

# zerotopandas-course-project

November 14, 2020

## 1 FIFA19 Players Data Analysis

FIFA19 is the official football game of EA Sports. Queries in the following project are based on the Player Dataset of FIFA19. This Dataset is available in Kaggle which is a hub of datasets. This dataset consists of details of players and their stats in the year 2019. This can be used to determine success ratio, ratings, top players etc.

In this project I have used Numpy, Pandas, Matplotlib and Seaborn. This project is part of the course [Data Analysis with Python: Zero to Pandas](#). It is a mandatory project to mark the success of the course (untill evaluated to PASS).

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#### 1.1.1 How to run the code

This is an executable [Jupyter notebook](#) hosted on [Jovian.ml](#), a platform for sharing data science projects. You can run and experiment with the code in a couple of ways: *using free online resources* (recommended) or *on your own computer*.

**Option 1: Running using free online resources (1-click, recommended)** The easiest way to start executing this notebook is to click the “Run” button at the top of this page, and select “Run on Binder”. This will run the notebook on [mybinder.org](https://mybinder.org), a free online service for running Jupyter notebooks. You can also select “Run on Colab” or “Run on Kaggle”.

## Option 2: Running on your computer locally

1. Install Conda by [following these instructions](#). Add Conda binaries to your system PATH, so you can use the `conda` command on your terminal.
2. Create a Conda environment and install the required libraries by running these commands on the terminal:

```
conda create -n zerotopandas -y python=3.8
conda activate zerotopandas
pip install jovian jupyter numpy pandas matplotlib seaborn opendatasets --upgrade
```

3. Press the “Clone” button above to copy the command for downloading the notebook, and run it on the terminal. This will create a new directory and download the notebook. The command will look something like this:

```
jovian clone notebook-owner/notebook-id
```

4. Enter the newly created directory using `cd directory-name` and start the Jupyter notebook.

```
jupyter notebook
```

You can now access Jupyter’s web interface by clicking the link that shows up on the terminal or by visiting `http://localhost:8888` on your browser. Click on the notebook file (it has a `.ipynb` extension) to open it.

## 1.2 1. Downloading the Dataset

Datasets can be downloaded withing Jupyter using the `opendatasets` Python Library.

```
[1]: !pip install jovian opendatasets --upgrade --quiet
```

Let’s begin by downloading the data, and listing the files within the dataset.

```
[2]: dataset_url = 'https://www.kaggle.com/karangadiya/fifa19'
```

```
[3]: import opendatasets as od
      od.download(dataset_url)
```

Please provide your Kaggle credentials to download this dataset. Learn more:

`http://bit.ly/kaggle-creds`

Your Kaggle username: `gauravbisht005`

Your Kaggle Key: `.....`

100%| | 2.18M/2.18M [00:00<00:00, 103MB/s]

Downloading `fifa19.zip` to `./fifa19`

The dataset has been downloaded and extracted.

```
[4]: data_dir = './fifa19'
```

```
[5]: import os
      os.listdir(data_dir)
```

```
[5]: ['data.csv']
```

Let us save and upload our work to Jovian before continuing.

```
[6]: project_name = "FIFA19-Player-Data-Analysis"
```

```
[7]: !pip install jovian --upgrade -q
```

```
[8]: import jovian
```

```
[9]: jovian.commit(project=project_name)
```

```
<IPython.core.display.Javascript object>
```

```
[jovian] Attempting to save notebook..
```

```
[jovian] Please enter your API key ( from https://jovian.ml/ ):
```

```
API KEY: .....
```

```
[jovian] Updating notebook "gauravbisht005/fifa19-player-data-analysis" on
https://jovian.ml/
```

```
[jovian] Uploading notebook..
```

```
[jovian] Capturing environment..
```

```
[jovian] Committed successfully! https://jovian.ml/gauravbisht005/fifa19-player-
data-analysis
```

```
[9]: 'https://jovian.ml/gauravbisht005/fifa19-player-data-analysis'
```

## 1.3 2. Data Preparation and Cleaning

Data Cleaning is the process of finding and correcting inaccurate or incomplete records in dataset by replacing/ modifying/ removing those records so that the dataset is prepared to be operated.

```
[10]: import pandas as pd
```

```
[11]: fifa19_df = pd.read_csv(data_dir + "/data.csv")
```

```
[12]: fifa19_df
```

```
[12]:
```

	Unnamed: 0	ID	Name	Age	\
0	0	158023	L. Messi	31	
1	1	20801	Cristiano Ronaldo	33	

2	2	190871	Neymar Jr	26
3	3	193080	De Gea	27
4	4	192985	K. De Bruyne	27
...	...	...	...	...
18202	18202	238813	J. Lundstram	19
18203	18203	243165	N. Christoffersson	19
18204	18204	241638	B. Worman	16
18205	18205	246268	D. Walker-Rice	17
18206	18206	246269	G. Nugent	16

			Photo	Nationality	\
0			<a href="https://cdn.sofifa.org/players/4/19/158023.png">https://cdn.sofifa.org/players/4/19/158023.png</a>	Argentina	
1			<a href="https://cdn.sofifa.org/players/4/19/20801.png">https://cdn.sofifa.org/players/4/19/20801.png</a>	Portugal	
2			<a href="https://cdn.sofifa.org/players/4/19/190871.png">https://cdn.sofifa.org/players/4/19/190871.png</a>	Brazil	
3			<a href="https://cdn.sofifa.org/players/4/19/193080.png">https://cdn.sofifa.org/players/4/19/193080.png</a>	Spain	
4			<a href="https://cdn.sofifa.org/players/4/19/192985.png">https://cdn.sofifa.org/players/4/19/192985.png</a>	Belgium	
...			...	...	
18202			<a href="https://cdn.sofifa.org/players/4/19/238813.png">https://cdn.sofifa.org/players/4/19/238813.png</a>	England	
18203			<a href="https://cdn.sofifa.org/players/4/19/243165.png">https://cdn.sofifa.org/players/4/19/243165.png</a>	Sweden	
18204			<a href="https://cdn.sofifa.org/players/4/19/241638.png">https://cdn.sofifa.org/players/4/19/241638.png</a>	England	
18205			<a href="https://cdn.sofifa.org/players/4/19/246268.png">https://cdn.sofifa.org/players/4/19/246268.png</a>	England	
18206			<a href="https://cdn.sofifa.org/players/4/19/246269.png">https://cdn.sofifa.org/players/4/19/246269.png</a>	England	

