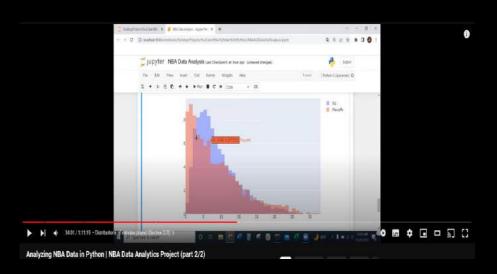




- RQ1: Who is the greatest scorer, playmaker, and defender in NBA history?
- RQ2: What factors impact draft/trade strategies the most?
- RQ3: Has the level of the NBA gotten better or worse?
- RQ4: Which performance stats contribute to determining a team's success the most?

RELATED WORK



⁷ NBA Sports Betting Using Machine Learning 🏀



Cleveland Cavaliers vs Brooklyn Nets (58.7%): UNDER 225.5 (72.3%) Indiana Pacers (73.0%) vs Dallas Mavericks: UNDER 219.5 (58.6%) Philadelphia 76ers (51.5%) vs Boston Celtics: UNDER 221.5 (64.7%) Atlanta Hawks (85.1%) vs Detroit Pistons: UNDER 220 (59.3%) Toronto Raptors vs Miami Heat (52.1%): UNDER 222 (57.0%) Minnesota Timberwolves vs Orlando Magic (58.5%): OVER 218 (50.1%) Houston Rockets vs Phoenix Suns (80.9%): UNDER 218 (62.1%) Golden State Warriors vs San Antonio Spurs (51.7%): OVER 228.5 (50.6%) LA Clippers (98.0%) vs Sacramento Kings: OVER 230 (75.1%) Portland Trail Blazers (59.9%) vs Memphis Grizzlies: UNDER 223 (61.0%)

A machine learning AI used to predict the winners and under/overs of NBA games. Takes all team data from the 2007-08 season to current season, matched with odds of those games, using a neural network to predict winning bets for today's games. Achieves ~69% accuracy on money lines and ~55% on under/overs. Outputs expected value for teams money lines to provide better insight. The fraction of your bankroll to bet based on the Kelly Criterion is also outputted. Note that a popular, less risky approach is to bet 50% of the stake recommended by the Kelly Criterion.

NBA-analytics

Short, offhand analyses of the NBA

Note: All stats are from stats.nba.com, unless noted.

Topics Covered

- Underrated Assisters
- Consistent Players
- Why is Miami the worst team at drives?
- Which teams run the wrong plays?
- Shot Clock Analysis
- Westbrook Rebounding
- Defender distance
- · Aging player's efficiency
- Steal efficiency
- Timeout Conversion
- Rim Protection
- Tanking
- Empirical Bayes estimation of NBA statistics
- · James-Stein estimation of NBA statistics
- Player Efficiency Rating (PER)

DATASET(S)



FIGERIES EVEN

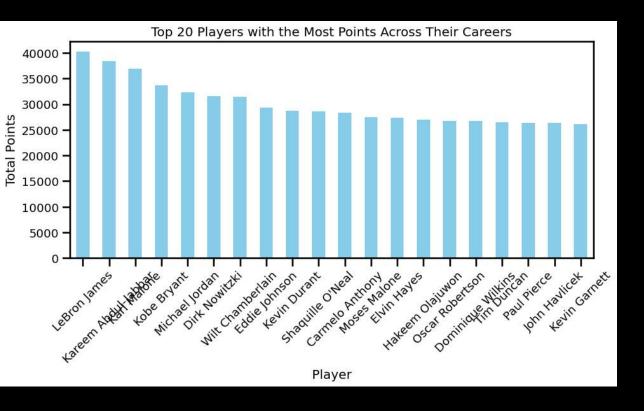
RQ1: WHO IS THE BEST SCORER,
PLAYMAKER, AND DEFENDER OF
ALL TIME?

