

# Footie: Web Simulator of Football Leagues and Championships

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## Abstract

The goal of this project is to create a web football simulator, which would simulate realistic matches of the football world. The concept of the whole simulator is complex. It generates future results based on real historical match results. However, unpredictable and crazy results are also a part of the fun. Each user has its own simulation and he or she can have completely different results, which are more and more diverse year after year of the simulation. Because of the complexity of the football world, the simulator supports only selected football nations. The main focus of the project is to improve the simulation of matches. Calculation of match results is based on real historical results and it uses neural networks. It is mixed with random factors described in the paper. The project focuses on the group of football fans who would like to only watch simulations of the football world without the need to manage teams and players. This is something that is not available in the current market. Every game focuses on management mainly, not the simulation. The core of the simulator could be also used for trying to predict real football matches because it is based on real data. This paper describes base concepts of predicting a football match. It presents football systems and shows, how the simulator game is different from existing solutions.

**Keywords:** Football — simulator — machine learning — match prediction

**Supplementary Material:** N/A

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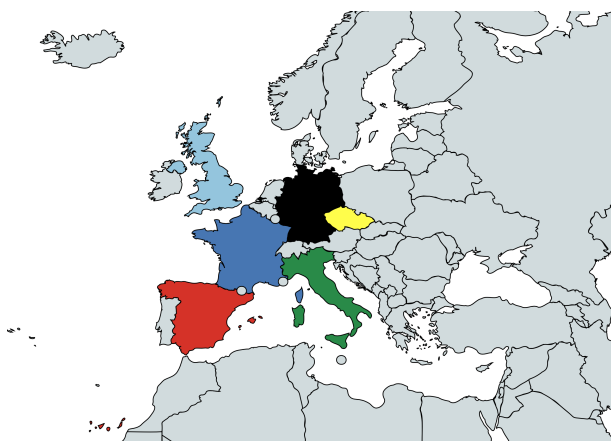
## 1. Introduction

The main value of this project is a system predicting match results based on real data (real results). It is connected to a simulating system which includes random factors and provides a unique evolution of the football world, based on current data. These random factors are not just random functions. They are based on some potential motivating factors, which exist in the real sports world. A user who would like to see scenarios of the next football season can have satisfaction in this game same as the user who likes sci-fi scenarios and would like to see generated football world for example

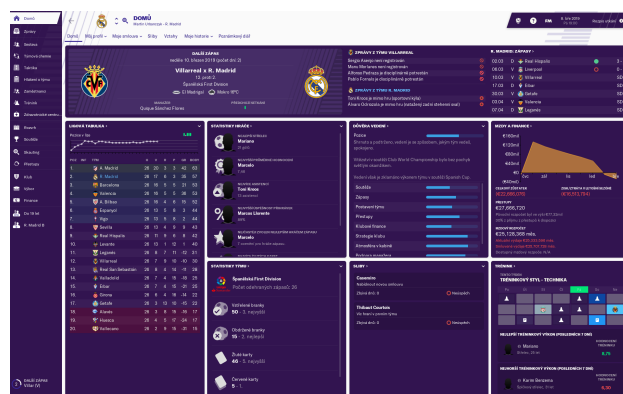
in 2100. The core of the prediction can be also used for prediction of match results in the current season in the real world.

The current version of the system simulates the biggest European football nations (England, Spain, France, Italy, and Germany) and the Czech Republic. It also contains the biggest club competitions like Champions League or Europa League, which use the UEFA coefficients[1] for recalculation of nations' ranking after each season. The ranking is used to calculate the number of participating clubs for every nation. Hence, it is possible that some smaller nations might have

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**Figure 1.** UEFA nations which are included



**Figure 2.** FM 2019 main dashboard [2]

more teams the competitions in future seasons. The data of all world leagues are included, but the project is pretty complex. I chose to focus on the match prediction. Figure 1 present map of UEFA nations, which are included in detail simulation.

