

NBA Shot Selection



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01

Research Question + Motivation

How has the distribution and efficiency of NBA shot selection evolved across eras, and what do player clustering patterns reveal about the most optimal team compositions in each era?

- Player tendencies vary widely — some specialize in restricted area finishes, others thrive on midrange pull-ups or catch-and-shoot 3s.
- Team strategies often emphasize certain zones (e.g., Moreyball: 3s and layups), creating tension or synergy between a player's habits and a team's system.
- Efficiency tradeoffs: A player shooting from their comfort zone may be more efficient even if it deviates from team norms — but alignment can lead to more seamless offensive execution.
- Clustering analysis helps us identify groups of players based on shot profiles (e.g., paint-heavy slashers vs. perimeter snipers), and assess which groups yield higher FG% in specific eras.
- Era-based trends reveal how the value of shot locations has shifted — e.g., a midrange-heavy style in the 2000s vs. modern 3PT-heavy spacing.



02

Data and Methods

Data Sources



HoopR

Package data from the NBA Stats API
Past 30 years (1996-2024)



Basketball Reference

1996 scoring efficiency (team-level) not available
in hoopR

Data frame was manually created for this year

Used in scoring efficiency regression



Data Manipulation

Preprocessing

- Calculate Standardized Field Goal Attempt Frequencies
- Merge team and player data
- Player bias scores were calculated by comparing individual shot frequencies to their team's average shot frequencies



Clustering

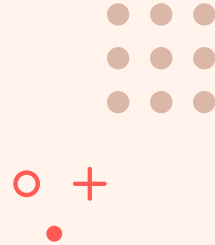
- Player Shot Location (Restricted_Area_FGA, Corner_3_FGA)
- Team Shot Location (Similar to player, but for teams)
- Created for 1996-2024

Regression

- ESPN Team Stats (hoopR) - Scoring Efficiency
- Clustering data was aggregated to determine count of player types per team (ex: num_paint_players)
- Scoring efficiency regressed count of players based on cluster



Key Variables



Shot Frequency Variables

Restricted_Area_Freq_Player = $\text{Restricted_Area_FGA} / \text{Total_FGA}$

Paint_Freq_Player = $\text{In_The_Paint_Non_RA_FGA} / \text{Total_FGA}$

Mid_Range_Freq_Player = $\text{Mid_Range_FGA} / \text{Total_FGA}$

Corner_3_Freq_Player = $(\text{Left_Corner_3_FGA} + \text{Right_Corner_3_FGA}) / \text{Total_FGA}$

Above_Break_3_Freq_Player = $\text{Above_the_Break_3_FGA} / \text{Total_FGA}$

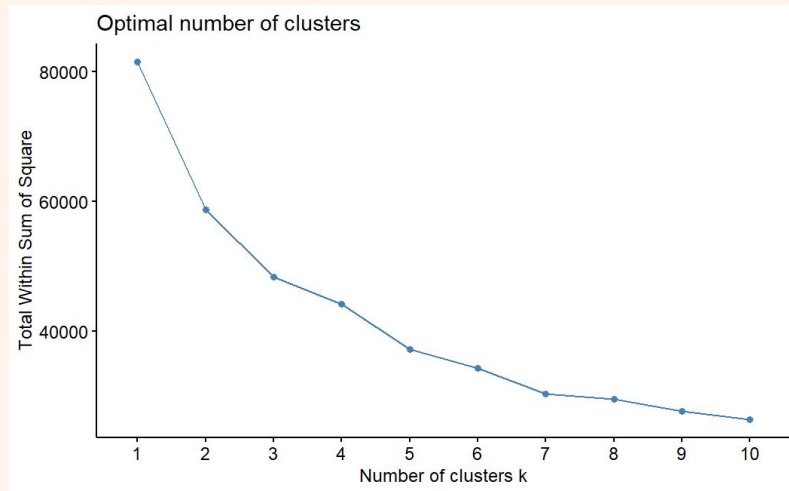
Backcourt_Freq_Player = $\text{Backcourt_FGA} / \text{Total_FGA}$



