



TOP SHOT

COMPREHENSIVE ANALYSIS OF THE FUTURE
PRICE OF NBA TOP SHOT MOMENTS

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Abstract

- NBA Top Shot moments are NFTs built on the FLOW blockchain and created by Dapper Labs in collaboration with the NBA.
- This project aims to:
 - ❖ Provide a statistical analysis that could yield correlations of the characteristics of a moment and its price.
 - ❖ Provide a prediction of the price of moments with the use of machine learning regression models.
- Expand the up to date analysis and research and serve as a foundation for future research, as well as provide helpful and practical information about the valuation of moments.
- Provide future possible application of this information to other similar highlight-oriented sport NFTs like NFL AllDay or UFC Strike.

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Introduction: What are NFTs?

- NFTs are tokens that we can use to represent ownership of unique items.
 - Only one official Owner.
 - Secured by the blockchain.
- Examples: Art, collectibles, real estate.



Introduction: NBA Top Shot

- Marketplace where users can buy/sell NFTs of NBA/WNBA highlight clips.
- Types of highlights:
 - Block
 - Dunk
 - Assist
 - Jump Shot
 - 3 point
 - Handles
 - Steal
 - Layup
- The modern version of trading cards.
- NBA Top Shot NFTs are known as “Moments”.



Introduction: Research Purpose

- Provide a statistical analysis that could yield correlations of the characteristics of a moment and its price, and a prediction of the price of moments.
- Explore a new research area.
- Serve as a foundation for future research of this project and other future highlight-oriented sports NFTs.

Introduction: Research Benefits

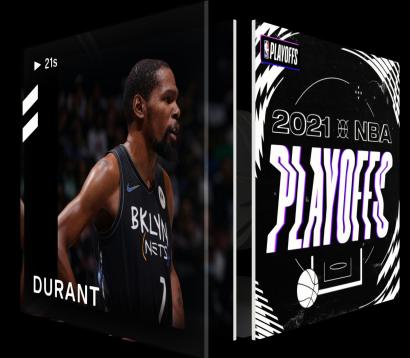
- Provide information to more than 1.1M registered users & other NFT owners / potential owners.
 - Discovery of relationships between features.
 - Research of a model that can predict the price of moments.

Background: Literature Review

- Scarcity of in-depth analysis of NBA Top Shot moments.
- No scholarly application.
- NBA Top Shot analysis found in GitHub and Medium.

Background: Literature Review (Cont.)

- Previous NBA Top Shot analysis found:
 - “NBA TopShot Presentazione” by Riccardo Del Chin.
- Study if the skill of the player appearing in the moment, the rarity and mint size of moments and the results of NBA games could have a direct impact on the price of a moment.



Background: Literature Review (Cont.)

- Relevant Results found by Del Chin's Analysis:
 - Higher prices correlated with higher in game-efficiency of a player.
 - Point difference of a moment's game final score did not have a relevant impact in the price.
 - Team's position in the league's standings did not affect the moment's price of players in that team.
 - Rarity, scarcity and listings of a moment impacted the price.



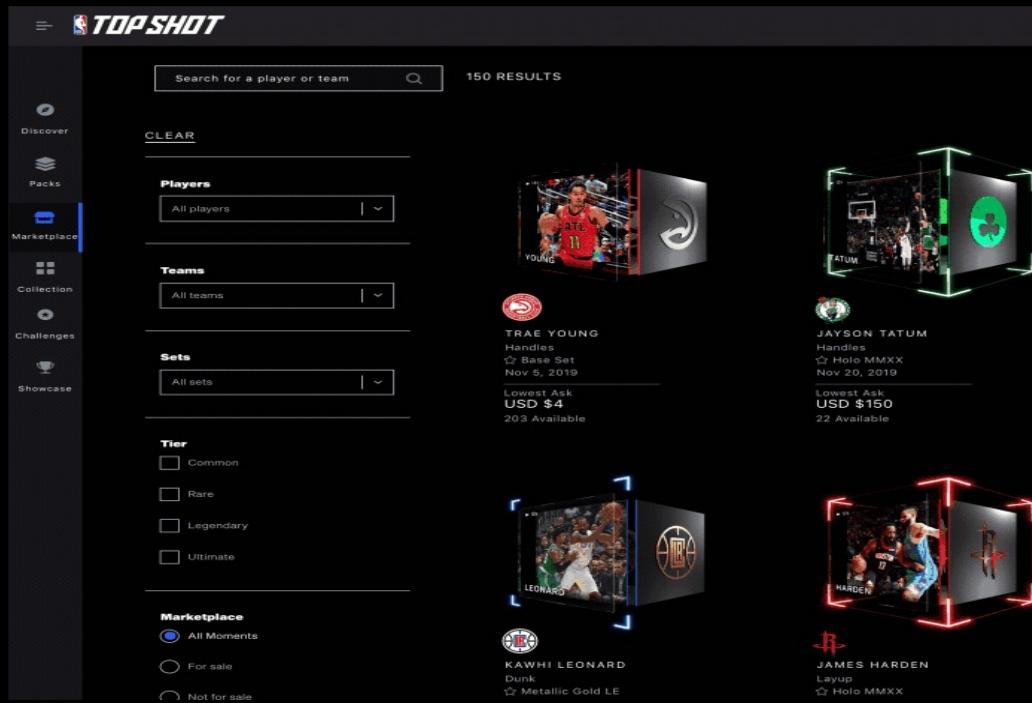
Background: Literature Review (Cont.)

- Features not included in his analysis.
 - ❖ Type of Play
 - ❖ Set
 - ❖ Series
- Characteristics added post June 2021:
 - ❖ New rarity: Fandom
 - ❖ Introduction of badges.