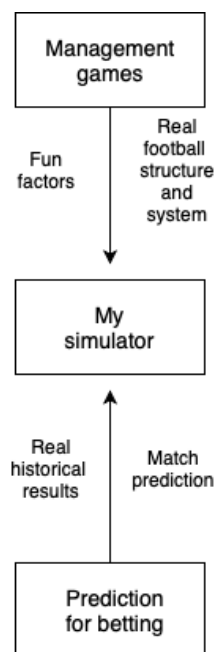
The system needs to be entertaining and playable for football stats maniacs as well as for casual fans who would like to see how their favourite team evolves in the nearby future. The project does not guarantee fun for every user. It focuses on the people who are interested in football and its future. It offers something to either casual fans and bigger experts who have experience with playing manager games and who are interested in deeper stats. Who knows, maybe in your simulation Zbrojovka Brno will play Champions League final in the year 2100. It is also planned to create a community hub where users can share their results and compare the state of their world.

The crucial task of the project was to find the key features of the prediction. I also had to find the *soft* features which might also have an impact on match (e. g., motivation, frustration, form, or match importance).

## 2. Analysis

Most of my inspiration came from playing management games and detailed watching of the football world. The biggest football manager is the game from SEGA called Football Manager 2019 [2]. It is a pretty realistic game but it focuses more on the management factor. No matter which team you choose, the world automatically changes due to your actions. I took some inspiration from its user interface (Figure 2). I also used to play Czech Soccer Manager [3] in the past. The game is not so huge and complex as FM 2019, but I took some inspiration from it's easy going UI and its simplicity.

None of these games and projects focuses on realistic simulation. You can get some crazy results



**Figure 3.** Difference between existing solutions and my simulator

practically immediately. It is alright when you play a game just for fun. On the other hand, there are also some projects which focus just on match prediction mainly for betting [4]. These projects do not focus on the simulation of a hypothetical world. Figure 3 presents how I mixed my project from these types of existing solutions.

## 3. Design

The solution is on the halfway to both analyzed categories. It uses real prediction of the results. Then, it adds the random factor to make the game more entertaining. Results of the matches are close to the predicted results but they might be different for every player. This helps to make the evolution of the football world diverse and crazy in the future. The simulator, however, does not force players to plan tactics, buy or sell proper players or negotiate about their salaries. It focuses more on the simulation of the games than any

80 other game does.

81 For example, if players of Football Manager go for  
82 one year lasting holiday (the active management of the  
83 team is interrupted), they will see more diverse results  
84 in the league they chose. Just because they chose it.  
85 For example, you took the manager seat in England.  
86 Even if you do nothing, there will be some shocking  
87 results that are not so realistic. However, in a league  
88 where you are not a manager (for example Germany),  
89 results are still going to be similar (for example Bayern  
90 Munchen will win 10 titles in 10 next seasons). There  
91 will be some unpredictable results but not more in  
92 some region just because you choose it.

93 If the player starts a new game in Football Man-  
94 ager, they will see more diverse results in the league  
95 they chose. Just because they chose it. Other leagues  
96 evolve too, but the change is far slower. So it is not  
97 quite realistic. But it is mainly a game, so it is un-  
98 derstandable because players want to go higher fast.  
99 For example, when a player begins a game in England,  
100 weaker teams can finish higher in England. Although,  
101 in Germany or Spain it is different. Every time it takes  
102 more than a decade when some weaker team wins the  
103 league.

104 However, I plan to add some extensions in the fu-  
105 ture, that will allow players to add their teams, players  
106 to real leagues, or make some transfers to see how  
107 things might evolve (e. g., What would happen if  
108 Messi and Ronaldo play in the same team that is on  
109 the bottom of the table).

110 All of the biggest football manager games are avail-  
111 able for the only desktop platform. I decided to make  
112 a cloud solution, so you can connect everywhere and  
113 it play it on (almost) every device, that has an internet  
114 browser.

