Project 1: FIFA 21 Players Data Analysis and Visualization



Project Description

Welcome to this data analysis and visualization project! In this project, we will be exploring the 'FIFA 21 Players' dataset which comprises a rich variety of information about soccer players from around the world. The data extracted are in 'csv' file format.

Your task is to clean and prepare the data for analysis, and then leverage your data visualization skills to uncovers insights about FIFA 21 players, teams, and overall value and wage dynamics.

Here are your tasks outlined in detail:

- 1. Data Cleaning: The initial dataset contains some unnecessary and redundant information. Your first task is to clean and prepare it for analysis. In doing so:
 - Remove the 'Unnamed: 0' column from the dataset.
 - · Removing the Newline Characters.
 - Look for columns which contain star ('★') characters. Remove these characters from all such columns.
 - Fill in the missing values using an appropriate method.
 - In the 'Value', 'Wage', and 'Release Clause' columns, remove the '€', 'K', and 'M' characters and convert the currency value to 'int'. For eg, "M" in value column is Million, so multiply the row values by 1,000,000, etc.
- 2. Data Visualization: After the data has been cleaned and prepared, use appropriate data visualization methods to represent the data as follows:
 - Create a Pie Chart to display the preferred foot ('Left' or 'Right') among the FIFA 21 players.
 - Use a Bar Plot to represent the top 3 FIFA 21 players.
 - Visualize the best FIFA teams from 100 players based on the 'OVA' score using a Bar Plot.
 - Use a Bar Plot to display the top 10 most valuable FIFA teams.
 - Create a Scatter Plot to show players who are highly valuable but still underpaid.

Ensure all plots are fully labeled with titles, x and y variables, and a legend if necessary.

Remember, the goal of this project is not just about getting the right answer but also about presenting your findings in a clear, logical, and visually appealing manner. Good luck and enjoy the analysis!

Read File

In [39]: import pandas as pd

fifa_df = pd.read_csv('fifa21_raw_data.csv', low_memory=False) fifa_df.head()

:		Unnamed (ID	Name	LongName	Nationality	Age	↓OVA	РОТ	Club	Contract	•••	A/W	D/W	П
	0	()	158023	L. Messi	Lionel Messi	Argentina	33	93	93	\n\n\n\nFC Barcelona	2004 ~ 2021		Medium	Low	1
	1		1	20801	Cristiano Ronaldo	C. Ronaldo dos Santos Aveiro	Portugal	35	92	92	\n\n\n\nJuventus	2018 ~ 2022		High	Low	*
	2	2	2	200389	J. Oblak	Jan Oblak	Slovenia	27	91	93	\n\n\n\nAtlético Madrid	2014 ~ 2023		Medium	Medium	;
	3	3	3	192985	K. De Bruyne	Kevin De Bruyne	Belgium	29	91	91	\n\n\n\nManchester City	2015 ~ 2023		High	High	4
	4	2	1	190871	Neymar Jr	Neymar da Silva Santos Jr.	Brazil	28	91	91	\n\n\n\nParis Saint- Germain	2017 ~ 2022	•••	High	Medium	1

Data Cleaning

 $5 \text{ rows} \times 35 \text{ columns}$

Out[39]

Task 1: Removing Unuseful Columns

	ID	Name	LongName	Nationality	Age	↓OVA	РОТ	Club	Contract	Positions		A/W	D/W	IR
0	158023	L. Messi	Lionel Messi	Argentina	33	93	93	\n\n\n\nFC Barcelona	2004 ~ 2021	RW, ST, CF		Medium	Low	5 ★
	20801	Cristiano Ronaldo	C. Ronaldo dos Santos Aveiro	Portugal	35	92	92	\n\n\n\nJuventus	2018 ~ 2022	ST, LW		High	Low	5 ★
2	00389	J. Oblak	Jan Oblak	Slovenia	27	91	93	\n\n\n\nAtlético Madrid	2014 ~ 2023	GK		Medium	Medium	3 ★
	192985	K. De Bruyne	Kevin De Bruyne	Belgium	29	91	91	\n\n\n\nManchester City	2015 ~ 2023	CAM, CM	•••	High	High	4 ★
	190871	Neymar Jr	Neymar da Silva Santos Jr.	Brazil	28	91	91	\n\n\n\nParis Saint- Germain	2017 ~ 2022	LW, CAM		High	Medium	5 ★

