

FIFA Sports Analysis and Visualization

```
In [1]: import pandas as pd #for linear algebra
import numpy as np #data processing, csv file
import seaborn as sns
sns.set_style('whitegrid')
import matplotlib.pyplot as plt

import warnings
warnings.filterwarnings('ignore')

%matplotlib inline
```

FIFA-Players Analysis

```
In [2]: fifa19 = pd.read_csv(r'C:\Users\Affan\OneDrive\Desktop\FSDS Course NIT\Prakash S
fifa19
```

Out[2]:

	ID	Name	Age	Photo	Nationality
0	158023	L. Messi	31	https://cdn.sofifa.org/players/4/19/158023.png	Argentina
1	20801	Cristiano Ronaldo	33	https://cdn.sofifa.org/players/4/19/20801.png	Portugal
2	190871	Neymar Jr	26	https://cdn.sofifa.org/players/4/19/190871.png	Brazil
3	193080	De Gea	27	https://cdn.sofifa.org/players/4/19/193080.png	Spain
4	192985	K. De Bruyne	27	https://cdn.sofifa.org/players/4/19/192985.png	Belgium
...
18202	238813	J. Lundstram	19	https://cdn.sofifa.org/players/4/19/238813.png	England
18203	243165	N. Christoffersson	19	https://cdn.sofifa.org/players/4/19/243165.png	Sweden
18204	241638	B. Worman	16	https://cdn.sofifa.org/players/4/19/241638.png	England
18205	246268	D. Walker-Rice	17	https://cdn.sofifa.org/players/4/19/246268.png	England
18206	246269	G. Nugent	16	https://cdn.sofifa.org/players/4/19/246269.png	England

18207 rows × 88 columns



In [3]: `fifa19.head()`

Out[3]:

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

5 rows × 88 columns

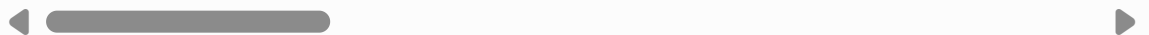


In [4]: `fifa19.tail()`

Out[4]:

	ID	Name	Age	Photo	Nationality	
18202	238813	J. Lundstram	19	https://cdn.sofifa.org/players/4/19/238813.png	England	
18203	243165	N. Christoffersson	19	https://cdn.sofifa.org/players/4/19/243165.png	Sweden	
18204	241638	B. Worman	16	https://cdn.sofifa.org/players/4/19/241638.png	England	
18205	246268	D. Walker-Rice	17	https://cdn.sofifa.org/players/4/19/246268.png	England	
18206	246269	G. Nugent	16	https://cdn.sofifa.org/players/4/19/246269.png	England	

5 rows × 88 columns



In [5]: `fifa19.info()`

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

Index: 18207 entries, 0 to 18206

Data columns (total 88 columns):

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

```

55 HeadingAccuracy      18159 non-null float64
56 ShortPassing         18159 non-null float64
57 Volleys              18159 non-null float64
58 Dribbling            18159 non-null float64
59 Curve                18159 non-null float64
60 FKAccuracy           18159 non-null float64
61 LongPassing          18159 non-null float64
62 BallControl          18159 non-null float64
63 Acceleration         18159 non-null float64
64 SprintSpeed          18159 non-null float64
65 Agility              18159 non-null float64
66 Reactions            18159 non-null float64
67 Balance              18159 non-null float64
68 ShotPower            18159 non-null float64
69 Jumping              18159 non-null float64
70 Stamina              18159 non-null float64
71 Strength             18159 non-null float64
72 LongShots            18159 non-null float64
73 Aggression           18159 non-null float64
74 Interceptions        18159 non-null float64
75 Positioning          18159 non-null float64
76 Vision               18159 non-null float64
77 Penalties            18159 non-null float64
78 Composure            18159 non-null float64
79 Marking              18159 non-null float64
80 StandingTackle       18159 non-null float64
81 SlidingTackle        18159 non-null float64
82 GKDividing           18159 non-null float64
83 GKHandling           18159 non-null float64
84 GKKicking            18159 non-null float64
85 GKPositioning        18159 non-null float64
86 GKReflexes           18159 non-null float64
87 Release Clause       16643 non-null object
dtypes: float64(38), int64(5), object(45)
memory usage: 12.4+ MB

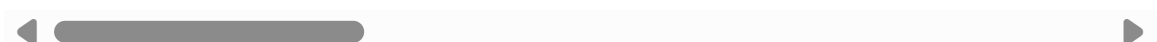
```

In [6]: `fifa19.describe()`

Out[6]:

	ID	Age	Overall	Potential	Special	Internat Reput:
count	18207.000000	18207.000000	18207.000000	18207.000000	18207.000000	18159.00
mean	214298.338606	25.122206	66.238699	71.307299	1597.809908	1.11
std	29965.244204	4.669943	6.908930	6.136496	272.586016	0.39
min	16.000000	16.000000	46.000000	48.000000	731.000000	1.00
25%	200315.500000	21.000000	62.000000	67.000000	1457.000000	1.00
50%	221759.000000	25.000000	66.000000	71.000000	1635.000000	1.00
75%	236529.500000	28.000000	71.000000	75.000000	1787.000000	1.00
max	246620.000000	45.000000	94.000000	95.000000	2346.000000	5.00

8 rows × 43 columns



```
In [7]: fifa19.columns
```

```
Out[7]: Index(['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')
```

```
In [8]: fifa19['Nationality'].unique()
```

```
Out[8]: 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)
```

```
In [9]: fifa19['Body Type'].value_counts()
```

```
Out[9]: Body Type
Normal      10595
Lean        6417
Stocky      1140
Messi       1
C. Ronaldo  1
Neymar      1
Courtois    1
PLAYER_BODY_TYPE_25  1
Shaqiri     1
Akinfenwa   1
Name: count, dtype: int64
```

```
In [10]: fifa19.columns.isna()
```

```
Out[10]: array([False, False, False, False, False, False, False, False, False,
        False, False, False, False, False, False, False, False, False, False,
        False, False, False, False, False, False, False, False, False,
        False, False, False, False, False, False, False, False, False, False,
        False, False, False, False, False, False, False, False, False, False,
        False, False, False, False, False, False, False, False, False, False,
        False, False, False, False, False, False, False, False, False, False,
        False, False, False, False, False, False, False, False, False, False,
        False, False, False, False, False, False, False, False])
```

```
In [11]: fifa19.info()
```

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

Index: 18207 entries, 0 to 18206

Data columns (total 88 columns):

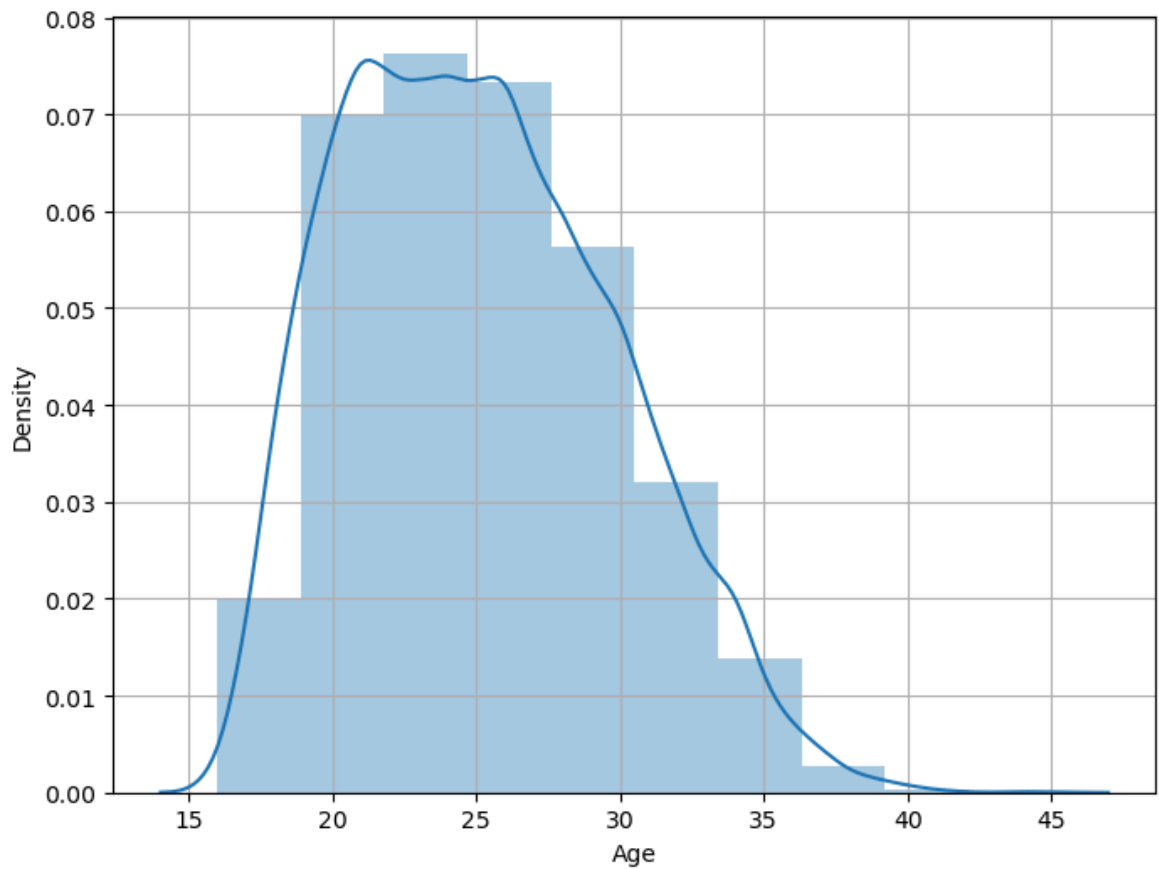
#	Column	Non-Null	Count	Dtype
0	ID	18207	non-null	int64
1	Name	18207	non-null	object
2	Age	18207	non-null	int64
3	Photo	18207	non-null	object
4	Nationality	18207	non-null	object
5	Flag	18207	non-null	object
6	Overall	18207	non-null	int64
7	Potential	18207	non-null	int64
8	Club	17966	non-null	object
9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

