Predicting Soccer Player Market Value

Dav King, Luke Thomas, Thomas Barker, Harry Liu

Introduction and Motivation

- Tens of thousands of soccer players are transferred on the international transfer market each year
- Players often cost clubs millions of dollars and they play an integral role in team success
- Identifying undervalued or overvalued players is desirable from a management perspective



Data Introduction

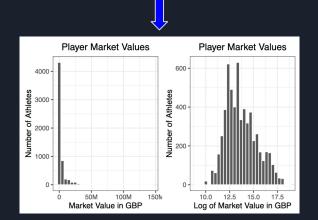
- Data downloaded from Kaggle
- Original data collected from TransferMarkt
- Combined three datasets
- Predictors from on-field performance include each player's total goals, assists, minutes played, and yellow cards during the 2018-19 season
- Other relevant predictors include each
 player's position, nationality, age, and height
- Response variable is market value (GBP) at the end of the 2018-19 season



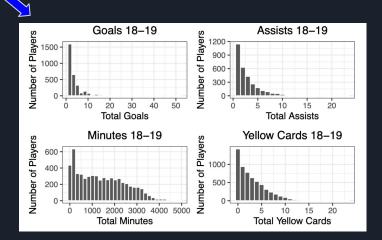


Exploratory Data Analysis

- Moderate correlations between minutes played, goals, and assists
 - Consider interaction effects
- On-field performance predictors very right-skewed, with most observations by 0
 - o Consider log-transform
- Player market value also very right skewed
 - Consider log-transform







Final Model

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\begin{array}{lll} final\_value &=& e^{13.7} \times e^{0.00049 \times age\_in\_days} \times e^{0.0602 \times total\_goals} \times e^{0.0671 \times total\_assists} \times e^{0.00062 \times total\_minutes} \times e^{-0.00787 \times total\_yellow} \times e^{0.00988 \times height\_in\_cm} \times e^{0.35308 \times citizenship\_France} \times e^{0.45405 \times citizenship\_Spain} \times e^{-0.21465 \times citizenship\_Other} \times e^{-0.02893 \times position\_Defender} \times e^{0.00323 \times position\_Goalkeeper} \times e^{0.11777 \times position\_Midfield} \end{array}
```

term	estimate	std.error	statistic	p.value	conf.low	conf.high
(Intercept)	13.72449	0.01826	751.58388	0.00000	13.68869	13.76029
age_in_days	-0.00049	0.00001	-36.45455	0.00000	-0.00051	-0.00046
total_goals	0.06020	0.00689	8.73898	0.00000	0.04670	0.07371
total_assists	0.06710	0.01012	6.62931	0.00000	0.04726	0.08695
total_minutes	0.00062	0.00003	21.94142	0.00000	0.00056	0.00067
total_yellow	-0.00787	0.00833	-0.94541	0.34450	-0.02419	0.00845
height_in_cm	0.00988	0.00134	7.34644	0.00000	0.00724	0.01251
country_of_citizenship_France	0.35308	0.10759	3.28187	0.00104	0.14216	0.56400
country_of_citizenship_Spain	0.45405	0.10250	4.42998	0.00001	0.25311	0.65500
country_of_citizenship_other	-0.21465	0.07754	-2.76832	0.00566	-0.36666	-0.06264
position_Defender	-0.02893	0.05151	-0.56164	0.57439	-0.12991	0.07205
position_Goalkeeper	0.00330	0.08308	0.03971	0.96832	-0.15958	0.16618
position_Midfield	0.11777	0.05404	2.17930	0.02936	0.01182	0.22372

Interesting Findings

- All effects except for total yellow cards, being a defender, or being a goalkeeper were significant
- Being younger, taller, and responsible for more goals, assists, and minutes all predicted an increase in a player's market value
- Model performed well on testing data R² =
 0.391 (0.411 on training data) and RMSE =
 1.264 (1.213 on training data)



Conclusions and Future Work

- The ability to predict a player's market value is extremely useful for soccer clubs
- Significant effects:
 - Minutes played
 - Total goals
 - Total assists
 - > Age
 - Height
 - Midfielder
 - Nation of citizenship (Brazil, France, Spain)
- Limitations:
 - Skewed explanatory variables
 - Statistical assumptions
- Future work:
 - Multiple seasons
 - More predictors



Citation

Data Source:

Cariboo, David. *Football Data from Transfermarkt.* (data file). Kaggle, 2022. Web. 03 Nov 2022. https://www.kaggle.com/datasets/davidcariboo/player-scores