Task 2: Removing Newline Character

```
In [41]: fifa_df.replace(to_replace=r'(\n)+', value='', regex=True, inplace=True)
fifa_df.T
```

:		0	1	2	3	4	5	6	7	8	9		1896
	ID	158023	20801	200389	192985	190871	188545	209331	212831	231747	192448		25793
	Name	L. Messi	Cristiano Ronaldo	J. Oblak	K. De Bruyne	Neymar Jr	R. Lewandowski	M. Salah	Alisson	K. Mbappé	M. ter Stegen	•••	Son Yu
ı	_ongName	Lionel Messi	C. Ronaldo dos Santos Aveiro	Jan Oblak	Kevin De Bruyne	Neymar da Silva Santos Jr.	Robert Lewandowski	Mohamed Salah	Alisson Ramses Becker	Kylian Mbappé	Marc- André ter Stegen		Yu Son
1	Nationality	Argentina	Portugal	Slovenia	Belgium	Brazil	Poland	Egypt	Brazil	France	Germany		Chin P
	Age	33	35	27	29	28	31	28	27	21	28		2
	↓OVA	93	92	91	91	91	91	90	90	90	90		4
	РОТ	93	92	93	91	91	91	90	91	95	93		4
	Club	FC Barcelona	Juventus	Atlético Madrid	Manchester City	Paris Saint- Germain	FC Bayern München	Liverpool	Liverpool	Paris Saint- Germain	FC Barcelona		Tianji TED. Fi
	Contract	2004 ~ 2021	2018 ~ 2022	2014 ~ 2023	2015 ~ 2023	2017 ~ 2022	2014 ~ 2023	2017 ~ 2023	2018 ~ 2024	2018 ~ 2022	2014 ~ 2022		2020 202
	Positions	RW, ST, CF	ST, LW	GK	CAM, CM	LW, CAM	ST	RW	GK	ST, LW, RW	GK		CI
	Height	170cm	187cm	188cm	181cm	175cm	184cm	175cm	191cm	178cm	187cm		185cr
	Weight	72kg	83kg	87kg	70kg	68kg	80kg	71kg	91kg	73kg	85kg		79k
	Preferred Foot	Left	Right	Right	Right	Right	Right	Left	Right	Right	Right		Righ
	BOV	93	92	91	91	91	91	90	90	91	90	•••	4
	Best Position	RW	ST	GK	CAM	LW	ST	RW	GK	ST	GK		RI
	Joined	Jul 1, 2004	Jul 10, 2018	Jul 16, 2014	Aug 30, 2015	Aug 3, 2017	Jul 1, 2014	Jul 1, 2017	Jul 19, 2018	Jul 1, 2018	Jul 1, 2014		Aug [*] 202
	Loan Date End	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN		Nal
	Value	€103.5M	€63M	€120M	€129M	€132M	€111M	€120.5M	€102M	€185.5M	€110M	•••	€45
	Wage	€560K	€220K	€125K	€370K	€270K	€240K	€250K	€160K	€160K	€260K		€2
	Release Clause	€138.4M	€75.9M	€159.4M	€161M	€166.5M	€132M	€144.3M	€120.3M	€203.1M	€147.7M		€47
-	Total Stats	2231	2221	1413	2304	2175	2195	2211	1389	2147	1442	•••	135
ı	Base Stats	466	464	489	485	451	457	470	490	466	484	•••	29
	W/F	4 ★	4 ★	3 ★	5 ★	5 ★	4 ★	3 ★	3 ★	4 ★	4 ★		2 🕽
	SM	4★	5★	1★	4★	5★	4★	4★	1★	5★	1★	•••	27
	A/W	Medium	High	Medium	High	High	High	High	Medium	High		•••	Loı
	D/W	Low	Low	Medium	High	Medium	Medium	Medium	Medium	Low	Medium	•••	Lo
	IR	5 ★	5 ★	3 ★	4 ★	5 ★	4 ★	3 ★	3 ★	3 ★		•••	1 7
	PAC	85	89	87	76	91	78	93	86	96	88	•••	6
	SHO	92	93	92	86	85	91	86	88	86	85	•••	3
	PAS	91	81	78	93	86	78	81	85	78	88	•••	4
	DRI DEF	95	89	90	88	94	85	90	89 51	91	90	•••	4
	PHY	38	35 77	52 90	64 78	36 59	43 82	45 75	51 91	39 76	45 88	•••	4 5
	Hits	65 771	562	150	207	595	248	75 246	120	76 1.6K	130		Nal