Methodology

Clustering Players by Shot Selection Preferences

- Players grouped into clusters using K-Means based on their shot frequency profiles
- Used elbows to determine the optimal number of clusters
- Clusters represented distinct player types



Methodology

Multi-Season Analysis

- Comparisons across different eras in the NBA
- Statistical significance and directionality of coefficients were interpreted from regressions



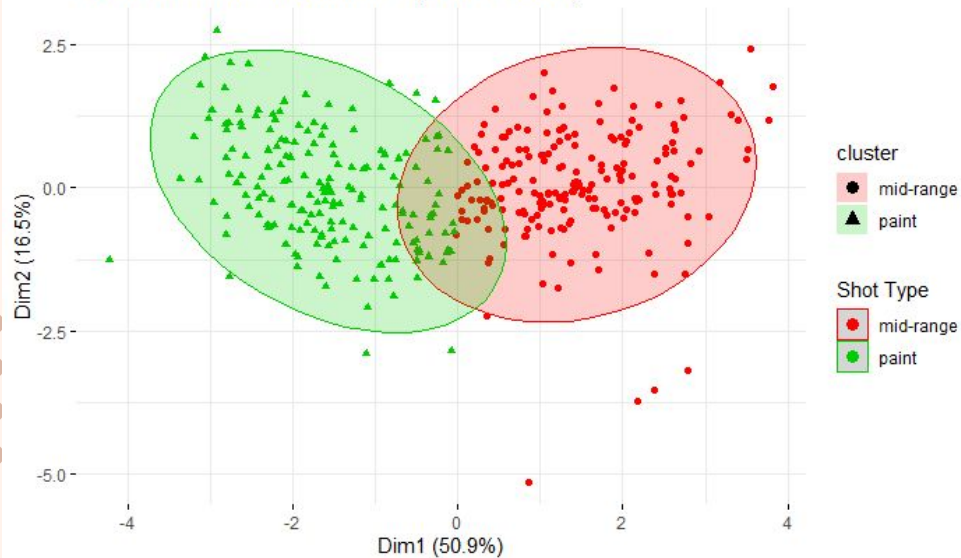
03

Results



1996

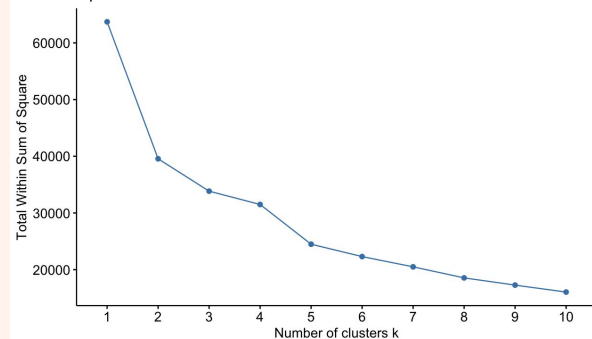
K-Means Cluster Visualization (1996 Season)



