# Player Positions in the NBA

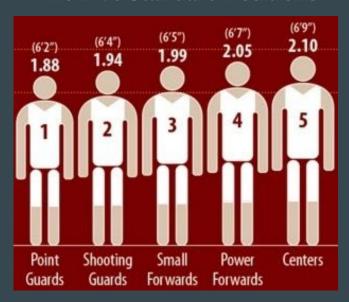
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#### Challenge

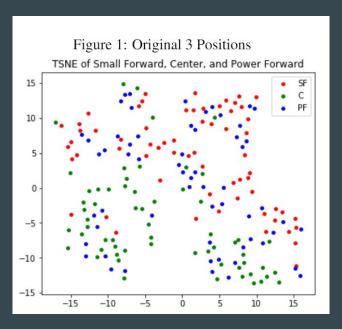
- Attributes vs. Skills
- Old NBA play style:
  - Defined by attributes (linked to skills)
- NBA play style has changed:
  - Defined by skills, unlinked to attributes
- Proposal:
  - Redefine roles based on skill types
  - Unsupervised learning problem

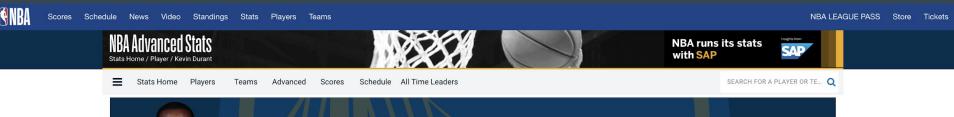
#### The Five Standard Positions



#### **Dataset**

- NBA Players Positions and Per-Game Statistics
- 16 Per-Game Statistic attributes and 1 Position attribute
- 185 Total Players who are labelled as SF, C, PF
- Scraped from NBA Stats and ESPN







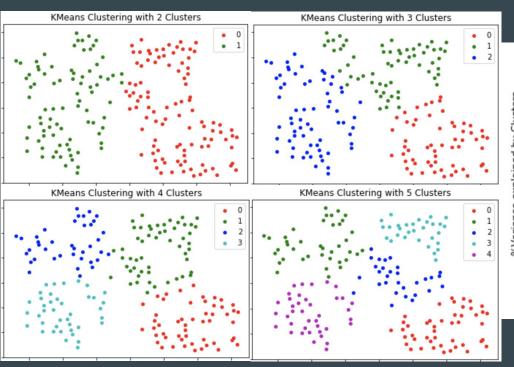
Profile -

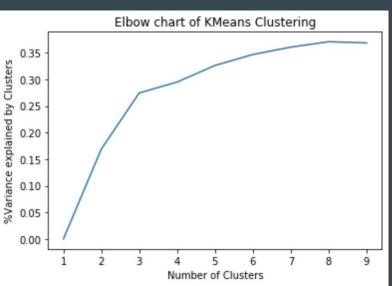
SEASON TYPE PER MODE Playoffs Per Game

■ GLOSSARY SHARE

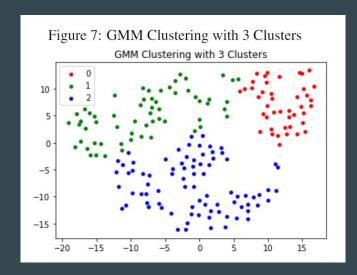
Traditional Splits BY YEAR TEAM FTM OREB DREB STL BLK DD2 2017-18 GSW 91.5 1.7 0 2016-17 GSW 15 35.5 28.5 17.9 55.6 2.5 5.7 44.2 6.9 89.3 1.1 7.9 1.3 2.6 0 11.8 2015-16 OKC 18 40.3 28.4 9.7 22.6 43.0 1.7 6.1 28.2 7.2 8.1 89.0 0.7 7.1 3.3 3.6 1.0 2.1 44.3 0 5.6 2013-14 OKC 29.6 22.2 81.0 19 42.9 10.2 2.2 0.6 2012-13 OKC 11 44.1 30.8 22.4 45.5 31.4 10.2 83.0 1.1 2.4 54.2 1.4 2011-12 OKC 20 41.8 28.5 19.2 51.7 2.1 5.5 37.3 6.7 7.7 86.4 0.7 7.4 3.2 1.5 1.2 2.6 47.7 0 3.5 2010-11 OKC 17 42.5 28.6 9.1 20.3 44.9 2.2 6.4 33.9 8.2 9.8 83.8 1.1 7.1 8.2 2.8 2.5 0.9 1.1 3.1 46.4 0 2.9 2009-10 OKC 6 38.5 25.0 7.2 20.5 35.0 1.7 5.8 28.6 9.0 10.3 87.1 1.3 6.3 7.7 2.3 3.7 0.5 1.3 2.8 39.5

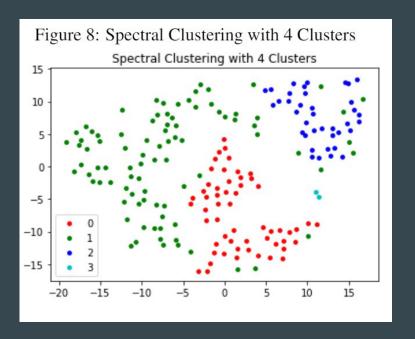
## **Model: K-Means**





# Models: GMM & Spectral Clustering





#### Results

**FGM** 

- ANOVA on K-Means 3 clusters (using player aggregate stats shown previously)
- Significance:
  - Cluster 0:
    - High Steals, More fouls (Defensively minded players)

FΤΔ

- Cluster 1:
  - Low Output Overall (Worse Players)

FTM

• Cluster 2:

FG3M

FG3A

FGA

■ More Shooting, High Volume (Offensively minded players)

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kmeans															
(	0.130228	0.269485	0.023075	0.069619	0.048777	0.070670	0.058993	0.164659	0.063088	0.044442	0.036097	0.028197	0.016488	0.106904	0.068111
	0.137950	0.278375	0.019179	0.057459	0.051151	0.073784	0.069674	0.162281	0.055731	0.054756	0.026841	0.027757	0.017918	0.100597	0.070343
	0.400744	0.440700	0.040000	0.404404	0.000700	0.440000	0.047400	0.470000	0.077407	0.050445	0.007044	0.000005	0.004500	0.004455	0.000400

DRFR

AST

TOV

STI

RIK

ORFR

BLKA

PF

PFD

## Next Steps

- Find ways to evaluate results better
- Further feature engineering to find better class separation
- Integrate College Basketball Statistics