Task 3: Removing Star Characters

```
In [42]: for col in ['IR', 'W/F', 'SM']:
         fifa_df[col] = fifa_df[col].str.replace('*', '')

fifa_df.T
```

		0	1	2	3	4	5	6	7	8	9		1896
	ID	158023	20801	200389	192985	190871	188545	209331	212831	231747	192448		25793
N	ame	L. Messi	Cristiano Ronaldo	J. Oblak	K. De Bruyne	Neymar Jr	R. Lewandowski	M. Salah	Alisson	K. Mbappé	M. ter Stegen		Son Yu
LongN	ame	Lionel Messi	C. Ronaldo dos Santos Aveiro	Jan Oblak	Kevin De Bruyne	Neymar da Silva Santos Jr.	Robert Lewandowski	Mohamed Salah	Alisson Ramses Becker	Kylian Mbappé	Marc- André ter Stegen		Yu Son
Nation	ality	Argentina	Portugal	Slovenia	Belgium	Brazil	Poland	Egypt	Brazil	France	Germany		Chin P
	Age	33	35	27	29	28	31	28	27	21	28		2
1	OVA	93	92	91	91	91	91	90	90	90	90		4
	РОТ	93	92	93	91	91	91	90	91	95	93		4
(Club	FC Barcelona	Juventus	Atlético Madrid	Manchester City	Paris Saint- Germain	FC Bayern München	Liverpool	Liverpool	Paris Saint- Germain	FC Barcelona		Tianji TED. F
Cont	ract	2004 ~ 2021	2018 ~ 2022	2014 ~ 2023	2015 ~ 2023	2017 ~ 2022	2014 ~ 2023	2017 ~ 2023	2018 ~ 2024	2018 ~ 2022	2014 ~ 2022		2020 202
Posit	ions	RW, ST, CF	ST, LW	GK	САМ, СМ	LW, CAM	ST	RW	GK	ST, LW, RW	GK		CI
He	ight	170cm	187cm	188cm	181cm	175cm	184cm	175cm	191cm	178cm	187cm		185cr
We	ight	72kg	83kg	87kg	70kg	68kg	80kg	71kg	91kg	73kg	85kg		79k
Prefe I	rred Foot	Left	Right	Right	Right	Right	Right	Left	Right	Right	Right		Righ
I	BOV	93	92	91	91	91	91	90	90	91	90		4
l Posi	Best ition	RW	ST	GK	CAM	LW	ST	RW	GK	ST	GK		RI
Joi	ined	Jul 1, 2004	Jul 10, 2018	Jul 16, 2014	Aug 30, 2015	Aug 3, 2017	Jul 1, 2014	Jul 1, 2017	Jul 19, 2018	Jul 1, 2018	Jul 1, 2014	•••	Aug ² 202
Loan [Date End	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN	NaN		Nal
V	alue	€103.5M	€63M	€120M	€129M	€132M	€111M	€120.5M	€102M	€185.5M	€110M		€45
W	/age	€560K	€220K	€125K	€370K	€270K	€240K	€250K	€160K	€160K	€260K		€2
	ease ause	€138.4M	€75.9M	€159.4M	€161M	€166.5M	€132M	€144.3M	€120.3M	€203.1M	€147.7M		€47
Total S	tats	2231	2221	1413	2304	2175	2195	2211	1389	2147	1442	•••	135
Base S	tats	466	464	489	485	451	457	470	490	466	484	•••	29
	W/F	4	4	3	5	5	4	3	3	4	4	•••	
	SM	4	5	1	4	5	4	4	1	5		•••	
	A/W	Medium	High	Medium	High	High	High	High	Medium	High	Medium		Loı
	D/W	Low	Low	Medium	High	Medium	Medium	Medium	Medium	Low	Medium		Lo
	IR	5	5	3	76	5	4	3	3	3	3	•••	6
	PAC SHO	85 92	89 93	87 92	76 86	91 85	78 91	93 86	86 88	96 86	88 85		6
	PAS	91	81	78	93	86	78	81	85	78	88		4
	DRI	95	89	90	88	94	85	90	89	91	90		4
	DEF	38	35	52	64	36	43	45	51	39	45		4
	PHY	65	77	90	78	59	82	75	91	76	88		5
	11:4-	774	F.C.2	150	207	FOF	240	0.40	100	1.01/	100		NI-1

