

# PRACTICE-PROJECT2: FIFA\_PALYER\_2019\_ANALYSIS\_WITH\_EDA

## FIFA\_PALYER\_2019\_ANALYSIS\_WITH\_EDA:

in this project we visualize players Age, Nationality, Value, Preferred Foot, International Reputation, Weak Foot, Position, Wage, Skill Moves, Height, Weight, Work Rate, Special, Potential, Overall, Body Type. we will visualize this with the help of barplot and pie chart

## import libraries `

```
In [1]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
```

## Load the Datasets

```
In [2]: fifa_data = pd.read_csv('Footballer.csv')
```

```
In [3]: fifa_data.head()
```

```
Out[3]:
```

	Unnamed: 0	ID	Name	Age	Photo	Nationality	
0	0	158023	L. Messi	31	<a href="https://cdn.sofifa.org/players/4/19/158023.png">https://cdn.sofifa.org/players/4/19/158023.png</a>	Argentina	<a href="https://cdn.sofifa.org/fla">https://cdn.sofifa.org/fla</a>
1	1	20801	Cristiano Ronaldo	33	<a href="https://cdn.sofifa.org/players/4/19/20801.png">https://cdn.sofifa.org/players/4/19/20801.png</a>	Portugal	<a href="https://cdn.sofifa.org/fla">https://cdn.sofifa.org/fla</a>
2	2	190871	Neymar Jr	26	<a href="https://cdn.sofifa.org/players/4/19/190871.png">https://cdn.sofifa.org/players/4/19/190871.png</a>	Brazil	<a href="https://cdn.sofifa.org/fla">https://cdn.sofifa.org/fla</a>
3	3	193080	De Gea	27	<a href="https://cdn.sofifa.org/players/4/19/193080.png">https://cdn.sofifa.org/players/4/19/193080.png</a>	Spain	<a href="https://cdn.sofifa.org/fla">https://cdn.sofifa.org/fla</a>
4	4	192985	K. De Bruyne	27	<a href="https://cdn.sofifa.org/players/4/19/192985.png">https://cdn.sofifa.org/players/4/19/192985.png</a>	Belgium	<a href="https://cdn.sofifa.org/fla">https://cdn.sofifa.org/fla</a>

5 rows × 89 columns

```
In [4]: fifa_data.shape
```

```
Out[4]: (18207, 89)
```

```
In [5]: fifa_data.describe().T#we will arrange describe function in vertically
```

Out[5]:		count	mean	std	min	25%	50%	75%	max
	Unnamed: 0	18207.0	9103.000000	5256.052511	0.0	4551.5	9103.0	13654.5	18206.0
	ID	18207.0	214298.338606	29965.244204	16.0	200315.5	221759.0	236529.5	246620.0
	Age	18207.0	25.122206	4.669943	16.0	21.0	25.0	28.0	45.0
	Overall	18207.0	66.238699	6.908930	46.0	62.0	66.0	71.0	94.0
	Potential	18207.0	71.307299	6.136496	48.0	67.0	71.0	75.0	95.0
	Special	18207.0	1597.809908	272.586016	731.0	1457.0	1635.0	1787.0	2346.0
	International Reputation	18159.0	1.113222	0.394031	1.0	1.0	1.0	1.0	5.0
	Weak Foot	18159.0	2.947299	0.660456	1.0	3.0	3.0	3.0	5.0
	Skill Moves	18159.0	2.361308	0.756164	1.0	2.0	2.0	3.0	5.0
	Jersey Number	18147.0	19.546096	15.947765	1.0	8.0	17.0	26.0	99.0
	Crossing	18159.0	49.734181	18.364524	5.0	38.0	54.0	64.0	93.0
	Finishing	18159.0	45.550911	19.525820	2.0	30.0	49.0	62.0	95.0
	HeadingAccuracy	18159.0	52.298144	17.379909	4.0	44.0	56.0	64.0	94.0
	ShortPassing	18159.0	58.686712	14.699495	7.0	54.0	62.0	68.0	93.0
	Volleys	18159.0	42.909026	17.694408	4.0	30.0	44.0	57.0	90.0
	Dribbling	18159.0	55.371001	18.910371	4.0	49.0	61.0	68.0	97.0
	Curve	18159.0	47.170824	18.395264	6.0	34.0	48.0	62.0	94.0
	FKAccuracy	18159.0	42.863153	17.478763	3.0	31.0	41.0	57.0	94.0
	LongPassing	18159.0	52.711933	15.327870	9.0	43.0	56.0	64.0	93.0
	BallControl	18159.0	58.369459	16.686595	5.0	54.0	63.0	69.0	96.0
	Acceleration	18159.0	64.614076	14.927780	12.0	57.0	67.0	75.0	97.0
	SprintSpeed	18159.0	64.726967	14.649953	12.0	57.0	67.0	75.0	96.0
	Agility	18159.0	63.503607	14.766049	14.0	55.0	66.0	74.0	96.0
	Reactions	18159.0	61.836610	9.010464	21.0	56.0	62.0	68.0	96.0
	Balance	18159.0	63.966573	14.136166	16.0	56.0	66.0	74.0	96.0
	ShotPower	18159.0	55.460047	17.237958	2.0	45.0	59.0	68.0	95.0
	Jumping	18159.0	65.089432	11.820044	15.0	58.0	66.0	73.0	95.0
	Stamina	18159.0	63.219946	15.894741	12.0	56.0	66.0	74.0	96.0
	Strength	18159.0	65.311967	12.557000	17.0	58.0	67.0	74.0	97.0
	LongShots	18159.0	47.109973	19.260524	3.0	33.0	51.0	62.0	94.0
	Aggression	18159.0	55.868991	17.367967	11.0	44.0	59.0	69.0	95.0
	Interceptions	18159.0	46.698276	20.696909	3.0	26.0	52.0	64.0	92.0
	Positioning	18159.0	49.958478	19.529036	2.0	38.0	55.0	64.0	95.0
	Vision	18159.0	53.400903	14.146881	10.0	44.0	55.0	64.0	94.0
	Penalties	18159.0	48.548598	15.704053	5.0	39.0	49.0	60.0	92.0
	Composure	18159.0	58.648274	11.436133	3.0	51.0	60.0	67.0	96.0
	Marking	18159.0	47.281623	19.904397	3.0	30.0	53.0	64.0	94.0
	StandingTackle	18159.0	47.697836	21.664004	2.0	27.0	55.0	66.0	93.0
	Tackle	18159.0	45.661435	21.289135	3.0	24.0	52.0	64.0	91.0

	count	mean	std	min	25%	50%	75%	max
<b>GK Diving</b>	18159.0	16.616223	17.695349	1.0	8.0	11.0	14.0	90.0
<b>GK Handling</b>	18159.0	16.391596	16.906900	1.0	8.0	11.0	14.0	92.0
<b>GK Kicking</b>	18159.0	16.232061	16.502864	1.0	8.0	11.0	14.0	91.0
<b>GK Positioning</b>	18159.0	16.388898	17.034669	1.0	8.0	11.0	14.0	90.0
<b>GK Reflexes</b>	18159.0	16.710887	17.955119	1.0	8.0	11.0	14.0	94.0

show all data-types

```
In [6]: fifa_data.info()
```

<class 'pandas.core.frame.DataFrame'>

RangeIndex: 18207 entries, 0 to 18206

Data columns (total 89 columns):

#	Column	Non-Null Count	Dtype
0	Unnamed: 0	18207 non-null	int64
1	ID	18207 non-null	int64
2	Name	18207 non-null	object
3	Age	18207 non-null	int64
4	Photo	18207 non-null	object
5	Nationality	18207 non-null	object
6	Flag	18207 non-null	object
7	Overall	18207 non-null	int64
8	Potential	18207 non-null	int64
9	Club	17966 non-null	object
10	Club Logo	18207 non-null	object
11	Value	18207 non-null	object
12	Wage	18207 non-null	object
13	Special	18207 non-null	int64
14	Preferred Foot	18159 non-null	object
15	International Reputation	18159 non-null	float64
16	Weak Foot	18159 non-null	float64
17	Skill Moves	18159 non-null	float64
18	Work Rate	18159 non-null	object
19	Body Type	18159 non-null	object
20	Real Face	18159 non-null	object
21	Position	18147 non-null	object
22	Jersey Number	18147 non-null	float64
23	Joined	16654 non-null	object
24	Loaned From	1264 non-null	object
25	Contract Valid Until	17918 non-null	object
26	Height	18159 non-null	object
27	Weight	18159 non-null	object
28	LS	16122 non-null	object
29	ST	16122 non-null	object
30	RS	16122 non-null	object
31	LW	16122 non-null	object
32	LF	16122 non-null	object
33	CF	16122 non-null	object
34	RF	16122 non-null	object
35	RW	16122 non-null	object
36	LAM	16122 non-null	object
37	CAM	16122 non-null	object
38	RAM	16122 non-null	object
39	LM	16122 non-null	object
40	LCM	16122 non-null	object
41	CM	16122 non-null	object
42	RCM	16122 non-null	object
43	RM	16122 non-null	object
44	LWB	16122 non-null	object
45	LDM	16122 non-null	object
46	CDM	16122 non-null	object
47	RDM	16122 non-null	object
48	RWB	16122 non-null	object
49	LB	16122 non-null	object
50	LCB	16122 non-null	object
51	CB	16122 non-null	object
52	RCB	16122 non-null	object
53	RB	16122 non-null	object
54	Crossing	18159 non-null	float64
55	Finishing	18159 non-null	float64
56	HeadingAccuracy	18159 non-null	float64
57	ShortPassing	18159 non-null	float64
58	Volleys	18159 non-null	float64

59	Dribbling	18159	non-null	float64
60	Curve	18159	non-null	float64
61	FKAccuracy	18159	non-null	float64
62	LongPassing	18159	non-null	float64
63	BallControl	18159	non-null	float64
64	Acceleration	18159	non-null	float64
65	SprintSpeed	18159	non-null	float64
66	Agility	18159	non-null	float64
67	Reactions	18159	non-null	float64
68	Balance	18159	non-null	float64
69	ShotPower	18159	non-null	float64
70	Jumping	18159	non-null	float64
71	Stamina	18159	non-null	float64
72	Strength	18159	non-null	float64
73	LongShots	18159	non-null	float64
74	Aggression	18159	non-null	float64
75	Interceptions	18159	non-null	float64
76	Positioning	18159	non-null	float64
77	Vision	18159	non-null	float64
78	Penalties	18159	non-null	float64
79	Composure	18159	non-null	float64
80	Marking	18159	non-null	float64
81	StandingTackle	18159	non-null	float64
82	SlidingTackle	18159	non-null	float64
83	GKDividing	18159	non-null	float64
84	GKHandling	18159	non-null	float64
85	GK Kicking	18159	non-null	float64
86	GK Positioning	18159	non-null	float64
87	GK Reflexes	18159	non-null	float64
88	Release Clause	16643	non-null	object

dtypes: float64(38), int64(6), object(45)  
memory usage: 12.4+ MB

## null values in percentage

```
In [7]: percent = 100*(fifa_data.isnull().sum()/len(fifa_data))
```

```
In [8]: percent[:40]
```

```
Out[8]: Unnamed: 0      0.000000
        ID              0.000000
        Name            0.000000
        Age             0.000000
        Photo           0.000000
        Nationality      0.000000
        Flag            0.000000
        Overall          0.000000
        Potential        0.000000
        Club            1.323667
        Club Logo        0.000000
        Value            0.000000
        Wage            0.000000
        Special          0.000000
        Preferred Foot   0.263635
        International Reputation 0.263635
        Weak Foot        0.263635
        Skill Moves      0.263635
        Work Rate        0.263635
        Body Type        0.263635
        Real Face        0.263635
        Position         0.329544
        Jersey Number    0.329544
        Joined           8.529686
        Loaned From      93.057615
        Contract Valid Until 1.587302
        Height           0.263635
        Weight           0.263635
        LS              11.451639
        ST              11.451639
        RS              11.451639
        LW              11.451639
        LF              11.451639
        CF              11.451639
        RF              11.451639
        RW              11.451639
        LAM             11.451639
        CAM             11.451639
        RAM             11.451639
        LM              11.451639
        dtype: float64
```

```
In [9]: percent[41:]
```

```

Out[9]:
CM 11.451639
RCM 11.451639
RM 11.451639
LWB 11.451639
LDM 11.451639
CDM 11.451639
RDM 11.451639
RWB 11.451639
LB 11.451639
LCB 11.451639
CB 11.451639
RCB 11.451639
RB 11.451639
Crossing 0.263635
Finishing 0.263635
HeadingAccuracy 0.263635
ShortPassing 0.263635
Volleys 0.263635
Dribbling 0.263635
Curve 0.263635
FKAccuracy 0.263635
LongPassing 0.263635
BallControl 0.263635
Acceleration 0.263635
SprintSpeed 0.263635
Agility 0.263635
Reactions 0.263635
Balance 0.263635
ShotPower 0.263635
Jumping 0.263635
Stamina 0.263635
Strength 0.263635
LongShots 0.263635
Aggression 0.263635
Interceptions 0.263635
Positioning 0.263635
Vision 0.263635
Penalties 0.263635
Composure 0.263635
Marking 0.263635
StandingTackle 0.263635
SlidingTackle 0.263635
GKDividing 0.263635
GKHandling 0.263635
GK Kicking 0.263635
GKPositioning 0.263635
GKReflexes 0.263635
Release Clause 8.590103
dtype: float64