			Flag	Overall	Potential	\
0			<a href="https://cdn.sofifa.org/flags/52.png">https://cdn.sofifa.org/flags/52.png</a>	94	94	
1			<a href="https://cdn.sofifa.org/flags/38.png">https://cdn.sofifa.org/flags/38.png</a>	94	94	
2			<a href="https://cdn.sofifa.org/flags/54.png">https://cdn.sofifa.org/flags/54.png</a>	92	93	
3			<a href="https://cdn.sofifa.org/flags/45.png">https://cdn.sofifa.org/flags/45.png</a>	91	93	
4			<a href="https://cdn.sofifa.org/flags/7.png">https://cdn.sofifa.org/flags/7.png</a>	91	92	
...			...	...	...	
18202			<a href="https://cdn.sofifa.org/flags/14.png">https://cdn.sofifa.org/flags/14.png</a>	47	65	
18203			<a href="https://cdn.sofifa.org/flags/46.png">https://cdn.sofifa.org/flags/46.png</a>	47	63	
18204			<a href="https://cdn.sofifa.org/flags/14.png">https://cdn.sofifa.org/flags/14.png</a>	47	67	
18205			<a href="https://cdn.sofifa.org/flags/14.png">https://cdn.sofifa.org/flags/14.png</a>	47	66	
18206			<a href="https://cdn.sofifa.org/flags/14.png">https://cdn.sofifa.org/flags/14.png</a>	46	66	

	Club	...	Composure	Marking	StandingTackle	\
0	FC Barcelona	...	96.0	33.0	28.0	
1	Juventus	...	95.0	28.0	31.0	
2	Paris Saint-Germain	...	94.0	27.0	24.0	
3	Manchester United	...	68.0	15.0	21.0	
4	Manchester City	...	88.0	68.0	58.0	
...	...	...	...	...	...	
18202	Crewe Alexandra	...	45.0	40.0	48.0	
18203	Trelleborgs FF	...	42.0	22.0	15.0	
18204	Cambridge United	...	41.0	32.0	13.0	
18205	Tranmere Rovers	...	46.0	20.0	25.0	

18206	Tranmere Rovers	...	43.0	40.0	43.0
	SlidingTackle	GKDividing	GKHandling	GKKicking	GKPositioning \
0	26.0	6.0	11.0	15.0	14.0
1	23.0	7.0	11.0	15.0	14.0
2	33.0	9.0	9.0	15.0	15.0
3	13.0	90.0	85.0	87.0	88.0
4	51.0	15.0	13.0	5.0	10.0
...	...	...	...	...	...
18202	47.0	10.0	13.0	7.0	8.0
18203	19.0	10.0	9.0	9.0	5.0
18204	11.0	6.0	5.0	10.0	6.0
18205	27.0	14.0	6.0	14.0	8.0
18206	50.0	10.0	15.0	9.0	12.0
	GKReflexes	Release	Clause		
0	8.0	€226.5M			
1	11.0	€127.1M			
2	11.0	€228.1M			
3	94.0	€138.6M			
4	13.0	€196.4M			
...	...	...			
18202	9.0	€143K			
18203	12.0	€113K			
18204	13.0	€165K			
18205	9.0	€143K			
18206	9.0	€165K			

[18207 rows x 89 columns]

### 1.3.1 2.1 Determining the number of attributes in the dataset

```
[13]: fifa19_df.columns
```

```
[13]: Index(['Unnamed: 0', 'ID', 'Name', 'Age', 'Photo', 'Nationality', 'Flag',
'Overall', 'Potential', 'Club', 'Club Logo', 'Value', 'Wage', 'Special',
'Preferred Foot', 'International Reputation', 'Weak Foot',
'Skill Moves', 'Work Rate', 'Body Type', 'Real Face', 'Position',
'Jersey Number', 'Joined', 'Loaned From', 'Contract Valid Until',
'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', 'FKAccuracy', 'LongPassing', 'BallControl', 'Acceleration',
'SprintSpeed', 'Agility', 'Reactions', 'Balance', 'ShotPower',
'Jumping', 'Stamina', 'Strength', 'LongShots', 'Aggression',
'Interceptions', 'Positioning', 'Vision', 'Penalties', 'Composure',
```

```
'Marking', 'StandingTackle', 'SlidingTackle', 'GKDividing', 'GKHandling',
'GKkicking', 'GKPositioning', 'GKReflexes', 'Release Clause'],
dtype='object')
```

### 1.3.2 2.2 Cleaning the dataset

```
[14]: missing_data = pd.isna(fifa19_df.columns).sum()
missing_data
```

```
[14]: 0
```

Fortunately, there was no missing data in the dataset!!

```
[15]: fifa19_df.drop("Unnamed: 0",axis=1, inplace= True)
fifa19_df
```

```
[15]:
```

	ID	Name	Age	\
0	158023	L. Messi	31	
1	20801	Cristiano Ronaldo	33	
2	190871	Neymar Jr	26	
3	193080	De Gea	27	
4	192985	K. De Bruyne	27	
...	...	...	...	
18202	238813	J. Lundstram	19	
18203	243165	N. Christoffersson	19	
18204	241638	B. Worman	16	
18205	246268	D. Walker-Rice	17	
18206	246269	G. Nugent	16	