Hits

Task 4: Filling Missing Values

771

562

150

207

595

248

246

120

1.6K

130 ...

	0	1	2	3	4	5	6	7	8	9	•••	189
ID	158023	20801	200389	192985	190871	188545	209331	212831	231747	192448		2579
Name	L. Messi	Cristiano Ronaldo	J. Oblak	K. De Bruyne	Neymar Jr	R. Lewandowski	M. Salah	Alisson	K. Mbappé	M. ter Stegen		So Y
LongName	Lionel Messi	C. Ronaldo dos Santos Aveiro	Jan Oblak	Kevin De Bruyne	Neymar da Silva Santos Jr.	Robert Lewandowski	Mohamed Salah	Alisson Ramses Becker	Kylian Mbappé	Marc- André ter Stegen	•••	Y So
Nationality	Argentina	Portugal	Slovenia	Belgium	Brazil	Poland	Egypt	Brazil	France	Germany		China
Age	33	35	27	29	28	31	28	27	21	28		
↓OVA	93	92	91	91	91	91	90	90	90	90		
РОТ	93	92	93	91	91	91	90	91	95	93		
Club	FC Barcelona	Juventus	Atlético Madrid	Manchester City	Paris Saint- Germain	FC Bayern München	Liverpool	Liverpool	Paris Saint- Germain	FC Barcelona		Tiar TEDA
Contract	2004 ~ 2021	2018 ~ 2022	2014 ~ 2023	2015 ~ 2023	2017 ~ 2022	2014 ~ 2023	2017 ~ 2023	2018 ~ 2024	2018 ~ 2022	2014 ~ 2022		202(20
Positions	RW, ST, CF	ST, LW	GK	CAM, CM	LW, CAM	ST	RW	GK	ST, LW, RW	GK		(
Height	170cm	187cm	188cm	181cm	175cm	184cm	175cm	191cm	178cm	187cm		185
Weight	72kg	83kg	87kg	70kg	68kg	80kg	71kg	91kg	73kg	85kg		79
Preferred Foot	Left	Right	Right	Right	Right	Right	Left	Right	Right	Right		Riç
BOV	93	92	91	91	91	91	90	90	91	90		
Best Position	RW	ST	GK	CAM	LW	ST	RW	GK	ST	GK		ŀ
Joined	Jul 1, 2004	Jul 10, 2018	Jul 16, 2014	Aug 30, 2015	Aug 3, 2017	Jul 1, 2014	Jul 1, 2017	Jul 19, 2018	Jul 1, 2018	Jul 1, 2014		Auç 20
Loan Date End	Not on Loan	Not on Loan	Not on Loan	Not on Loan	Not on Loan	Not on Loan	Not on Loan	Not on Loan	Not on Loan	Not on Loan		Not Lc
Value	€103.5M	€63M	€120M	€129M	€132M	€111M	€120.5M	€102M	€185.5M	€110M		€4
Wage	€560K	€220K	€125K	€370K	€270K	€240K	€250K	€160K	€160K	€260K	•••	€
Release Clause	€138.4M	€75.9M	€159.4M	€161M	€166.5M	€132M	€144.3M	€120.3M	€203.1M	€147.7M		€4
Total Stats	2231	2221	1413	2304	2175	2195	2211	1389	2147	1442		13
Base Stats	466	464	489	485	451	457	470	490	466	484	•••	2
W/F	4	4	3	5	5	4	3	3	4	4	•••	
SM	4	5	1	4	5	4	4	1	5		•••	
A/W	Medium	High	Medium	High	High	High	High	Medium	High	Medium		L
D/W IR	Low 5	Low 5	Medium 3	High 4	Medium 5	Medium	Medium 3	Medium 3	Low	Medium 3		L
PAC	85	89	87	76	91	78	93	86	3 96	88	•••	
SHO	92	93	92	86	85	91	86	88	86	85		
PAS	91	81	78	93	86	78	81	85	78			
DRI	95	89	90	88	94	85	90	89	91	90		
DEF	38	35	52	64	36	43	45	51	39	45		
PHY	65	77	90	78	59	82	75	91	76	88		
Hits	771	562	150	207	595	248	246	120	1.6K	130		Unkno