```

## cleaned dataset

```
In [10]: fifa_data=fifa_data.dropna(axis=0, subset=['Club'])#club null values is 1.32% so we drop
```

```
In [11]: fifa_data=fifa_data.dropna(axis=0, subset=['Preferred Foot'])#some columns are related
```

```
In [12]: null_Val = fifa_data.isnull().sum().sort_values(ascending=False)
```

```
In [13]: null_Val[:40]#print remaining null values
```

```
Out[13]:
```

Loaned From	16654
LWB	1992
LCM	1992
RS	1992
LW	1992
LF	1992
CF	1992
RF	1992
RW	1992
LAM	1992
CAM	1992
RAM	1992
LM	1992
CM	1992
LS	1992
RCM	1992
RM	1992
LDM	1992
CDM	1992
RDM	1992
RWB	1992
LB	1992
LCB	1992
CB	1992
RCB	1992
RB	1992
ST	1992
Release Clause	1275
Joined	1264
Weight	0
Aggression	0
FKAccuracy	0
LongPassing	0
BallControl	0
Acceleration	0
SprintSpeed	0
Agility	0
Reactions	0
Balance	0
ShotPower	0

dtype: int64

fill some important columns null values

```
In [14]: fifa_data['Loaned From'].dtype
```

```
Out[14]: dtype('O')
```

```
In [15]: fifa_data['Loaned From'].fillna('None',inplace=True)
```

```
In [16]: fifa_data['Loaned From'].value_counts()
```



```
Out[16]: None 16654
         Atalanta 20
         Sassuolo 18
         Juventus 17
         SL Benfica 17
         ...
         Sheffield Wednesday 1
         Club Necaxa 1
         Royal Excel Mouscron 1
         Rayo Vallecano 1
         Neuchâtel Xamax 1
         Name: Loaned From, Length: 342, dtype: int64
```

```
In [17]: fifa_data['Joined'].dtype
```

```
Out[17]: dtype('O')
```

```
In [18]: fifa_data['Joined'].mode()
```

```
Out[18]: 0    Jul 1, 2018
         Name: Joined, dtype: object
```

```
In [19]: fifa_data['Joined'].fillna('Jul 1, 2018', inplace=True)
```

```
In [20]: fifa_data['Joined'].value_counts()
```

```
Out[20]: Jul 1, 2018    2802
         Jul 1, 2017    1133
         Jan 1, 2018     635
         Jul 1, 2016     614
         Jul 1, 2015     368
         ...
         Jan 4, 2014      1
         Apr 20, 2017     1
         Nov 10, 2017     1
         Dec 11, 2013     1
         Jul 4, 2012      1
         Name: Joined, Length: 1736, dtype: int64
```

```
In [21]: fifa_data['Joined'].head(100)
```

```
Out[21]: 0    Jul 1, 2004
         1    Jul 10, 2018
         2    Aug 3, 2017
         3    Jul 1, 2011
         4    Aug 30, 2015
         ...
         95   Jul 14, 2011
         96   Aug 6, 2018
         97   Sep 1, 2015
         98   Jul 1, 2015
         99   Jul 1, 2015
         Name: Joined, Length: 100, dtype: object
```

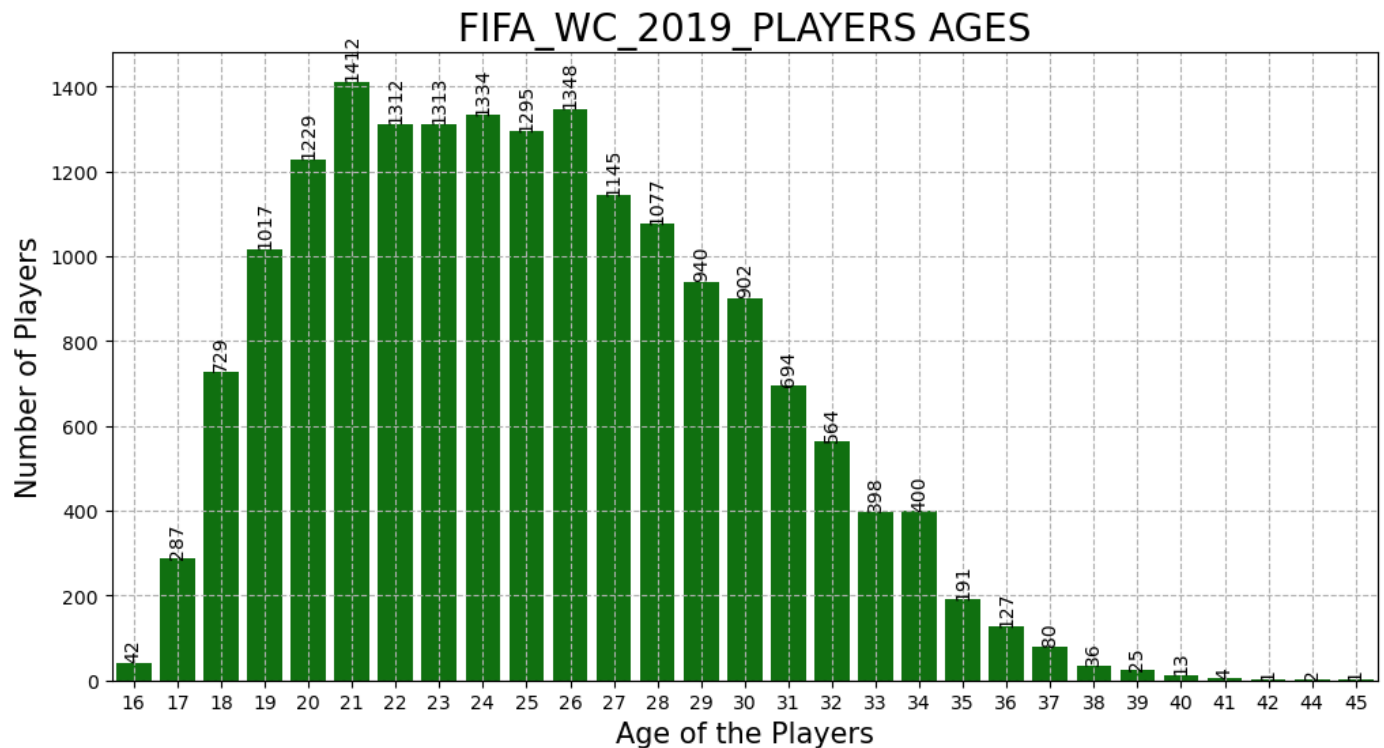
```
In [22]: fifa_data.columns
```

```
Out[22]: 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',
      'GK Kicking', 'GK Positioning', 'GK Reflexes', 'Release Clause'],
      dtype='object')
```

## visualization

### 1.Age

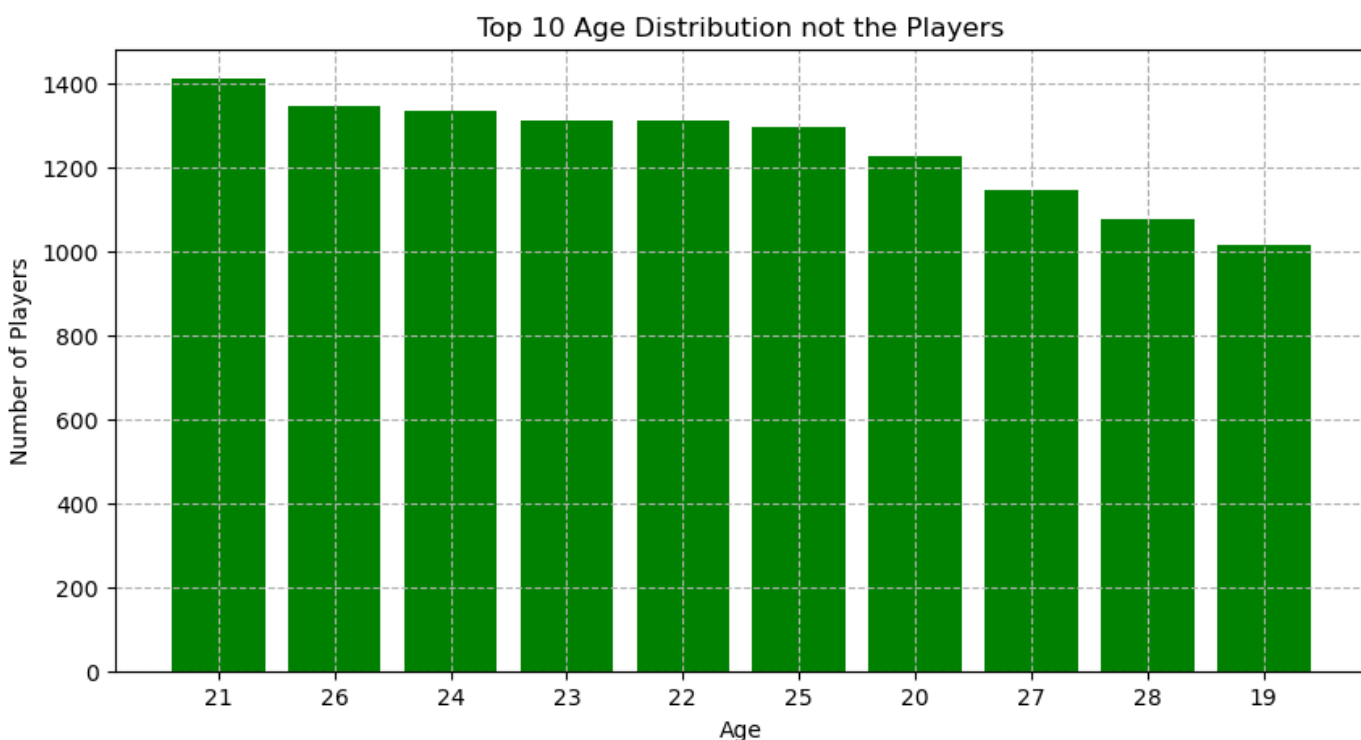
```
In [23]: plt.figure(figsize=(12,6))
ax = sns.countplot(data = fifa_data, x='Age',color = 'green')
for p in ax.containers:
    ax.bar_label(p, rotation=90)
plt.grid(True, linestyle='--')
plt.xlabel('Age of the Players', fontsize=15)
plt.ylabel('Number of Players', fontsize=15)
plt.title('FIFA_WC_2019_PLAYERS AGES', fontsize=20)
plt.show()
```



```
In [24]: age = fifa_data['Age'].value_counts()[:10]
age
```

```
Out[24]: 21    1412
          26    1348
          24    1334
          23    1313
          22    1312
          25    1295
          20    1229
          27    1145
          28    1077
          19    1017
          Name: Age, dtype: int64
```

```
In [25]: plt.figure(figsize=(10,5))
          plt.bar(age.index.astype(str),age.values, color='green')
          plt.xlabel('Age')
          plt.ylabel('Number of Players' )
          plt.title('Top 10 Age Distribution not the Players')
          plt.grid(True, linestyle='--');
```



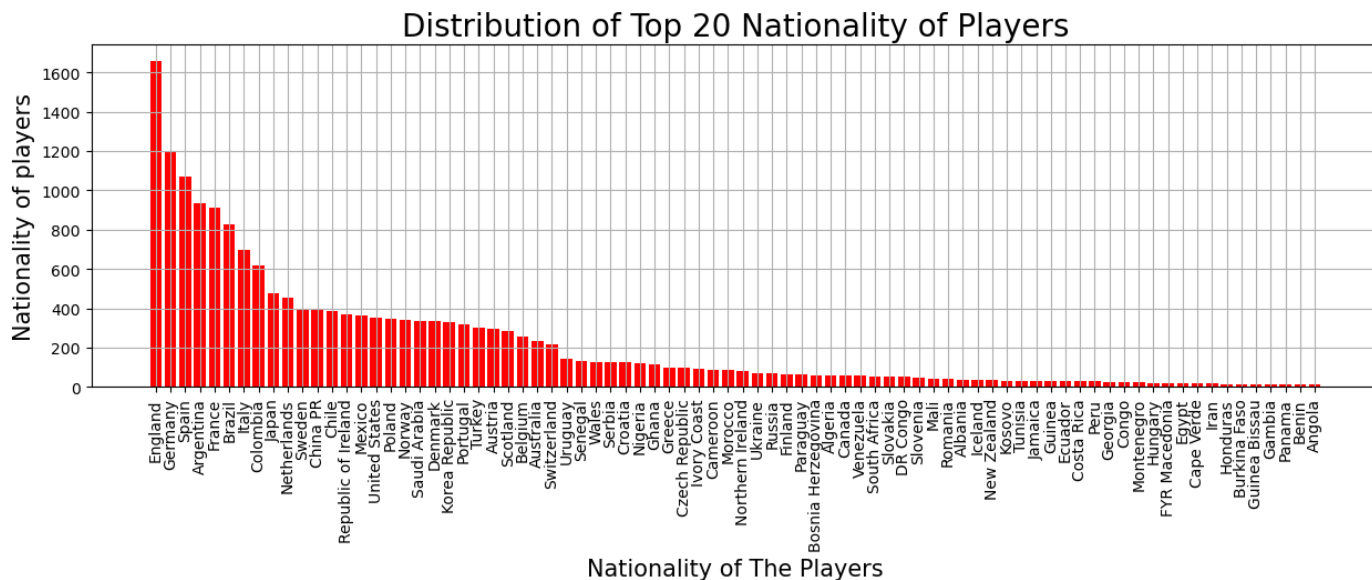
## 2.Nationality

```
In [26]: top_Nationality_playes = fifa_data['Nationality'].value_counts().head(80)
          top_Nationality_playes
```

```
Out[26]: England    1657
          Germany    1195
          Spain      1071
          Argentina   935
          France      911
          ...
          Guinea Bissau 15
          Gambia       15
          Panama       15
          Benin        15
          Angola       15
          Name: Nationality, Length: 80, dtype: int64
```

