data Sources: Kaggle NBA database
 - 30 teams
 - 4800+ players
 - 65,000+ games (every game since the inaugural 1946-47 NBA season)

MIN: Minutes

PTS: Points

OR: Offensive Rebounds

FGM-A: Field Goals Made-Attempted

FG%: Field Goal Percentage

3PM-A: 3-Point Field Goals Made-Attempted

3PM-A: 3-Point Field Goal Percentage

BLK: Blocks

FTM-A: Free Throws Made-Attempted

STL: Steals

• Data Preprocessing: Data Cleaning, Data Integration, Data Transformation



Feature Engineering

- Core Features for Prediction:
 - Player Performance: FG% (Field goal percentage), 3P% (3-point shooting percentage), FT% (Free throw percentage), TRB, AST, PTS, GmSc
- Creating New Features:
 - Recent Trends: 5-game rolling averages, Exponentially Weighted Averages (EWA)
 - Career Trends: Cumulative statistics over seasons (e.g., Career PTS, Career GmSc)

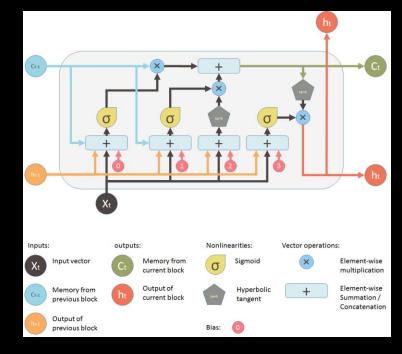


Predictive Modeling

Time-Series Forecasting (Predicting Future Performance)

 LSTM (Long Short-Term Memory Networks): Capture career trends by analyzing player performance over multiple seasons and predicting

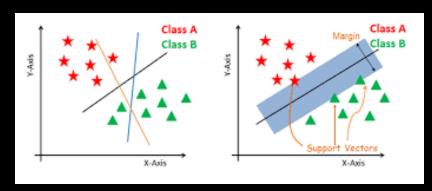
future trajectories.





Predictive Modeling

- Regression Models (Predicting Long-Term Performance)
 - XGBoost: Use performance metrics (e.g., FG%, 3P%, PTS) and historical features (e.g., WAR/82, career averages) to predict long-term performance.
- Clustering & Classification (Career Stage Identification)
 - SVM (Support Vector Machine): Classify players into "emerging stars," "consistent performers," or "declining veterans".





Model Evaluation & Interpretation

- Evaluation Metrics:
 - For Time-Series Models: RMSE (Root Mean Squared Error), MAE (Mean Absolute Error)
 - For Regression Models: R², MSE (Mean Squared Error)
 - For Classification Models: Accuracy, F1–Score to evaluate career stage classification
- Visualizing Career Trajectories:
 - Player Performance Trends: Plot the career trajectory of key players (e.g., average points per game, Game Score, FG%)
 - Comparative Analysis: Compare young players to past legends by analyzing the progression curves for players with similar stats or backgrounds



Expected Results

1. Career Longevity Prediction:

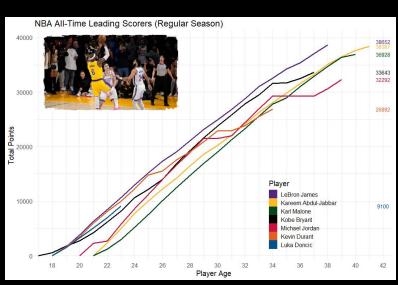
- statistical estimates of how long a player will remain based on factors like injury history, and performance metrics.

2. Role Transition Probabilities:

- Likelihood of players shifting roles (e.g., star \rightarrow role player, rookie \rightarrow bench) based on early-career stats.

3. Career Trajectory Summary

- Plot describing players' performance over the age axis



Weekly Plans

Week 6-7	Week 8-9	Week 10-11	Week 12-13
Data Gathering and Processing	Exploratory Analysis & Feature Engineering	Model Development & Training	Model Refinement & Final Reporting