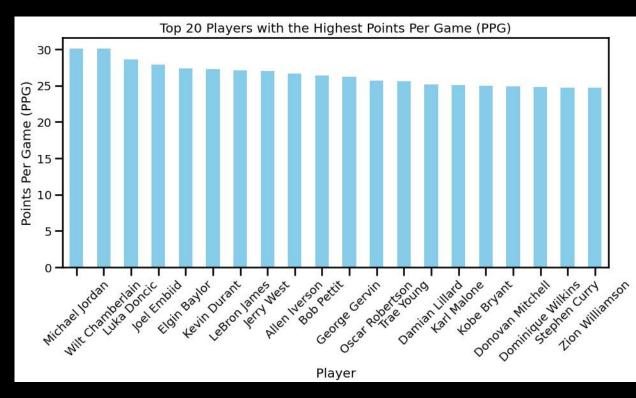
Michael Jordan

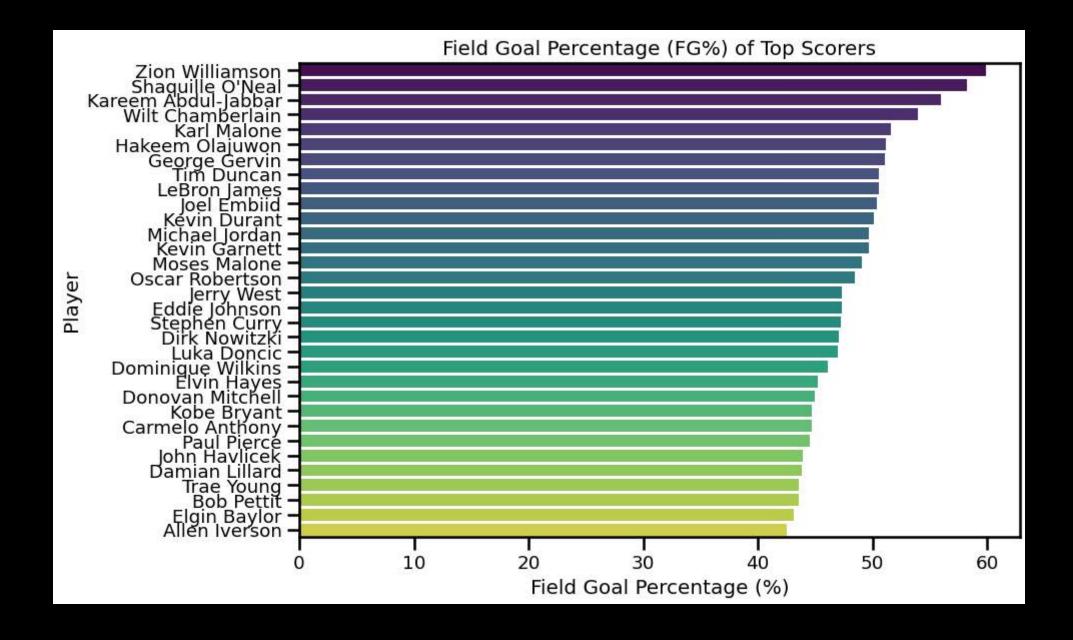
LeBron James Magic Johnson

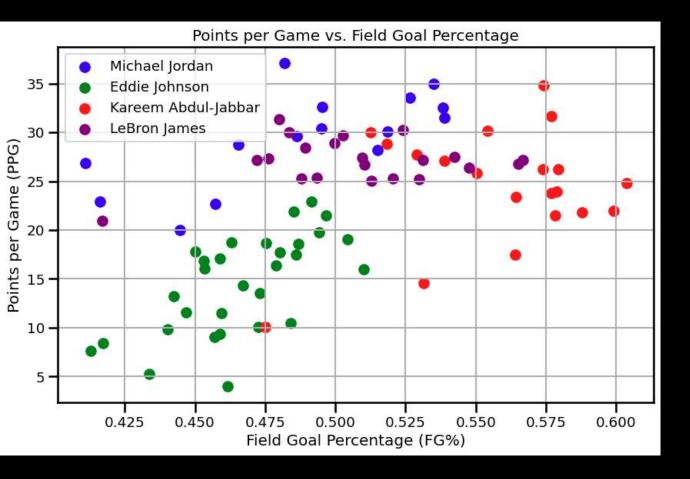
Kareem Abdul-Jabbar

SCORERS



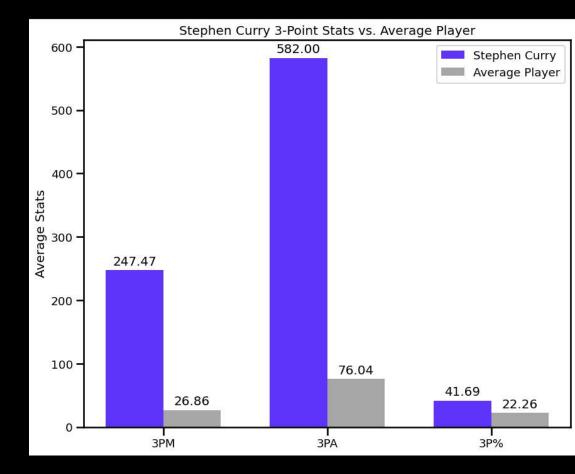




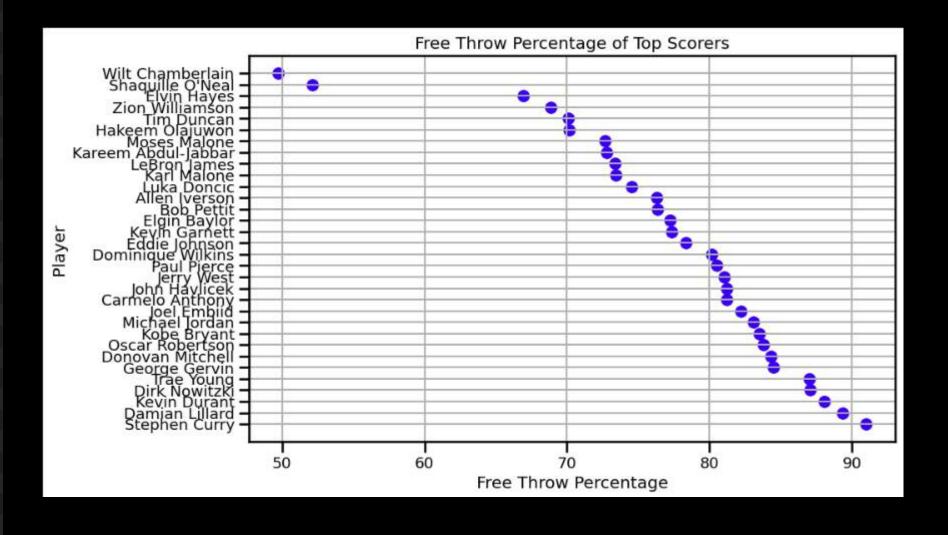




Each dot above represents a season that the player played



	three_point_pct
player_name	
Stephen Curry	0.416933
Kevin Durant	0.395706
Dirk Nowitzki	0.374714
Damian Lillard	0.368750
Paul Pierce	0.368105
Donovan Mitchell	0.366286
Zion Williamson	0.362000
Trae Young	0.352667
LeBron James	0.347143
Carmelo Anthony	0.345300
Allen Iverson	0.344882
Luka Doncic	0.344000
Joel Embiid	0.343750
Kobe Bryant	0.317700
Eddie Johnson	0.297857
Michael Jordan	0.283933
Bob Pettit	0.273773
Oscar Robertson	0.273773
John Havlicek	0.273773
Jerry West	0.273773
Elgin Baylor	0.273773
Wilt Chamberlain	0.273773
Dominique Wilkins	0.273188
George Gervin	0.263000
Kevin Garnett	0.207136
Karl Malone	0.201158
Elvin Hayes	0.146200
Tim Duncan	0.138474
Hakeem Olajuwon	0.100611
Moses Malone	0.100312
Kareem Abdul-Jabbar	0.033300
Shaquille O'Neal	0.025000