Task 5: Cleaning and Converting Financial Data

```
In [44]:
    def convert_value(value):
        value = value.replace('€', '').replace('.', '').replace('K', '000').replace('M', '000000')
        return int(value)

for column in ['Value', 'Wage', 'Release Clause']:
        fifa_df[column] = fifa_df[column].apply(convert_value)
```

display(fifa_df.info())
display(fifa_df.T)

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 18979 entries, 0 to 18978
Data columns (total 34 columns):

#	Column	Non-Nu	ıll Count	Dtype
0	ID	18979	non-null	int64
1	Name	18979	non-null	object
2	LongName	18979	non-null	object
3	Nationality	18979	non-null	object
4	Age	18979	non-null	int64
5	↓0VA	18979	non-null	int64
6	P0T	18979	non-null	int64
7	Club	18979	non-null	object
8	Contract	18979	non-null	object
9	Positions	18979	non-null	object
10	Height	18979	non-null	object
11	Weight	18979	non-null	object
12	Preferred Foot	18979	non-null	object
13	BOV	18979	non-null	int64
14	Best Position	18979	non-null	object
15	Joined	18979	non-null	object
16	Loan Date End	18979	non-null	object
17	Value	18979	non-null	int64
18	Wage	18979	non-null	int64
19	Release Clause	18979	non-null	int64
20	Total Stats	18979	non-null	int64
21	Base Stats	18979	non-null	int64
22	W/F	18979	non-null	object
23	SM	18979	non-null	object
24	A/W	18979	non-null	object
25	D/W	18979	non-null	object
26	IR	18979	non-null	object
27	PAC	18979	non-null	int64
28	SH0	18979	non-null	int64
29	PAS	18979	non-null	int64
30	DRI	18979	non-null	int64
31	DEF	18979	non-null	int64
32	PHY	18979	non-null	int64
33	Hits	18979		object
1.1	' ' (((()	1 / .		