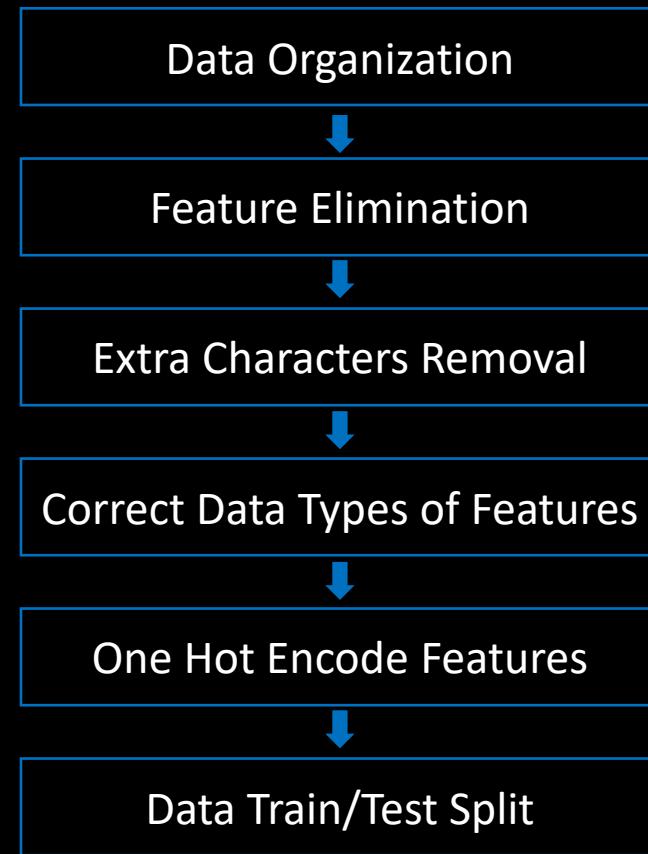
Methodology

- Dataset obtained from the marketplace of the NBA Top Shot platform.



Methodology (Cont.)

Data Cleaning & Processing



Methodology: Dataset Features

- Title: NBA or WNBA player appearing in the moment.
 - Example: Kevin Durant, Luka Doncic or Sue Bird



Methodology: Dataset Features (Cont.)

- Rookie Mint: Moment created during the rookie season of the player appearing in the moment.
- Rookie Year: The play that appears in the moment is from a game during the first season of a player.
- Rookie Premiere: Moments that have a play from the first career game of a player.



ROOKIE MINT
BADGE



ROOKIE YEAR
BADGE



ROOKIE PREMIERE
BADGE

Methodology: Dataset Features (Cont.)

- Rookie Three Stars: Moments that have a Rookie Mint, Rookie Year and Rookie Premiere badge.
- Top Shot Debut: Moments that contain a player's first play created in NBA Top Shot.
- Championship Year: Moments that contains a play from the season that a team won the NBA/WNBA Championship.



Methodology: Dataset Features (Cont.)

- Challenge Rewards: Exclusive moments that are distributed when a user is holding in an account specific moments by a date stated by NBA Top Shot.



Methodology: Dataset Features (Cont.)

- Rarity : Tier of the moment.



Common



Rare



Legendary



Fandom

- Scarcity: Mint size of the moments.

Methodology: Dataset Features (Cont.)

- Sets: Moments of the same tier that have an attribute in common, can be part of a collection.
- Series: Season when a moment was released.
- Supply: Number of listings in the marketplace for a specific moment.
- Team: Team the player appearing in the moment plays for.
- Play Type: Play that is displayed in the highlight.

Methodology: Predictive Modeling Metrics

- Mean Absolute Error (MAE)
- R-Squared
- Mean Squared Error (MSE)
- Root Mean Squared Error (RMSE)

Methodology: Predictive Modeling Models

- Multiple Linear Regression
 - Ridge Regression
 - Lasso Regression
 - Elastic Net
- Stochastic Gradient Descent Regressor
- Tree Based models
 - Decision Trees
 - Bagging
 - Random Forest
 - ADABoost
 - Gradient Boosting Regressor
- K-Neighbors Regressor
- Gaussian Process Regressor
- Partial Least Squares Regressor
- Support Vector Regressor
 - Nu Support Vector Regressor
 - Linear Support Vector Regressor
- Deep Learning models
 - Multi-Layer Perceptron Regressor
 - Recurrent Neural Networks
- Voting Regressor