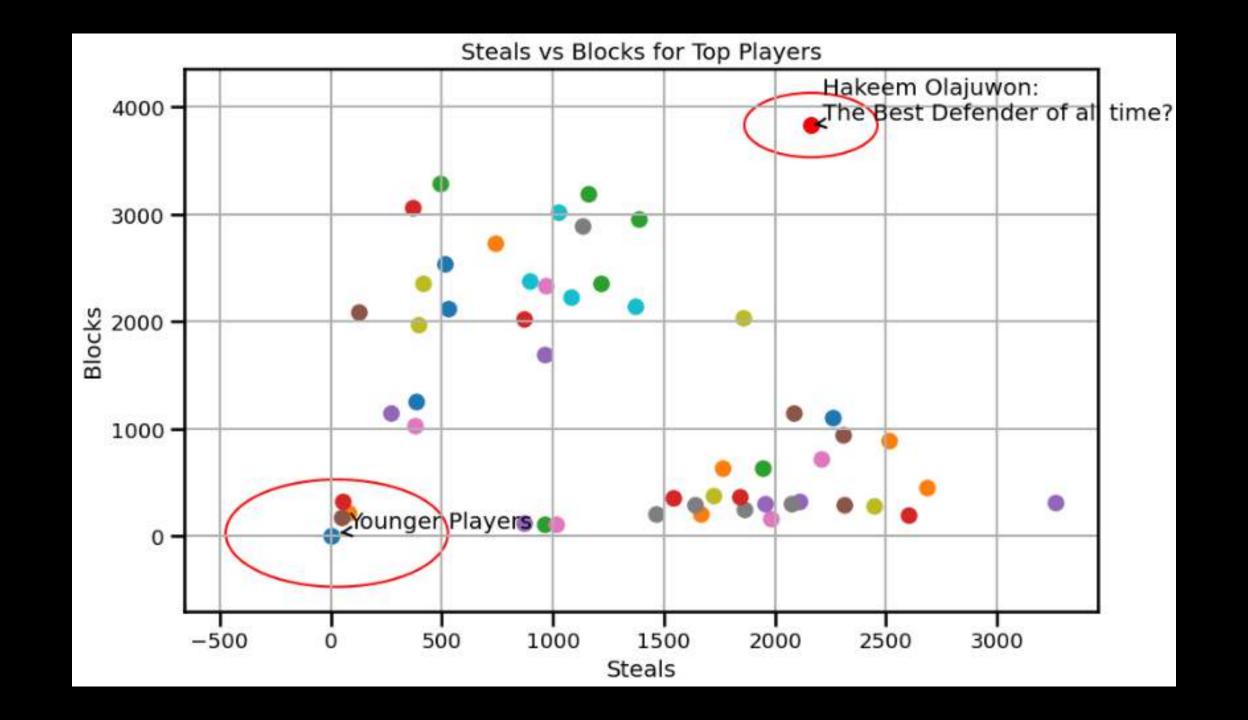


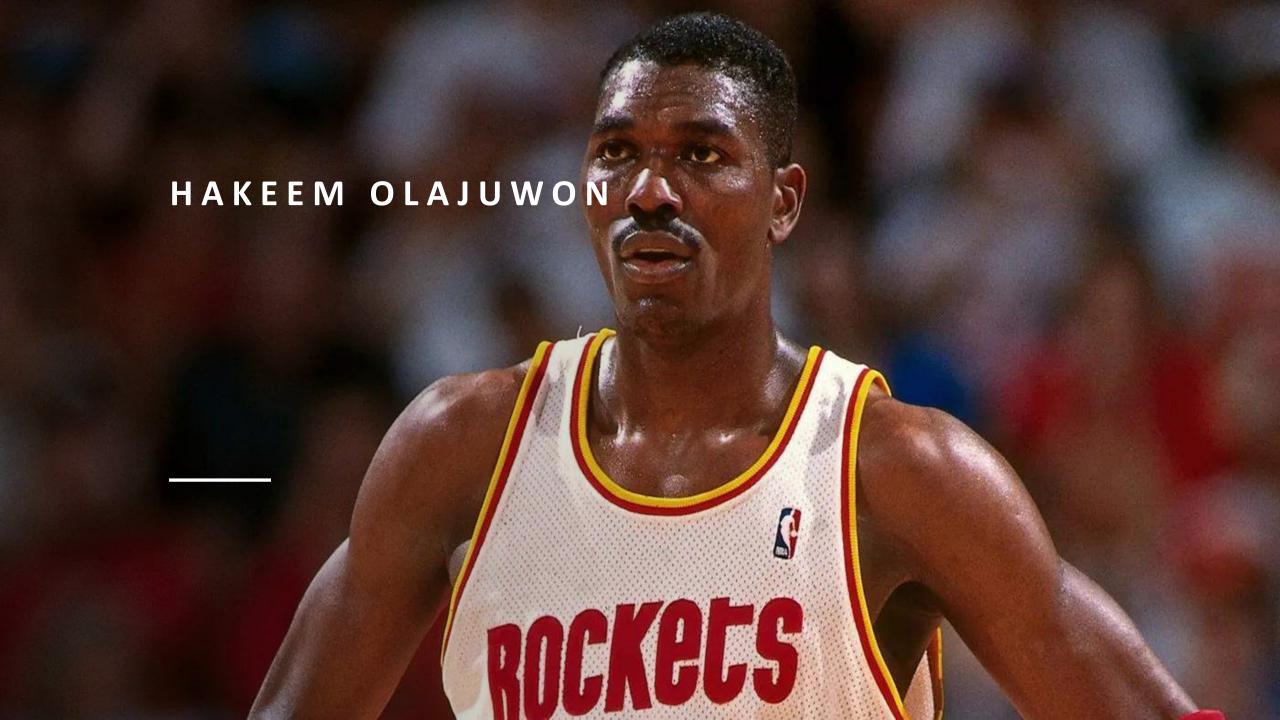




DEFENDERS

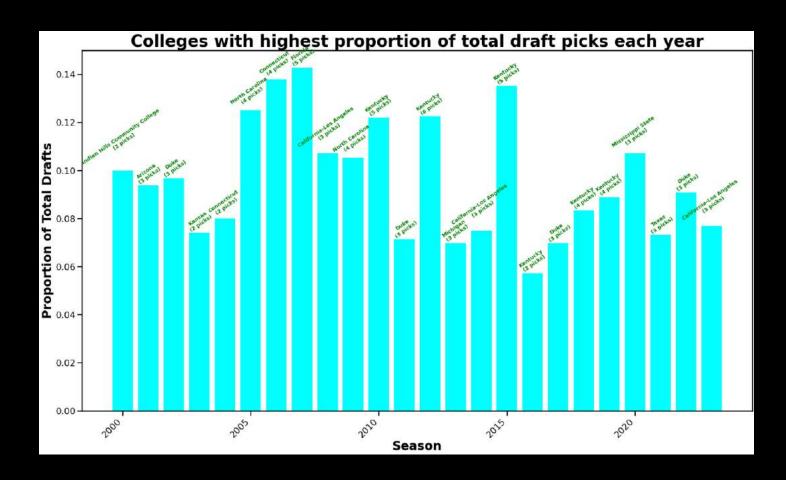
	steals		Player	SPG		blocks			Player	BPG
player_name		0	Alvin Robertson	2.711168	player_name			0	Mark Eaton	3.501714
John Stockton	3265.0	1	Micheal Ray Richardson	2.631295	Hakeem Olajuwon	3830.0		1	Victor Wembanyama	3.430769
Jason Kidd	2684.0	2	Michael Jordan	2.345149	Dikembe Mutombo	3289.0		2	Manute Bol	3.342949
Chris Paul	2600.0	3	Mookie Blaylock	2.334083	Kareem Abdul-Jabbar	3189.0		3	Hakeem Olajuwon	3.093700
Michael Jordan	2514.0	4	Lafayette Lever	2.215426	Mark Eaton	3064.0		4	David Robinson	2.992908
Gary Payton	2445.0	5	Slick Watts	2.199085	Tim Duncan	3020.0		5	Alonzo Mourning	2.811456
Maurice Cheeks	2310.0	6	John Stockton	2.170878	David Robinson	2954.0		6	Dikembe Mutombo	2.750000
Scottie Pippen	2307.0	7	Allen Iverson	2.169584	Patrick Ewing			7	Shawn Bradley	2.546875
LeBron James	2262.0	8	Maurice Cheeks	2.098093	Shaquille O'Neal			8	Patrick Ewing	2.432773
Clyde Drexler	2207.0	9		2.058591	Tree Rollins			9	Theo Ratliff	2.429630
Hakeem Olajuwon	2162.0	10	Clyde Drexler		George Johnson		1	0	Chet Holmgren	2.397260
Alvin Robertson	2112.0	11	Ralph Jackson		Robert Parish		1	1	Marcus Camby	2.395683
Karl Malone	2085.0	12	Renaldo Major		Alonzo Mourning			2	Walker Kessler	2.375000
Mookie Blaylock	2075.0	13		1.985455	Marcus Camby			3	Anthony Davis	2.319616
Allen Iverson	1983.0	14	Scottie Pippen	1.958404	Dwight Howard			4	Shaquille O'Neal	2.263463
Derek Harper	1957.0				Ben Wallace			5	Hassan Whiteside	2.236791
Kobe Bryant	1944.0	15		1.955769	Shawn Bradley					
Isiah Thomas	1861.0	16		1.939732	Manute Bol			6 -	Myles Turner	
Kevin Garnett	1859.0	17	Nate McMillan	1.939698	Kevin Garnett			7	Bill Walton	2.209402
Russell Westbrook	1841.0	18	Jason Kidd	1.929547	Larry Nance		1	8	Larry Nance	2.203261
Andre Iguodala	1765.0	19	Magic Johnson	1.902870	Theo Ratliff	1968.0	1	9	Tree Rollins	2.198962