## 115 4. Implementation

### 116 4.1 Architecture

117 The architecture of the system is displayed in Figure  
118 4. The system is structured into modules which are  
119 not dependent on each other. Each functionality has it  
120 separate control (i.e. team screen - controls for play-  
121 ers, fixtures, stats). This makes the system extensible  
122 and scalable. In the future, it is planned to provide  
123 subsets of the modules to users according to their pur-  
124 chase plans. They will be able to choose the modules  
125 according to their interests (e. g. leagues, nations).

126 I implemented the system as a web application  
127 in PHP 7.2 (Nette framework) and JavaScript mainly  
128 for asynchronous loading and user actions. It has a  
129 lightweight user interface. It is mobile friendly and  
130 it does not overwhelm users with stats and data (the

detailed data are accessible in specific pages). The sys- 131  
tem uses the MVC architecture and it supports REST 132  
API written in node.js for the connection with the 133  
database. It is expected that native mobile applications 134  
will be implemented in the future. 135

The system uses the standard MySQL database. 136  
The database has two parts. The first part consists of 137  
“hard” data which will not be changed in the future 138  
and are same for all users. For example, it contains 139  
example names of players and teams, the number of 140  
teams in each league etc. The second part differs for 141  
each user. It consists of actual data of a simulation 142  
(e. g., player goals, team history etc.). The whole 143  
database contains a lot of data (for example, around 144  
80 000 players). 145

Also, I needed to do a lot of parsing and merging of 146  
data, because of different sources. It was necessary to 147  
merge the Football Manager database, my system and 148  
database and api-football data. I used mainly Python 149  
scripts for this. 150

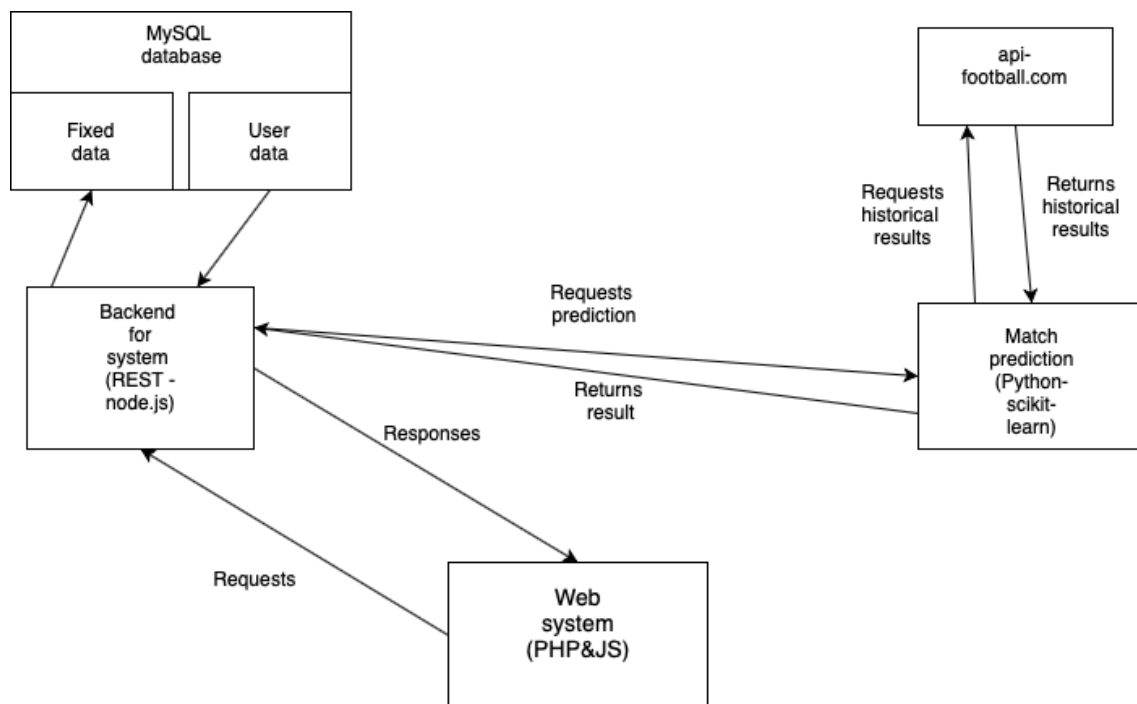
### 4.2 Match Prediction 151

The match prediction is provided by using the Python 152  
scikit-learn library [5]. It uses the neural network 153  
which was trained with the data gathered from the api- 154  
football.com service [6]. The service provides a REST 155  
API for getting real results. The results are mixed 156  
with the results of the simulation (in the first years of 157  
simulation). The main features of machine learning 158  
are: 159