Analysis & Discussion: Exploratory Data Analysis

Badge/Non Badge Features Analysis

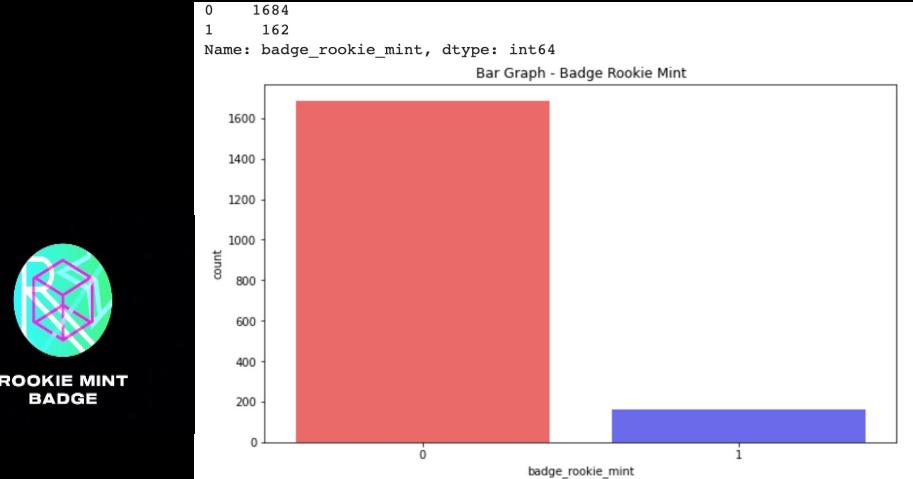


Figure 1 – Count of Moments without/with the Rookie Mint Badge

Rookie Badges

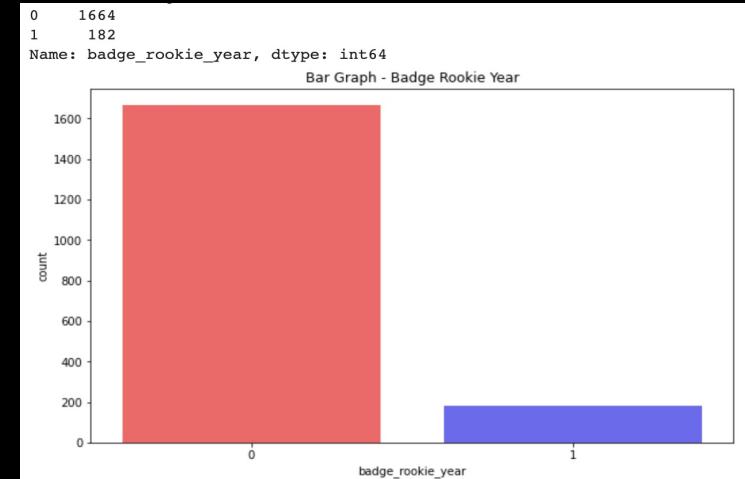


Figure 3 – Count of Moments without/with the Rookie Year Badge

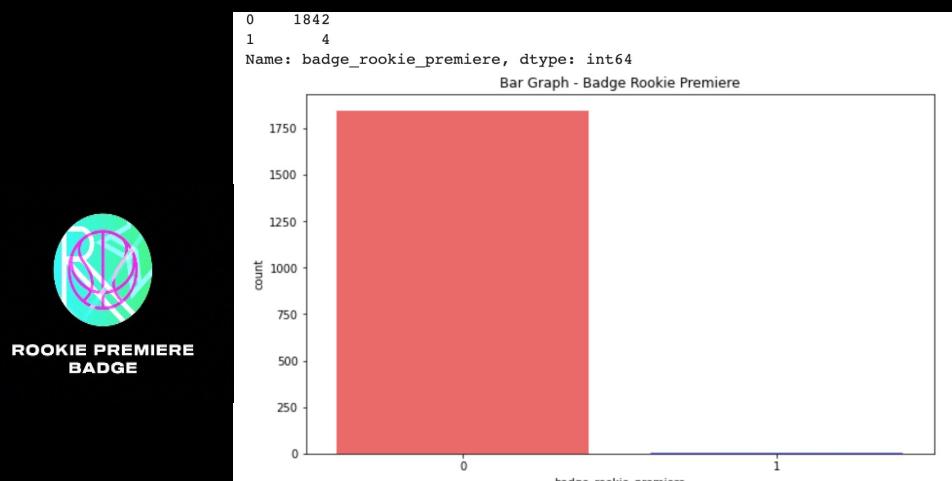


Figure 2 – Count of Moments without/with the Rookie Premiere Badge

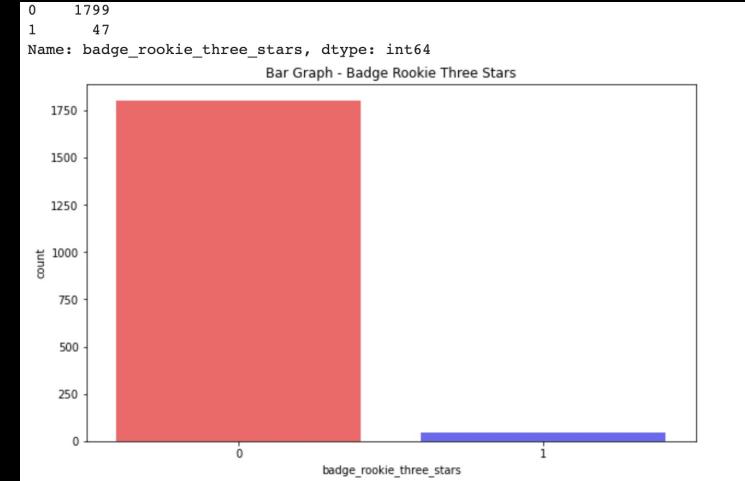


Figure 4 – Count of Moments without/with the Three-Star Rookie Badge

Analysis & Discussion: Exploratory Data Analysis (Cont.)

Badge/Non Badge Features Analysis Rookie Badges

	Badge - Avg. Price	No-Badge - Avg. Price
 ROOKIE MINT BADGE	\$5,550.14	\$2,729.19
 ROOKIE PREMIERE BADGE	\$574.25	\$2,981.97
 ROOKIE YEAR BADGE	\$5,020.80	\$2,753.18
 THREE-STAR ROOKIE BADGE	\$515.45	\$3,041.05

Table 1 – Average price of Moments with/without each Rookie Badge

Analysis & Discussion: Exploratory Data Analysis (Cont.)