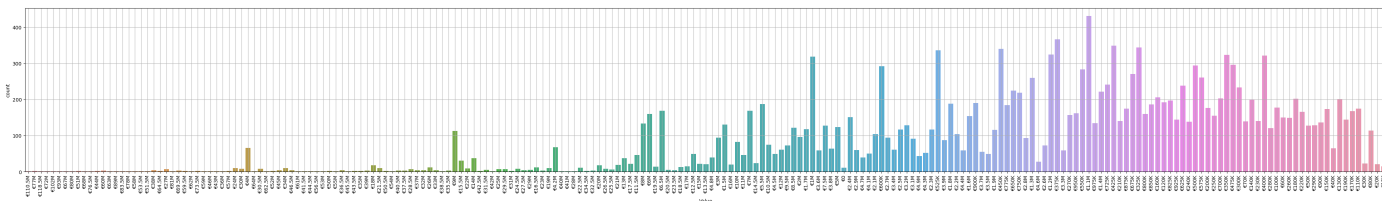
```
In [27]: plt.figure(figsize=(15,4))
          plt.bar(top_Nationality_playes.index.astype(str), top_Nationality_playes.values, color =
```

```
plt.xlabel('Nationality of The Players', fontsize=15)
plt.ylabel('Nationality of players', fontsize=15)
plt.title('Distribution of Top 20 Nationality of Players', fontsize=20)
plt.xticks(rotation=90)
plt.grid(True);
```



### 3.Value

```
In [28]: plt.figure(figsize=(50,6))
sns.countplot(data=fifa_data, x='Value')
plt.xticks(rotation=90)
plt.grid(True)
```

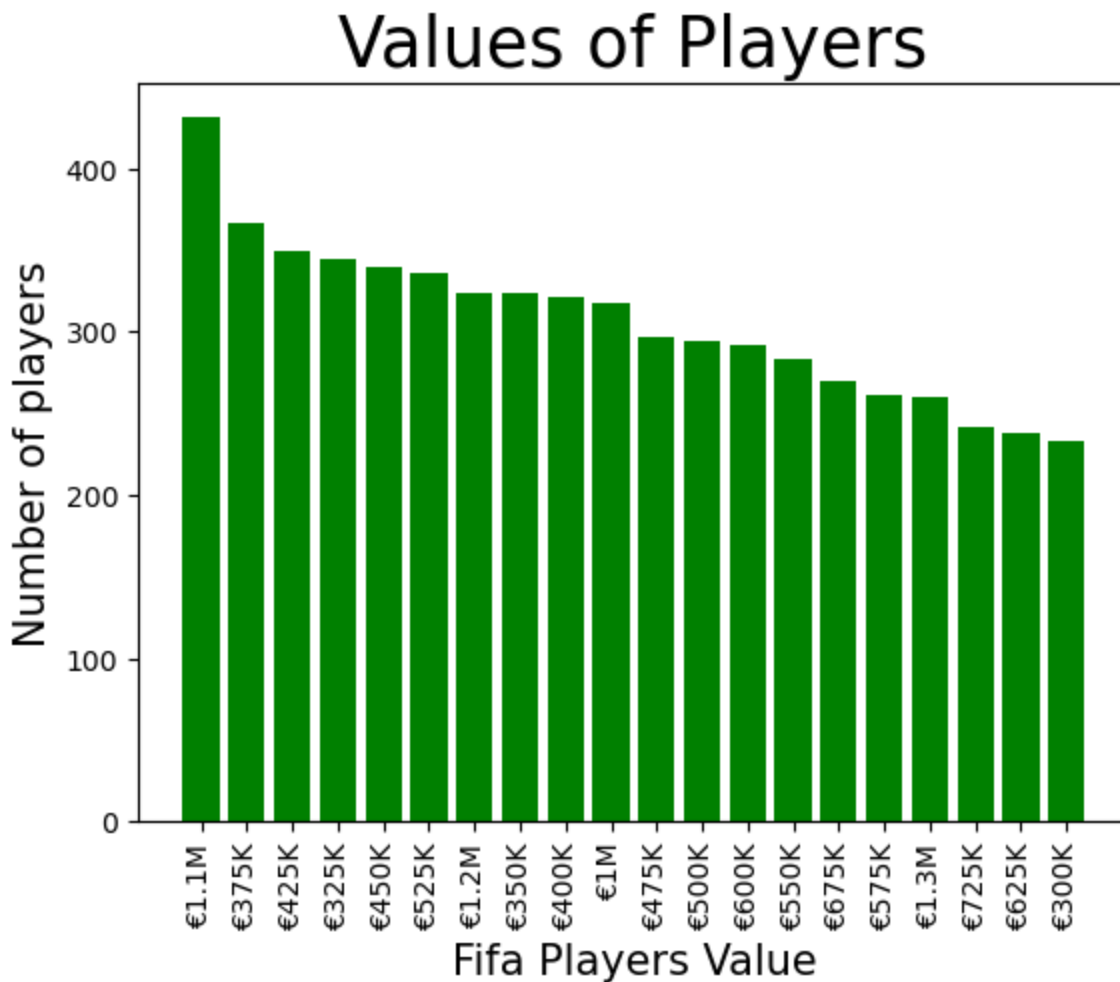


```
In [29]: value_of_players = fifa_data['Value'].value_counts()[0:20]
value_of_players
```

```
Out[29]: €1.1M      431
€375K      366
€425K      349
€325K      344
€450K      340
€525K      336
€1.2M      324
€350K      323
€400K      321
€1M        318
€475K      296
€500K      294
€600K      292
€550K      283
€675K      270
€575K      261
€1.3M      260
€725K      241
€625K      238
€300K      233
```

dtype: int64

```
In [30]: plt.bar(value_of_players.index.astype(str), value_of_players.values,color='green')
plt.xlabel('Fifa Players Value', fontsize=15)
plt.ylabel('Number of players', fontsize=15)
plt.title('Values of Players', fontsize=25)
plt.xticks(rotation=90);
```

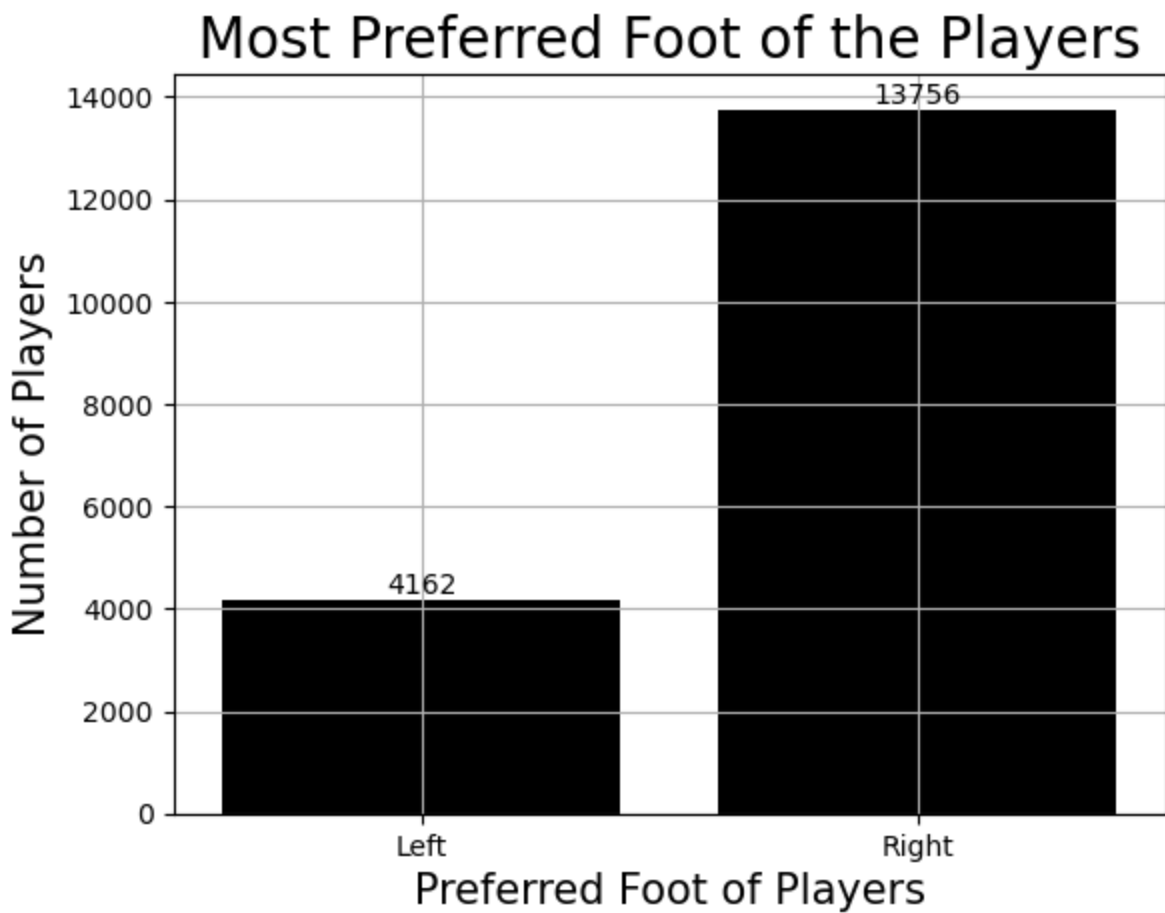


## 4.Preferred Foot

```
In [31]: fifa_data['Preferred Foot'].value_counts()
```

```
Out[31]: Right    13756
Left       4162
Name: Preferred Foot, dtype: int64
```

```
In [32]: ax = sns.countplot(data=fifa_data,x='Preferred Foot',color='black')
plt.grid(True)
plt.xlabel('Preferred Foot of Players',fontsize=15)
plt.ylabel('Number of Players',fontsize=15)
plt.title('Most Preferred Foot of the Players',fontsize=20)
for p in ax.containers:
    ax.bar_label(p)
```



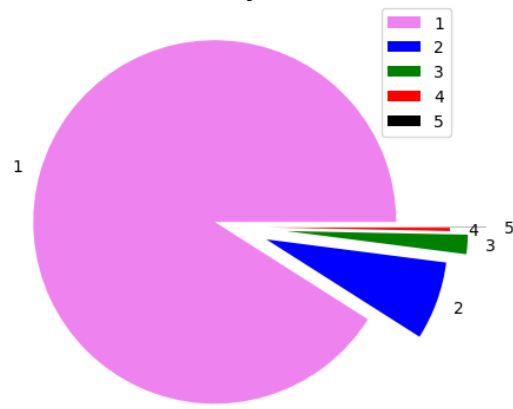
## 5. International Reputation

```
In [33]: fifa_data['International Reputation'].value_counts()
```