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

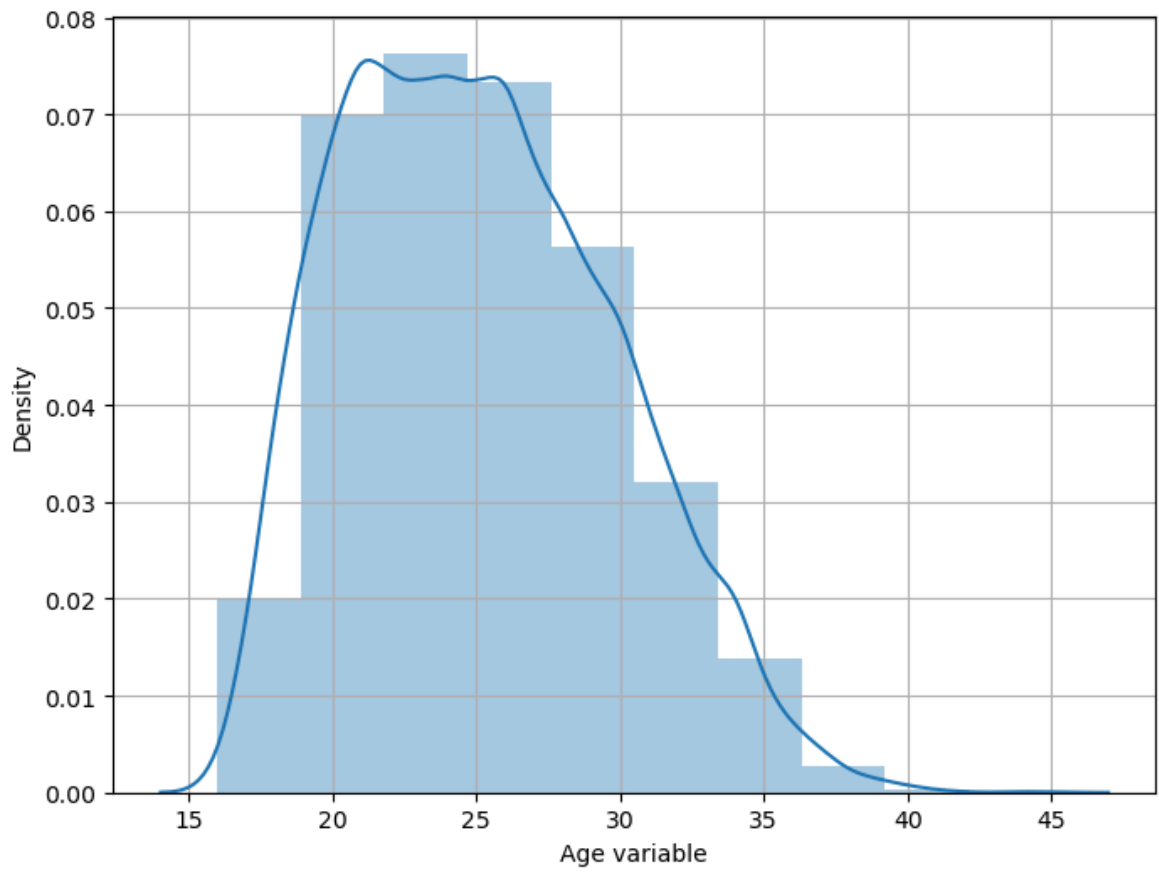
dtypes: float64(38), int64(5), object(45)
memory usage: 12.4+ MB

from the above info we can say that this dataset has 89 total variables out of which 44 are numerical var from which 38 are float, 6 are int dtype and remaning 45 out of 89 var is char dtype ----Below we are going to explore about AGE variable----

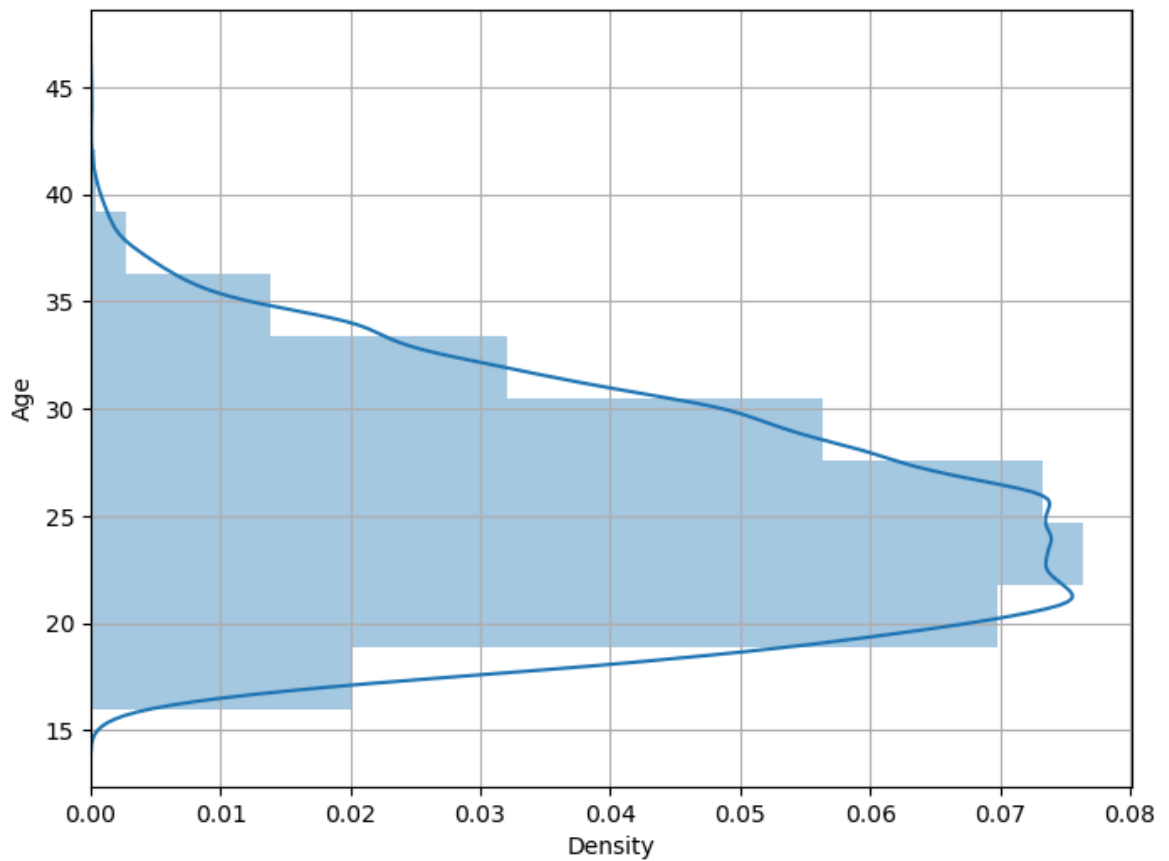
```
In [12]: f,ax = plt.subplots(figsize=(8,6))
x=fifa19['Age']
ax=sns.distplot(x,bins=10)
plt.grid(ax)
plt.show()
```

```
In [13]: f,ax = plt.subplots(figsize=(8,6))
x=fifa19['Age']
x=pd.Series(x,name='Age variable')
ax=sns.distplot(x,bins=10)
plt.grid(ax)
plt.show()
```