Badge/Non Badge Features Analysis

Challenge Rewards, Top Shot Debut and Championship Year Badges

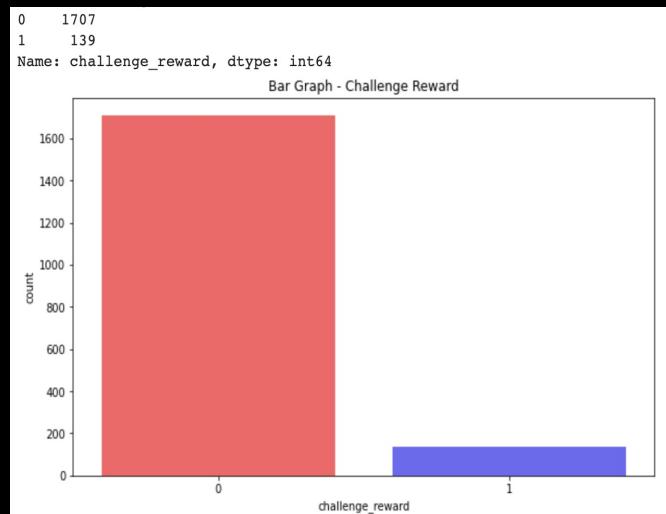


Figure 5 – Count of Moments without/with the Challenge Reward Badge

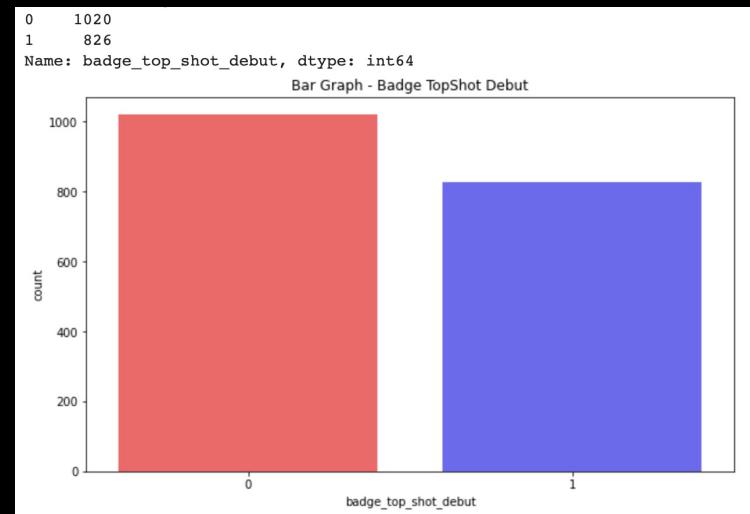


Figure 6 – Count of Moments without/with the Top Shot Debut Badge

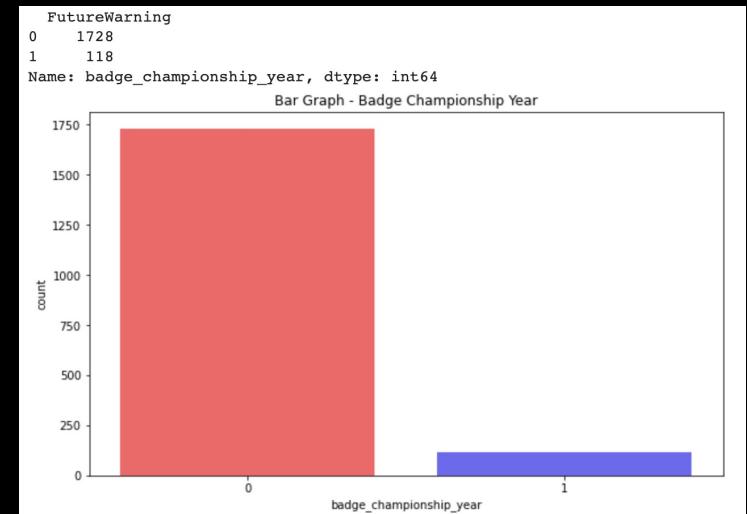


Figure 7 – Count of Moments without/with the Championship Year Badge



Analysis & Discussion: Exploratory Data Analysis (Cont.)

Badge/Non Badge Features Analysis

Challenge Rewards, Top Shot Debut and Championship Year Badges

Average price of Non-Challenge Reward and Challenge Reward moments.

challenge_reward

0	2296.200351
1	11334.266187



Average price of Non-Top Shot Debut and Top Shot Debut moments.

badge_top_shot_debut

0	2212.459804
1	3920.542373



Average price of Non-Championship Year and Championship Year moments.

badge_championship_year

0	2424.736111
1	11060.449153



Analysis & Discussion: Exploratory Data Analysis (Cont.)