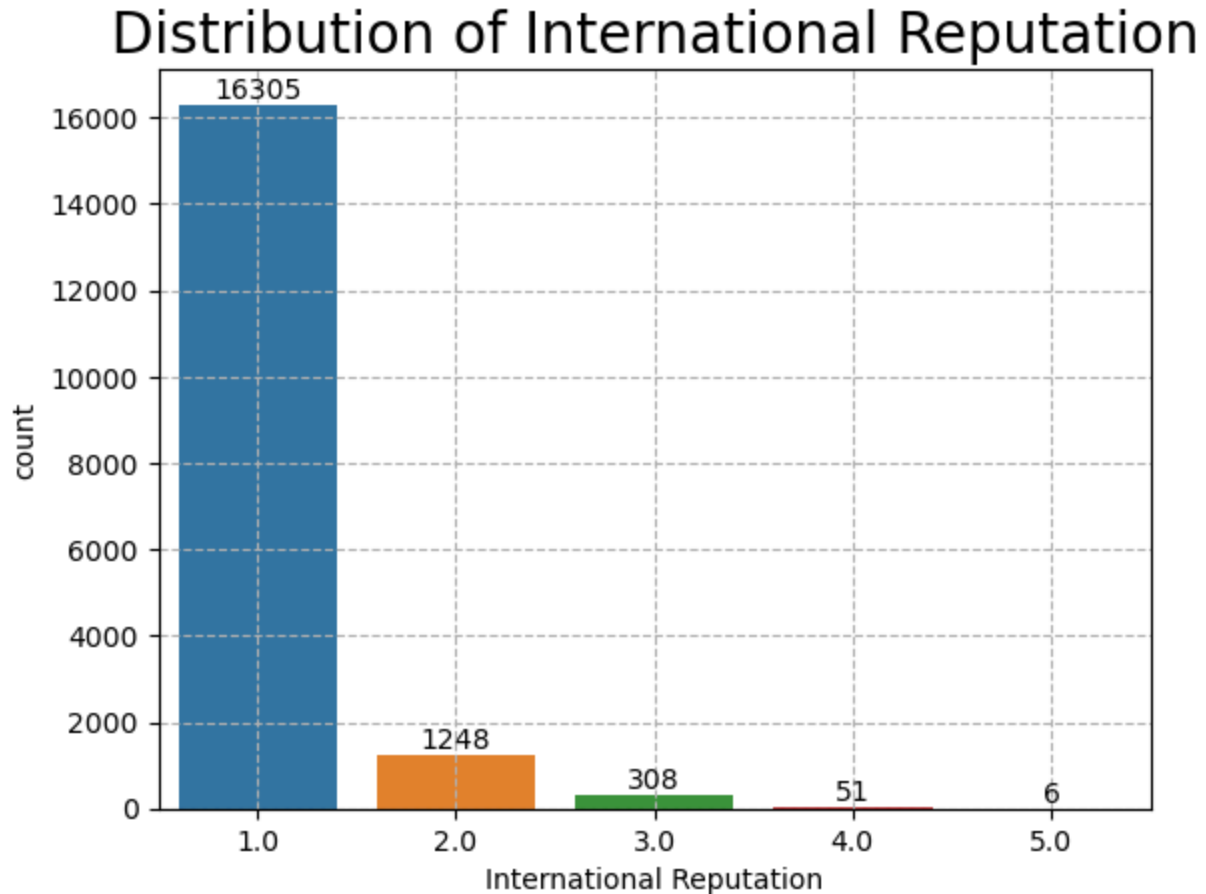
```
Out[33]: 1.0    16305
          2.0    1248
          3.0     308
          4.0      51
          5.0       6
          Name: International Reputation, dtype: int64
```

```
In [34]: # plot a pie chart of above data
values = ['1', '2', '3', '4', '5']
sizes = ['16305', '1248', '308', '51', '6']
colors = ['violet', 'blue', 'green', 'red', 'black']
explode = [.1, .2, .3, .2, .4]
plt.figure(figsize=(5,5))
plt.pie(sizes, labels = values, colors=colors, explode=explode)
plt.title('A pie chart of Internation Reputation of Football Players', fontsize=30)
plt.legend()
plt.show();
```

# A pie chart of International Reputation of Football Players



```
In [35]: ax = sns.countplot(x='International Reputation', data=fifa_data)
for bars in ax.containers:
    ax.bar_label(bars)
plt.title('Distribution of International Reputation', fontsize=20)
plt.grid(True, linestyle='--')
plt.show()
```



## 6.Weak Foot

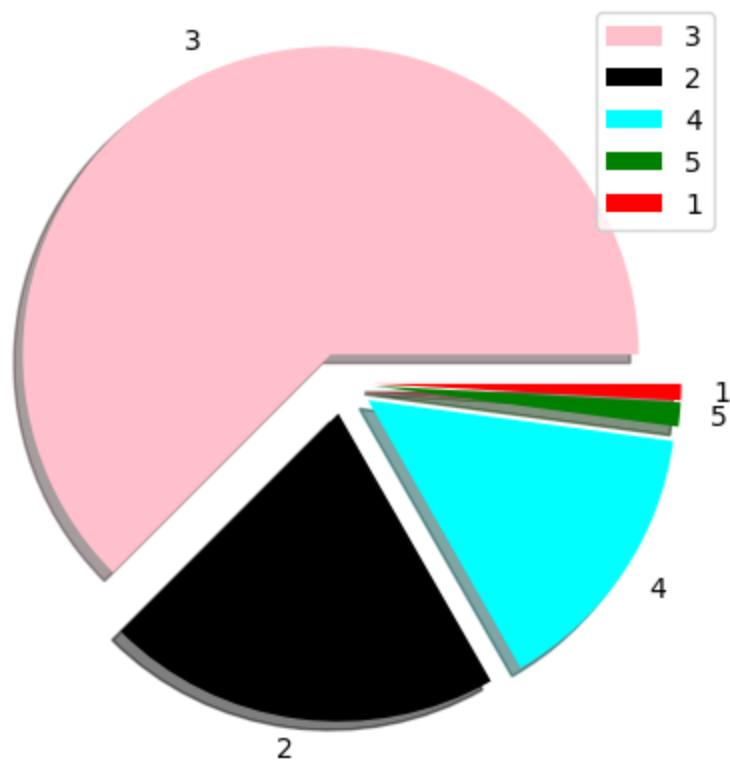
```
In [36]: fifa_data['Weak Foot'].value_counts()
```

```
Out[36]: 3.0    11201
          2.0    3715
          4.0    2622
          5.0     227
          1.0     153
          Name: Weak Foot, dtype: int64
```

```
In [37]: values = ['3', '2', '4', '5', '1']
          size = ['11201', '3715', '2622', '227', '153']
          color = ['pink', 'black', 'cyan', 'green', 'red']
          explode = [0.1, 0.1, 0.1, 0.1, 0.1]

          plt.figure(figsize=(5,5))
          plt.pie(size, labels=values, colors=color, explode=explode, shadow=True)
          plt.title('Distribution of WEAK_FOOT_PLAYERS', fontsize=20)
          plt.legend()
          plt.show();
```

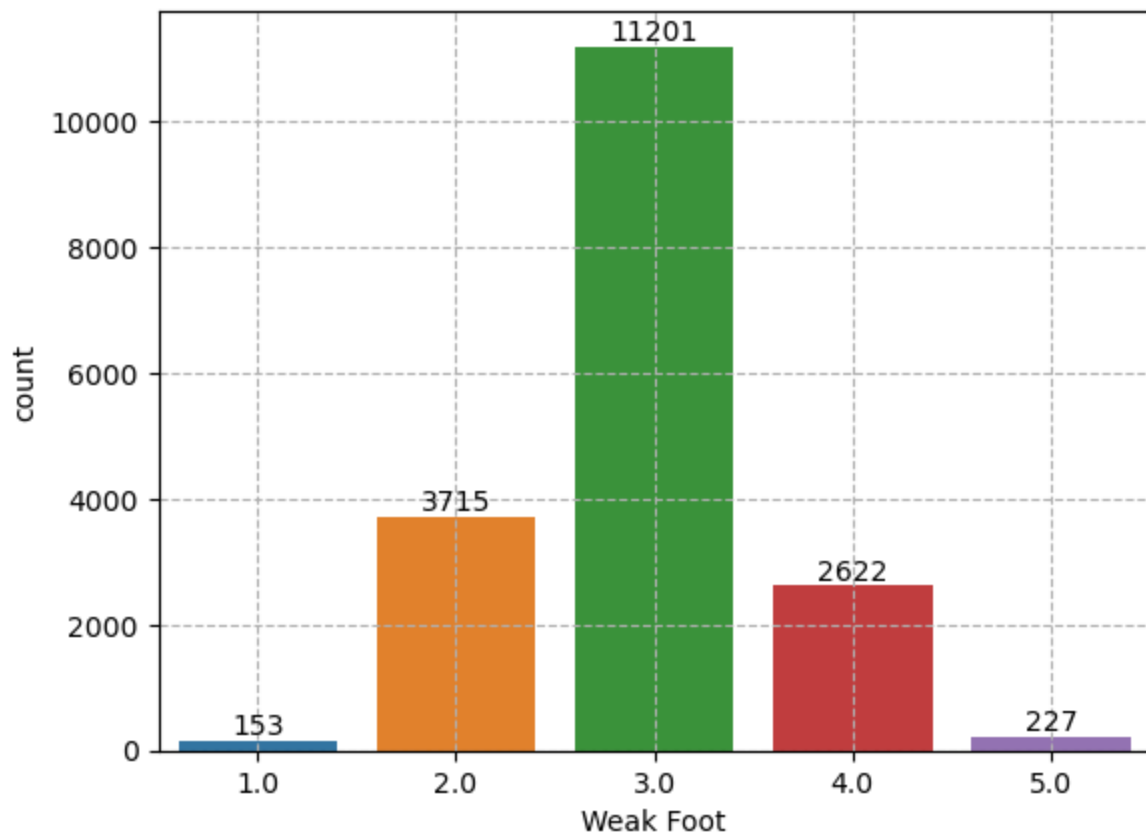
## Distribution of WEAK\_FOOT\_PLAYERS



```
In [38]: ax = sns.countplot(x='Weak Foot', data=fifa_data)
          for bars in ax.containers:
              ax.bar_label(bars)
          plt.title('Distribution of Weak Foot', fontsize=25)
          plt.grid(True, linestyle='--')
```



# Distribution of Weak Foot

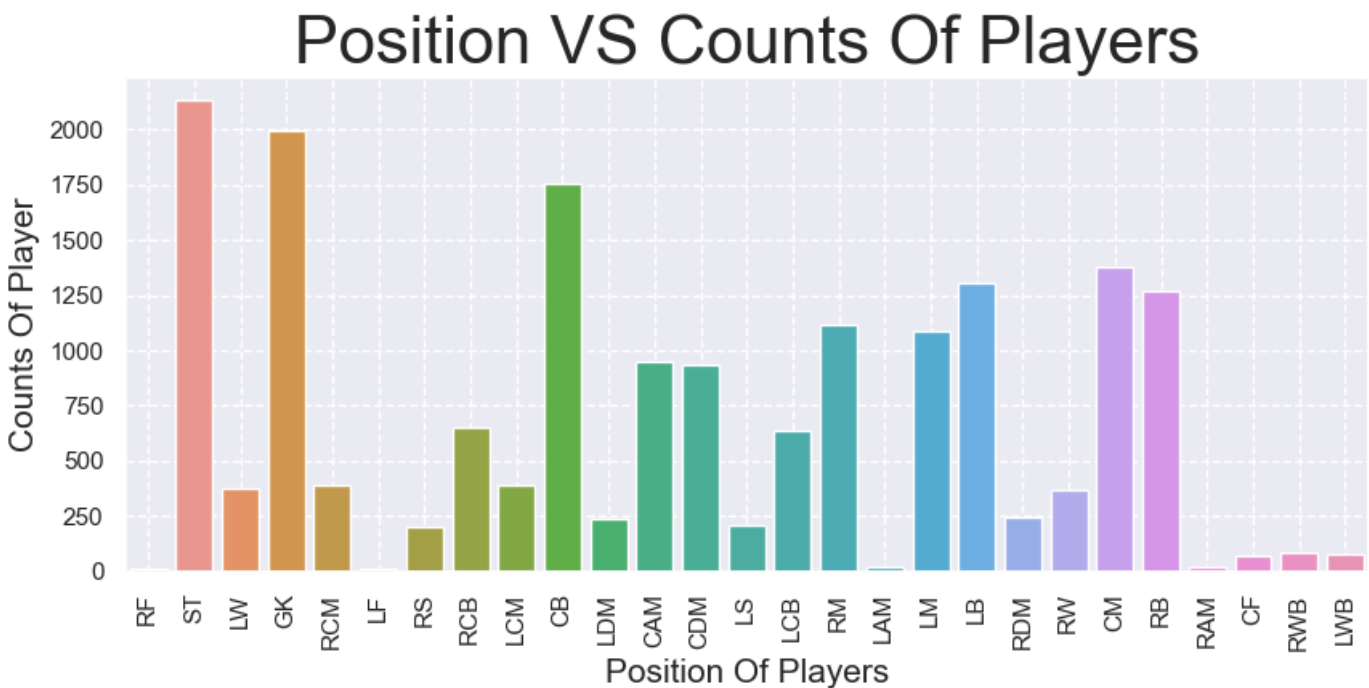


## 7.Position

```
In [39]: fifa_data['Position'].value_counts()
```

```
Out[39]: ST      2130
GK      1992
CB      1754
CM      1377
LB      1305
RB      1268
RM      1114
LM      1086
CAM      948
CDM      936
RCB      652
LCB      637
LCM      389
RCM      387
LW      374
RW      365
RDM      246
LDM      239
LS       206
RS       201
RWB       87
LWB       78
CF        74
LAM        21
RAM        21
RF         16
LF         15
Name: Position, dtype: int64
```

```
In [40]: plt.figure(figsize=(10,4))
sns.set(style = 'dark', palette = 'colorblind', color_codes = True)
sns.countplot(data = fifa_data, x = 'Position')
plt.xlabel('Position Of Players', fontsize=15)
plt.ylabel('Counts Of Player', fontsize=15)
plt.title('Position VS Counts Of Players', fontsize=30)
plt.grid(True, linestyle='--')
plt.xticks(rotation=90);
```