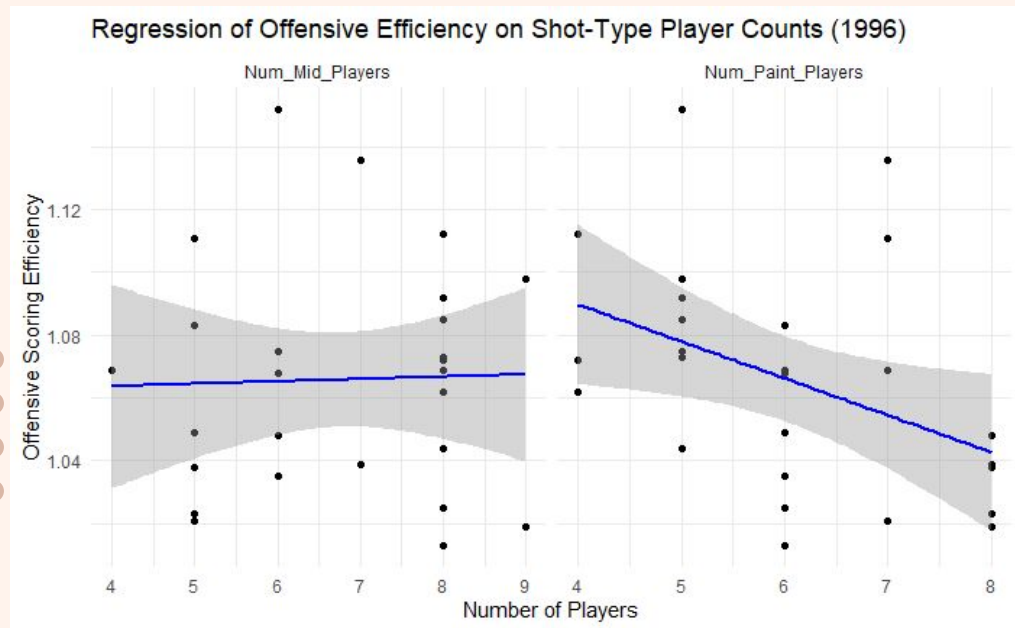
Average Shot Frequency by Cluster (1996)



Optimal number of clusters



Scoring Efficiency - Regression (1996)



Residuals:

Min	1Q	Median	3Q	Max
-0.046796	-0.026016	-0.001939	0.010251	0.085467

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	1.190144	0.060322	19.730	6.5e-16 ***
Num_Paint_Players	-0.014604	0.005808	-2.515	0.0194 *
Num_Mid_Players	-0.005341	0.005151	-1.037	0.3106

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

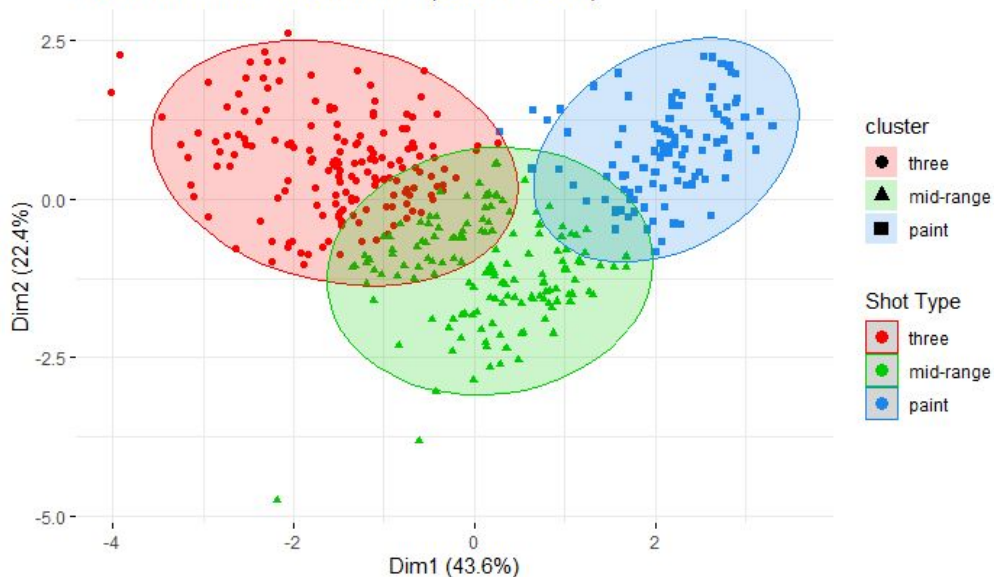
Residual standard error: 0.0336 on 23 degrees of freedom
(3 observations deleted due to missingness)