	Photo	Nationality	\
0	<a href="https://cdn.sofifa.org/players/4/19/158023.png">https://cdn.sofifa.org/players/4/19/158023.png</a>	Argentina	
1	<a href="https://cdn.sofifa.org/players/4/19/20801.png">https://cdn.sofifa.org/players/4/19/20801.png</a>	Portugal	
2	<a href="https://cdn.sofifa.org/players/4/19/190871.png">https://cdn.sofifa.org/players/4/19/190871.png</a>	Brazil	
3	<a href="https://cdn.sofifa.org/players/4/19/193080.png">https://cdn.sofifa.org/players/4/19/193080.png</a>	Spain	
4	<a href="https://cdn.sofifa.org/players/4/19/192985.png">https://cdn.sofifa.org/players/4/19/192985.png</a>	Belgium	
...	...	...	
18202	<a href="https://cdn.sofifa.org/players/4/19/238813.png">https://cdn.sofifa.org/players/4/19/238813.png</a>	England	
18203	<a href="https://cdn.sofifa.org/players/4/19/243165.png">https://cdn.sofifa.org/players/4/19/243165.png</a>	Sweden	
18204	<a href="https://cdn.sofifa.org/players/4/19/241638.png">https://cdn.sofifa.org/players/4/19/241638.png</a>	England	
18205	<a href="https://cdn.sofifa.org/players/4/19/246268.png">https://cdn.sofifa.org/players/4/19/246268.png</a>	England	
18206	<a href="https://cdn.sofifa.org/players/4/19/246269.png">https://cdn.sofifa.org/players/4/19/246269.png</a>	England	

	Flag	Overall	Potential	\
0	<a href="https://cdn.sofifa.org/flags/52.png">https://cdn.sofifa.org/flags/52.png</a>	94	94	
1	<a href="https://cdn.sofifa.org/flags/38.png">https://cdn.sofifa.org/flags/38.png</a>	94	94	
2	<a href="https://cdn.sofifa.org/flags/54.png">https://cdn.sofifa.org/flags/54.png</a>	92	93	
3	<a href="https://cdn.sofifa.org/flags/45.png">https://cdn.sofifa.org/flags/45.png</a>	91	93	

4	<a href="https://cdn.sofifa.org/flags/7.png">https://cdn.sofifa.org/flags/7.png</a>	91	92
...	...	...	...
18202	<a href="https://cdn.sofifa.org/flags/14.png">https://cdn.sofifa.org/flags/14.png</a>	47	65
18203	<a href="https://cdn.sofifa.org/flags/46.png">https://cdn.sofifa.org/flags/46.png</a>	47	63
18204	<a href="https://cdn.sofifa.org/flags/14.png">https://cdn.sofifa.org/flags/14.png</a>	47	67
18205	<a href="https://cdn.sofifa.org/flags/14.png">https://cdn.sofifa.org/flags/14.png</a>	47	66
18206	<a href="https://cdn.sofifa.org/flags/14.png">https://cdn.sofifa.org/flags/14.png</a>	46	66

	Club	Club Logo \
0	FC Barcelona	<a href="https://cdn.sofifa.org/teams/2/light/241.png">https://cdn.sofifa.org/teams/2/light/241.png</a>
1	Juventus	<a href="https://cdn.sofifa.org/teams/2/light/45.png">https://cdn.sofifa.org/teams/2/light/45.png</a>
2	Paris Saint-Germain	<a href="https://cdn.sofifa.org/teams/2/light/73.png">https://cdn.sofifa.org/teams/2/light/73.png</a>
3	Manchester United	<a href="https://cdn.sofifa.org/teams/2/light/11.png">https://cdn.sofifa.org/teams/2/light/11.png</a>
4	Manchester City	<a href="https://cdn.sofifa.org/teams/2/light/10.png">https://cdn.sofifa.org/teams/2/light/10.png</a>
...	...	...
18202	Crewe Alexandra	<a href="https://cdn.sofifa.org/teams/2/light/121.png">https://cdn.sofifa.org/teams/2/light/121.png</a>
18203	Trelleborgs FF	<a href="https://cdn.sofifa.org/teams/2/light/703.png">https://cdn.sofifa.org/teams/2/light/703.png</a>
18204	Cambridge United	<a href="https://cdn.sofifa.org/teams/2/light/1944.png">https://cdn.sofifa.org/teams/2/light/1944.png</a>
18205	Tranmere Rovers	<a href="https://cdn.sofifa.org/teams/2/light/15048.png">https://cdn.sofifa.org/teams/2/light/15048.png</a>
18206	Tranmere Rovers	<a href="https://cdn.sofifa.org/teams/2/light/15048.png">https://cdn.sofifa.org/teams/2/light/15048.png</a>

	...	Composure	Marking	StandingTackle	SlidingTackle	GKDividing \
0	...	96.0	33.0	28.0	26.0	6.0
1	...	95.0	28.0	31.0	23.0	7.0
2	...	94.0	27.0	24.0	33.0	9.0
3	...	68.0	15.0	21.0	13.0	90.0
4	...	88.0	68.0	58.0	51.0	15.0
...	...	...	...	...	...	...
18202	...	45.0	40.0	48.0	47.0	10.0
18203	...	42.0	22.0	15.0	19.0	10.0
18204	...	41.0	32.0	13.0	11.0	6.0
18205	...	46.0	20.0	25.0	27.0	14.0
18206	...	43.0	40.0	43.0	50.0	10.0

	GKHandling	GKKicking	GKPositioning	GKReflexes	Release Clause
0	11.0	15.0	14.0	8.0	€226.5M
1	11.0	15.0	14.0	11.0	€127.1M
2	9.0	15.0	15.0	11.0	€228.1M
3	85.0	87.0	88.0	94.0	€138.6M
4	13.0	5.0	10.0	13.0	€196.4M
...	...	...	...	...	...
18202	13.0	7.0	8.0	9.0	€143K
18203	9.0	9.0	5.0	12.0	€113K
18204	5.0	10.0	6.0	13.0	€165K
18205	6.0	14.0	8.0	9.0	€143K
18206	15.0	9.0	12.0	9.0	€165K