1. **Team rating:** the actual rating from the database; 160  
it is calculated after each season from the results. 161
2. **Actual team form:** the last 10 matches played 162  
by each team. 163
3. **Form at home/away stadium:** the last 5 matches 164  
played at home and away from fields, depending 165  
on which team is the home team and which is 166  
away team. 167
4. **Last matches between the two teams:** the 168  
matches between the two teams during last five 169  
years (not the 5 last matches, because it is not 170  
relevant if these two teams played together 15 171  
years ago under completely different circum- 172  
stances). 173

I made a dataset of historical results and tried to 174  
predict some older games. I have a lot of different 175  
data from all over the world. I did a lot of experimen- 176  
tation with a different type of matches. I took also 177  
some official odds for comparison. It was provided 178  
by the api-football API, but I also did a comparison 179  
with Fortuna’s odds [7]. I had around 85 percent of 180

181	successfully predicted results of the matches contain-	231
182	ing a strong favorite. The success of prediction means	232
183	the right prediction of win/draw/loses. It was worse	233
184	for mixed datasets, where it was around 65 percent of	234
185	successful predictions. Then, I did a competition with	235
186	Fortuna on near planned football matches through the	236
187	whole weekend in the biggest football leagues, which	237
188	are mentioned upper. The most important comparison	238
189	here was between my prediction and odd prediction	
190	(my prediction versus the lowest odd on each match).	
191	Betting company is still a winner, but not too heavily.	
192	In each iteration, they beat me just by a few percents	
193	(average is 6 percent, which is 3 matches on average—	
194	from 49 weekend matches). I chose a polynomial	
195	regression model and the rate between training and	
196	testing data is 0.8/0.2. The final model is extracted,	
197	but it requires preprocessed data from the system and	
198	api-football API.	
199	<b>4.3 Prediction with Randomness</b>	
200	The simulation of future matches uses mostly histor-	
201	ical simulator's data (it calculates new results from	
202	the results of the last 10 matches and last 5 years).	
203	Some random factors can have an impact in later years.	
204	These main random factors are used mainly to simu-	
205	late extra motivation, or frustration, or importance of	
206	upcoming matches which can produce an unexpected	
207	score. It generates the rank of "randomness".	
208	Then, the final result is mixed with the result gener-	
209	ated by the neural network. The algorithm works with	
210	a tape of numbers representing the random difference	
211	from the predicted results. In the beginning, there are	
212	mostly the numbers 0 and 1 (but there are also higher	
213	numbers). If the level of randomness should be higher	
214	than I add higher numbers (2, 3, 4, ...). Some of the	
215	numbers of this tape is chosen for each team. The cho-	
216	sen values can add or remove some goals which were	
217	predicted by the neural network. Mostly, the result	
218	will be similar to the prediction, but there can be a	
219	situation (which happens also in real football world)	
220	when a result is shocking. The goal is not to have still	
221	the same results, but it is not going to be completely	
222	random. Hence, I take the result of the prediction, gen-	
223	erate the random value and use the value to change the	
224	originally predicted result.	
225	<b>5. Testing</b>	
226	<b>5.1 Target Groups</b>	
227	Last year, before the beginning of the work, I spoke	
228	with a group of different people who are involved and	
229	deeply interested in football or football management	
230	games. I asked them for help with testing the sys-	
	tem. I tried to ask both experts and casuals. As an	231
	expert, I mean a person who deeply understands foot-	232
	ball leagues structure and watches more than just his	233
	favourite teams. As a casual, I mean a person who is	234
	just a fan of some teams and is not really interested in	235
	complex statistics etc. The website is going to be free	236
	for all, however, if users would like to edit the date for	237
	their own there will be a small fee for this addition.	238
	<b>5.2 Actual testing</b>	239
	Firstly, when I finished the UI templates, I asked the	240
	users for feedback and they gave me a lot of hints.	241
	Currently, they are giving me feedback about the actual	242
	functionality of system and prediction. It helped me to	243
	fix some bugs.	244
	Many of the users made long-lasting simulation	245
	so they gave me feedback on how the football world	246
	can evolve in the future. Such testing would be time-	247
	consuming for one person. Their simulations com-	248
	pletely differ in later years but had some common	249
	results in the nearby future. After decades of sim-	250
	ulation, some football giants have fallen into lower	251
	leagues and new giants were born. It is quite different	252
	for everybody, but this is part of the fun. The system is	253
	currently not publicly accessible. It is planned before	254
	the deadline of the thesis (end of May). So I gave users	255
	my laptop for a few days to check the simulator.	256
	<b>6. Plans for the Future</b>	257
	The project is developed as part of a master thesis but	258
	I plan to continue with the development afterwards.	259
	Firstly, I plan to add all European leagues (at least first	260
	and second level). Secondly, other football competi-	261
	tions will be included—e. g., international football	262
	(World Cups, European Championships and its qual-	263
	ifiers). And also, as I said before, I plan to add some	264
	customizers for users. It will be possible to edit/add	265
	their own teams, players or make some transfers so	266
	the users can check how the world changes after. In	267
	the year 2020, I am going to move the application to	268
	other platforms (Windows: desktop application, iOS	269
	and Android). It is possible because I made universal	270
	back-end using REST API. However, users need to be	271
	connected to the Internet on all platforms.	272
	<b>7. Conclusion</b>	273
	This paper provides information about Footie—the	274
	web football simulator. The simulator generates match,	275
	league and season results and is not limited by a num-	276
	ber of seasons. It is practically limited just by the	277
	amount of data which can be stored.	278