dtypes: int64(16), object(18) memory usage: 4.9+ MB

None

	0	1	2	3	4	5	6	7	8
ID	158023	20801	200389	192985	190871	188545	209331	212831	231747
Name	L. Messi	Cristiano Ronaldo	J. Oblak	K. De Bruyne	Neymar Jr	R. Lewandowski	M. Salah	Alisson	K. Mbappé
LongName	Lionel Messi	C. Ronaldo dos Santos Aveiro	Jan Oblak	Kevin De Bruyne	Neymar da Silva Santos Jr.	Robert Lewandowski	Mohamed Salah	Alisson Ramses Becker	Kylian Mbappé
Nationality	Argentina	Portugal	Slovenia	Belgium	Brazil	Poland	Egypt	Brazil	France
Age	33	35	27	29	28	31	28	27	21
↓OVA	93	92	91	91	91	91	90	90	90
РОТ	93	92	93	91	91	91	90	91	95
Club	FC Barcelona	Juventus	Atlético Madrid	Manchester City	Paris Saint- Germain	FC Bayern München	Liverpool	Liverpool	Paris Saint- Germain
Contract	2004 ~ 2021	2018 ~ 2022	2014 ~ 2023	2015 ~ 2023	2017 ~ 2022	2014 ~ 2023	2017 ~ 2023	2018 ~ 2024	2018 ~ 2022
Positions	RW, ST, CF	ST, LW	GK	САМ, СМ	LW, CAM	ST	RW	GK	ST, LW, RW
Height	170cm	187cm	188cm	181cm	175cm	184cm	175cm	191cm	178cm
Weight	72kg	83kg	87kg	70kg	68kg	80kg	71kg	91kg	73kg
Preferred Foot	Left	Right	Right	Right	Right	Right	Left	Right	Right
BOV	93	92	91	91	91	91	90	90	91
Best Position	RW	ST	GK	CAM	LW	ST	RW	GK	ST
Joined	Jul 1, 2004	Jul 10, 2018	Jul 16, 2014	Aug 30, 2015	Aug 3, 2017	Jul 1, 2014	Jul 1, 2017	Jul 19, 2018	Jul 1, 2018
Loan Date End	Not on Loan	Not on Loan	Not on Loan	Not on Loan	Not on Loan	Not on Loan	Not on Loan	Not on Loan	Not on Loan
Value	1035000000	63000000	120000000	129000000	132000000	111000000	1205000000	102000000	1855000000
Wage	560000	220000	125000	370000	270000	240000	250000	160000	160000
Release Clause	1384000000	759000000	1594000000	161000000	1665000000	132000000	1443000000	1203000000	2031000000
Total Stats	2231	2221	1413	2304	2175	2195	2211	1389	2147
Base Stats	466	464	489	485	451	457	470	490	466
W/F	4	4	3	5	5	4	3	3	4
SM	4	5	1	4	5	4	4	1	5
A/W	Medium	High	Medium	High	High	High	High	Medium	High
D/W	Low	Low	Medium	High	Medium	Medium	Medium	Medium	Low
IR	5	5	3	4	5	4	3	3	3
PAC	85	89	87	76	91	78	93	86	96
SHO	92	93	92	86	85	91	86	88	86
PAS	91	81	78	93	86	78	81	85	78
DRI	95	89	90	88	94	85	90	89	91
DEF	38	35	52	64	36	43	45	51	39
PHY	65	77	90	78	59	82	75	91	76
Hits	771	562	150	207	595	248	246	120	1.6K

Data Visualization

Task 6: Pie Chart Creation - Preferred Foot

```
import matplotlib.pyplot as plt

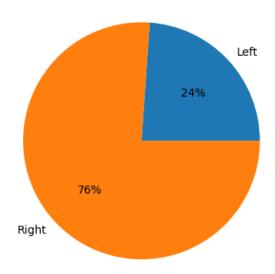
fifa_df_pref_foot = fifa_df.groupby('Preferred Foot').agg({'Name':'count'}).reset_index()

plt.pie(fifa_df_pref_foot.Name, labels=fifa_df_pref_foot['Preferred Foot'], autopct='%.0f%*')

plt.title("Player's Preferred Foot")

plt.show()
```

Player's Preferred Foot

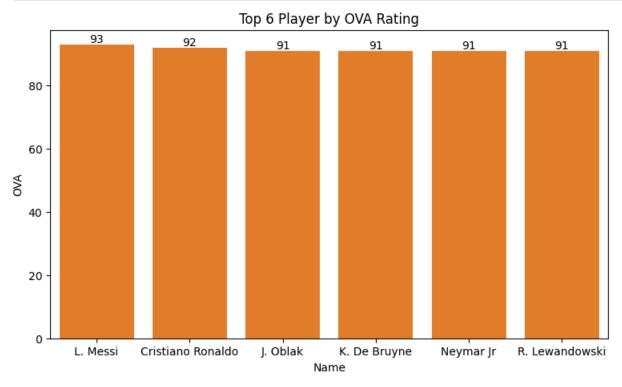


Task 7: Barplot Creation - Top 6 Players

```
In [46]: fifa_df.rename(columns = {'iOVA':'OVA'}, inplace=True)
    top_6_by_OVA = fifa_df.sort_values(by = 'OVA', ascending=False).head(6)[['Name', 'OVA']]
    top_6_by_OVA

import seaborn as sns

plt.figure(figsize=(9,5))
    sns.barplot(top_6_by_OVA, x='Name', y='OVA').bar_label(sns.barplot(top_6_by_OVA, x='Name', y='OVA').containers[0]
    plt.title('Top 6 Player by OVA Rating')
    plt.show()
```