[18207 rows x 88 columns]

There was no missing data in the dataset but an unnecessary attribute which has been removed above.

```
[16]: import jovian
```

```
[17]: jovian.commit()
```

```
<IPython.core.display.Javascript object>
```

```
[jovian] Attempting to save notebook..  
[jovian] Updating notebook "gauravbisht005/fifa19-player-data-analysis" on  
https://jovian.ml/  
[jovian] Uploading notebook..  
[jovian] Capturing environment..  
[jovian] Committed successfully! https://jovian.ml/gauravbisht005/fifa19-player-  
data-analysis
```

```
[17]: 'https://jovian.ml/gauravbisht005/fifa19-player-data-analysis'
```

### 1.4 3. Exploratory Analysis and Visualization

In the following segment, I have analysed datasets to summarise general main characteristics using visual methods with the help of Python Libraries such as Matplotlib.pyplot and Seaborn.

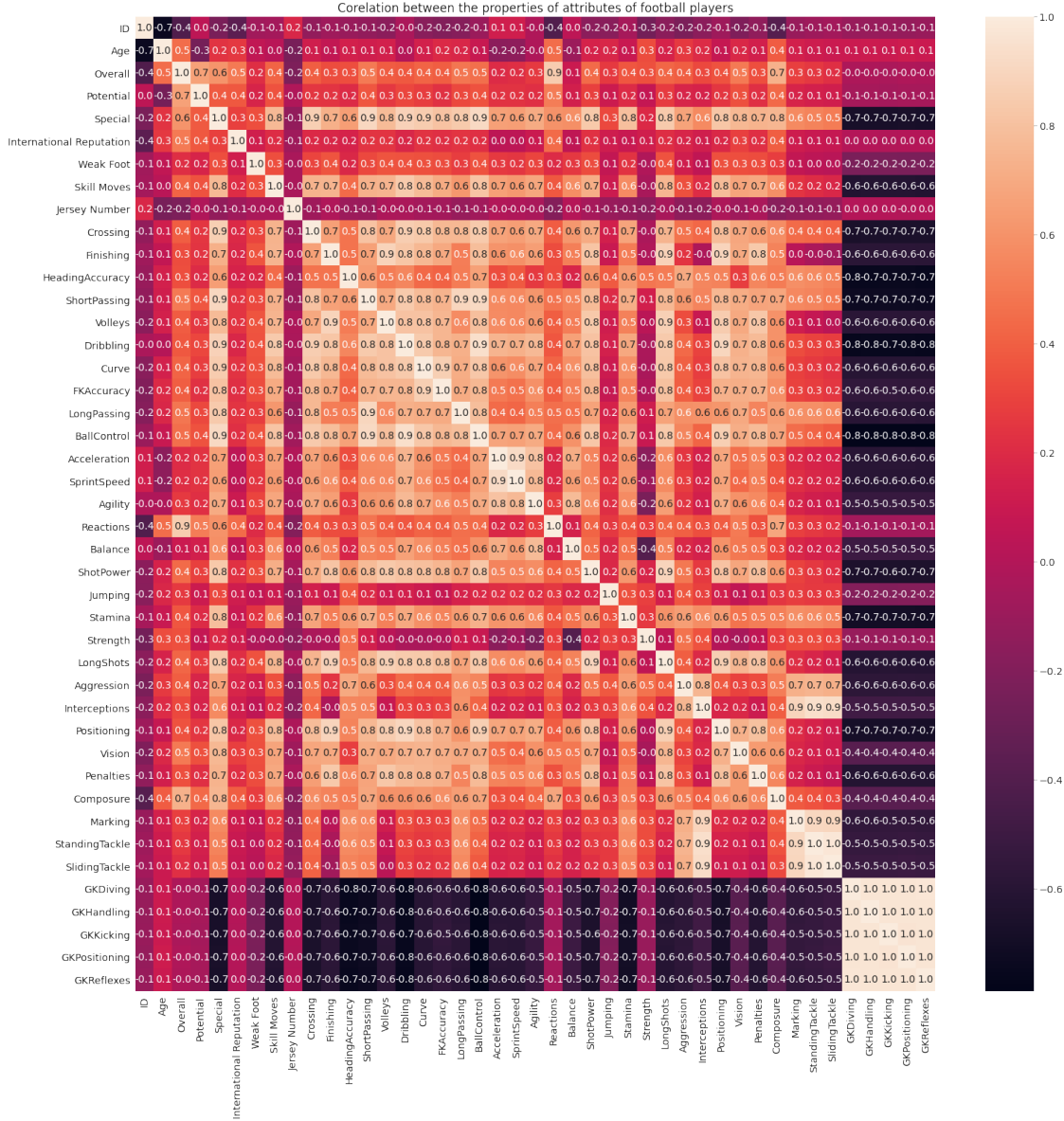
Let's begin by importing matplotlib.pyplot and seaborn.

```
[18]: import seaborn as sns  
import matplotlib  
import matplotlib.pyplot as plt  
%matplotlib inline  
  
sns.set_style('darkgrid')  
matplotlib.rcParams['font.size'] = 14  
matplotlib.rcParams['figure.figsize'] = (9, 5)  
matplotlib.rcParams['figure.facecolor'] = '#00000000'
```

#### 1.4.1 3.1 Heatmap representing relation between the properties of attributes of football players

```
[19]: plt.figure(figsize = (25, 25))  
sns.heatmap(fifa19_df.corr(), annot = True, fmt = '.1f')  
plt.title("Corelation between the properties of attributes of football players")  
plt.show()
```