Rarity

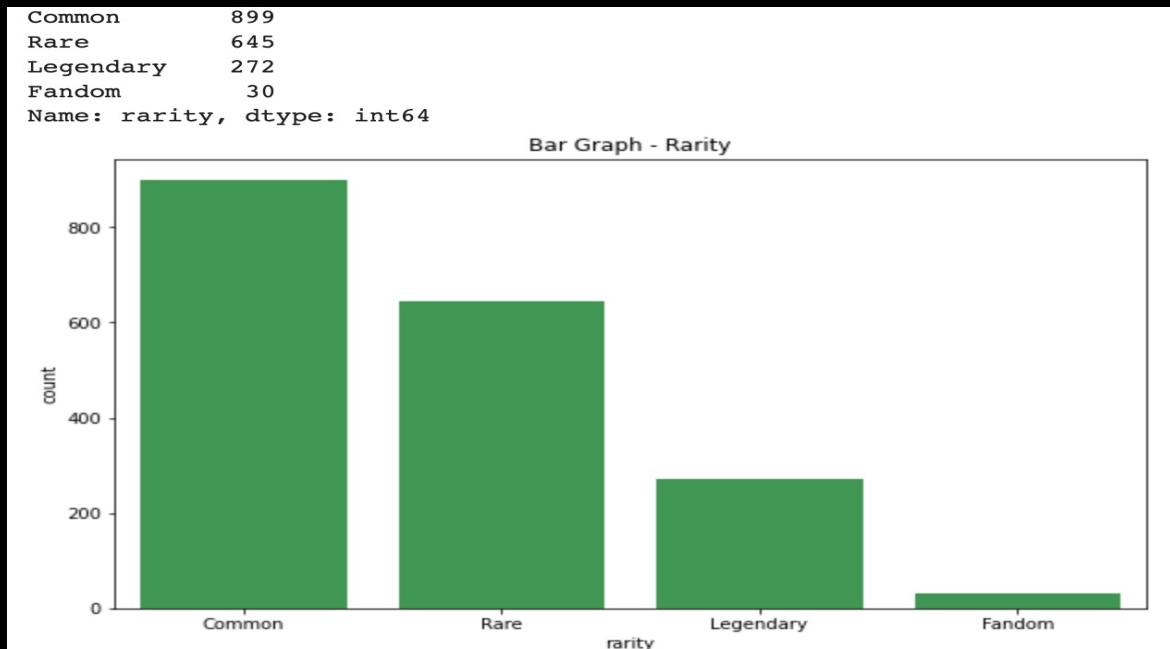


Figure 8 – Moment count per Rarity

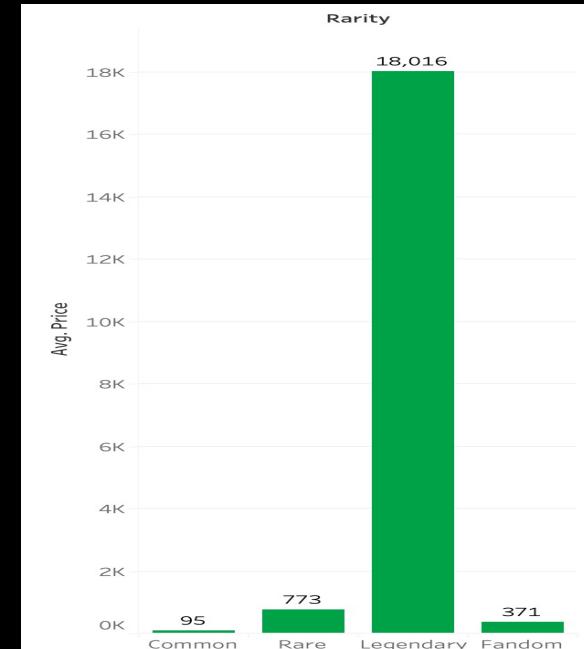


Figure 9 - Moment average price per Rarity

Analysis & Discussion: Exploratory Data Analysis (Cont.)

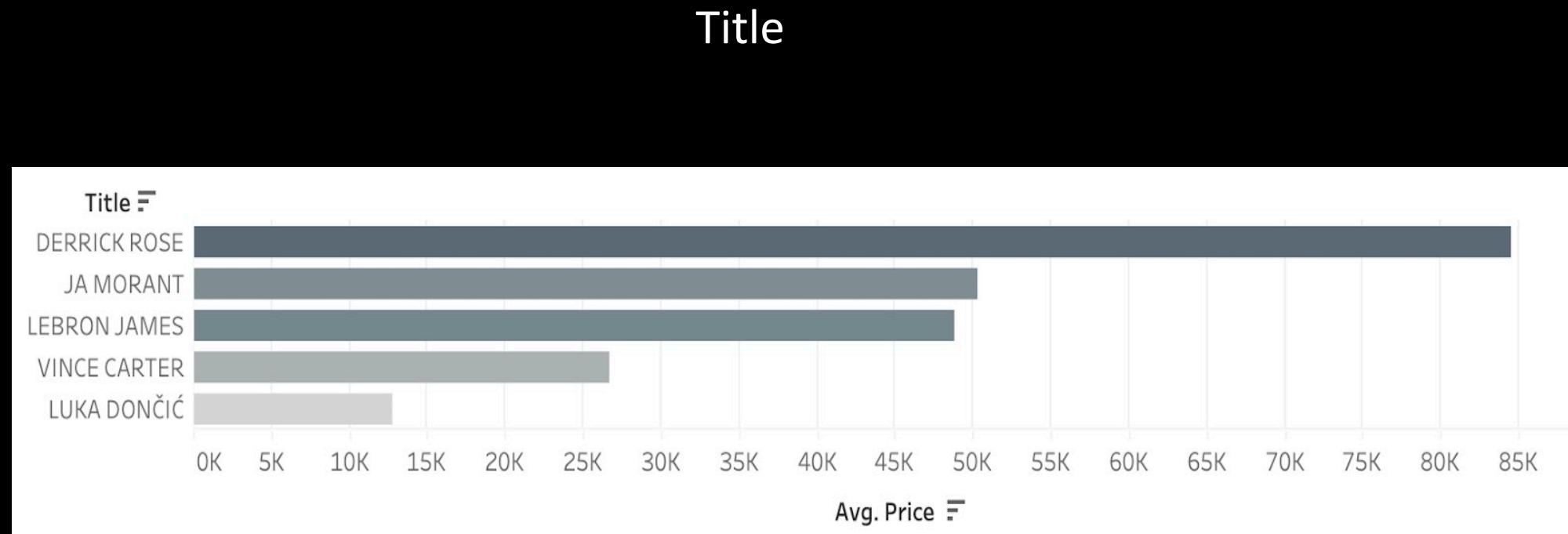


Figure 10 – Top 5 highest average price players

Analysis & Discussion: Exploratory Data Analysis (Cont.)