## 8.Weight

```
In [41]: fifa_data['Weight'].head()
```

```
Out[41]: 0    159lbs
1    183lbs
2    150lbs
3    168lbs
4    154lbs
Name: Weight, dtype: object
```

```
In [42]: # replace 'lbs' from weight column
def extract_value(value):
    out = value.replace('lbs', '')
    return float(out)
```

```
In [43]: fifa_data['Weight'] = fifa_data['Weight'].apply(lambda x : extract_value(x))
```

```
In [44]: fifa_data['Weight'].head()
```

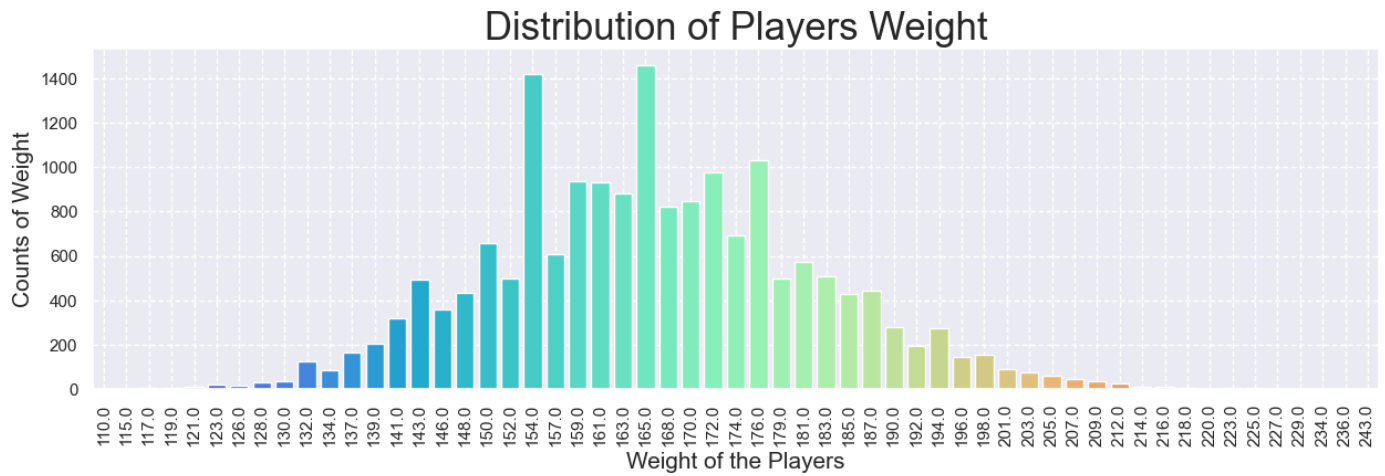
```
Out[44]: 0    159.0
1    183.0
2    150.0
3    168.0
4    154.0
Name: Weight, dtype: float64
```

```
In [45]: top_weight = fifa_data['Weight'].value_counts()
top_weight
```

```
Out[45]: 165.0    1461
          154.0    1418
          176.0    1031
          172.0     974
          159.0     936
          161.0     929
          163.0     883
          170.0     849
          168.0     823
          174.0     691
          150.0     655
          157.0     608
          181.0     573
          183.0     510
          179.0     499
          152.0     498
          143.0     494
          187.0     441
          148.0     432
          185.0     427
          146.0     361
          141.0     318
          190.0     281
          194.0     273
          139.0     206
          192.0     193
          137.0     165
          198.0     153
          196.0     142
          132.0     124
          201.0      91
          134.0      86
          203.0      75
          205.0      59
          207.0      44
          209.0      34
          130.0      33
          128.0      31
          212.0      23
          123.0      18
          126.0      14
          214.0      11
          121.0      10
          216.0       9
          117.0       6
          218.0       5
          119.0       4
          223.0       3
          225.0       3
          227.0       2
          236.0       2
          110.0       2
          243.0       1
          220.0       1
          229.0       1
          115.0       1
          234.0       1
          Name: Weight, dtype: int64
```

```
In [46]: plt.figure(figsize=(15,4))
          sns.countplot(x='Weight',data=fifa_data, palette='rainbow')
          plt.xlabel('Weight of the Players', fontsize=15)
          plt.ylabel('Counts of Weight', fontsize=15)
          plt.title('Distribution of Players Weight', fontsize=25)
```

```
plt.grid(True, linestyle='--')
plt.xticks(rotation=90);
```



## 9.Wage

```
In [47]: fifa_data['Wage'].unique()
```

```
Out[47]: array(['€565K', '€405K', '€290K', '€260K', '€355K', '€340K', '€420K',
        '€455K', '€380K', '€94K', '€205K', '€125K', '€285K', '€225K',
        '€145K', '€240K', '€315K', '€200K', '€130K', '€300K', '€215K',
        '€100K', '€255K', '€165K', '€265K', '€160K', '€150K', '€245K',
        '€110K', '€77K', '€115K', '€210K', '€195K', '€230K', '€250K',
        '€135K', '€155K', '€180K', '€175K', '€190K', '€185K', '€21K',
        '€82K', '€73K', '€92K', '€88K', '€96K', '€170K', '€66K', '€235K',
        '€28K', '€105K', '€38K', '€81K', '€57K', '€15K', '€63K', '€22K',
        '€84K', '€120K', '€90K', '€72K', '€93K', '€45K', '€74K', '€51K',
        '€42K', '€31K', '€75K', '€25K', '€140K', '€41K', '€78K', '€53K',
        '€95K', '€80K', '€43K', '€60K', '€85K', '€64K', '€67K', '€18K',
        '€70K', '€91K', '€20K', '€49K', '€87K', '€86K', '€26K', '€29K',
        '€55K', '€35K', '€33K', '€56K', '€30K', '€11K', '€59K', '€23K',
        '€46K', '€39K', '€32K', '€36K', '€98K', '€54K', '€68K', '€58K',
        '€27K', '€40K', '€44K', '€19K', '€1K', '€61K', '€50K', '€99K',
        '€17K', '€52K', '€62K', '€12K', '€10K', '€71K', '€14K', '€76K',
        '€48K', '€65K', '€69K', '€24K', '€34K', '€16K', '€37K', '€47K',
        '€89K', '€97K', '€79K', '€13K', '€83K', '€6K', '€3K', '€9K', '€8K',
        '€7K', '€4K', '€2K', '€5K'], dtype=object)
```

```
In [48]: def replace_value(value):
        out = value.replace('€', '')
        if 'M' in out:
            out = float(out.replace('M', ''))*1000000
        elif 'K' in value:
            out = float(out.replace('K', ''))*1000
        return float(out)
```

## 10.Skill Moves

```
In [49]: fifa_data['Skill Moves'].value_counts()
```

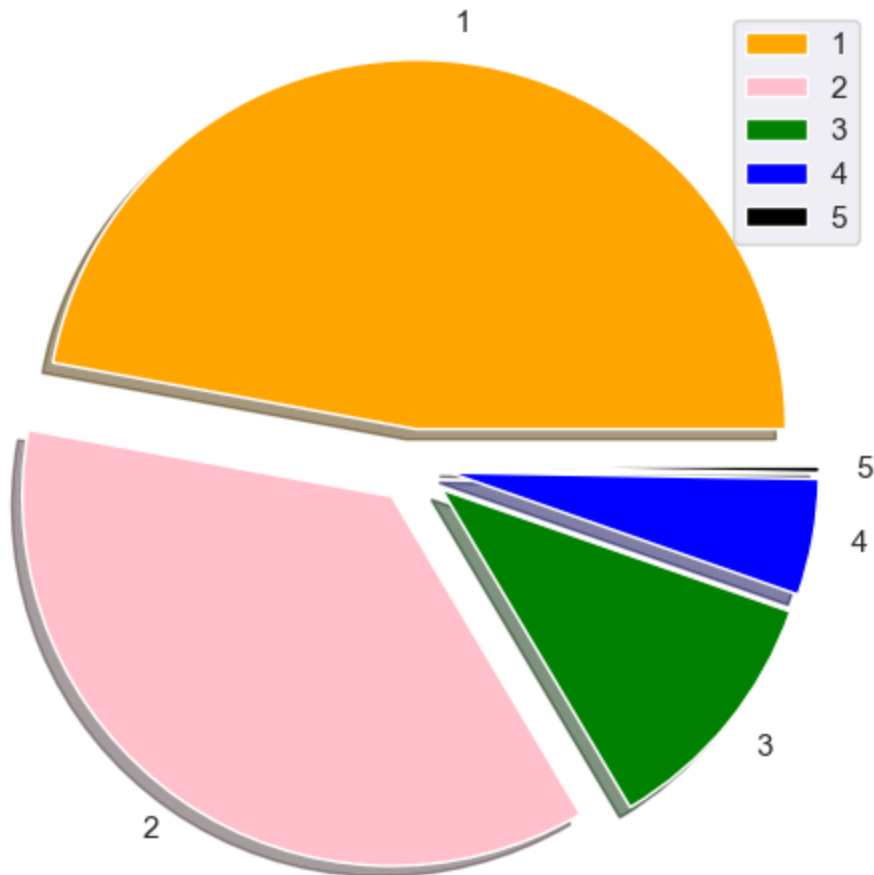
```
Out[49]: 2.0    8443
        3.0    6522
        1.0    1992
        4.0     911
        5.0     50
```

```
moves, dtype: int64
```

```
In [50]: values = ['1', '2', '3', '4', '5']
size = ['8443', '6522', '1992', '911', '50']
color = ['orange', 'pink', 'green', 'blue', 'black']
explods = [.1, .1, .1, .1, .1]

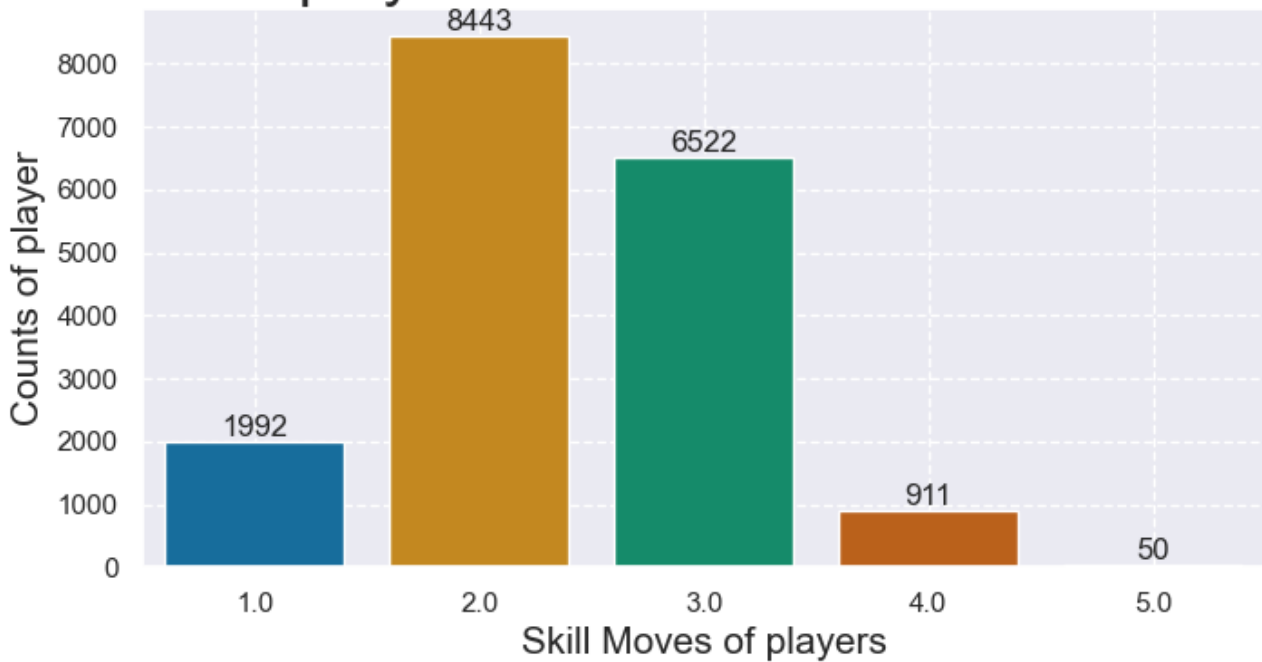
plt.figure(figsize=(6,6))
plt.pie(size, labels=values, colors=color, explode=explode, shadow=True )
plt.title('Dristribution of Skill Moves', fontsize=25)
plt.legend();
```

## Dristribution of Skill Moves



```
In [51]: plt.figure(figsize=(8,4))
ax = sns.countplot(x='Skill Moves', data=fifa_data)
for x in ax.containers:
    ax.bar_label(x)
plt.xlabel('Skill Moves of players', fontsize=15)
plt.ylabel('Counts of player', fontsize=15)
plt.title('Count of player on the bawe of their skill moves', fontsize=25)
plt.grid( linestyle='--')
```