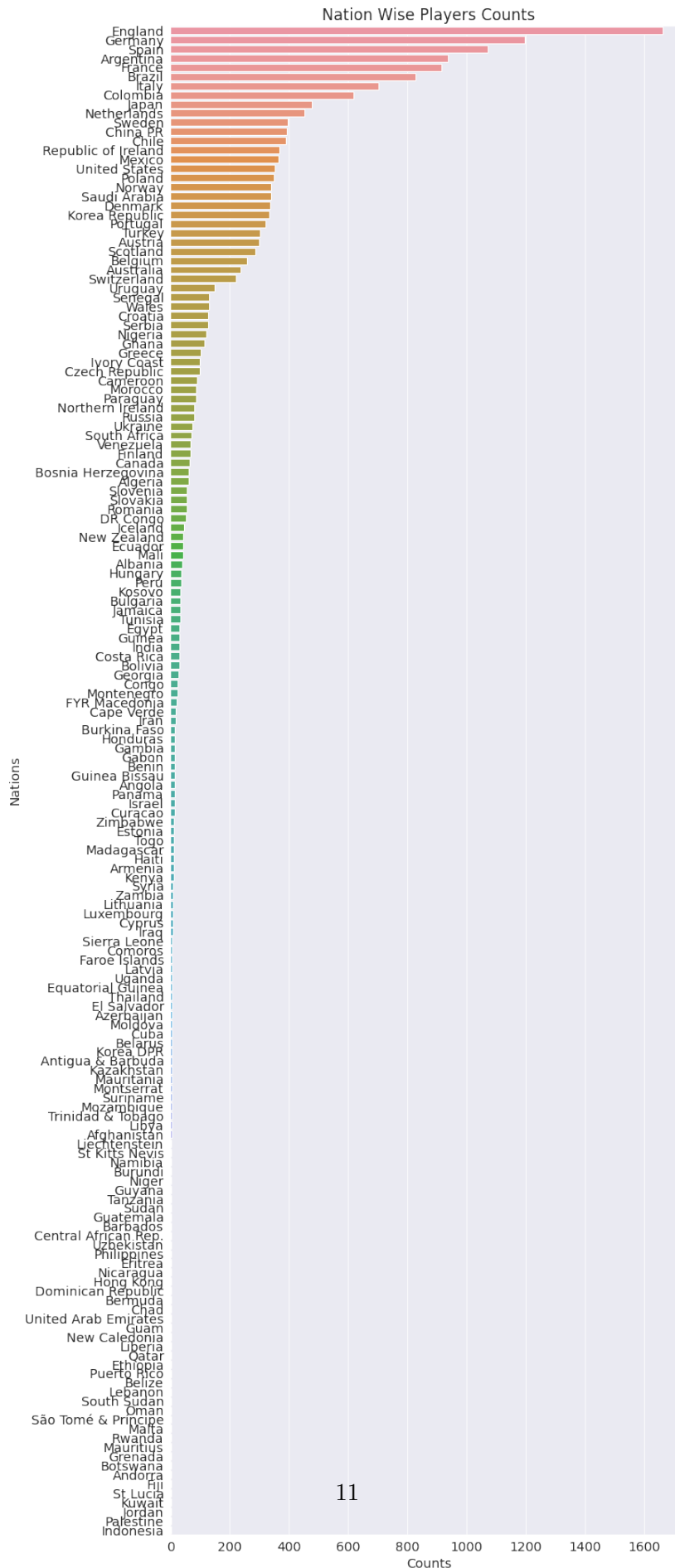


```
[20]:
```

	Nations	Counts
0	England	1662
1	Germany	1198
2	Spain	1072
3	Argentina	937
4	France	914
..	...	...
159	St Lucia	1
160	Kuwait	1
161	Jordan	1
162	Palestine	1
163	Indonesia	1

[164 rows x 2 columns]

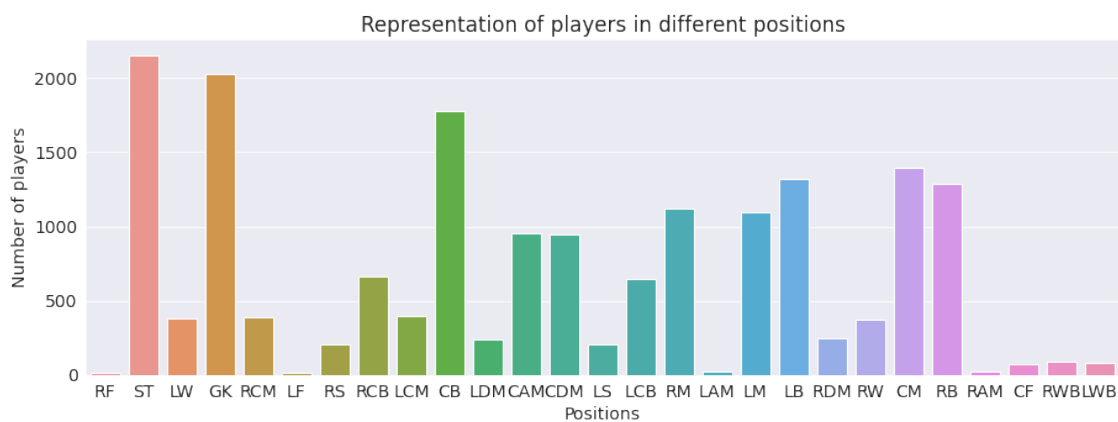
```
[21]: plt.figure(figsize = (10, 30))
sns.barplot(x = "Counts", y = "Nations", data = counts_Nationality)
plt.title("Nation Wise Players Counts");
plt.show()
```



The above graph represents that the maximum number of players in FIFA19 are from England.