Series

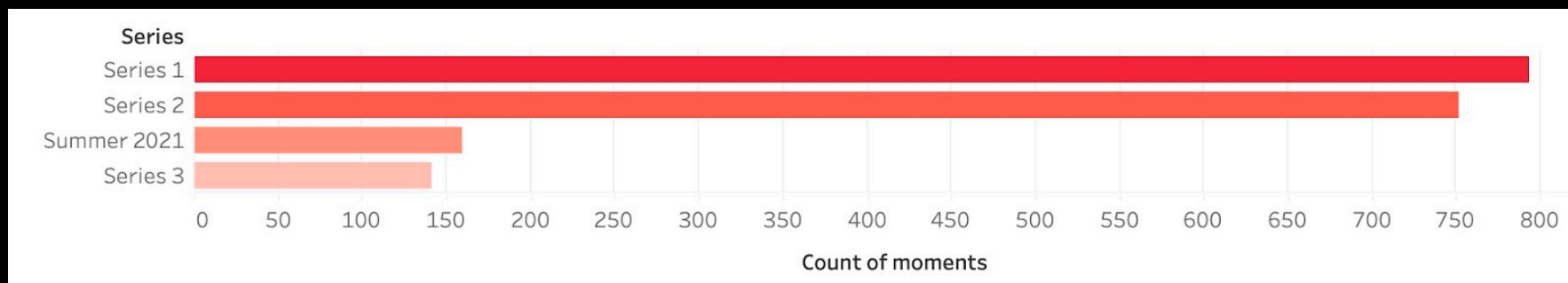


Figure 11 – Moment count per Series

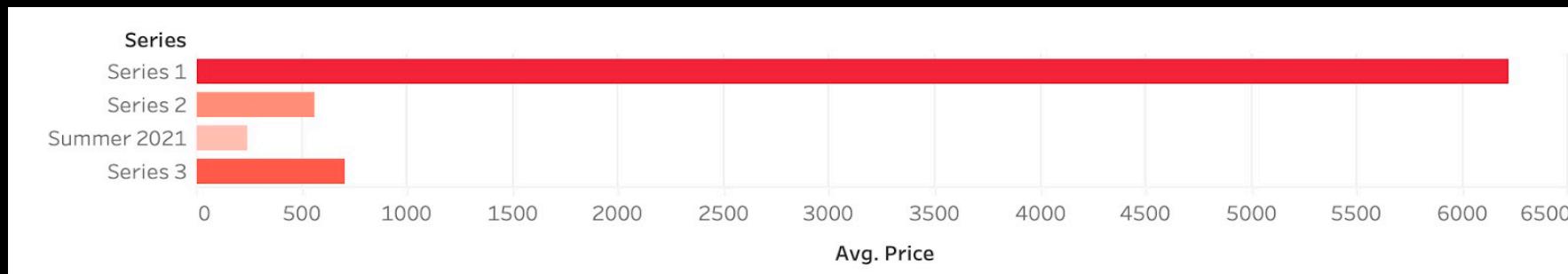


Figure 12 – Moment average price per Series

Analysis & Discussion: Exploratory Data Analysis (Cont.)