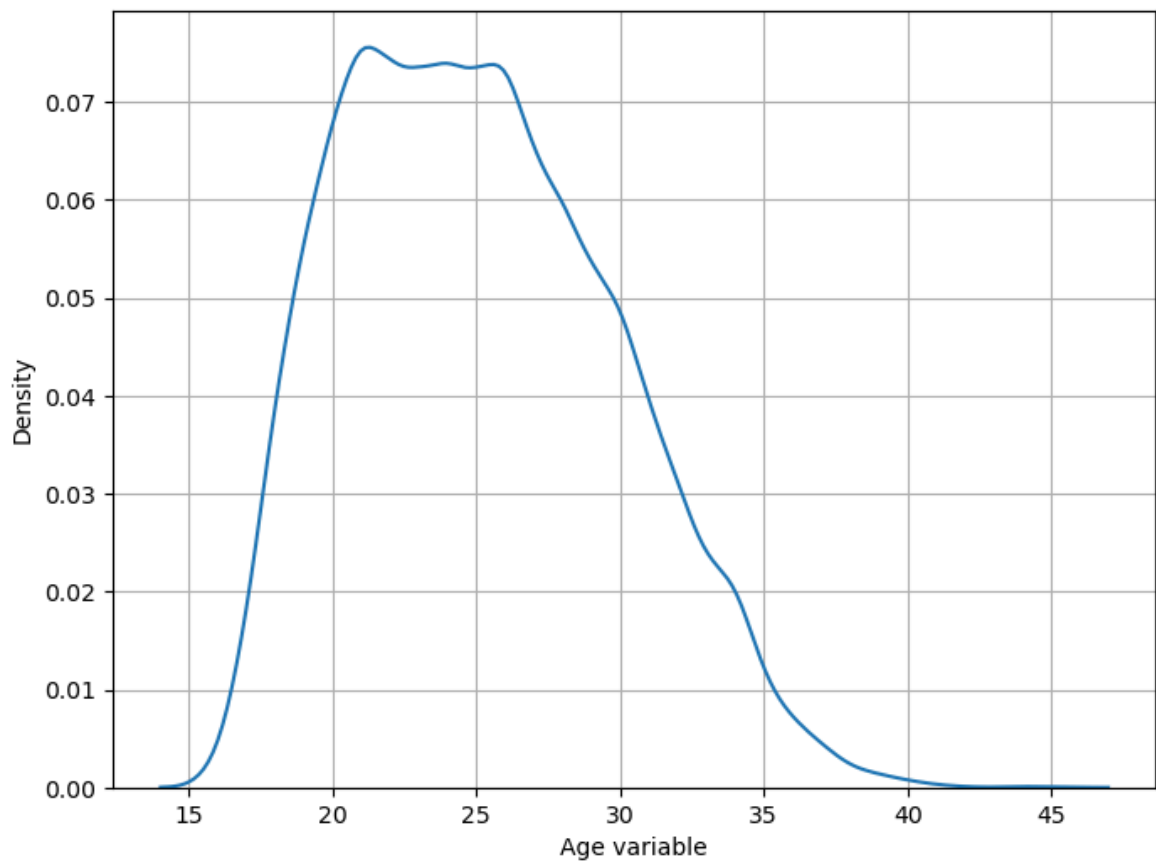
```
In [14]: f,ax = plt.subplots(figsize=(8,6))
x=fifa19['Age']
ax=sns.distplot(x,bins=10,vertical=True)
plt.grid(ax)
plt.show()
```



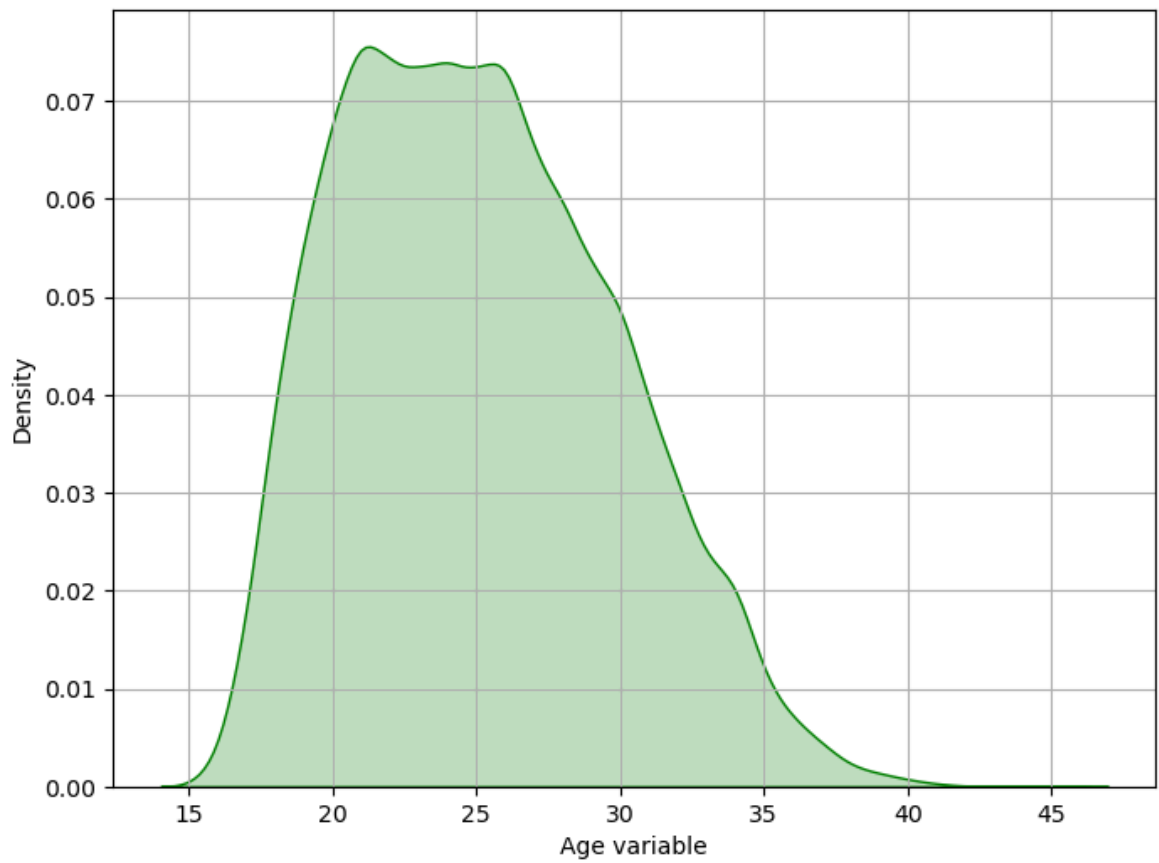
KDE PLOT

kde plot is used for plotting univariate and bivariate analysis

```
In [15]: f,ax = plt.subplots(figsize=(8,6))
x=fifa19['Age']
x=pd.Series(x,name='Age variable')
ax=sns.kdeplot(x)
plt.grid(ax)
plt.show()
```