### 1.4.3 3.3 Comparison of positions to number of players

```
[22]: plt.figure(figsize=(15,5))
sns.countplot('Position', data=fifa19_df)
plt.xlabel("Positions")
plt.ylabel("Number of players")
plt.title("Representation of players in different positions")
plt.show()
```



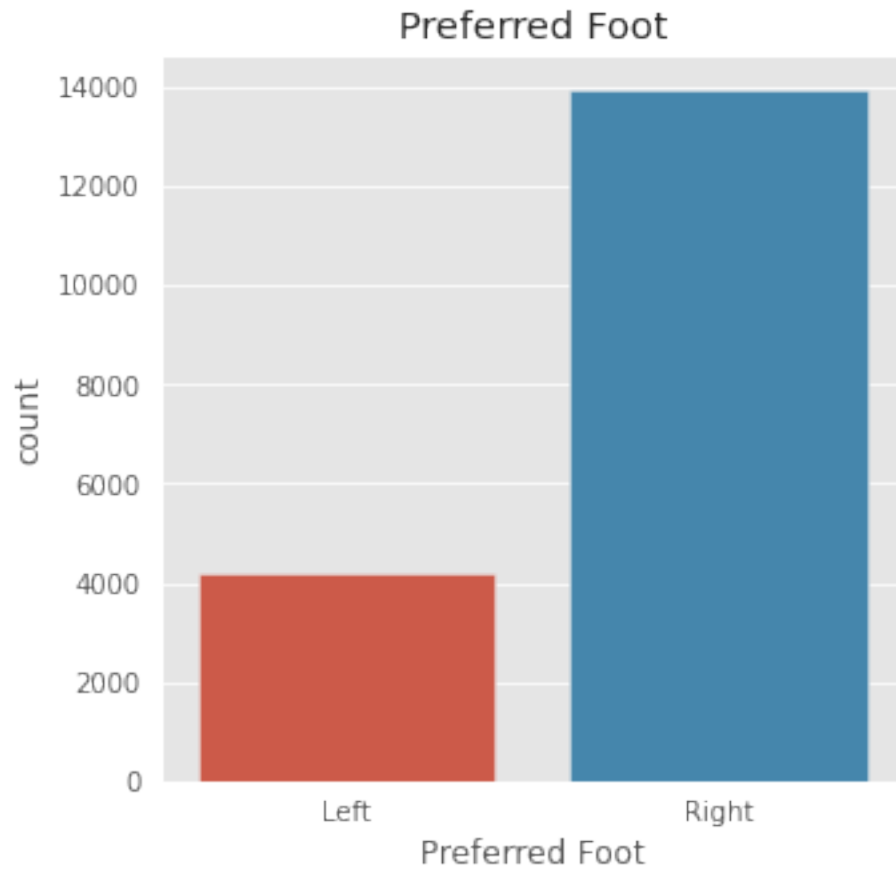
The above bar graph represents the number of players to the number of possible positions.

### 1.4.4 3.4 Preferred foot of players

```
[23]: preferred_foot = fifa19_df['Preferred Foot'].value_counts().reset_index()
preferred_foot.columns = ['Foot', "Total Players"]
preferred_foot = preferred_foot.set_index('Foot')
preferred_foot
```

```
[23]:      Total Players
Foot
Right      13948
Left       4211
```

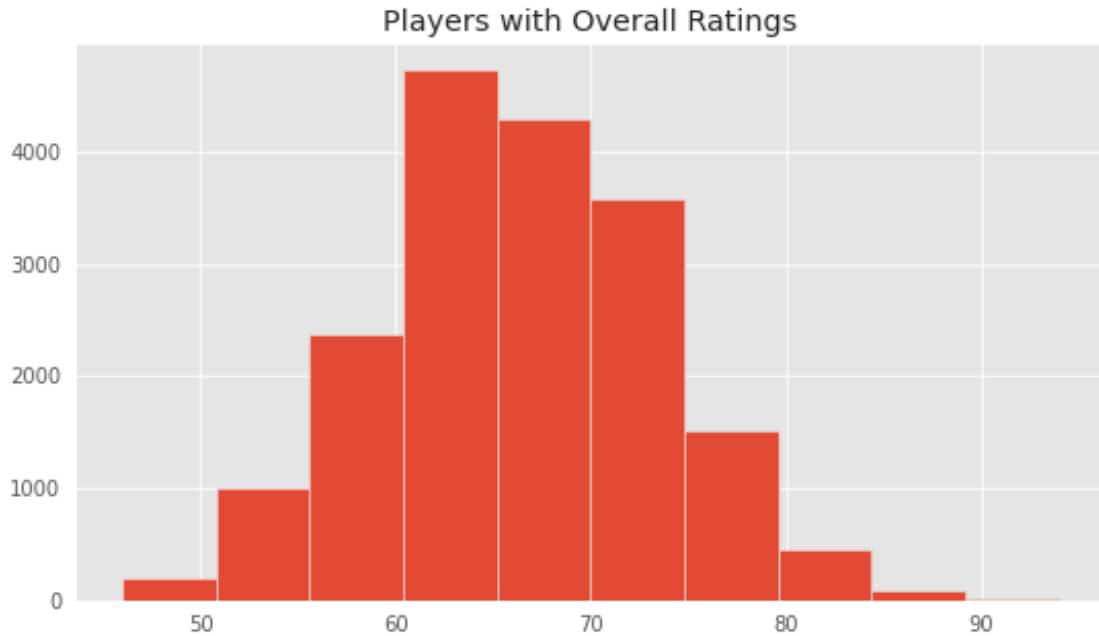
```
[24]: plt.figure(figsize = (5,5))
plt.style.use('ggplot')
sns.countplot(x=fifa19_df['Preferred Foot'])
plt.title("Preferred Foot")
plt.show()
```



The above bargraph represents the number of players which prefer right foot to the number of players which prefer left foot.

#### 1.4.5 3.5 Player's Overall (Rating) Distribution

```
[25]: plt.title("Players with Overall Ratings")  
      plt.hist(fifa19_df.Overall);
```



The above shown graph is a histogram which depicts the number of players having an Overall Rating from 1- 100 in FIFA19.