**Figure 4.** Architecture of the system

In the beginning, it was important to define the main idea of the system. I had a lot of concepts. I tried to compare my concepts with existing solutions (mainly football management games and match prediction for making bets). Then, I designed a which is probably on the halfway to both categories.

Then, I made an architecture that includes multiple modules. The main modules are the web system, back-end API, neural network and API with real historical results. The main goal was to create match predictor which combines the predicted results with regulated random factors in order to bring both realistic but sometimes unpredictable and crazy results (as real football world does). It was tested on both historical and real-time matches with the help of betting odds.

At the current stage, the system includes basic simulation of the biggest European leagues and European club cups. The system is composed of modules, so it is not a problem to add or remove any module. The web application is designed responsively so it is runnable on all devices which has a modern internet browser. Its database is optimized as much as possible because of the amount of data. It included around 280 leagues and 70 000 players. All these leagues are going to be playable in the future. Currently, the app is still in development so it can be only seen locally, but at least its first phase prototype is going to be released by the end of May 2019.

## References

- [1] Bert Kassies. *UEFA Coefficients calculation method*. <https://kassiesa.home.xs4all.nl/bert/uefa/calc.html>.
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- [4] Jaroslav Bezděk. *Aplikace metod strojového učení na predikci výsledků fotbalových zápasů*. 2015.
- [5] SciKit-Learn Developers. *scikit-learn user guide*. [Online; 16.3.2019].
- [6] Api football REST API. <https://www.api-football.com>.
- [7] Fortuna official website. <https://www.ifortuna.cz>.