# Count of player on the bawe of their skill moves



## 11.Height

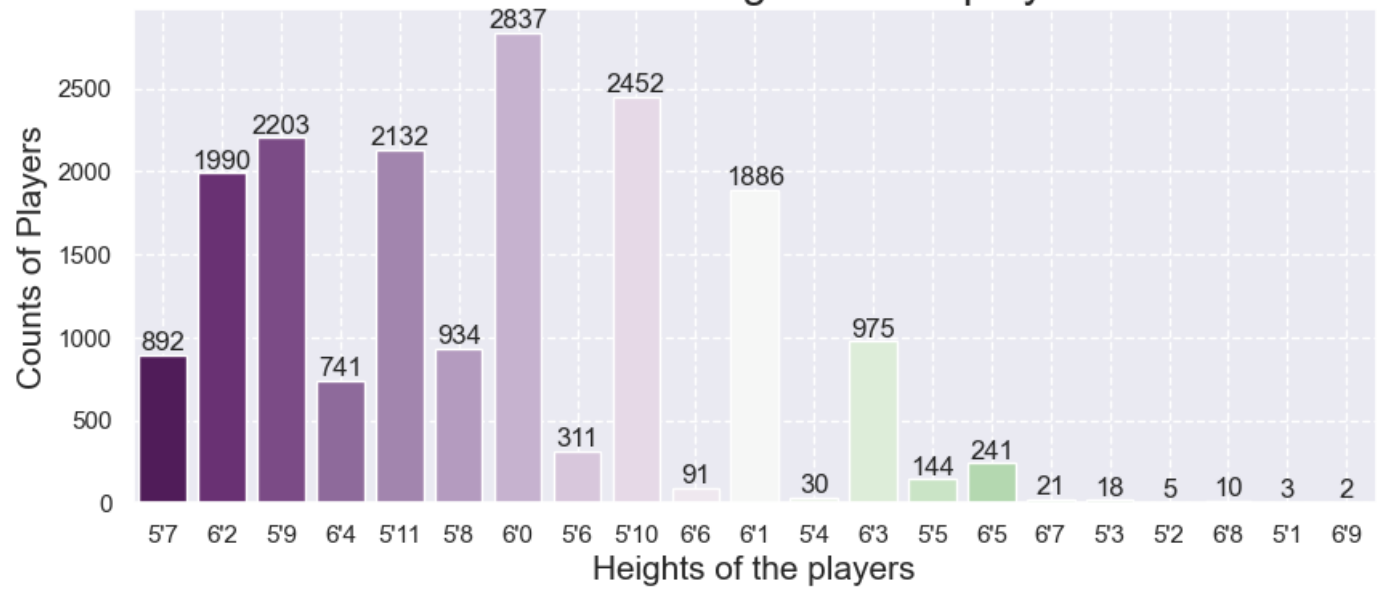
```
In [52]: fifa_data['Height'].value_counts()
```

```
Out[52]: 6'0      2837
5'10     2452
5'9      2203
5'11     2132
6'2      1990
6'1      1886
6'3       975
5'8       934
5'7       892
6'4       741
5'6       311
6'5       241
5'5       144
6'6        91
5'4        30
6'7        21
5'3        18
6'8        10
5'2         5
5'1         3
6'9         2
Name: Height, dtype: int64
```

```
In [53]: plt.figure(figsize=(10,4))
ax = sns.countplot(x='Height', data=fifa_data, palette='PRGn')
for x in ax.containers:
    ax.bar_label(x)

plt.xlabel('Heights of the players', fontsize=15)
plt.ylabel('Counts of Players', fontsize=15)
plt.title('Dristributions Heights of the players', fontsize=20)
plt.grid(True, linestyle='--')
```

## Distributions Heights of the players



## 12.Weight

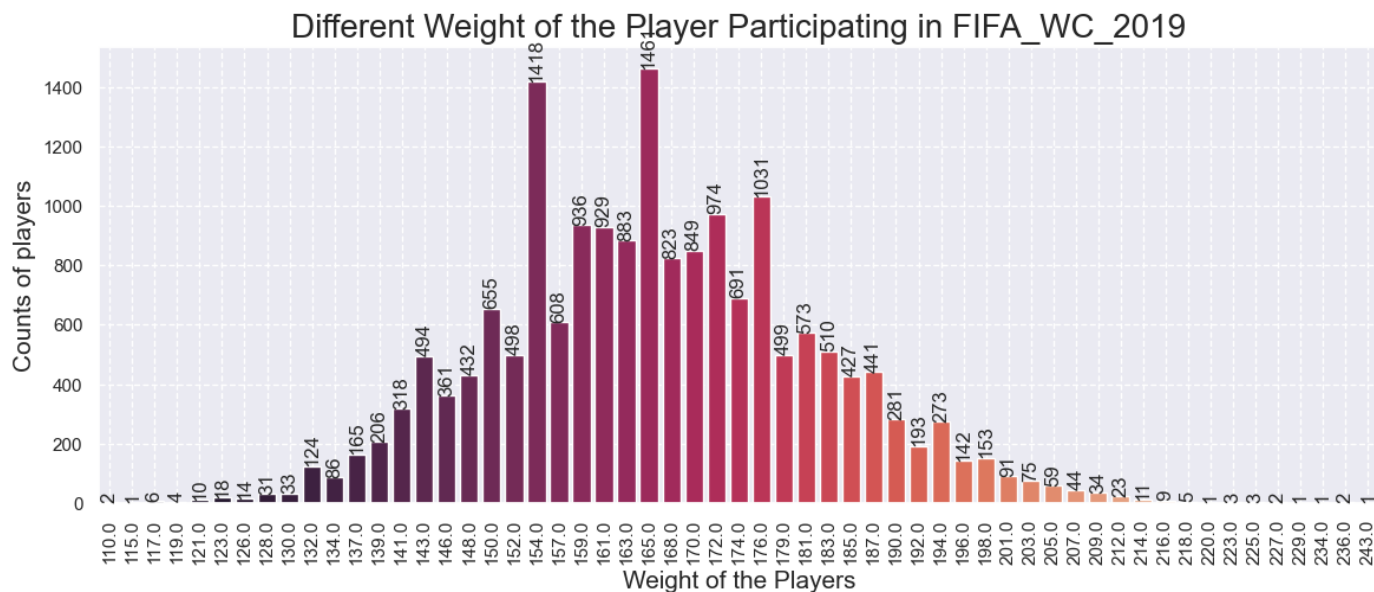
```
In [54]: fifa_data['Weight'].value_counts()
```

```
Out[54]: 165.0    1461
          154.0    1418
          176.0   1031
          172.0    974
          159.0    936
          161.0    929
          163.0    883
          170.0    849
          168.0    823
          174.0    691
          150.0    655
          157.0    608
          181.0    573
          183.0    510
          179.0    499
          152.0    498
          143.0    494
          187.0    441
          148.0    432
          185.0    427
          146.0    361
          141.0    318
          190.0    281
          194.0    273
          139.0    206
          192.0    193
          137.0    165
          198.0    153
          196.0    142
          132.0    124
          201.0     91
          134.0     86
          203.0     75
          205.0     59
          207.0     44
          209.0     34
          130.0     33
          128.0     31
          212.0     23
          123.0     18
          126.0     14
          214.0     11
          121.0     10
          216.0      9
          117.0      6
          218.0      5
          119.0      4
          223.0      3
          225.0      3
          227.0      2
          236.0      2
          110.0      2
          243.0      1
          220.0      1
          229.0      1
          115.0      1
          234.0      1
          Name: Weight, dtype: int64
```

```
In [55]: plt.figure(figsize=(14,5))
          ax = sns.countplot(x='Weight', data=fifa_data, palette='rocket')
          for bar in ax.containers:
              ax.bar_label(bar, rotation=90)
```



```
plt.xlabel('Weight of the Players', fontsize=15)
plt.ylabel('Counts of players', fontsize=15)
plt.title('Different Weight of the Player Participating in FIFA_WC_2019', fontsize=20)
plt.grid(True, linestyle='--')
plt.xticks(rotation=90);
```



## 13.Work Rate

```
In [56]: fifa_data['Work Rate'].value_counts()
```

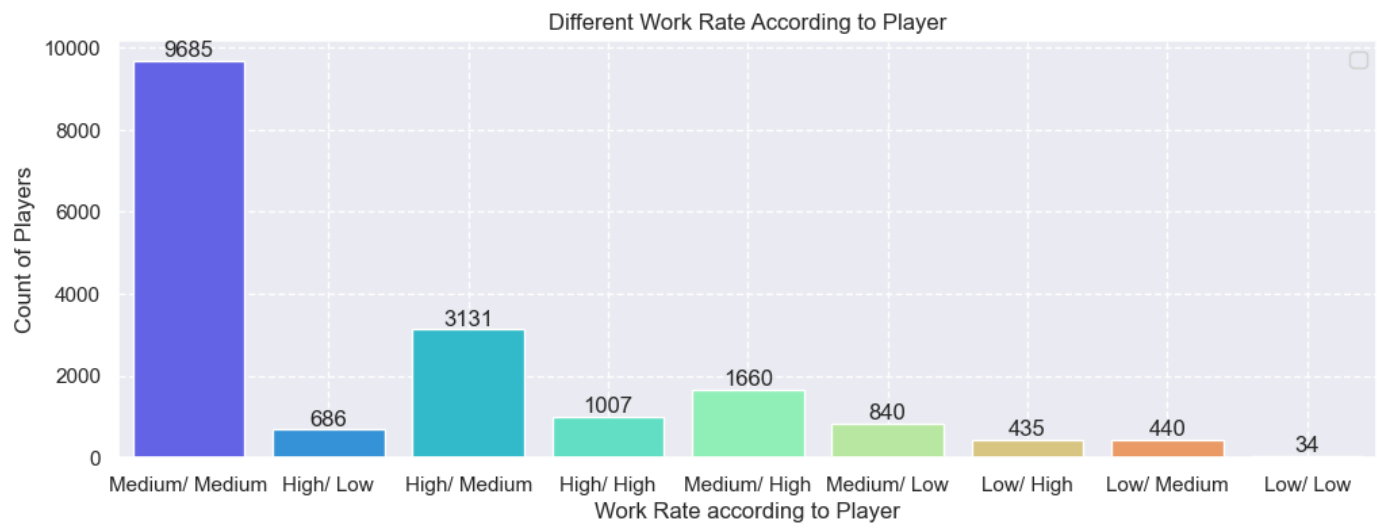
```
Out[56]: Medium/ Medium      9685
High/ Medium      3131
Medium/ High      1660
High/ High      1007
Medium/ Low      840
High/ Low      686
Low/ Medium      440
Low/ High      435
Low/ Low      34
Name: Work Rate, dtype: int64
```

```
In [57]: plt.figure(figsize=(12,4))
ax = sns.countplot(x='Work Rate', data = fifa_data, palette='rainbow')
for bars in ax.containers:
    ax.bar_label(bars)

plt.xlabel('Work Rate according to Player')
plt.ylabel('Count of Players')
plt.grid(linestyle='--')
plt.title('Different Work Rate According to Player')
plt.legend()
```

No artists with labels found to put in legend. Note that artists whose label start with an underscore are ignored when legend() is called with no argument.

```
Out[57]: <matplotlib.legend.Legend at 0x17a1ea92a10>
```



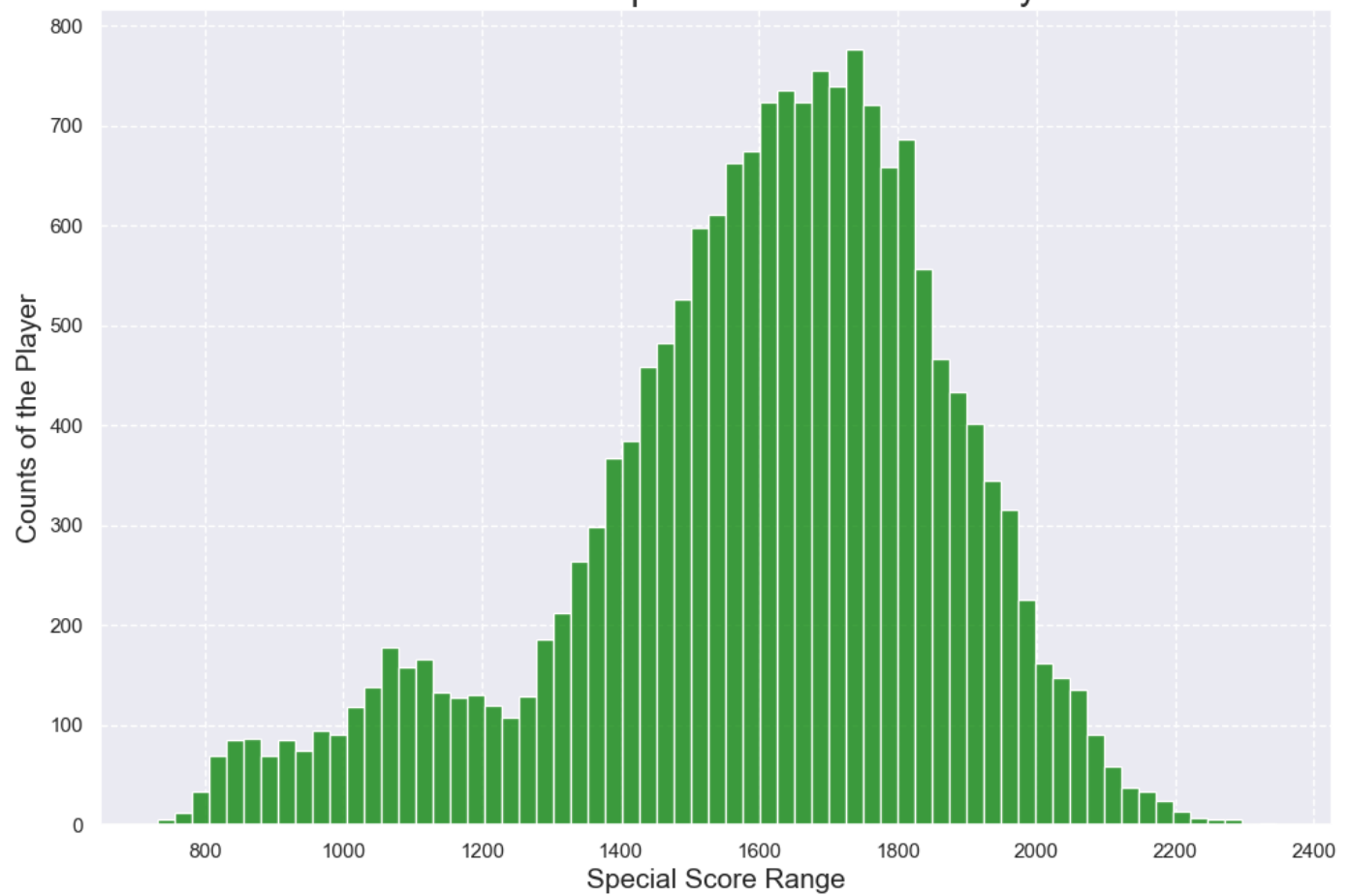
## 14.Special

```
In [58]: fifa_data['Special'].value_counts()
```

```
Out[58]: 1745    46
1628    45
1728    44
1655    42
1691    41
..
2127     1
2197     1
2104     1
2151     1
731      1
Name: Special, Length: 1421, dtype: int64
```

```
In [59]: plt.figure(figsize=(12,8))
sns.histplot(x='Special',bins=65,kde=False,data=fifa_data,color='green')
plt.xlabel('Special Score Range', fontsize=15)
plt.ylabel('Counts of the Player', fontsize=15)
plt.grid(linestyle='--')
plt.title('Distribution Of Special Score Of The Player', fontsize=20);
```