Let us save and upload our work to Jovian before continuing

```
[26]: import jovian
```

```
[27]: jovian.commit()
```

```
<IPython.core.display.Javascript object>
```

```
[jovian] Attempting to save notebook..
```

```
[jovian] Updating notebook "gauravbisht005/fifa19-player-data-analysis" on  
https://jovian.ml/
```

```
[jovian] Uploading notebook..
```

```
[jovian] Capturing environment..
```

```
[jovian] Committed successfully! https://jovian.ml/gauravbisht005/fifa19-player-  
data-analysis
```

```
[27]: 'https://jovian.ml/gauravbisht005/fifa19-player-data-analysis'
```

## 1.5 4. Asking and Answering Questions

In this segment I have answered some questions inferred from the earlier given data and the information gathered from the dataset.

**Q1: What are the different nations that are a part of FIFA?**

```
[28]: Nations = fifa19_df.Nationality.unique()
Nations
```

```
[28]: array(['Argentina', 'Portugal', 'Brazil', 'Spain', 'Belgium', 'Croatia',
'Uruguay', 'Slovenia', 'Poland', 'Germany', 'France', 'England',
'Italy', 'Egypt', 'Colombia', 'Denmark', 'Gabon', 'Wales',
'Senegal', 'Costa Rica', 'Slovakia', 'Netherlands',
'Bosnia Herzegovina', 'Morocco', 'Serbia', 'Algeria', 'Austria',
'Greece', 'Chile', 'Sweden', 'Korea Republic', 'Finland', 'Guinea',
'Montenegro', 'Armenia', 'Switzerland', 'Norway', 'Czech Republic',
'Scotland', 'Ghana', 'Central African Rep.', 'DR Congo',
'Ivory Coast', 'Russia', 'Ukraine', 'Iceland', 'Mexico', 'Jamaica',
'Albania', 'Venezuela', 'Japan', 'Turkey', 'Ecuador', 'Paraguay',
'Mali', 'Nigeria', 'Cameroon', 'Dominican Republic', 'Israel',
'Kenya', 'Hungary', 'Republic of Ireland', 'Romania',
'United States', 'Cape Verde', 'Australia', 'Peru', 'Togo',
'Syria', 'Zimbabwe', 'Angola', 'Burkina Faso', 'Iran', 'Estonia',
'Tunisia', 'Equatorial Guinea', 'New Zealand', 'FYR Macedonia',
'United Arab Emirates', 'China PR', 'Guinea Bissau', 'Bulgaria',
'Kosovo', 'South Africa', 'Madagascar', 'Georgia', 'Tanzania',
'Gambia', 'Cuba', 'Belarus', 'Uzbekistan', 'Benin', 'Congo',
'Mozambique', 'Honduras', 'Canada', 'Northern Ireland', 'Cyprus',
'Saudi Arabia', 'Curacao', 'Moldova', 'Bolivia',
'Trinidad & Tobago', 'Sierra Leone', 'Zambia', 'Chad',
'Philippines', 'Haiti', 'Comoros', 'Libya', 'Panama',
'São Tomé & Príncipe', 'Eritrea', 'Oman', 'Iraq', 'Burundi',
'Fiji', 'New Caledonia', 'Lithuania', 'Luxembourg', 'Korea DPR',
'Liechtenstein', 'St Kitts Nevis', 'Latvia', 'Suriname', 'Uganda',
'El Salvador', 'Bermuda', 'Kuwait', 'Antigua & Barbuda',
'Thailand', 'Mauritius', 'Guatemala', 'Liberia', 'Kazakhstan',
'Niger', 'Mauritania', 'Montserrat', 'Namibia', 'Azerbaijan',
'Guam', 'Faroe Islands', 'India', 'Nicaragua', 'Barbados',
'Lebanon', 'Palestine', 'Guyana', 'Sudan', 'St Lucia', 'Ethiopia',
'Puerto Rico', 'Grenada', 'Jordan', 'Rwanda', 'Qatar',
'Afghanistan', 'Hong Kong', 'Andorra', 'Malta', 'Belize',
'South Sudan', 'Indonesia', 'Botswana'], dtype=object)
```

**Q2: What is the highest number of players for a single position?**

```
[29]: position_count = fifa19_df.groupby('Position').count()
position_count = position_count['ID']
position_count
```

```
[29]: Position
CAM      958
CB       1778
CDM      948
```

```

CF      74
CM     1394
GK     2025
LAM      21
LB     1322
LCB     648
LCM     395
LDM     243
LF       15
LM     1095
LS      207
LW     381
LWB      78
RAM      21
RB     1291
RCB     662
RCM     391
RDM     248
RF       16
RM     1124
RS      203
RW     370
RWB      87
ST     2152
Name: ID, dtype: int64

```

Hence, from the information obtained above, we can come to conclusion that there are 2152 Strikers which is the position with most players in FIFA19.

### Q3: What is the dominant foot of players?

```
[30]: fifa19_df['Preferred Foot'].value_counts()
```

```

[30]: Right    13948
      Left     4211
      Name: Preferred Foot, dtype: int64

```

Hence, from the above information, we can state that the preferred foot of most of the players is Right Foot.

### Q4: What is the age distribution in FIFA19?

```
[31]: max_age = fifa19_df.Age.max()
      max_age
```

```
[31]: 45
```

The above query shows that the highest aged player in FIFA19 is 45 years old.