## 8. Screenshots of the application

2017/2018

CZECH REPUBLIC				ENGLAND			
Het Liga		Narodni Liga		English Premier Division		Sky Bet Championship	
1.	Bank Ostrava	0		1.	AFC Bournemouth	0	
2.	Viktoria Plzen	0		2.	Manchester United	0	
3.	Teplice	0		3.	Newcastle United	0	
4.	Sparta Prague	0		4.	Southampton	0	
5.	Slovan Liberec	0		5.	Stoke City	0	
6.	Slovakia	0		6.	Swansea City	0	
1.	Ceske Budejovice	0		1.	Aston Villa	0	
2.	Vlasim	0		2.	Middlesbrough	0	
3.	Vitkovice	0		3.	Millwall	0	
4.	Viktoria Ziskov	0		4.	Norwich City	0	
5.	Varnsdorf	0		5.	Nottingham Forest	0	
6.	Usti nad Labem	0		6.	Preston North End	0	

FRANCE				GERMANY			
French Under 17 League		Ligue 1 Conforama		German First Division		German Second Division	
1.	Association Sportive Orly	0		1.	Augsburg	0	
2.	Cavigal Nice Sports Football	0		2.	Stuttgart	0	
				3.	RBL	0	
				4.	Mainz	0	
1.	Amiens SC	0		1.	Aue	0	
2.	Montpellier Hérault SC	0		2.	SG Dynamo Dresden	0	
3.	OGC Nice	0		3.	Sandhausen	0	
4.	Olympique de Marseille	0		4.	Nürnberg	0	

Figure 5. Main screen - list of nations

English Premier Division

Tottenham Hotspur

0 - 0 - 0

PLAYERS

RESULTS

FIXTURES

TOP SCORERS

HISTORY

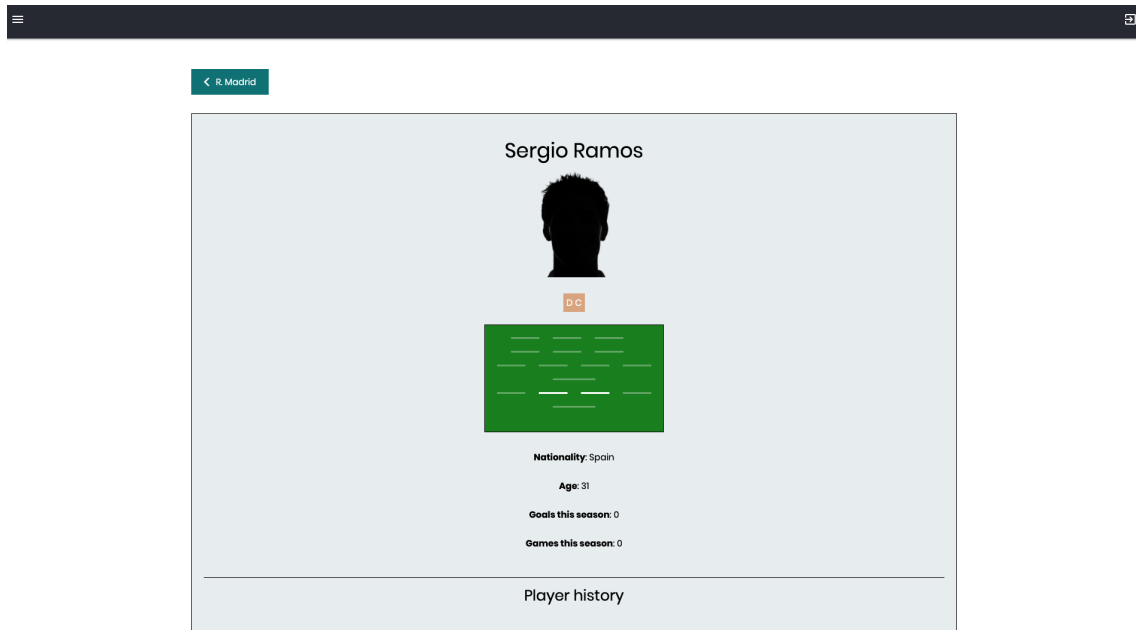
<div>Matchday 1</div> <div> </div> <div>EVERTON</div> <div>AWAY</div>	<div>Matchday 2</div> <div> </div> <div>CHELSEA</div> <div>HOME</div>	<div>Matchday 3</div> <div> </div> <div>BRIGHTON AND HOVE ALBION</div> <div>HOME</div>	<div>Matchday 4</div> <div> </div> <div>LIVERPOOL</div> <div>HOME</div>
<div>Matchday 5</div> <div> </div> <div>NEWCASTLE UNITED</div> <div>HOME</div>	<div>Matchday 6</div> <div> </div> <div>STOKE CITY</div> <div>HOME</div>	<div>Matchday 7</div> <div> </div> <div>AFC BOURNEMOUTH</div> <div>AWAY</div>	<div>Matchday 8</div> <div> </div> <div>WEST BROMWICH ALBION</div> <div>AWAY</div>