Series & Play Type

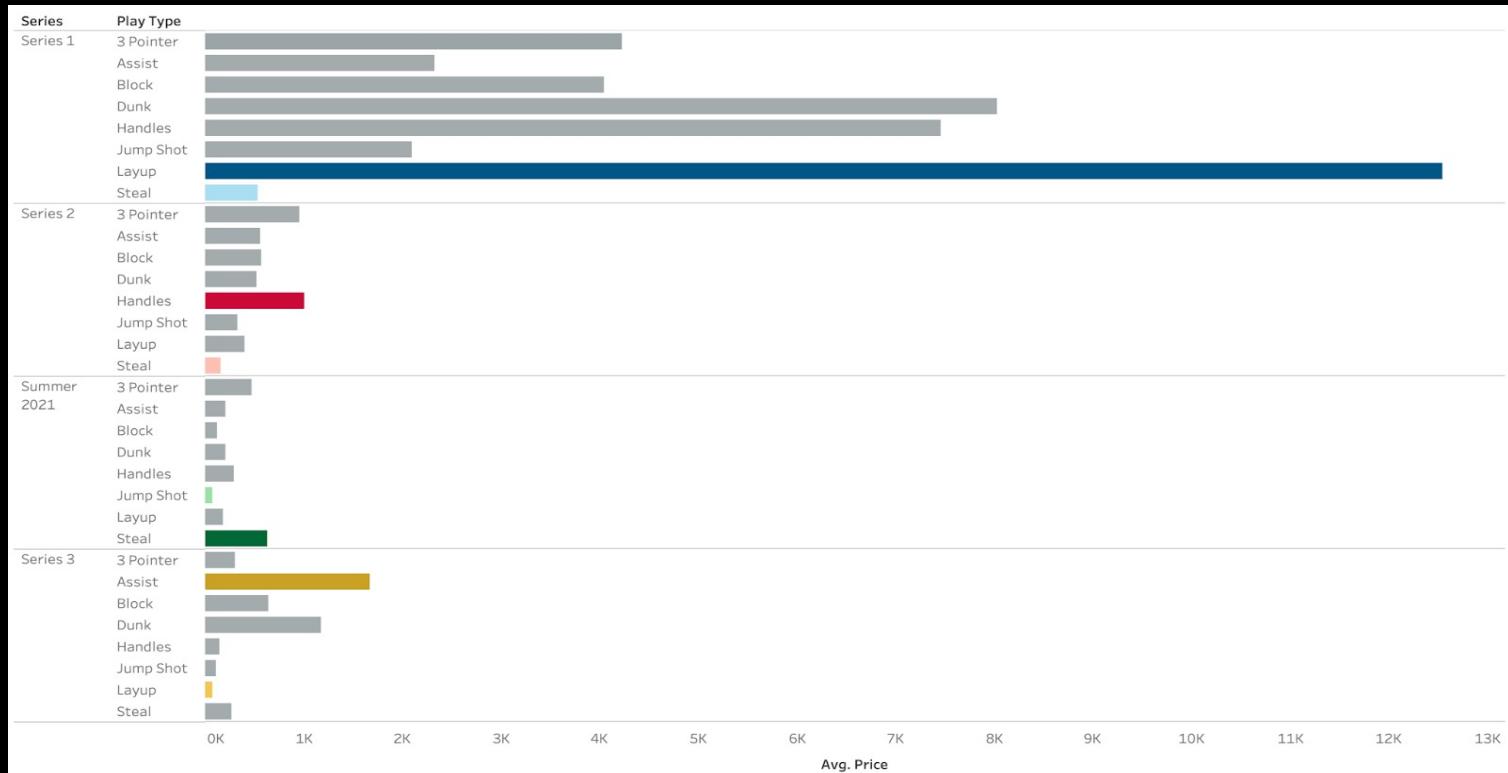


Figure 13 - Moment average price per Series and Play Type

Analysis & Discussion: Predictive Modeling Analysis

Models	Metrics			
	MAE	R-Squared	MSE	RMSE
Multiple Linear Regression	\$17,187,669,336.72	-\$89,909,312,409,256.00	\$17,879,103,578,396,200,000,000.00	\$133,712,765,203.61
Ridge Regression	\$3,502.76	\$0.17	\$165,001,064.94	\$12,845.27
Lasso Regression	\$10,077.46	-\$4.72	\$1,137,465,000.48	\$33,726.33
Elastic Net	\$3,491.62	\$0.14	\$170,501,707.62	\$13,057.63
Decision Trees	\$2,781.43	-\$2.34	\$663,198,153.40	\$25,752.63
Random Forest	\$2,601.31	-\$0.59	\$316,032,819.41	\$17,777.31
AdaBoost	\$11,462.49	-\$2.86	\$766,771,670.67	\$27,690.64
Gradient Boosting Regressor	\$1,394.27	\$0.18	\$162,275,503.63	\$12,738.74
SVR	\$2,145.43	-\$0.02	\$202,504,850.87	\$14,230.42
Nu SVR	\$2,151.03	-\$0.02	\$202,650,748.89	\$14,235.55
Linear SVR	\$2,228.00	-\$0.01	\$200,628,639.96	\$14,164.34
Bagging	\$3,106.05	-\$0.99	\$394,755,821.90	\$19,868.46
SGD Regressor	\$225,584,631,764,580,000.00	-\$1,047,226,085,839,020,000,000,000,000.00	\$208,248,324,417,022,000,000,000,000,000.00	\$456,342,332,484,093,000.00
K-Neighbors Regressor	\$2,066.63	\$0.12	\$175,146,975.09	\$13,234.31
Gaussian Process Regressor	\$2,055.42	-\$0.11	\$220,929,062.07	\$14,863.68
PLS Regression	\$9,323.07	-\$4.08	\$1,009,578,583.94	\$31,773.87
Voting Regressor	\$2,687.39	\$0.19	\$161,580,823.74	\$12,711.44
MLP Regressor	\$2,895.02	\$0.19	\$160,471,336.27	\$12,667.73
Sequential - RNN	\$4,772.39	-\$5.40	\$1,272,552,158.70	\$35,672.85

Table 2 – Machine Learning model results per performance metric

Analysis & Discussion: Predictive Modeling Analysis (Cont.)