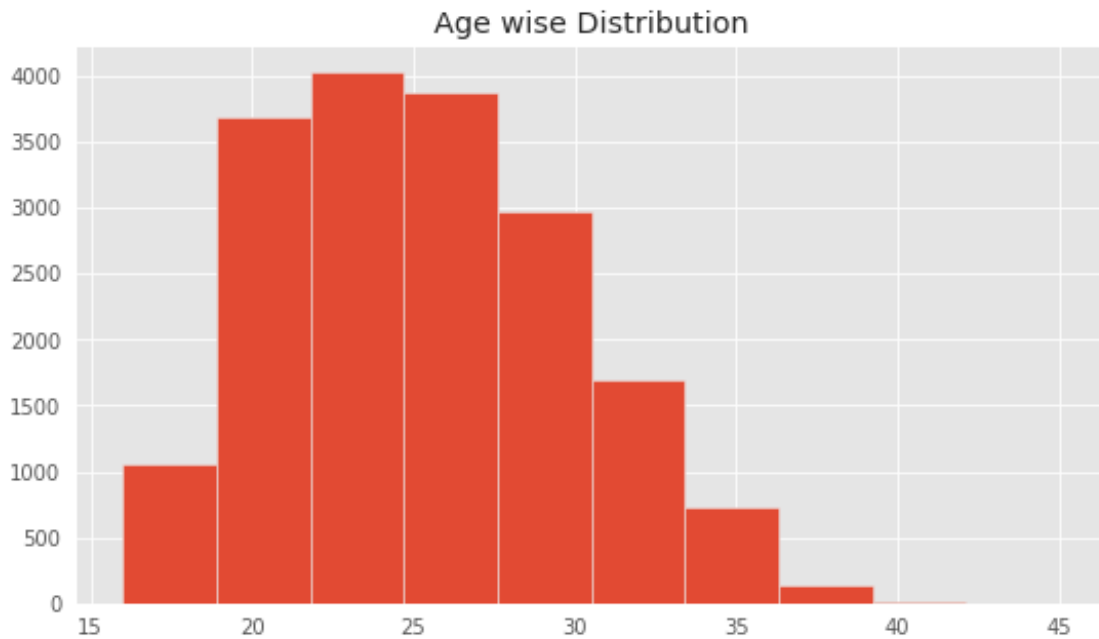


```
[32]: min_age = fifa19_df.Age.min()
min_age
```

[32]: 16

The above query shows that the youngest player in FIFA19 is just 16 years old.

```
[33]: plt.title("Age wise Distribution")
plt.hist(fifa19_df.Age);
```



Hence, the maximum number of players are of the age group 23-25. The youngest being 16 years old and the eldest being 45 years old.

**Q5: Which club has the highest Overall Rating?**

```
[34]: highest_overall_club = fifa19_df.groupby('Club').Overall.mean().reset_index().
      ↪sort_values(by='Overall', ascending=False)
highest_overall_club
```

```
[34]:
```

	Club	Overall
326	Juventus	82.280000
398	Napoli	80.000000
315	Inter	79.750000
470	Real Madrid	78.242424
382	Milan	78.074074
..	...	...
543	Sligo Rovers	56.631579

```

188      Derry City  55.777778
83      Bohemian FC 55.000000
361     Limerick FC 54.526316
92     Bray Wanderers 53.652174

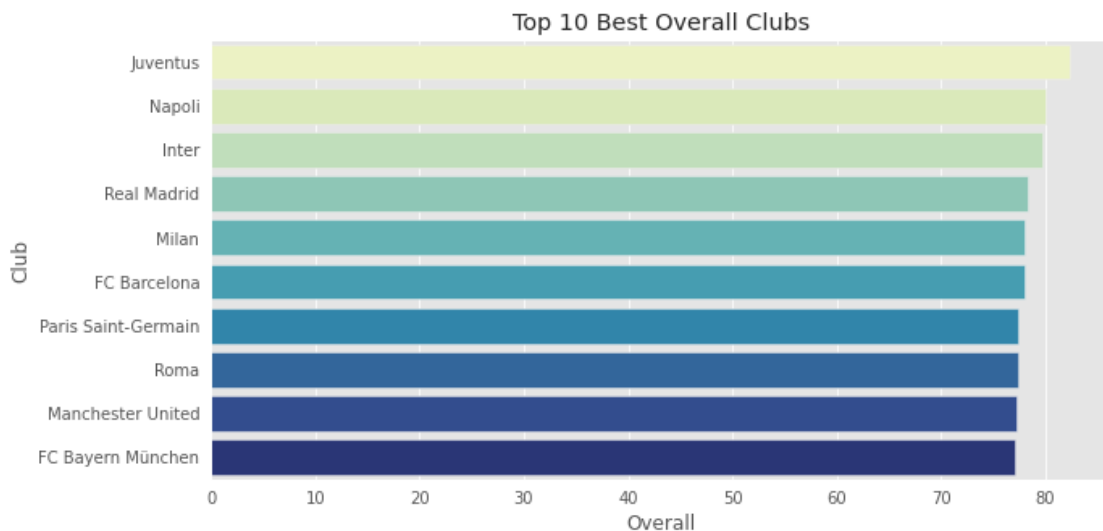
```

```
[651 rows x 2 columns]
```

```

[35]: top10_clubs = highest_overall_club.head(10)
plt.figure(figsize = (10,5))
sns.barplot(x=top10_clubs.Overall, y=top10_clubs['Club'], palette='YlGnBu')
plt.title("Top 10 Best Overall Clubs");

```



Hence, from the above information, we can say that ‘Juventus’ as a club has the highest Overall rating in FIFA19.

```
[36]: import jovian
```

```
[ ]: jovian.commit()
```

```
<IPython.core.display.Javascript object>
```

```
[jovian] Attempting to save notebook..
```

## 1.6 5. Inferences and Conclusion

From all the cleaning, preparation, analysis and visualisation we can successfully conclude few things such as: - GK in FIFA 19 have a closer mean average rating in contrast to the non-goal keeper players. - 164 countries are a part of this FIFA 19. - England constitutes the maximum number of players among any nation. - The game has maximum number of players who are strikers followed with goal keepers and Centre Backs. - More than 3/4th of the players are right footed or

prefer right foot. - Youngest player in the game is 16 years old and the oldest player is 45 years old. - Juventus is the highest overall rated club/ team in the game with an overall mean rating of 82.

```
[ ]: import jovian
```

```
[ ]: jovian.commit()
```