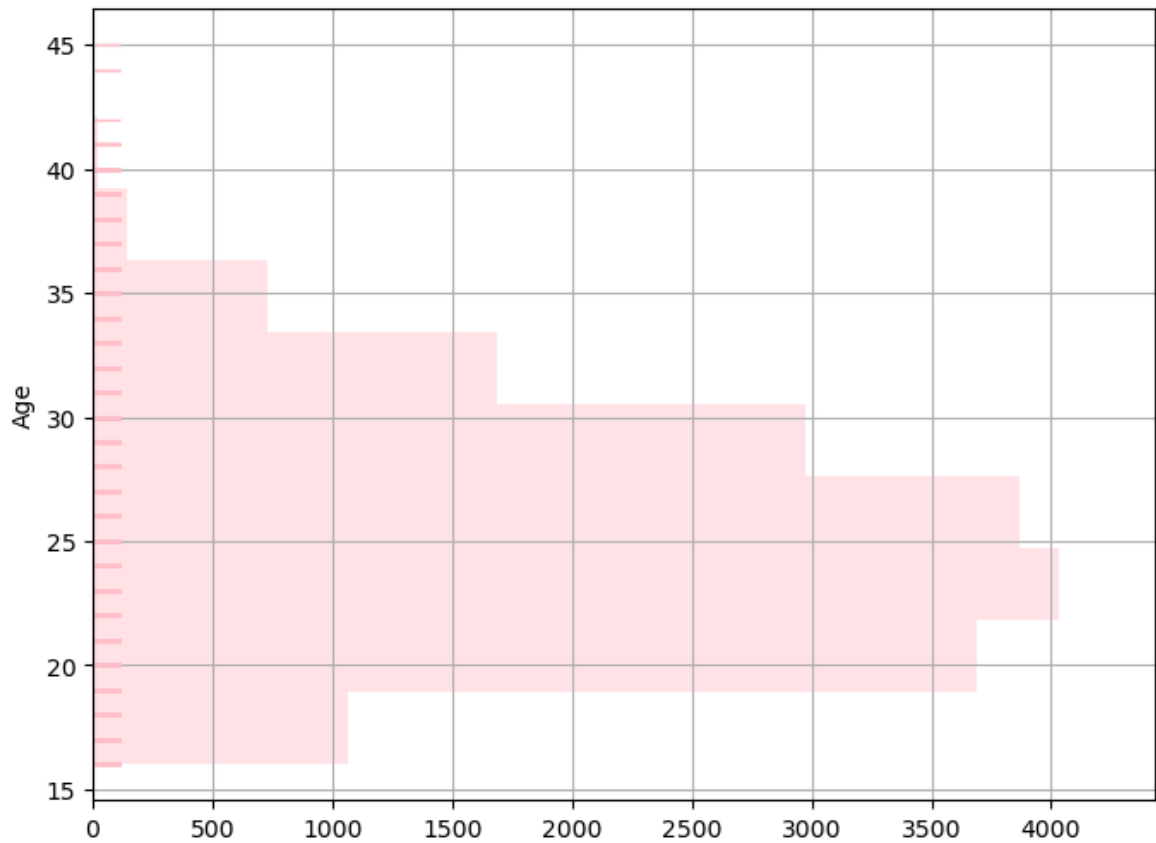


```
In [16]: f,ax = plt.subplots(figsize=(8,6))
x=fifa19['Age']
x=pd.Series(x,name='Age variable')
ax=sns.kdeplot(x,shade=True,color='g')
plt.grid(ax)
plt.show()
```



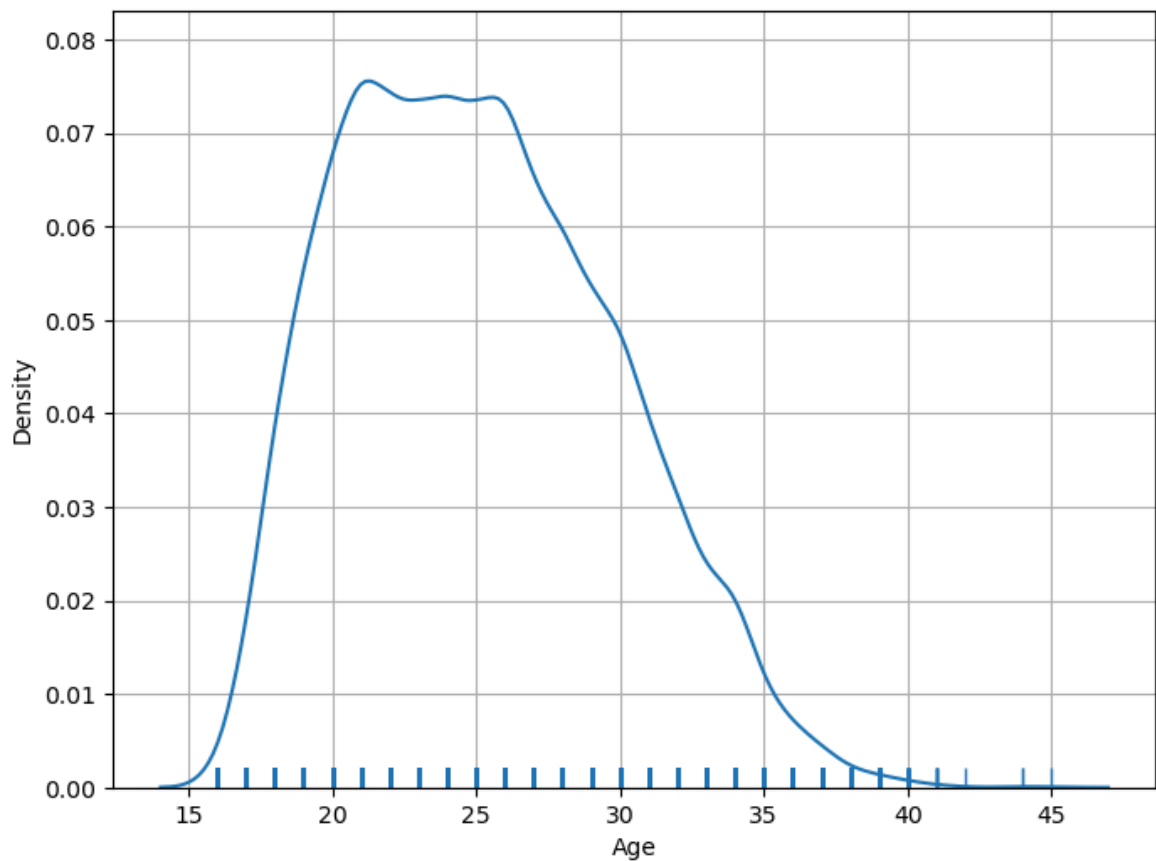
histogram

```
In [17]: f, ax = plt.subplots(figsize=(8,6))
x=fifa19['Age']
ax=sns.distplot(x,kde=False,rug=True,bins=10,color='pink',vertical=True)
plt.grid(ax)
plt.show()
```



alternate way to plot KDE as follows:

```
In [18]: f, ax = plt.subplots(figsize=(8,6))
x = fifa19['Age']
ax = sns.distplot(x, hist=False, rug=True, bins=10)
plt.grid(ax)
plt.show()
```



----Below we are going to explore about 'PREFERRED FOOT' variable----

```
In [19]: fifa19['Preferred Foot'].nunique() #no. of unique val
```

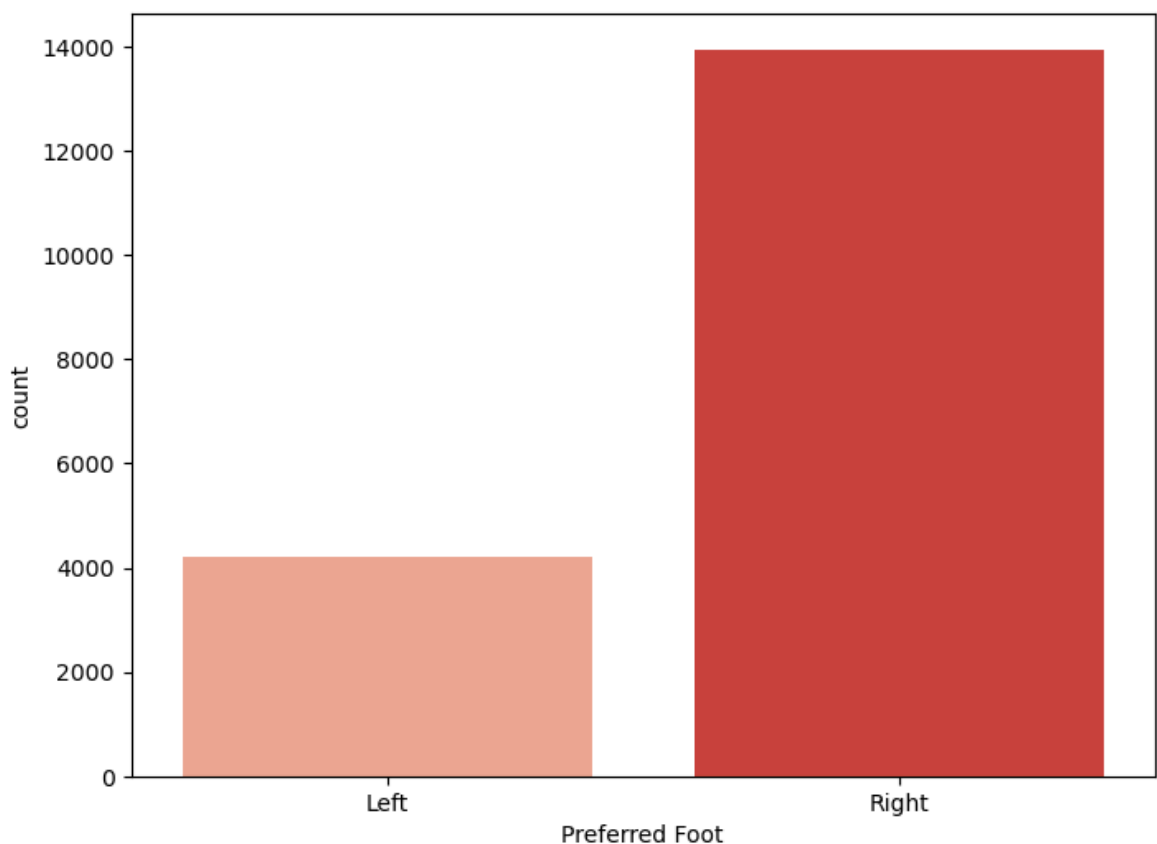
```
Out[19]: 2
```

```
In [20]: fifa19['Preferred Foot'].value_counts() #the var contains two value LEFT n RIGHT
```

```
Out[20]: Preferred Foot  
Right    13948  
Left     4211  
Name: count, dtype: int64
```