Multiple R-squared: 0.2164, Adjusted R-squared: 0.1482

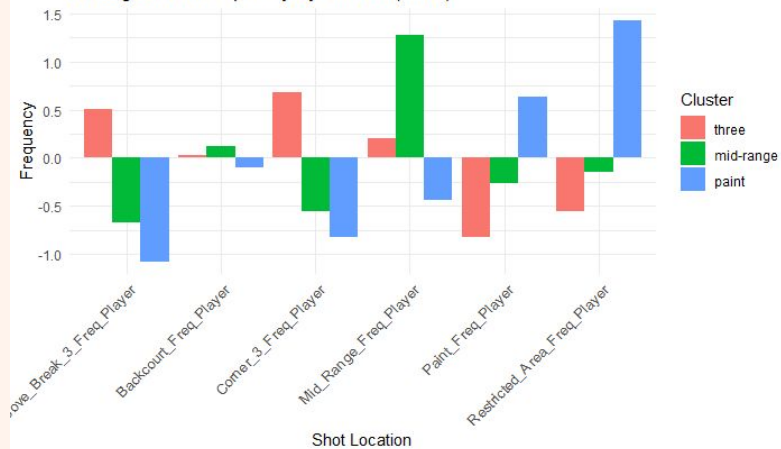
F-statistic: 3.175 on 2 and 23 DF, p-value: 0.06059

2006

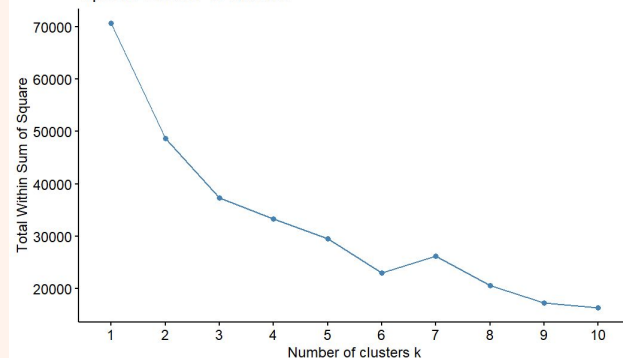
K-Means Cluster Visualization (2006 Season)



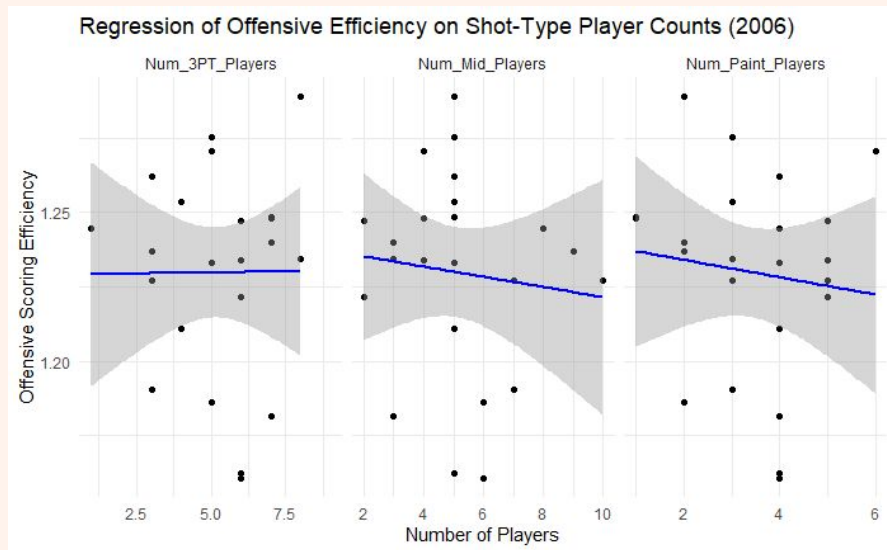
Average Shot Frequency by Cluster (2006)



Optimal number of clusters



Scoring Efficiency Regression (2006)