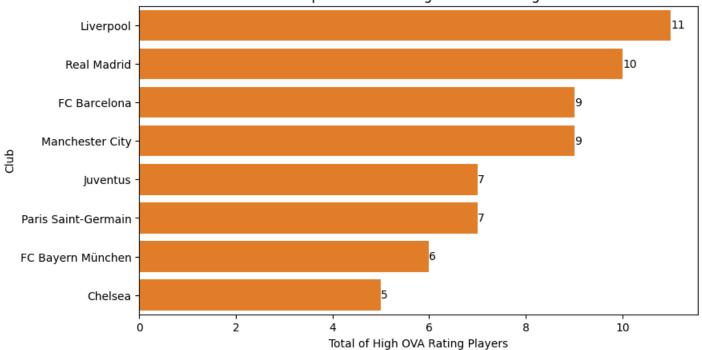


Task 8: Bar Plot Creation - Top 8 Clubs with Best Players

```
In [47]:
    top_100_by_0VA = fifa_df.sort_values(by = 'OVA', ascending=False).head(100)[['Name', 'OVA', 'Club']]
    top_8_club = top_100_by_0VA.groupby('Club').agg({'Name':'count'}).reset_index().sort_values(by = 'Name', ascending top_8_club

plt.figure(figsize=(9,5))
    sns.barplot(top_8_club, x='Name', y='Club').bar_label(sns.barplot(top_8_club, x='Name', y='Club').containers[0],
    plt.title('Top 8 Clubs with Highest OVA Rating')
    plt.xlabel('Total of High OVA Rating Players')
    plt.show()
```

Top 8 Clubs with Highest OVA Rating



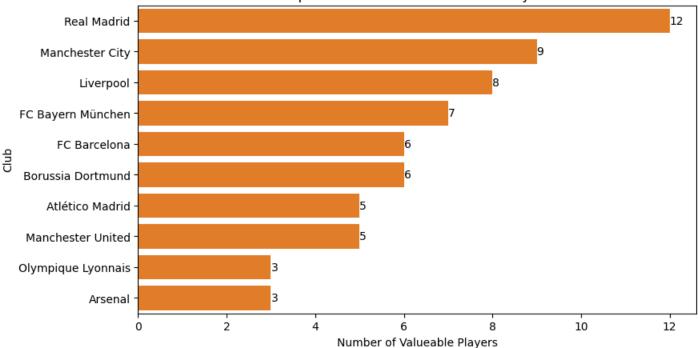
```
In [48]:
top_100_by_value = fifa_df.sort_values(by='Value', ascending=False)[['Name', 'Value', 'Club']].head(100)
top_10_club = top_100_by_value.groupby('Club').agg({'Name':'count'}).reset_index().sort_values(by='Name', ascending=False)[['Name', 'Value', 'Club']].head(100)
top_10_club
```

Out[48]:	Club	Name
25	Real Madrid	12
18	Manchester City	9
17	Liverpool	8
12	FC Bayern München	7
11	FC Barcelona	6
5	Borussia Dortmund	6
4	Atlético Madrid	5
19	Manchester United	5
22	Olympique Lyonnais	3
1	Arsenal	3

Task 9: Bar Plot Creation - Top 10 Teams with Most Valuable Player

```
In [49]:
top_100_by_value = fifa_df.sort_values(by='Value', ascending=False)[['Name', 'Value', 'Club']].head(100)
top_10_club = top_100_by_value.groupby('Club').agg({'Name':'count'}).reset_index().sort_values(by='Name', ascending=False)
plt.figure(figsize=(9,5))
sns.barplot(top_10_club, x='Name', y='Club').bar_label(sns.barplot(top_10_club, x='Name', y='Club').containers[0]
plt.title('Top 10 Clubs with Most Valuable Players')
plt.xlabel('Number of Valueable Players')
plt.show()
```

Top 10 Clubs with Most Valuable Players



Task 10: Scatter Plot Creation - High Value, Underpaid Players