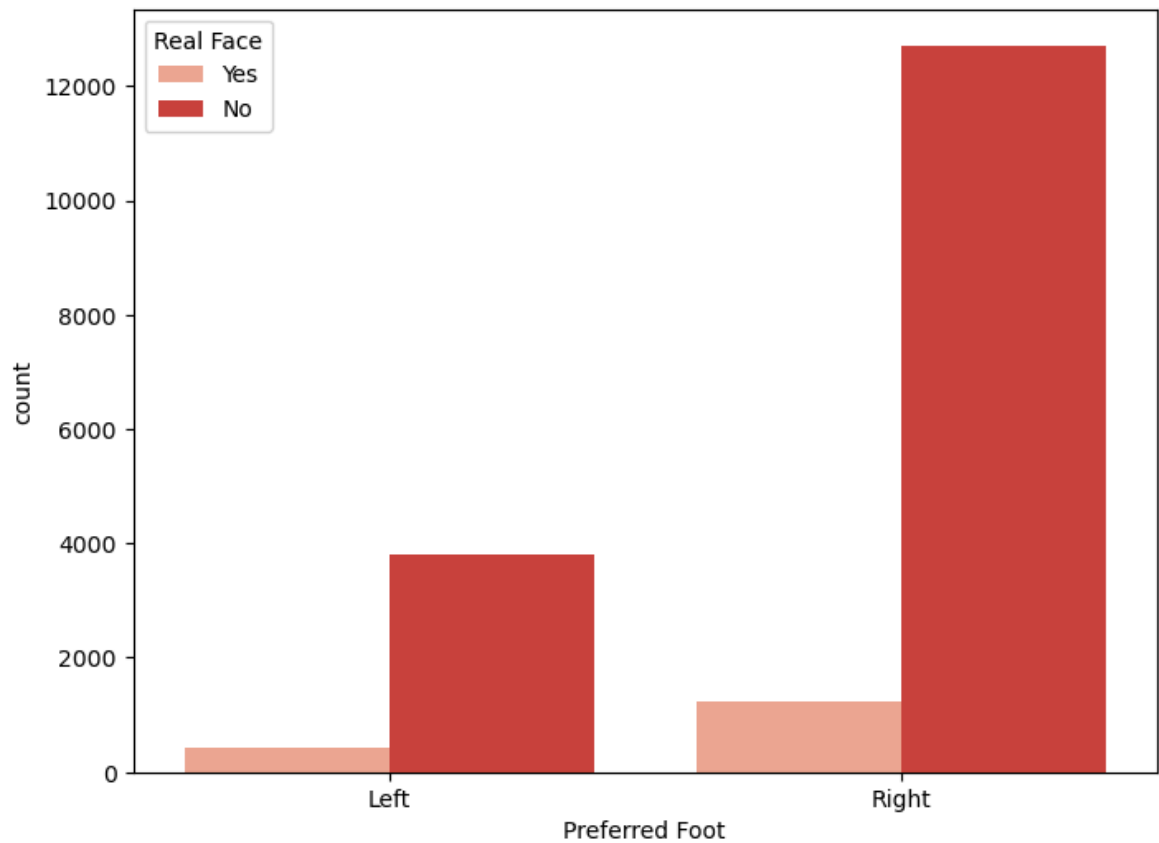
-----Here we are visualizing distribution of values with countplot()

```
In [21]: f,ax=plt.subplots(figsize=(8,6))  
sns.countplot(x='Preferred Foot', data=fifa19,color='c',palette='Reds',fill=True)  
plt.show()
```

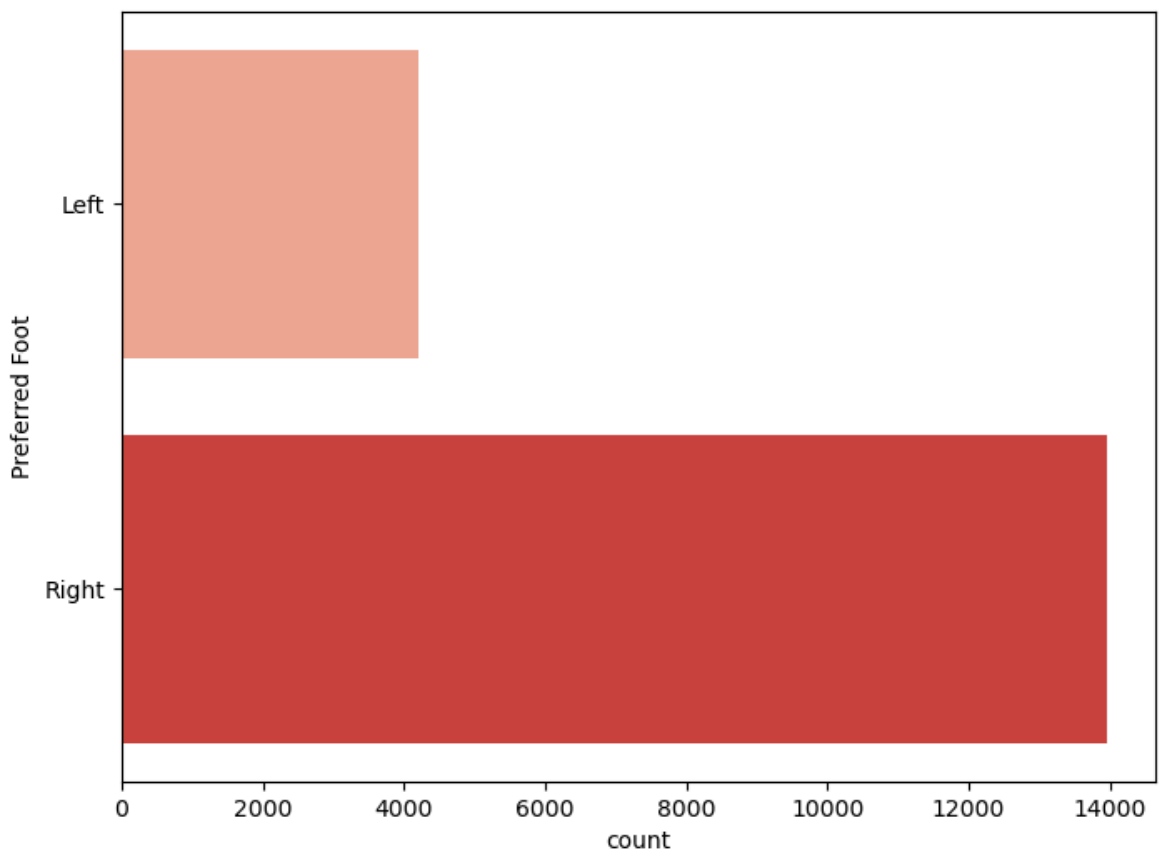


----we can also show value counts for two categorical vars as follows----

```
In [22]: f,ax=plt.subplots(figsize=(8,6))  
sns.countplot(x='Preferred Foot',hue='Real Face', data=fifa19,color='c',palette=  
plt.show()
```