Figure 6. Team screen - upcoming fixtures

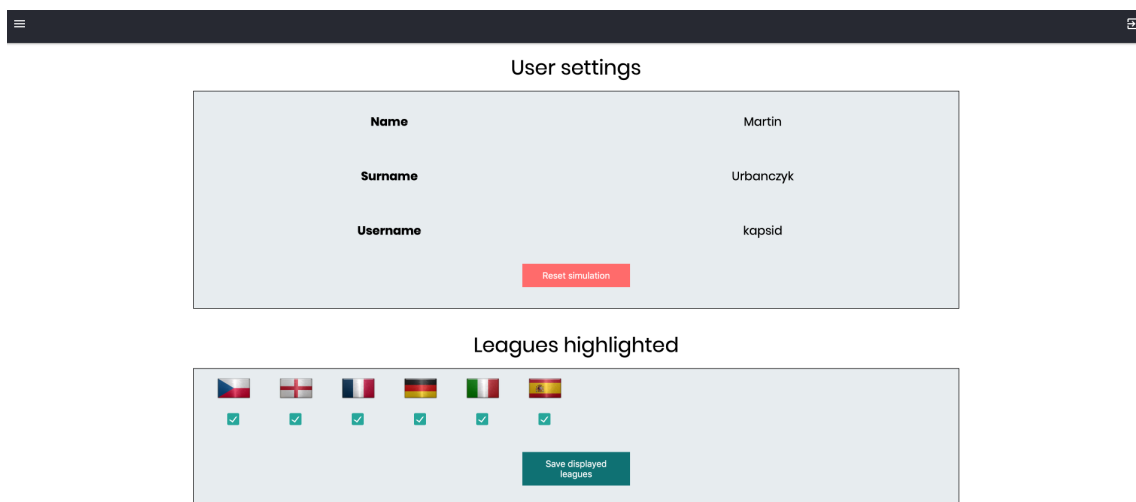
PLAYERS	RESULTS	FIXTURES	TOP SCORERS	HISTORY			
Name and Nation	Position	Value	Games	Goals	Assists	Yellow cards	Red cards
ALFIE WHITEMAN (England)	GK	£92k	0	0	0	0	0
ANTHONY GEORGIOU (England)	AM L	£350k	0	0	0	0	0
ARMANDO SHASHOUA (England)	AM C	£5k	0	0	0	0	0
BEN DAVIES (Wales)	D/WB L	£18m	0	0	0	0	0
BRANDON AUSTIN (England)	GK	£17k	0	0	0	0	0
BROOKLYN (Lyons)	D RLC	£22k	0	0	0	0	0
CHARLIE FREEMAN (England)	GK	£11k	0	0	0	0	0
CHRISTIAN ERIKSEN (Denmark)	AM LC	£58m	0	0	0	0	0
CHRISTIAN MAGHOMA (DR Congo)	D C	£52k	0	0	0	0	0
CY GODDARD (Japan)	M C	£17k	0	0	0	0	0
DANNY ROSE (England)	D/WB/M L	£26m	0	0	0	0	0
DAVINSON SÁNCHEZ (Colombia)	D C	£24m	0	0	0	0	0
DELE ADE (England)	AM RLC	£48m	0	0	0	0	0

Figure 7. Team screen - tables with the players

Figure 8. Competition screen - actual round results/fixtures



**Figure 9.** Player detail screen



**Figure 10.** Settings of simulated leagues