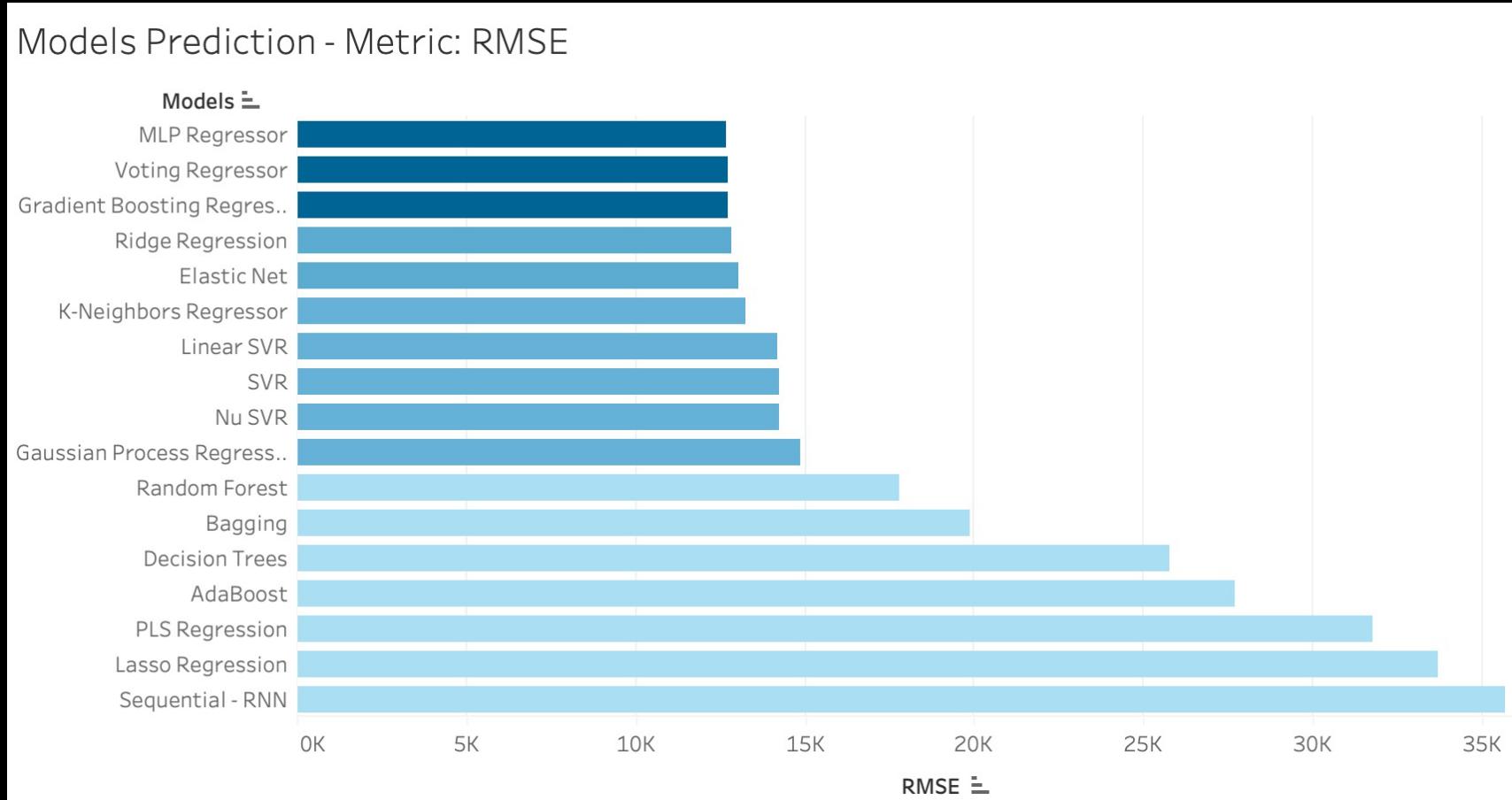


Figure 14 - Machine learning models performance using the metric RMSE

Analysis & Discussion: Predictive Modeling Analysis (Cont.)

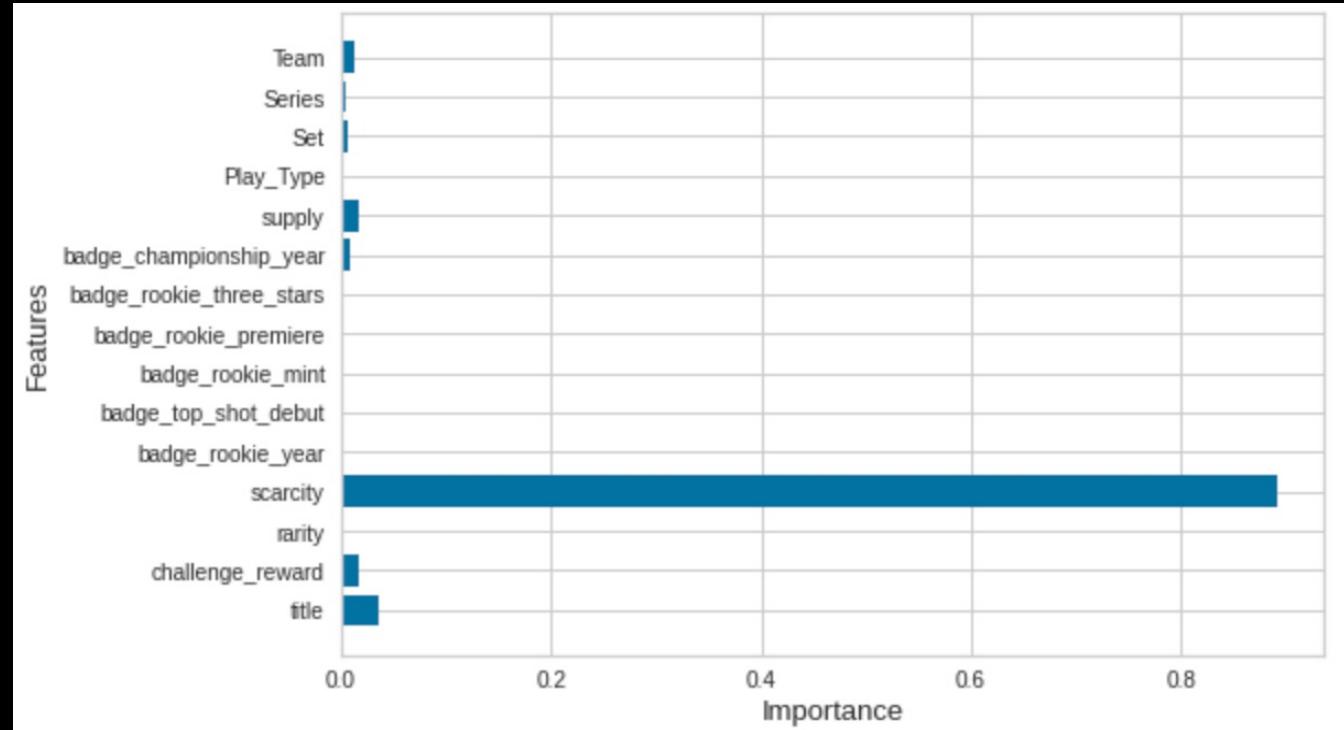


Figure 15 - Variable Importance plot of Gradient Boosting Regressor

Conclusions

- Relevant Results found in this Research

- ❖ Badges

- All 4 Rookie Badges scarce
 - Rookie Year & Rookie Mint more valuable on average than Rookie Premiere or Rookie Three Stars, and Non-Badge moments.
 - Challenge Rewards and Championship Year Scarce, Top Shot Debut less common.
 - All 3 Badges more valuable on average than Non-Badge moments.
 - Partial Supply & Demand economic relationship.

- ❖ Rarity

- Legendary tier is the most valuable on average.
 - Fandom tier's value is between Common and Rare Tiers.

- ❖ Title

- Top most expensive players: Elite level or recognized by basketball community.

Conclusions (Cont.)

- Relevant Results found in this Research

- ❖ Series

- Series 1, oldest series, highest value.
 - Series Summer 2021, least value and only off-season series.

- ❖ Series & Play Type

- Steals significantly lower value than rest of Play Types in Series 1.
 - Rest of Series 1 Play Types, significantly more valuable than other Series' Play Types.

Conclusions (Cont.)

- Relevant Results found in this Research
 - ❖ Best performer: prediction - Multiple-Layer Perceptron
 - ❖ Best performer: prediction + interpretability - Gradient Boosting Regressor
 - ❖ Most important features
 1. Scarcity
 2. Title
 3. Supply
 - ❖ Least important features
 1. Rarity
 2. Rookie Mint Badge
 3. Rookie Year Badge

Future Work

Research improvements

Eliminate unrealistic samples

Eliminate least important features

Apply model to specific data of interest

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QUESTIONS