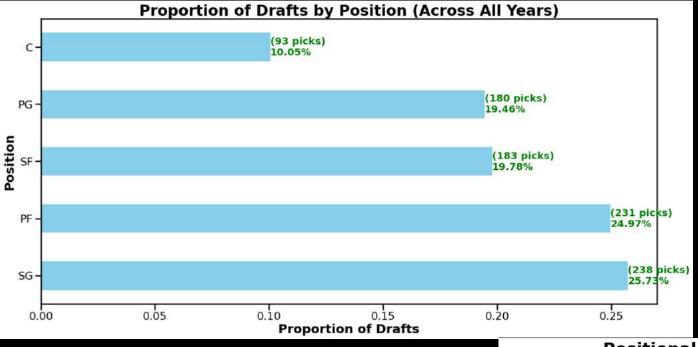




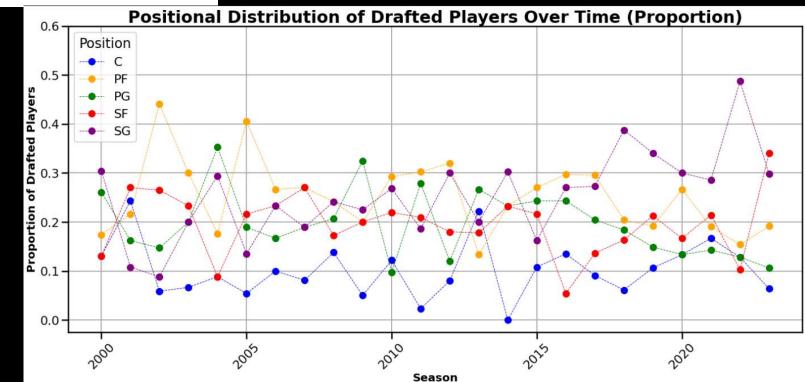
DO TEAMS ONLY DRAFT FROM CERTAIN COLLEGES?



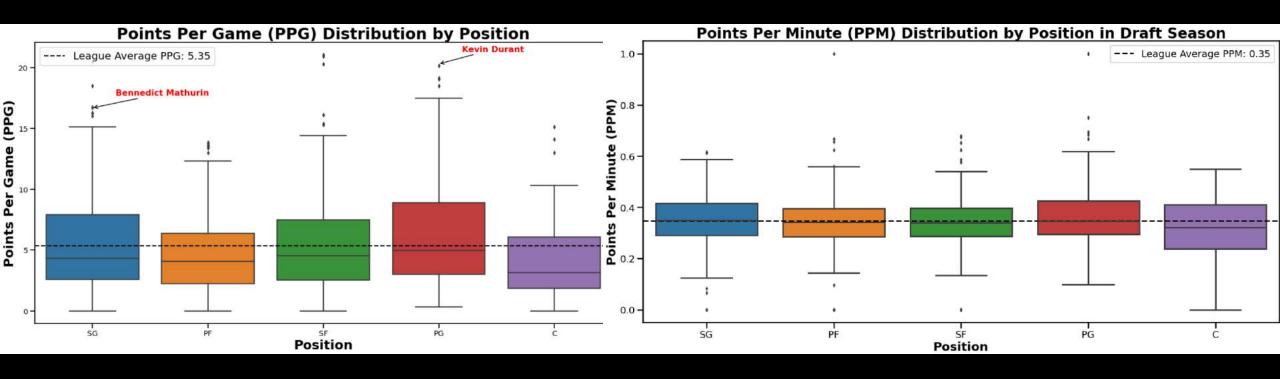
Kentucky	6
Duke	4
California-Los Angeles	3
Connecticut	2
North Carolina	2
Indian Hills Community College	1
Arizona	1
Kansas	1
Florida	1
Michigan	1
Mississippi State	1
Texas	1

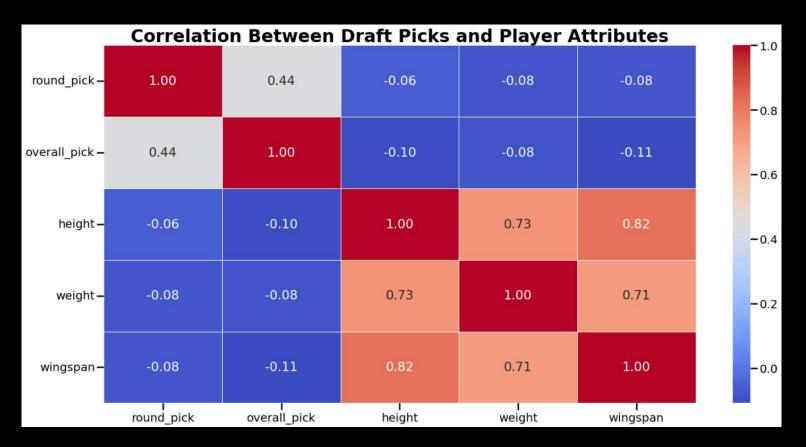


ARE SOME POSITIONS
MORE POPULAR THAN
OTHERS?



AVERAGE PERFORMANCE OF PLAYERS FROM EACH POSITION

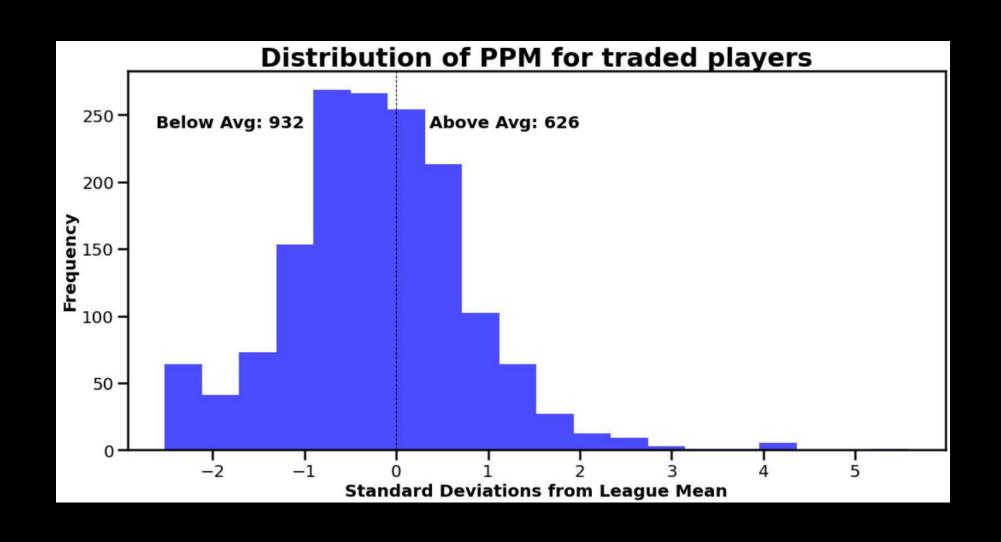




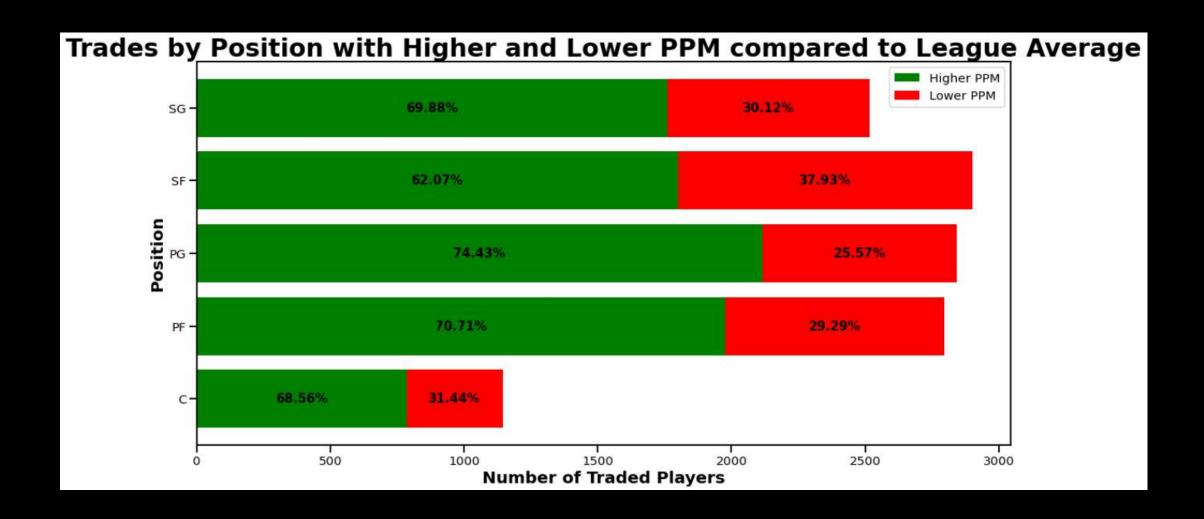
CONCLUSIONS:

- 1) colleges alone may not be sufficient indicators for optimizing draft strategies.
- 2) Positional analysis helps teams identify undervalued talent, maximize on-court efficiency, and build a balanced roster capable of competing at a high level in the NBA.
- 3) while physical attributes may be considered during the evaluation process, they do not necessarily dictate a player's draft/round pick

ARE PLAYERS PERFORMING GOOD OR BAD TRADED?



DISTRIBUTION OF TRADES BY POSITION



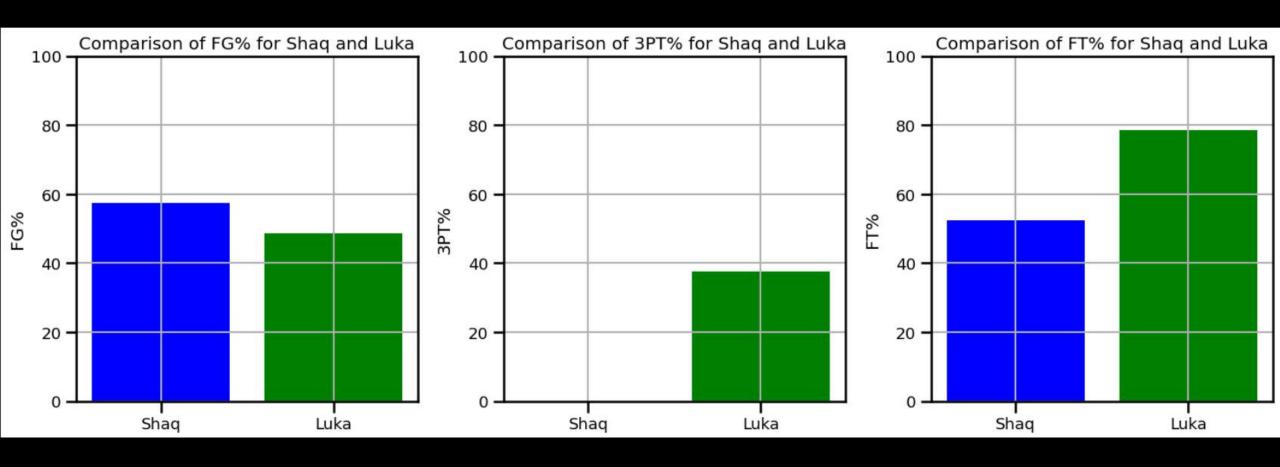
CONCLUSION

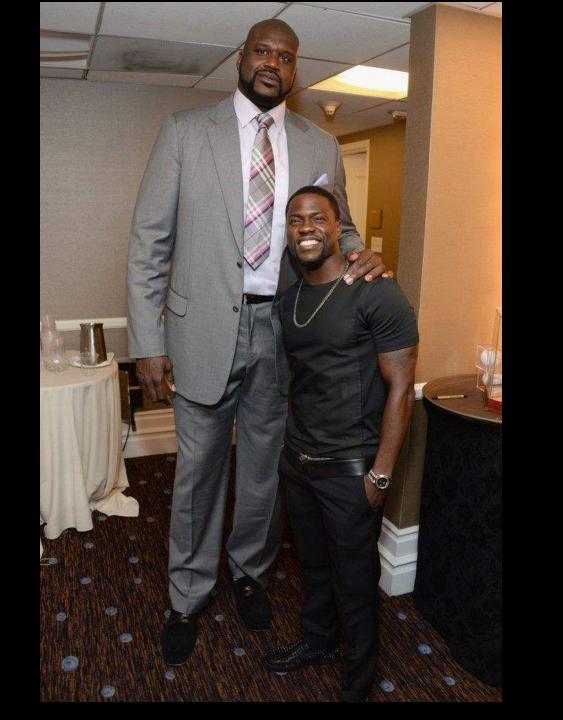
- 1) Contrary to expectations, a significant proportion of players traded to different teams have a PPM higher than the league average.
- 2) Centers (C) consistently emerge as the least traded position, while Small Forwards (SFs) and Point Guards (PGs) are the most frequently traded positions
- 3) Other Considerations:
 - ->Team Chemistry
 - ->Model to predict trade outcomes

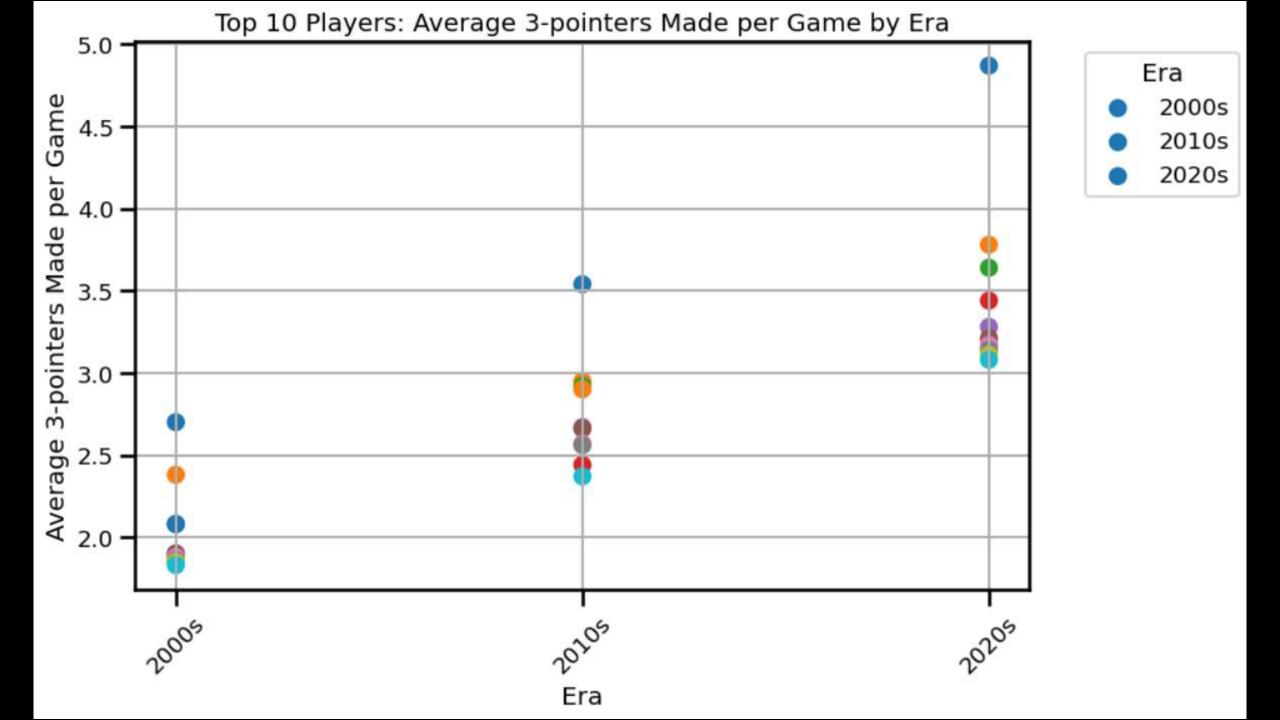


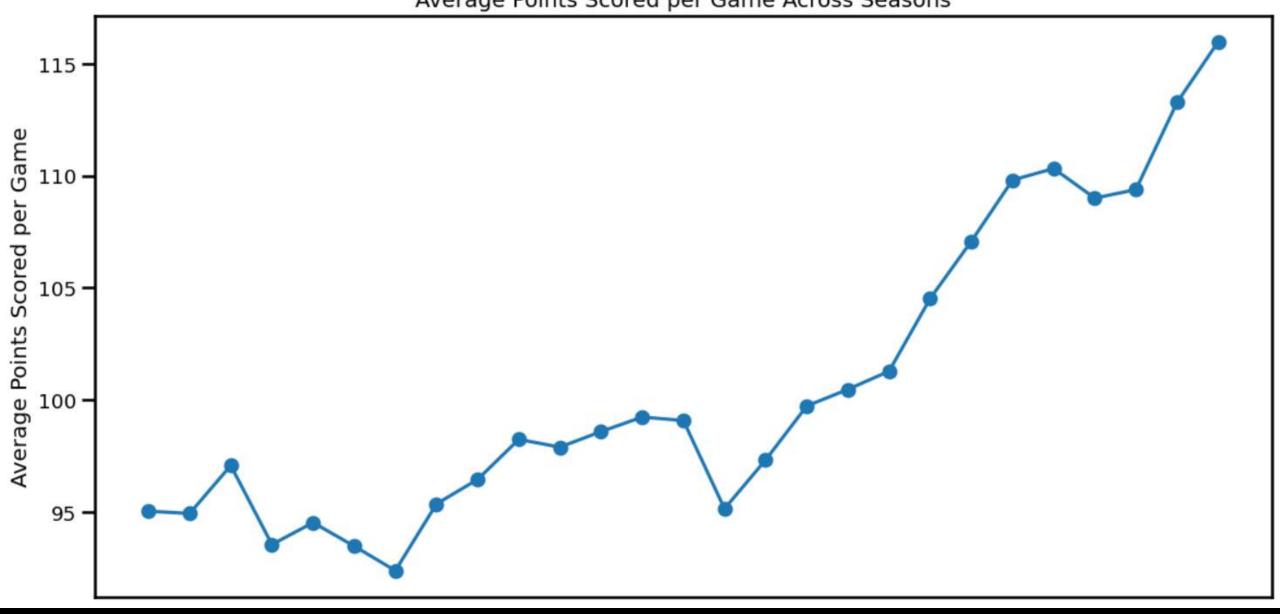


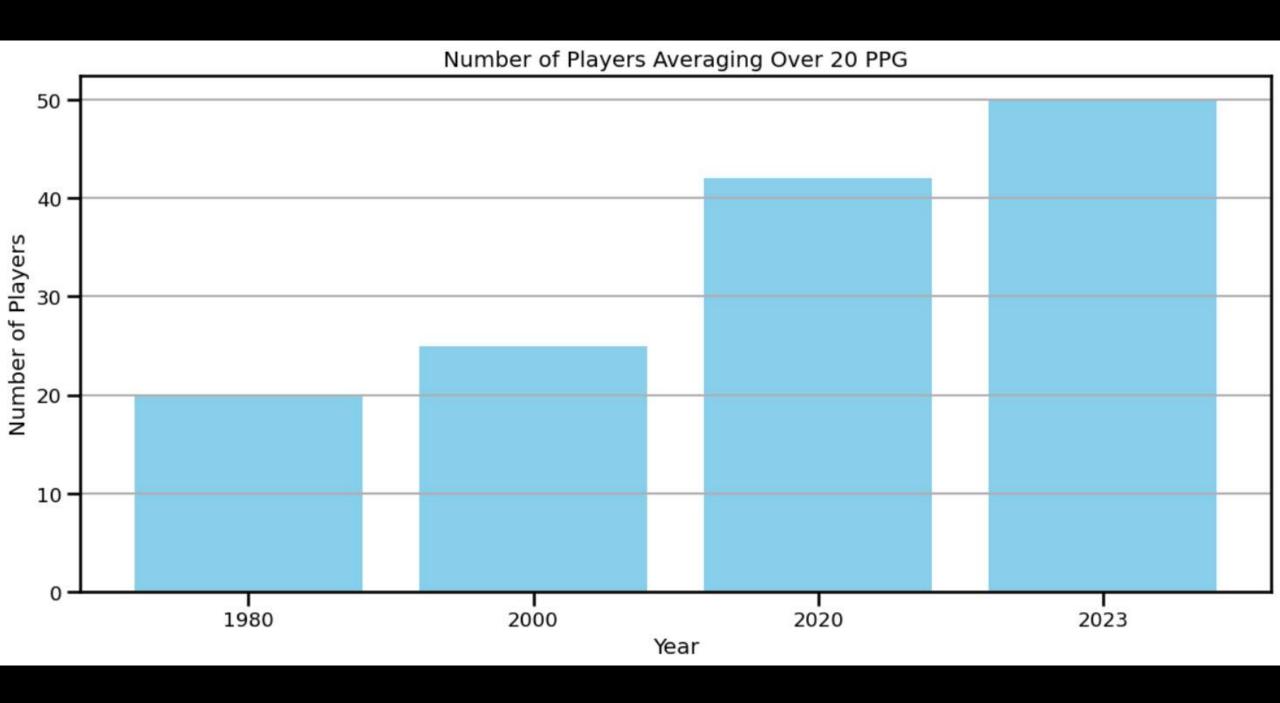
SCORING

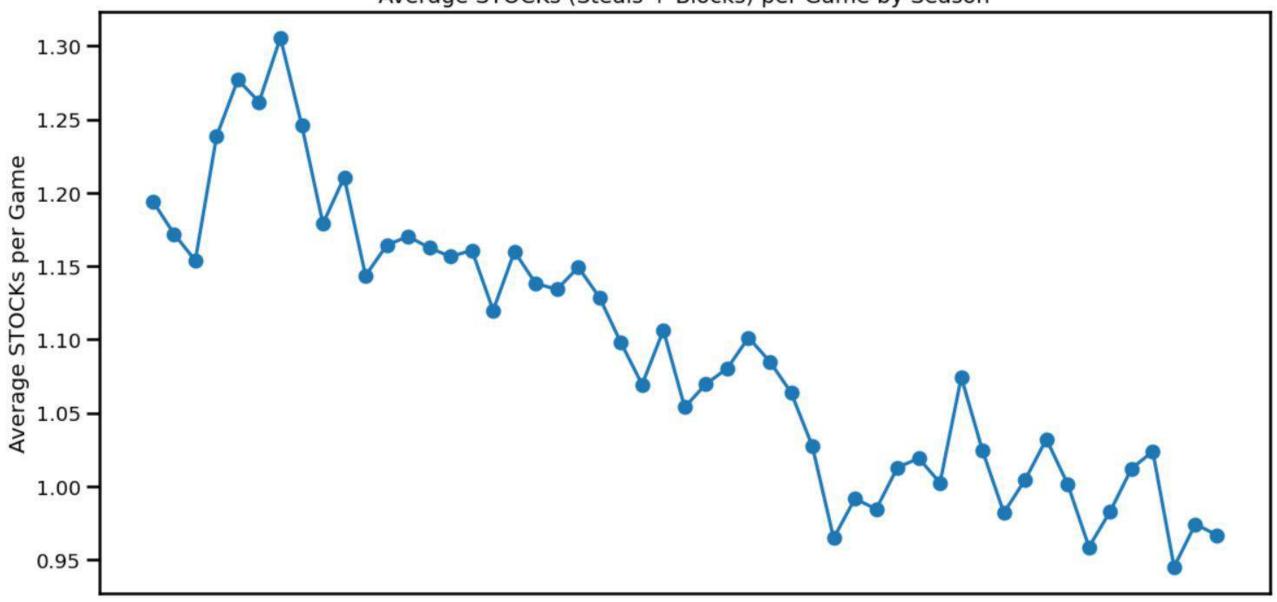












STEVE NASH CASE STUDY

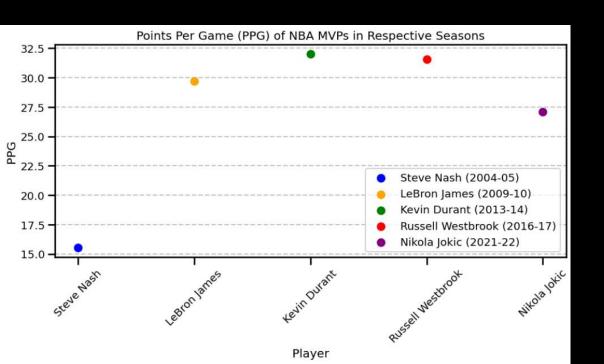
Steve Nash's PPG in the 2004-05 season: 15.533333333333333

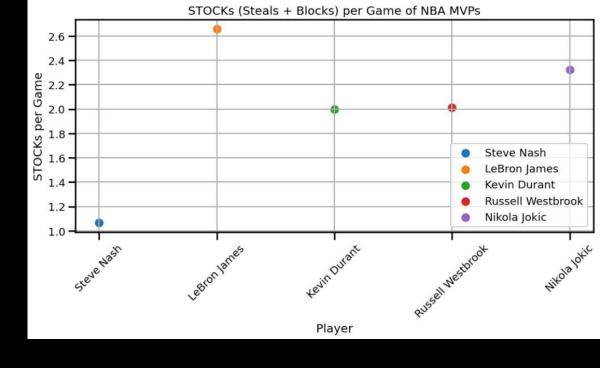
Steve Nash's FG% in the 2004-05 season: 50.2 %

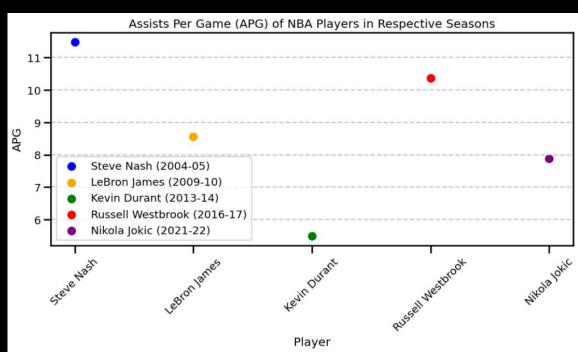
Average PPG of the top 5 scorers in 2004-05 season: 27.803807446939025 Average FG% of the top 5 players by PPG in the 2004-05 season: 45.0 %