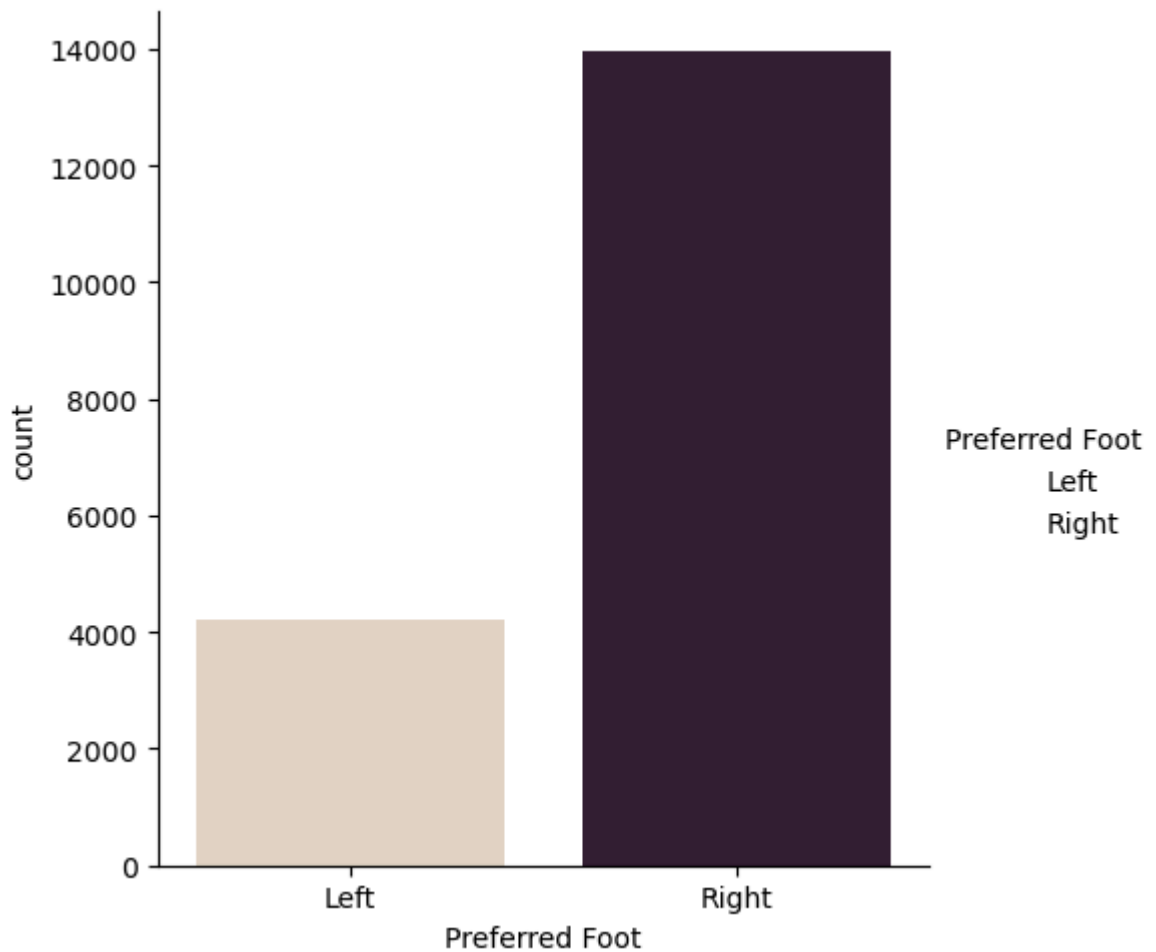


```
In [23]: #if i put 'y=' instead of 'x=' i can get the graph in vertical format  
f,ax=plt.subplots(figsize=(8,6))  
sns.countplot(y='Preferred Foot',data=fifa19,color='c',palette='Reds',fill=True,  
plt.show()
```



catplot() function

```
In [24]: g=sns.catplot(x='Preferred Foot',kind='count',palette='ch:.25',data=fifa19)
plt.show()
```



----exploring 'International Reputation' variable now----

```
In [25]: fifa19['International Reputation'].nunique()
```

```
Out[25]: 5
```

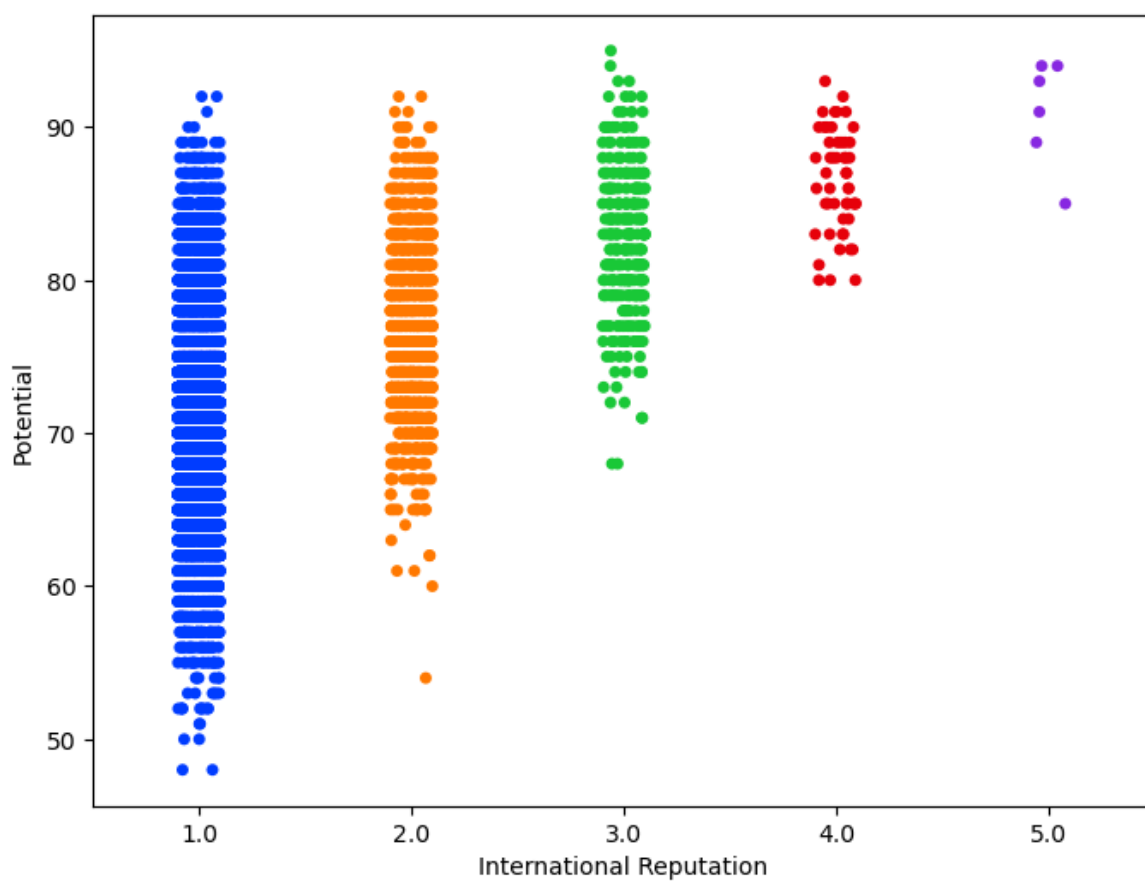
```
In [26]: fifa19['International Reputation'].value_counts()
```

```
Out[26]: International Reputation
1.0    16532
2.0     1261
3.0      309
4.0       51
5.0        6
Name: count, dtype: int64
```