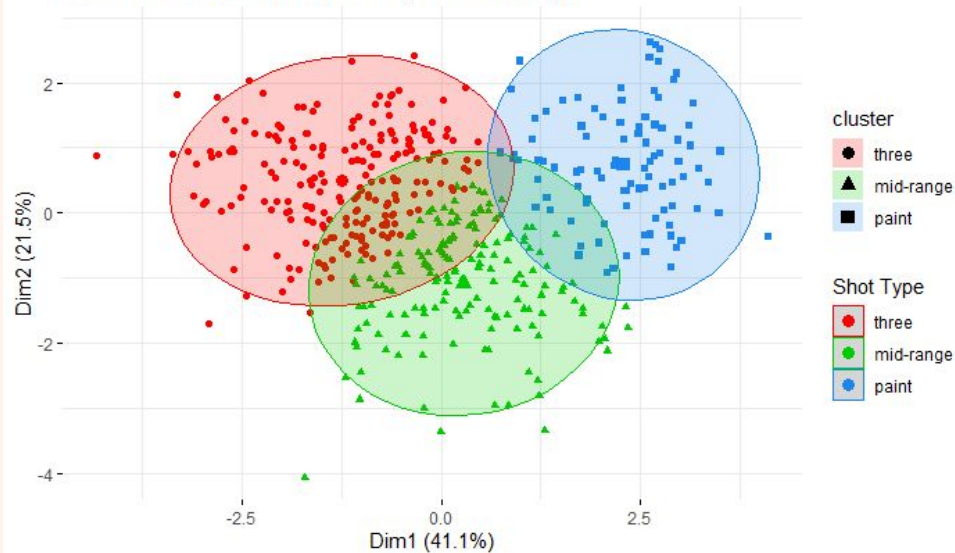
```
Residuals:
    Min       1Q   Median       3Q      Max
-0.05770 -0.01838  0.00840  0.01325  0.06737

Coefficients:
              Estimate Std. Error t value Pr(>|t|)
(Intercept)   1.331344   0.084036  15.843  2.1e-12 ***
Num_Paint_Players -0.007806  0.007100  -1.099   0.285
Num_Mid_Players  -0.007466  0.006467  -1.155   0.263
Num_3PT_Players  -0.007108  0.007148  -0.994   0.333
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Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

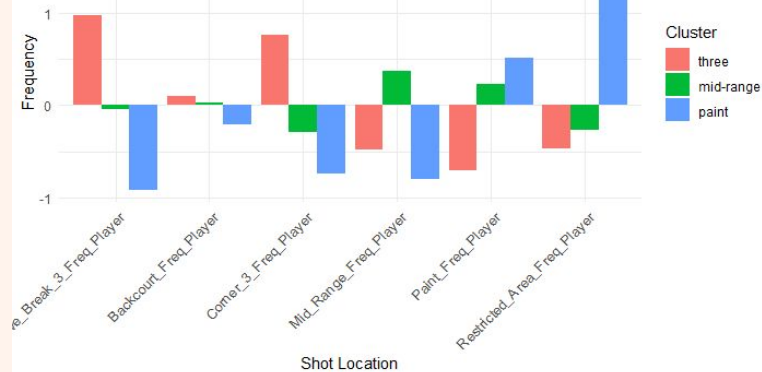
Residual standard error: 0.03533 on 19 degrees of freedom
(7 observations deleted due to missingness)
Multiple R-squared:  0.07821,    Adjusted R-squared:  -0.06734
F-statistic: 0.5373 on 3 and 19 DF,  p-value: 0.6624
```

2016

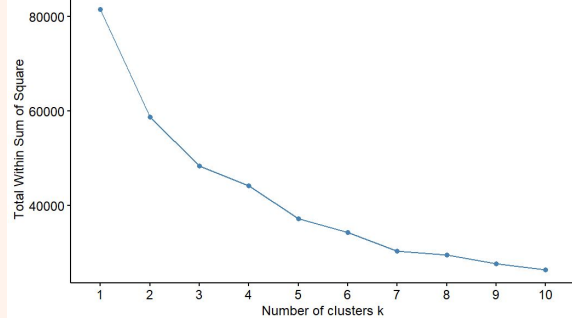
K-Means Cluster Visualization (2016 Season)