Distribution Of Special Score Of The Player



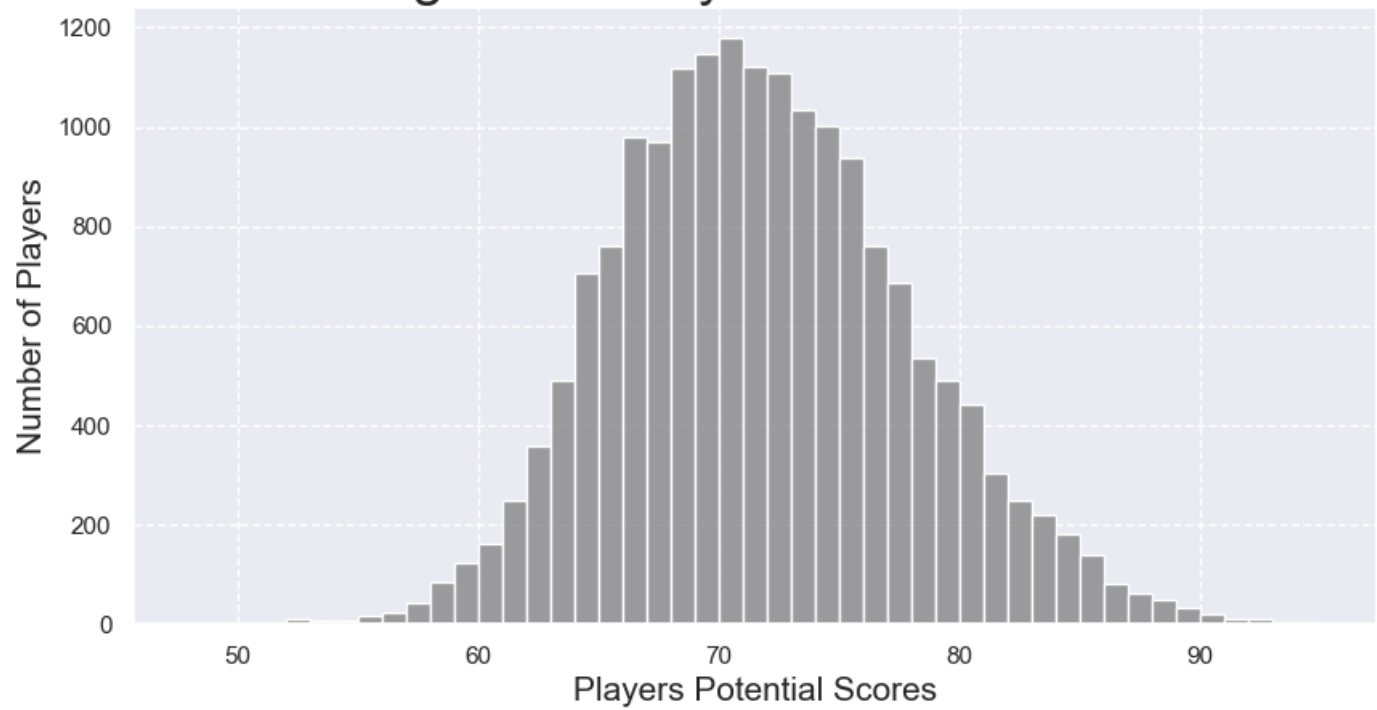
## 15.Potential

```
In [60]: fifa_data['Potential'].value_counts()
```

```
Out[60]: 70    1180
          69    1147
          71    1121
          68    1116
          72    1108
          73    1034
          74    1002
          66     980
          67     970
          75     937
          76     759
          65     759
          64     705
          77     687
          78     536
          79     490
          63     490
          80     442
          62     358
          81     304
          61     249
          82     248
          83     219
          84     180
          60     163
          85     138
          59     124
          58      85
          86      82
          87      61
          88      48
          57      43
          89      33
          56      23
          90      21
          55      18
          91      12
          52      10
          92       9
          54       7
          53       6
          93       4
          94       3
          51       2
          50       2
          48       2
          95       1
Name: Potential, dtype: int64
```

```
In [61]: plt.figure(figsize=(10,5))
sns.histplot(x='Potential',bins=47, data=fifa_data, color='gray')
plt.xlabel('Players Potential Scores', fontsize=15)
plt.ylabel('Number of Players', fontsize=15)
plt.title('Histogram of Players Potential Scores', fontsize=25)
plt.grid(linestyle='--')
```

# Histogram of Players Potential Scores



## 16.Overall

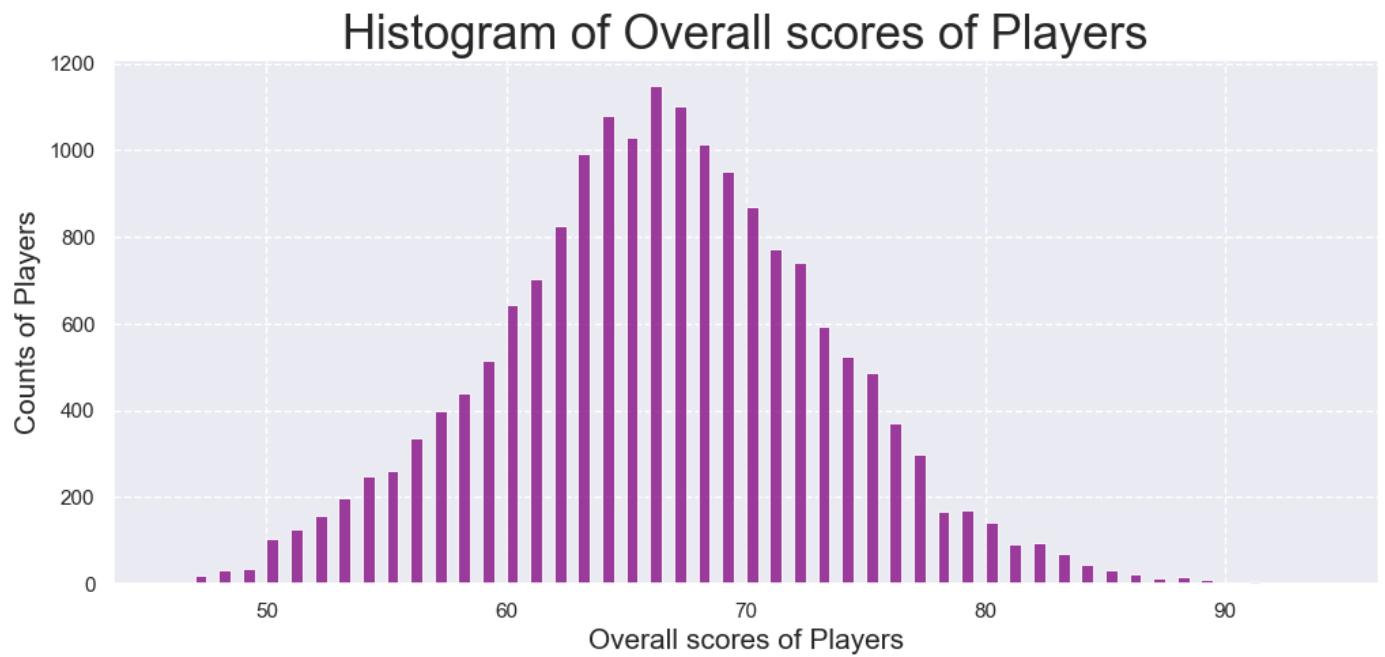
```
In [62]: fifa_data['Overall'].value_counts()
```

```
Out[62]: 66      1150
          67      1101
          64      1079
          65      1031
          68      1015
          63       993
          69       951
          70       870
          62       825
          71       772
          72       742
          61       703
          60       644
          73       594
          74       526
          59       515
          75       488
          58       439
          57       399
          76       371
          56       337
          77       300
          55       262
          54       248
          53       198
          79       169
          78       167
          52       159
          80       141
          51       125
          50       103
          82        94
          81        93
          83        70
          84        45
          49        36
          85        33
          48        32
          86        22
          47        20
          88        17
          87        13
          89        11
          91         6
          90         5
          94         2
          92         1
          46         1
Name: Overall, dtype: int64
```

```
In [63]: len(fifa_data['Overall'].value_counts())
```

```
Out[63]: 48
```

```
In [64]: plt.figure(figsize=(12,5))
sns.histplot(x='Overall',bins=96,kde=False, data=fifa_data, color='purple')
plt.xlabel('Overall scores of Players', fontsize=15)
plt.ylabel('Counts of Players', fontsize=15)
plt.title('Histogram of Overall scores of Players', fontsize=25)
plt.grid(linestyle='--')
```



## 17.Body Type

In [65]: `fifa_data['Body Type'].value_counts()`

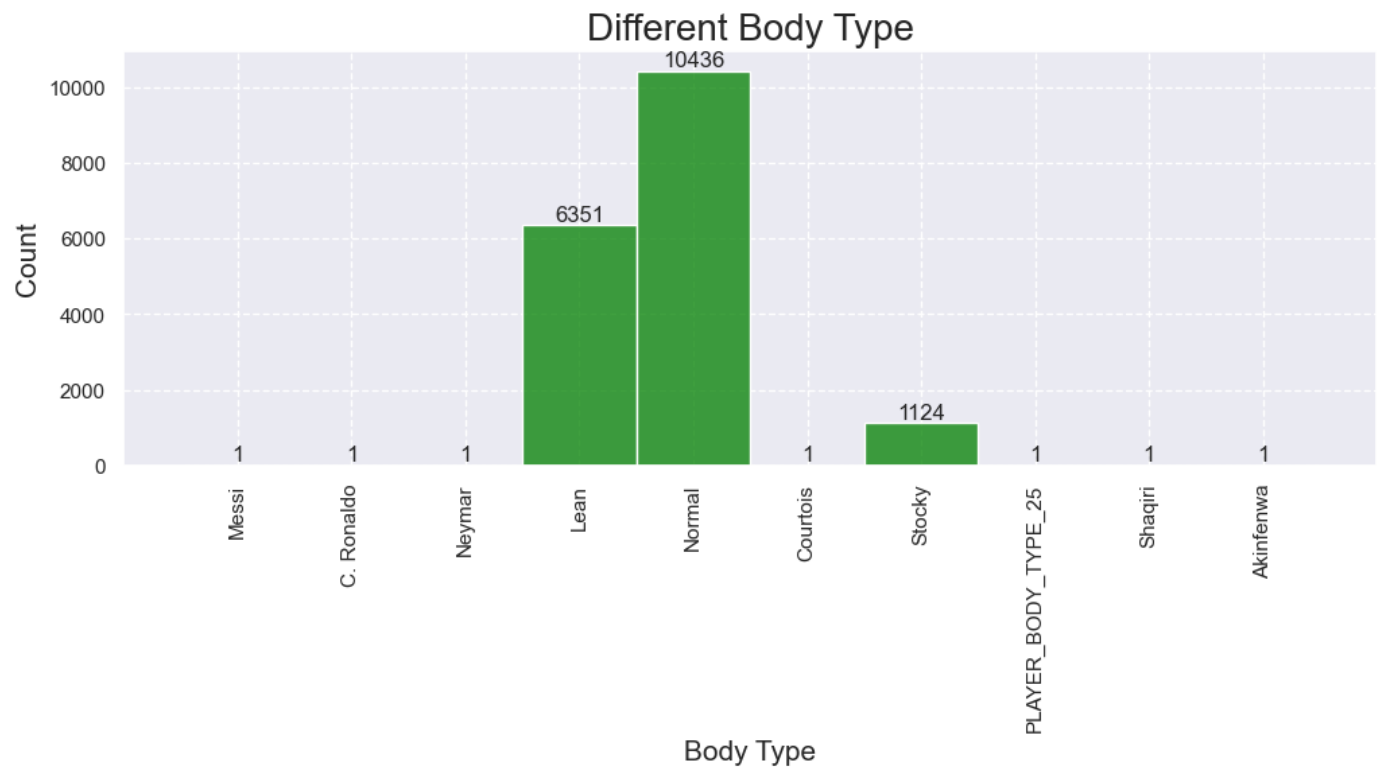
Out[65]:

Normal	10436
Lean	6351
Stocky	1124
Messi	1
C. Ronaldo	1
Neymar	1
Courtois	1
PLAYER_BODY_TYPE_25	1
Shaqiri	1
Akinfenwa	1

Name: Body Type, dtype: int64

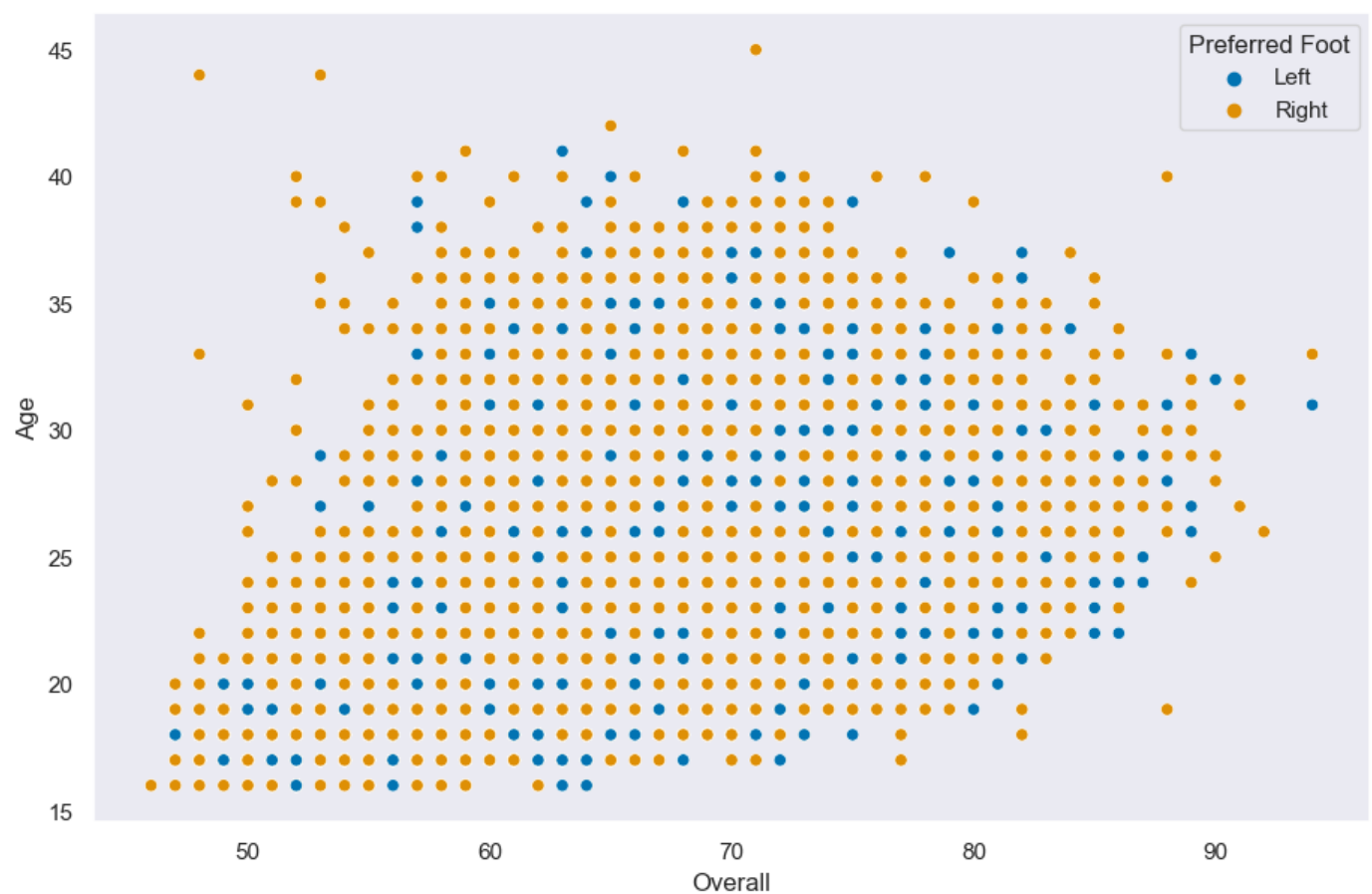
```
In [66]: plt.figure(figsize=(12,4))
ax = sns.histplot(fifa_data['Body Type'], color='green')
for bar in ax.containers:
    ax.bar_label(bar)

plt.xlabel('Body Type', fontsize=15)
plt.ylabel('Count', fontsize=15)
plt.title('Different Body Type', fontsize=20)
plt.grid(True, linestyle='--')
plt.xticks(rotation=90);
```



```
In [67]: plt.figure(figsize=(11,7))
sns.scatterplot(y='Age',x='Overall',hue='Preferred Foot',data=fifa_data)
```

```
Out[67]: <Axes: xlabel='Overall', ylabel='Age'>
```



```
In [68]: column_selected = ['Name', 'Age', 'Nationality', 'Overall', 'Potential', 'Club', 'Value',
                             'Wage', 'Special', 'Preferred Foot', 'International Reputation', 'Wea
                             'Skill Moves', 'Work Rate', 'Body Type', 'Position', 'Height', 'Weight',
                             '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', 'GKDivin  
'GKKicking', 'GKPositioning', 'GKReflexes', 'Release Clause']
```

```
selected_data = pd.DataFrame(fifa_data, columns=column_selected)  
selected_data.columns
```

```
Out[68]: Index(['Name', 'Age', 'Nationality', 'Overall', 'Potential', 'Club', 'Value',  
            'Wage', 'Special', 'Preferred Foot', 'International Reputation',  
            'Weak Foot', 'Skill Moves', 'Work Rate', 'Body Type', 'Position',  
            'Height', 'Weight', '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', 'GKDivin',  
            'GKHandling', 'GKKicking', 'GKPositioning', 'GKReflexes',  
            'Release Clause'],  
            dtype='object')
```

```
In [69]: selected_data.sample(10)
```

```
Out[69]:
```

	Name	Age	Nationality	Overall	Potential	Club	Value	Wage	Special	Preferred Foot	...	Composu
9068	D. Balanta	24	Colombia	66	72	Junior FC	€750K	€1K	1474	Right	...	54
7930	R. Di Gennaro	24	Italy	67	70	Inter	€675K	€11K	1068	Right	...	67
16779	C. Rojas	20	Colombia	56	65	Envigado FC	€140K	€1K	1366	Right	...	49
16877	T. Francois	17	Australia	56	74	Fulham	€180K	€3K	1438	Left	...	61
3660	R. Quioto	26	Honduras	72	72	Houston Dynamo	€3.3M	€7K	1818	Right	...	73
1132	L. Acosta	24	Argentina	77	79	DC United	€11M	€8K	1909	Right	...	74
4759	P. Šteinbors	32	Latvia	70	70	Arka Gdynia	€1M	€3K	1163	Right	...	59
16368	M. Ranmark	22	Norway	57	65	Molde FK	€110K	€1K	871	Left	...	38
1839	L. Deaux	29	France	75	75	En Avant de Guingamp	€5M	€21K	1838	Right	...	70
3423	Kim Jin Su	26	Korea Republic	72	73	Jeonbuk Hyundai Motors	€3M	€8K	1829	Left	...	63

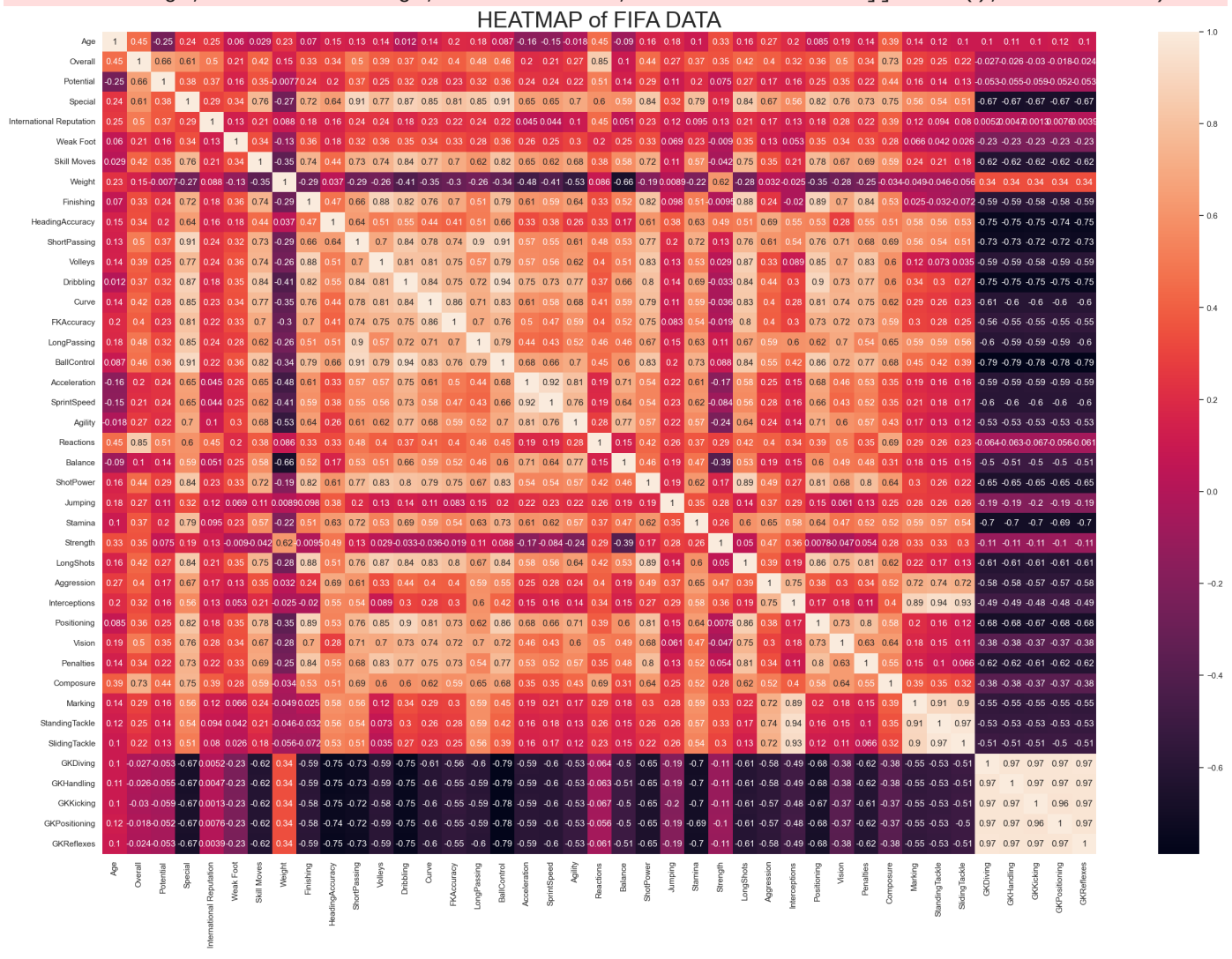
10 rows × 52 columns

```
In [70]: plt.figure(figsize=(30,20))  
sns.heatmap(selected_data[['Name', 'Age', 'Nationality', 'Overall', 'Potential', 'Club', 'Value',  
            'Wage', 'Special', 'Preferred Foot', 'International Reputation', 'Weak Foot',  
            'Skill Moves', 'Work Rate', 'Body Type', 'Position', 'Height', 'Weight', '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', 'GKDivin',  
            'GKHandling', 'GKKicking', 'GKPositioning', 'GKReflexes', 'Release Clause']])
```

```
'GKKicking', 'GKPositioning', 'GKReflexes', 'Release Clause']].corr(
plt.title('HEATMAP of FIFA DATA', fontsize=30);
```

C:\Users\Admin\AppData\Local\Temp\ipykernel\_320\933417802.py:11: FutureWarning: The default value of numeric\_only in DataFrame.corr is deprecated. In a future version, it will default to False. Select only valid columns or specify the value of numeric\_only to silence this warning.

```
'GKKicking', 'GKPositioning', 'GKReflexes', 'Release Clause']].corr(), annot=True)
```



In [ ]: