Machine Learning Predictions in the NBA

Presented by Dillon Medd

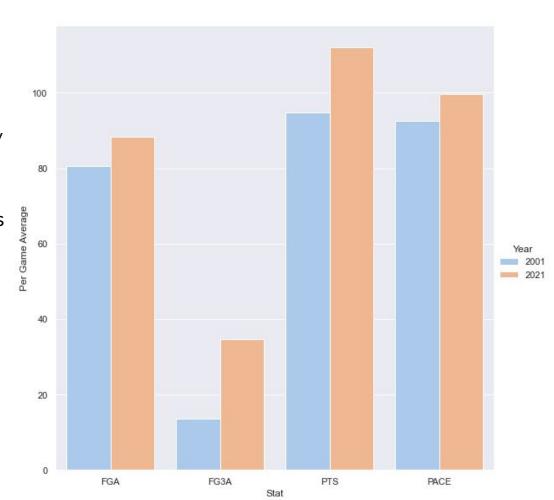
NBA

Data and Methods

- Goal: Predict scores of NBA games as accurately as possible
- 4,000 games worth of box score advanced statistics from over over the previous five regular seasons (2017-2021)
 - NBA.com API used to obtain data
- Two separate models were developed
 - Classification model to predict home team result (win/loss)
 - Regression model to predict margin of victory/defeat for home team
- Both models were trained using a combination of lagged team and player statistics

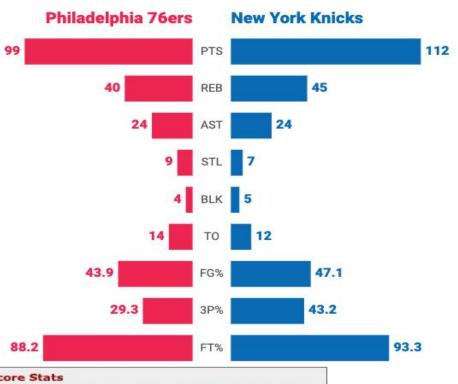
Analytics in the NBA

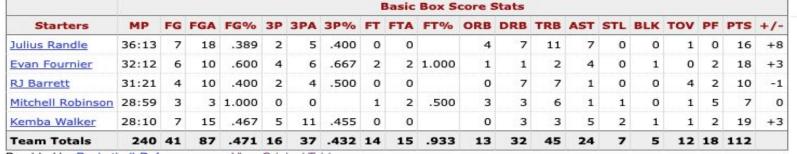
- "Analytics are part and parcel of virtually everything we do now" — NBA Commissioner Adam Silver
- Advanced statistics led to new strategies and evaluation metrics for teams and players
- Dramatic increase in 3 point attempts, points scored and pace of play in last 20 years



Traditional Box Score

- Traditional Box Scores are easy to follow and understand
- Traditional team and player stats have been tracked since the early days of the NBA
- Simple to calculate
 - Ex. (FG% = Makes / Attempts)



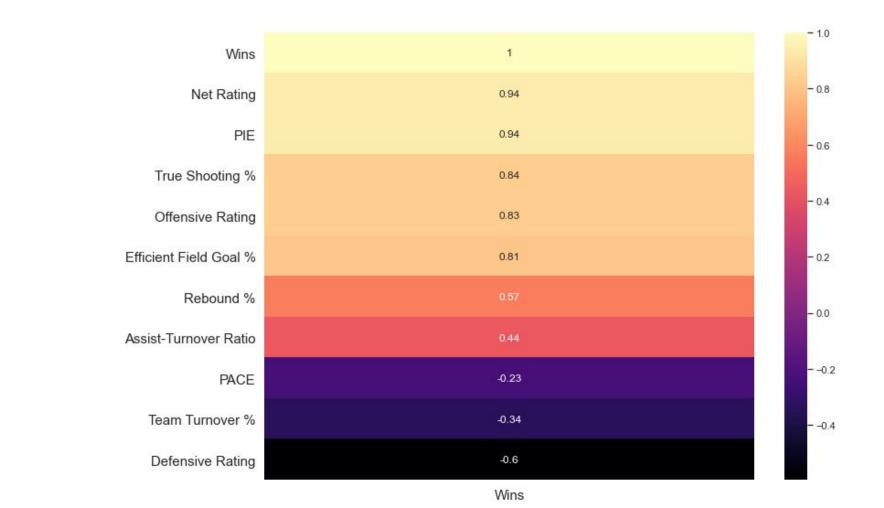


Advanced Box Score

- Advanced Statistics are commonly used by coaches, GMs, writers, etc. to evaluate players and teams in today's NBA
- Provides a better understanding how a player or team played than traditional statistics
- Requires more calculation:
 - Ex. True Shooting Percentage (TS%) = $(\frac{1}{2})$ * Points / (FGA + .44 FTA)

Starters	Advanced Box Score Stats															
	MP	TS%	eFG%	3PAr	FTr	ORB%	DRB%	TRB%	AST%	STL%	BLK%	TOV%	USG%	ORtg	DRtg	ВРМ
Julius Randle	36:13	.444	.444	.278	.000	12.9	21.1	17.2	29.2	0.0	0.0	5.3	23.8	120	113	0.4
Evan Fournier	32:12	.827	.800	.600	.200	3.6	3.4	3.5	18.6	0.0	3.6	0.0	15.4	182	118	8.8
RJ Barrett	31:21	.500	.500	.400	.000	0.0	24.4	12.6	4.4	0.0	0.0	28.6	20.3	74	112	-12.8
Mitchell Robinson	28:59	.902	1.000	.000	.667	12.1	11.3	11.7	4.6	1.9	0.0	20.5	7.7	155	113	-3.9
Kemba Walker	28:10	.633	.633	.733	.000	0.0	11.6	6.0	29.3	3.8	4.2	6.3	25.8	136	106	11.2
Team Totals	240	.598	.563	.425	.172	31.7	72.7	52.9	58.5	7.8	12.2	11.4	100.0	125.3	110.8	

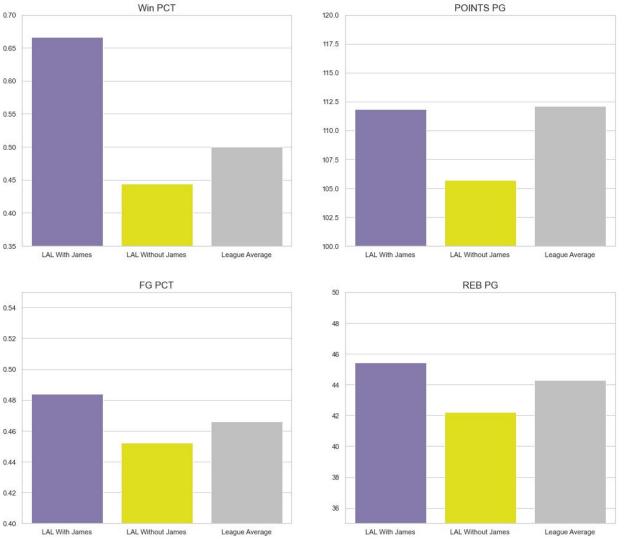
Advanced Statistics Correlation with Win Percentage



Plotly Graphs

https://chart-studio.plotly.com/~dmedd/7.embed

https://chart-studio.plotly.com/~dmedd/9/#/



Today's NBA sees star players rested often and injuries are unavoidable

results with and without James

in the lineup Emphasizes need to include

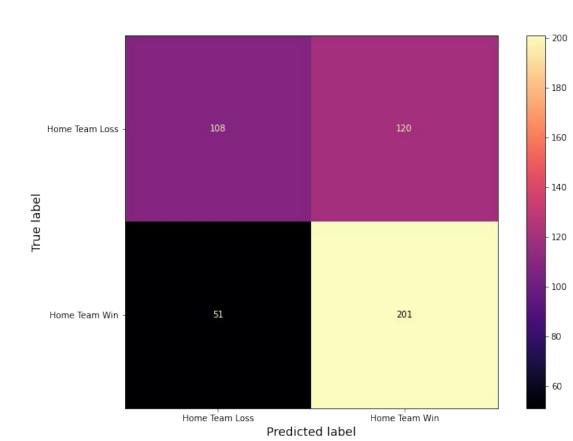
Lakers had very different

starting lineup information in model



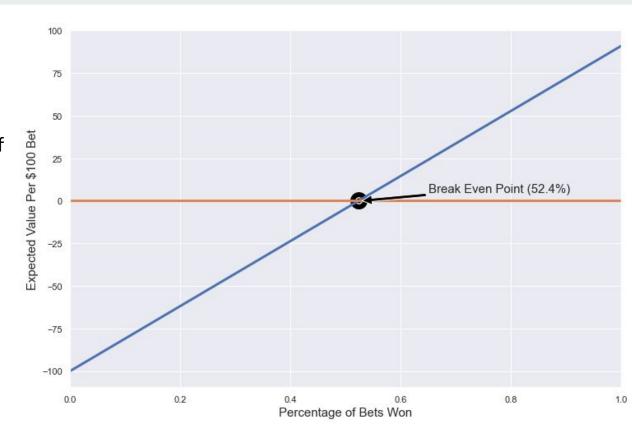
Classification Model Results

- Model predicted winning team correctly for 64.4% of test set
- Logistic Regression Classifier with 22 features
 - 16 Team Statistics
 - 6 Player Statistics
 - Standard Scaler applied



Regression Model Results

- Lasso Regression Model using 28 features returned a RMSE of 12.9
 - 20 Team Statistics
 - 8 Player Statistics
- Best measure of success is to test how it would fare against the 'spread'
- Due to standard 'vig', one must win 52.4% of the time to be profitable
- Model led to a successful bet for 56.1% of test set





Potential Next Steps

- More games, more testing
- Incorporate matchup data into model
- Neural Network
- Test accuracy of classification model predicted probabilities
- Predict exact scores

Thank You!

- dillonmedd1@gmail.com
- GitHub
- <u>LinkedIn</u>