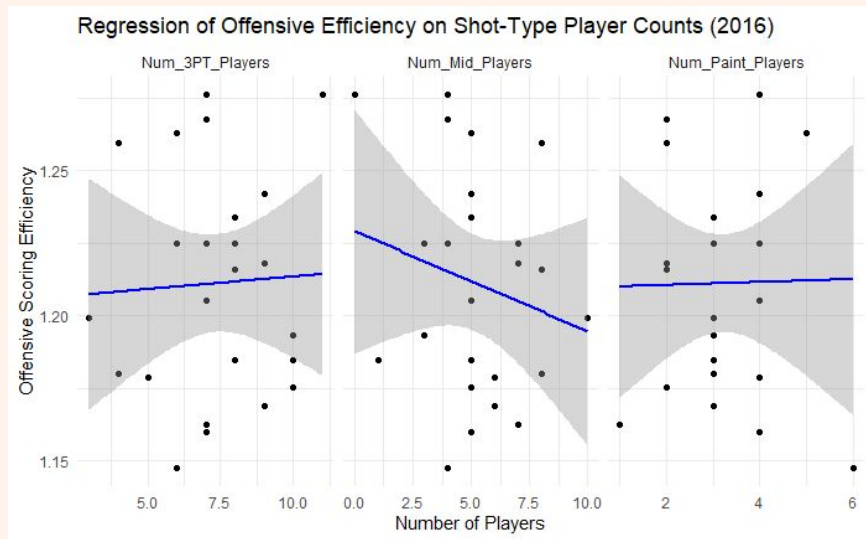
Average Shot Frequency by Cluster (2016)



Optimal number of clusters



Scoring Efficiency Regression (2016)



Residuals:

	Min	1Q	Median	3Q	Max
	-0.063994	-0.029137	-0.000105	0.028429	0.058647

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	1.307544	0.094956	13.770	1.15e-11 ***
Num_Paint_Players	-0.005664	0.009230	-0.614	0.546
Num_Mid_Players	-0.007606	0.006138	-1.239	0.230
Num_3PT_Players	-0.005274	0.006468	-0.815	0.424

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

Residual standard error: 0.04016 on 20 degrees of freedom
(6 observations deleted due to missingness)
Multiple R-squared: 0.07367, Adjusted R-squared: -0.06528
F-statistic: 0.5302 on 3 and 20 DF, p-value: 0.6667



04

Interpretation + Insights



Interpretation

1996

- Paint + Mid-Range Players dominated
- Inverse relationship between Paint players and Offensive Scoring Efficiency
 - P-Value: 0.0194 (statistically significant)



2006

- Mid-Range players on the rise
- Increase in paint shots outside of restricted area
- No clear relationship with offensive scoring efficiency



2016

- Dominated by 3PT and Paint Players
- Not statistically significant, but downward trend for mid-range players
- Could possibly go back to two clusters in the future



Insights

While a clear, direct relationship between the number of players in each cluster and offensive efficiency was not found there are still valuable insights from the analysis:

- Team's cannot have gaps in shooting
- 3 point shots/shooters are increasingly valuable
- Teams with inside players need shooters to bridge the gap





Estimated Best Team Composition



Estimating Best Team By Minimizing Loss Function Of Shooter And Paint Player Proportions:

2023-24: Est. Ideal Team Composition is: 33% Shooters, 22% Inside Players, 44% Versatile

Based by minimizing: $L(P_{\text{shooters}}, P_{\text{inside}}) = \sum E_i (P_i - P_{\text{shooter}})^2 + E_i (P_{ii} - P_{\text{inside}})^2$, Where E_i is the team's offensive efficiency and P is the proportion of shooters/inside players.

- Displays How Teams Need Outside Shooting Ability, Even With Strong Inside Players
- Close To Average Proportions, But Slightly More Shooters And Versatile Players

The End