Steve Nash's 3PT% in the 2004-05 season: 43.1 %

Average 3PT% of the top 5 players by PPG in the 2004-05 season: 36.44 %







Number of players with higher PPG than Steve Nash's 2005 season: 2010: 67 2015: 62 2020: 82 2024: 92

Number of players with higher FG% and higher PPG than Steve Nash's 2005 season: 2010: 16 2015: 7 2020: 15 2024: 26

Number of players with higher APG than Steve Nash's 2005 season:

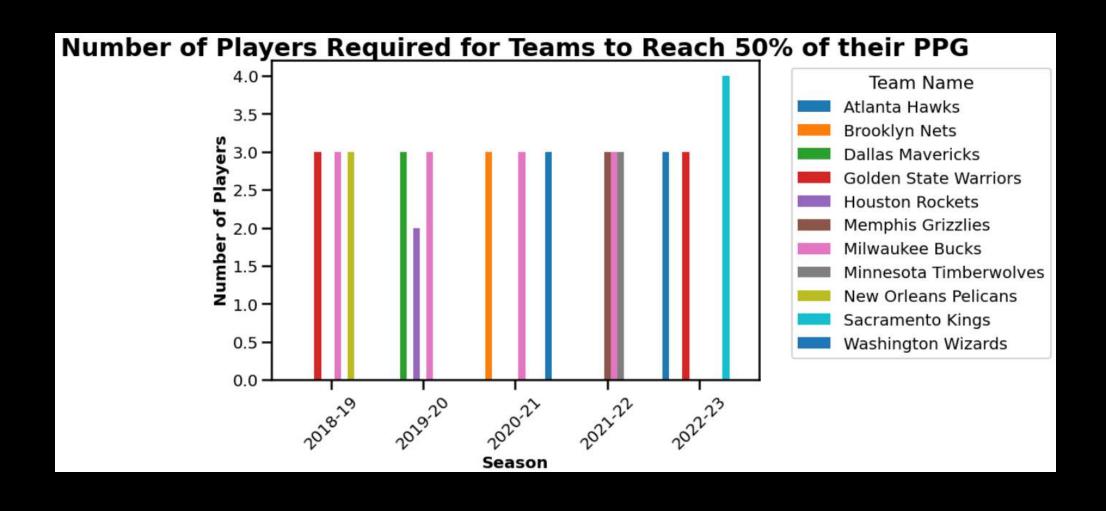
2010: 0 2015: 0 2020: 0 2024: 0 Number of players with higher AST/TO ratio than Steve Nash's 2005 season (minimum 5 APG): 2010: 3
2015: 3
2020: 1
2024: 12

Number of players with higher STOCKs than Steve Nash's 2005 season:

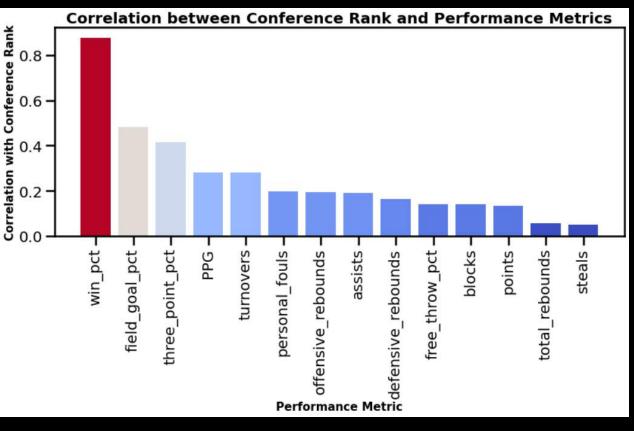
2010: 204 2015: 215 2020: 239 2024: 249



DO ALL PLAYERS CONTRIBUTE EQUALLY TO TEAM'S SUCCESS?



CAN WE PREDICT A TEAM'S SUCCESS?



- As expected, 'win_pct', 'field_goal_pct' and 'three_point_pct' seem to have the strongest correlation with 'conference_rank'
- surprising to see features like 'PPG' and 'points' have such a low correlation

Mean Absolute Error: 1.344382439772509

Mean Squared Error: 2.937034725051422

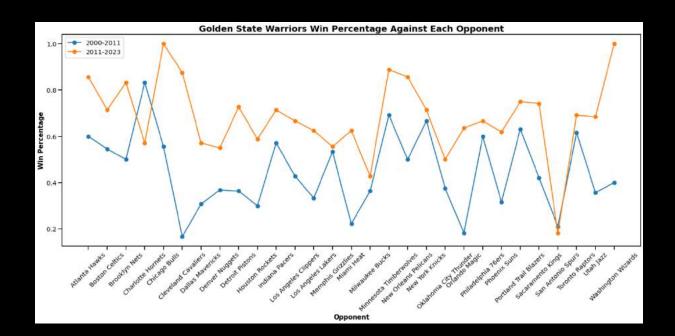
R-squared: 0.8418432518256858

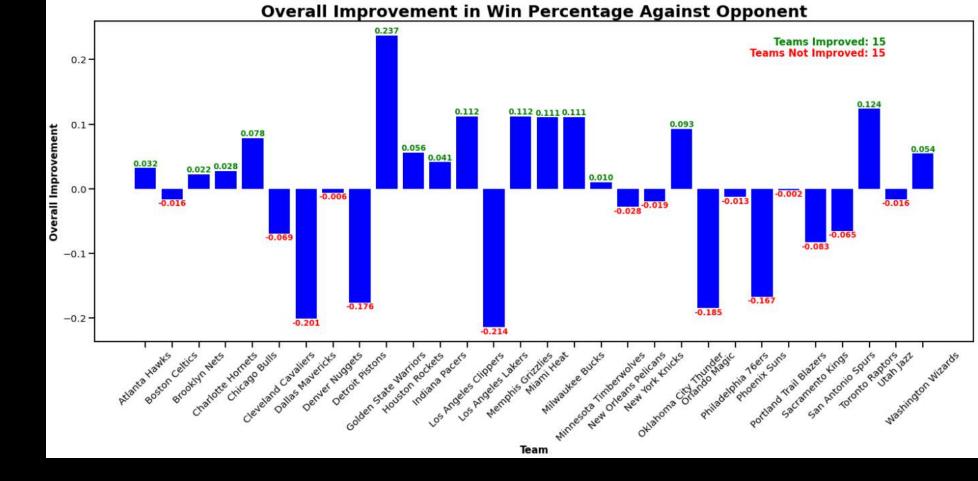
- Overall, the evaluation metrics suggest that the linear regression model performs reasonably well
- However, there are still a few limitations to consider

TEAM PERFORMANCE AGAINST SPECIFIC OPPONENTS

	team_performance_2000_to_2011 <pre></pre>									
	team_name	opponent_name	games_played	wins	losses	avg_points_scored	avg_field_goal_pct	win_pct	loss_pct	
	Atlanta Hawks	Boston Celtics	13		8	95.846154	0.451538	0.384615	0.615385	
	Atlanta Hawks	Brooklyn Nets	15		12	94.000000	0.441067	0.200000	0.800000	
2	Atlanta Hawks	Charlotte Hornets	19	14		95.473684	0.466789	0.736842	0.263158	
	Atlanta Hawks	Chicago Bulls	11			98.090909	0.433364	0.545455	0.454545	
4	Atlanta Hawks	Cleveland Cavaliers	19	8	11	92.578947	0.452368	0.421053	0.578947	
868	Washington Wizards	Portland Trail Blazers	9	4		88.444444	0.434000	0.444444	0.555556	
869	Washington Wizards	Sacaramento Kings		2	4	97.166667	0.431333	0.333333	0.666667	
870	Washington Wizards	San Antonio Spurs	11	4	7	91.181818	0.434818	0.363636	0.636364	
871	Washington Wizards	Toronto Raptors	18	10	8	96.333333	0.444667	0.555556	0.44444	
872	Washington Wizards	Utah Jazz	4			95.750000	0.407250	0.750000	0.250000	
873 ro	ws × 9 columns									

	1.0s								
	team_name	opponent_name	games_played	wins	losses	avg_points_scored	avg_field_goal_pct	win_pct	loss_pct
	Atlanta Hawks	Boston Celtics	12			102.666667	0.458750	0.416667	0.583333
	Atlanta Hawks	Brooklyn Nets	16		11	108.750000	0.446688	0.312500	0.687500
2	Atlanta Hawks	Charlotte Hornets	25	15	10	105.240000	0.458600	0.600000	0.400000
	Atlanta Hawks	Chicago Bulls	16	6	10	100.687500	0.437500	0.375000	0.625000
4	Atlanta Hawks	Cleveland Cavaliers	25	14	11	109.320000	0.467040	0.560000	0.440000
865	Washington Wizards	Portland Trail Blazers	5		2	105.800000	0.447200	0.600000	0.400000
866	Washington Wizards	Sacaramento Kings			4	111.222222	0.481889	0.555556	0.444444
867	Washington Wizards	San Antonio Spurs	10		8	101.700000	0.455400	0.200000	0.800000
868	Washington Wizards	Toronto Raptors	18		15	98.888889	0.430944	0.166667	0.833333
869	Washington Wizards	Utah Jazz			4	96.857143	0.444286	0.428571	0.571429





Conclusions:

- Team success in sports is intricately tied to star player performance, team metrics, and performance against specific opponents
- While these factors certainly do influence a team's success, there are additional elements that may impact outcomes

