# Machine Learning Approach to Soccer Player's Overall Rating Prediction

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#### Abstract

A abstract sums up your work in very few sentences: (i) state the problem you are addressing; (ii) say why it's an interesting problem, and which issues are hard to tackle; (iii) give your approach towards solving the problem; (iv) say Why and how well your approach solves the problem.

#### Introduction

Your introduction briefly explains the problem you address, and what you've achieved towards solving the problem. It's an edited and updated version of your introduction and objective from your topic proposal

#### Dataset

#### 0.1 Dateset overall

Our project dataset is FIFA 2019 players attributes dataset, which is collected from Kaggle. This dataset includes 89 attributes of a soccer player, and our project only uses some of these attributes concerning only player capability. Besides, considering that goalkeepers have a totally different set of capability metrics compared with other players, we will only consider data records of nongoalkeepers in our model.

#### 0.2 Data processing

Now we beginning the data processing. There are 18206 players for which 89 features each are provided. Firstly, we should drop the duplicate entries. Then we check the variables:

 $ID, Name, Age, Photo, Nationality, Flag, Overall, Potential, Club, ClubLogo, Value, \\ Wage, Special, PreferredFoot, International Reputation, WeakFoot, SkillMoves, \\ WorkRate, BodyType, RealFace, Position, JerseyNumber, \\ Joined, LoanedFrom, ContractValidUntil, Height, Weight, \\ LS, ST, RS, LW, LF, CF, RF, RW, LAM, CAM, RAM, LM, LCM, CM, \\ RCM, RM, LWB, LDM, CDM, RDM, RWB, LB, LCB, CB, RCB, RB, \\ Crossing, Finishing, HeadingAccuracy, ShortPassing, Volleys, Dribbling, \\ Curve, FK Accuracy, LongPassing, BallControl, Acceleration, \\ SprintSpeed, Agility, Reactions, Balance, ShotPower, Jumping, Stamina, Strength, \\ LongShots, Aggression, Interceptions, Positioning, Vision, Penalties, Composure, \\ Marking, StandingTackle, SlidingTackle, \\ GKDiving, GK Handling, GK Kicking, GK Positioning, GK Reflexes, ReleaseClause$ 

Now we have to delete the missing value which will distribing our model.

Index	column name	Total missing	Percent missing
0	Loaned From	16943	0.930576
1	LWB	2085	0.114516
2	LM	2085	0.114516
23	RS	2085	0.114516
24	ST	2085	0.114516
25	LS	2085	0.114516
26	RCB	2085	0.114516
27	Release Clause	1564	0.085901
28	Joined	1553	0.085297
29	Contract Valid Until	289	0.015873
58	Finishing	48	0.002636
59	Crossing	48	0.002636
60	Positioning	48	0.002636
61	LongPassing	48	0.002636
62	BallControl	48	0.002636
63	Acceleration	48	0.002636
64	SprintSpeed	48	0.002636
65	FKAccuracy	48	0.002636
66	Reactions	48	0.002636
67	ShotPower	48	0.002636
68	GKReflexes	48	0.002636
69	Jumping	48	0.002636
70	Stamina	48	0.002636
71	Strength	48	0.002636
72	LongShots	48	0.002636
73	Aggression	48	0.002636
74	Balance	48	0.002636
75	Interceptions	48	0.002636
76	Flag	0	0.000000
77	Name	0	0.000000
78	Age	0	0.000000
79	Photo	0	0.000000
80	Nationality	0	0.000000
81	Value	0	0.000000
82	Overall	0	0.000000
83	Potential	0	0.000000
84	Club Logo	0	0.000000
85	Wage	0	0.000000
86	Special	0	0.000000
87	ID	0	0.000000

Table 1: Missing value percent

Considering these missing value we have below comments:

- Since a big chunk of players are not loaned to other clubs, hence there are 16943 missing values out of 18206 i.e. they play for their own club.
- There is a consistency (11.45%) in missing values for the position (they play on), this needs to be explored.
- Release clause, joined, contract valid until, club, position, jersey number have missing values and need to be explored.
- There is a consistency (0.26%) in missing values for the players rating attributes.

Now we will drop unnecessary columns, which can address the problem simply, reference the FIFA official guideline (https://www.fifauteam.com/fifa-19-attributes-guide/). we only keep significant values that have highly contributed to the model.

Age, Overall, Potential, Weak Foot, Position, Crossing, Finishing, Heading Accuracy, Short Passing, Volleys, Dribbling, Curve, FK Accuracy, Long Passing, Ball Control, Acceleration, Sprint Speed, Agility, Reactions, Balance, Shot Power, Jumping, Stamina, Strength, Long Shots, Aggression, Interceptions, Positioning, Vision, Penalties, Composure, Marking, Standing Tackle, Sliding Tackle, GK Diving, GK Handling, GK Kicking, GK Positioning, GK Reflexes

We have to delete the missing value which will disturbing our model and then we will filter data with null values for position:

Index	column name	Total missing	Percent missing
0	Position	60	0.003295
1	GKReflexes	48	0.002636
2	Curve	48	0.002636
30	Positioning	48	0.002636
31	Interceptions	48	0.002636
32	Aggression	48	0.002636
33	LongShots	48	0.002636
34	Strength	48	0.002636
35	Stamina	48	0.002636
36	Potential	0	0.000000
37	Overall	0	0.000000
38	Age	0	0.000000

Table 2: Missing value percent in useful attribute

After handling these missing values we have 18147 rows thought, subtracting the 60 missing values. Besides, we have to transform some *string* type values to *int* type, which guarantees our model works.

At last, we partition original player data according to positions, we mainly divide players into 4 positions: Forward, Midfielder, Back, and Goalkeeper.

### Solution

The solution section covers all of your model design, algorithms, formulas, findings etc. It explains in detail each contribution, if possible with figures/schematics.

## Results and Discussion

The results section details your metrics and experiments for the assessment of your solution. It allows you to compare your idea with other approaches you've tested.