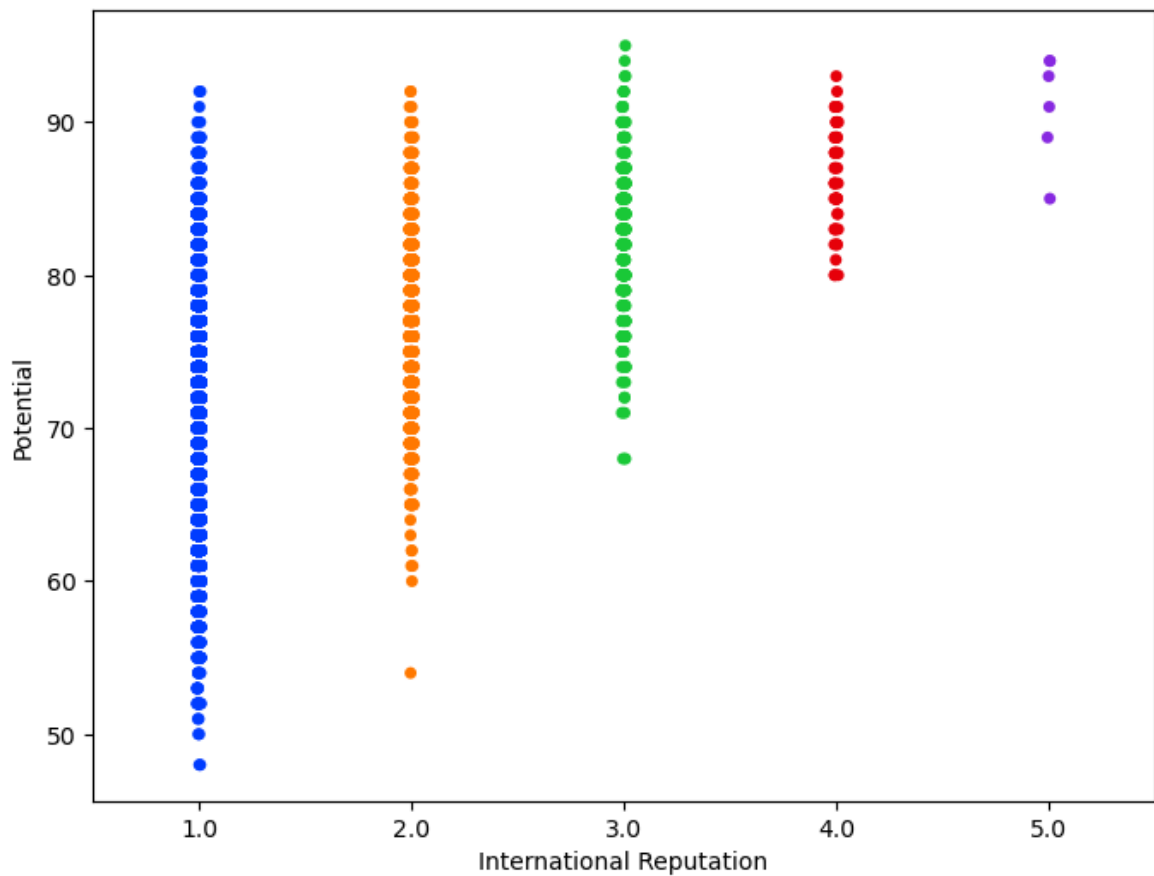
using Stripplot() function

this function draws a scatterplot where one var is categorical

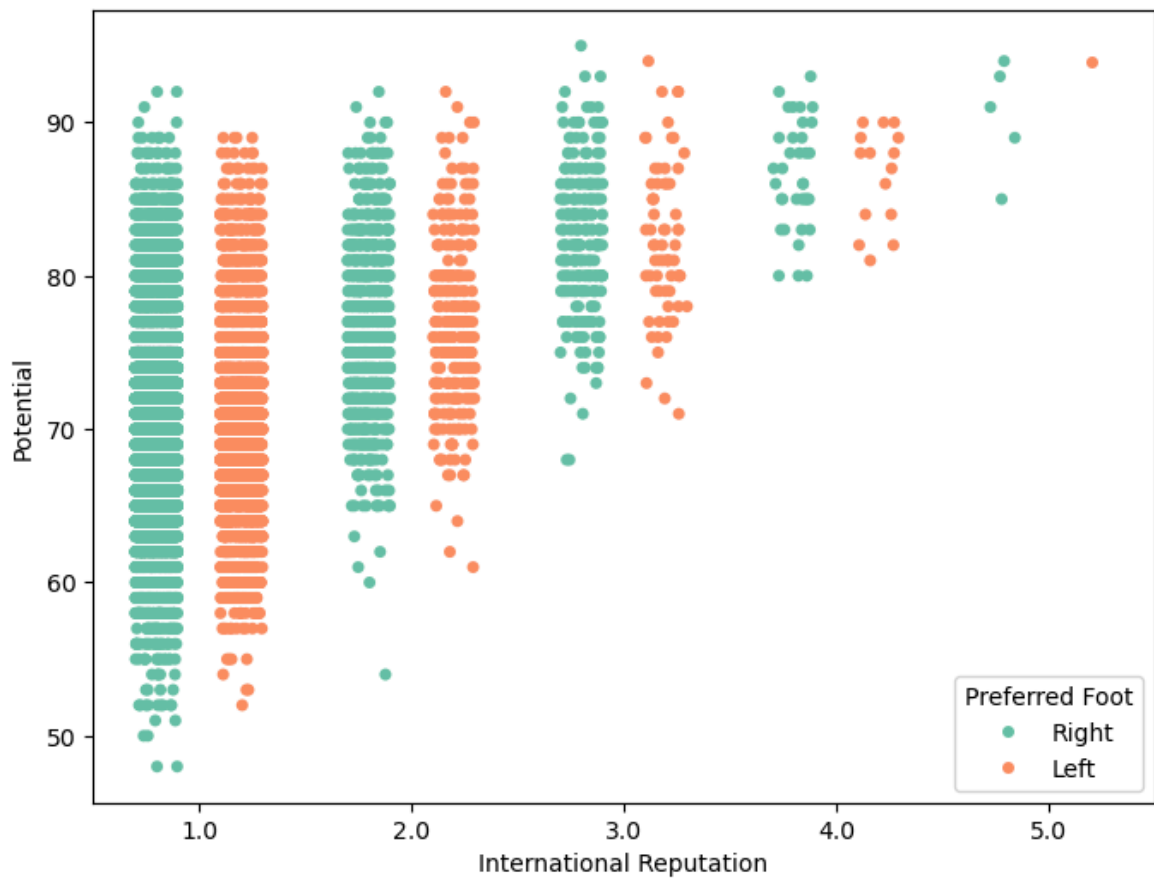

```
In [27]: f,ax=plt.subplots(figsize=(8,6))
sns.stripplot(x='International Reputation',y='Potential',data=fifa19,palette='br
plt.show()
```



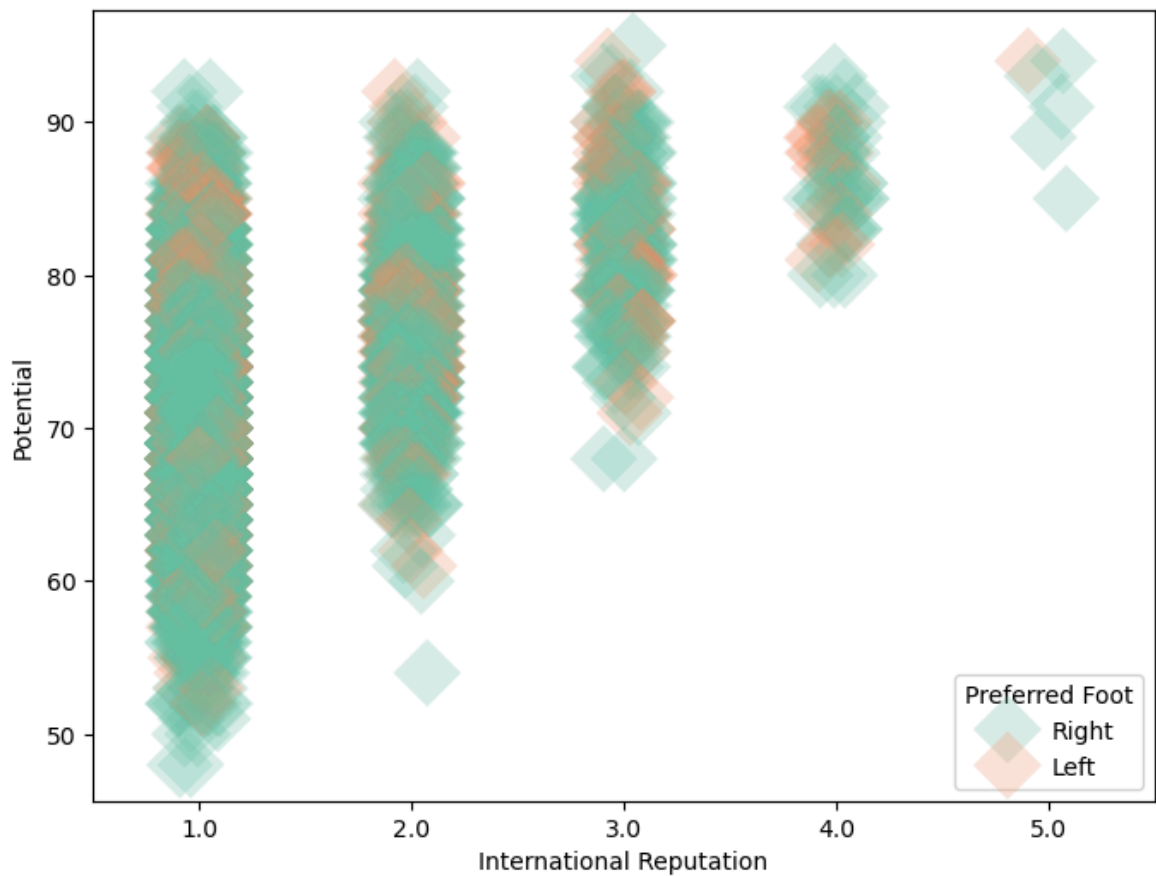
```
In [28]: f, ax = plt.subplots(figsize=(8, 6))
sns.stripplot(x="International Reputation", y="Potential", data=fifa19, jitter=0
plt.show()
```



```
In [29]: f, ax = plt.subplots(figsize=(8, 6))
sns.stripplot(x="International Reputation", y="Potential", hue="Preferred Foot",
              data=fifa19, jitter=0.2, palette="Set2", dodge=True)
plt.show()
```

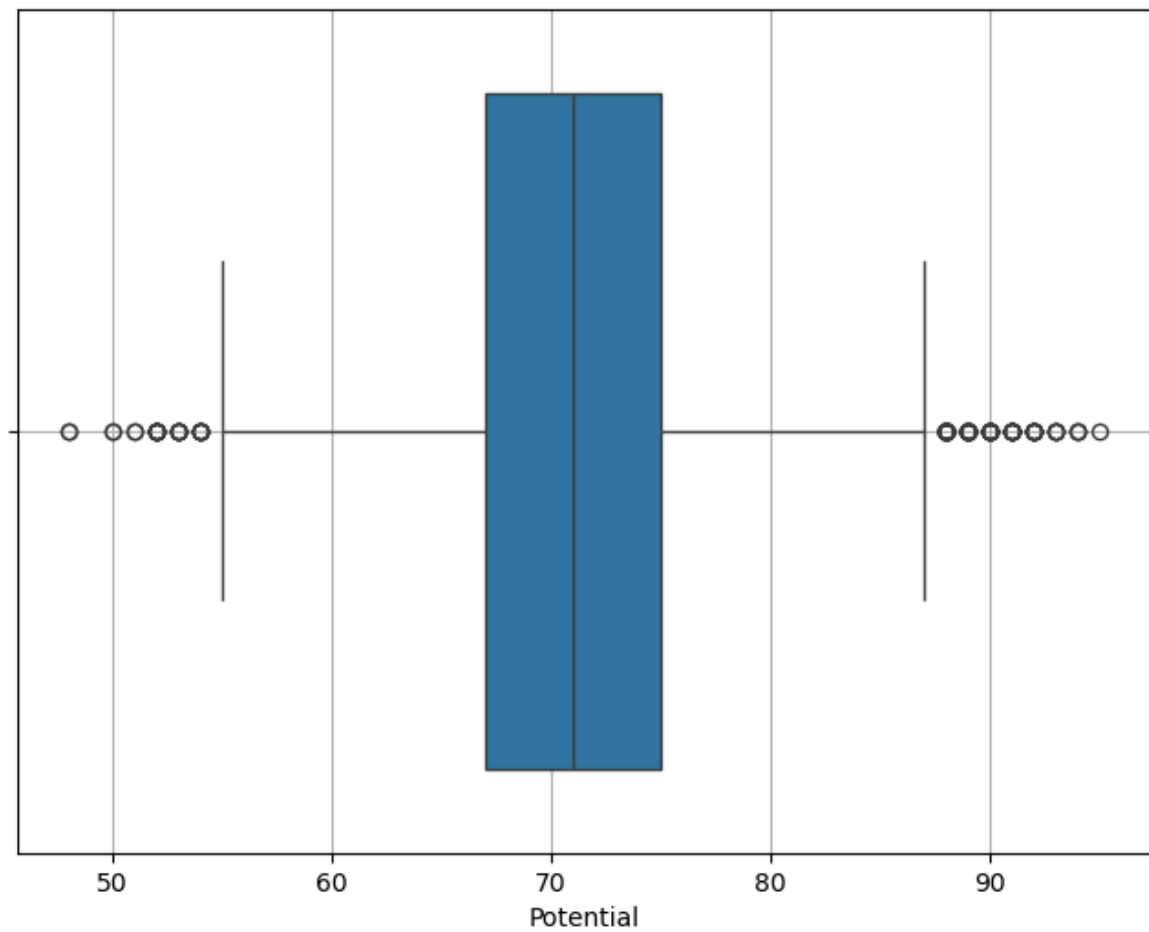


```
In [30]: #we can draw diff strips with large points and diff aesthetics as follows
f, ax = plt.subplots(figsize=(8, 6))
sns.stripplot(x="International Reputation", y="Potential", hue="Preferred Foot",
              data=fifa19, palette="Set2", size=20, marker="D",
              edgecolor="gray", alpha=.25)
plt.show()
```

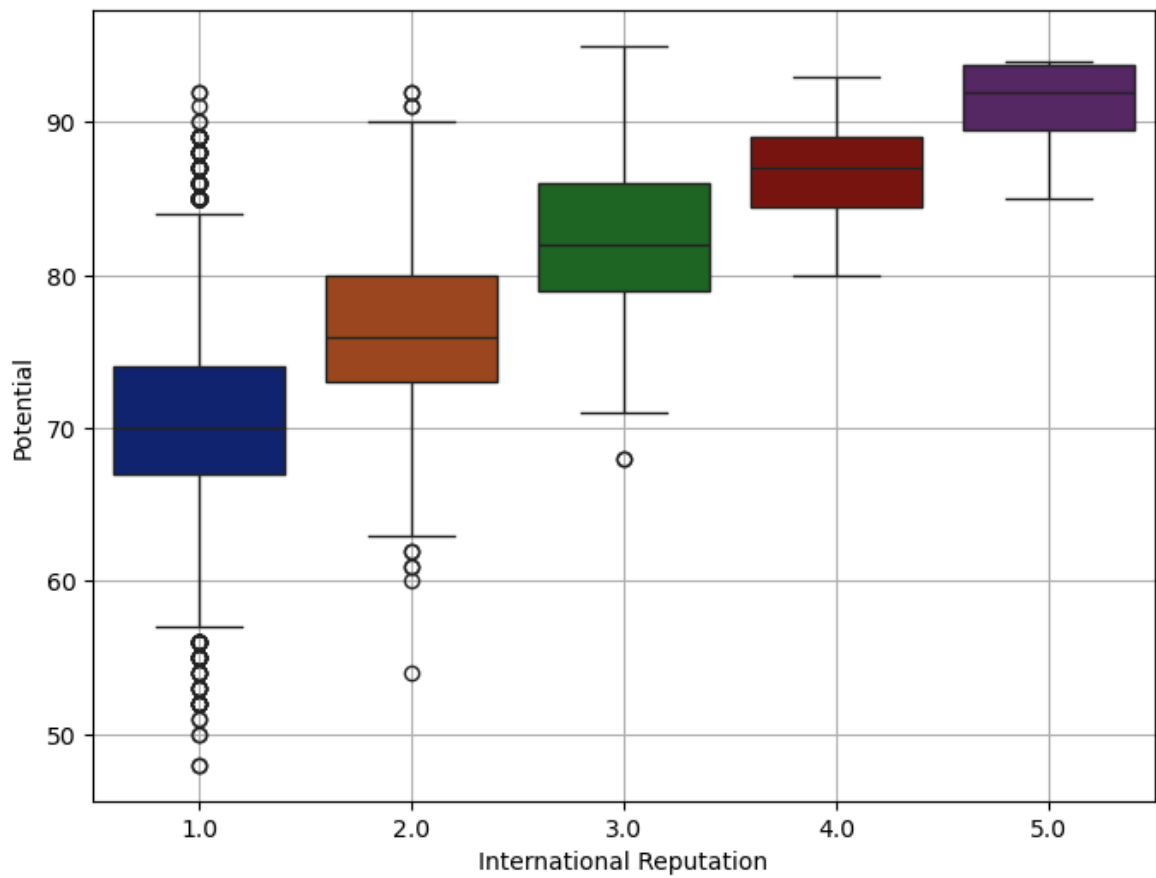


boxplot function

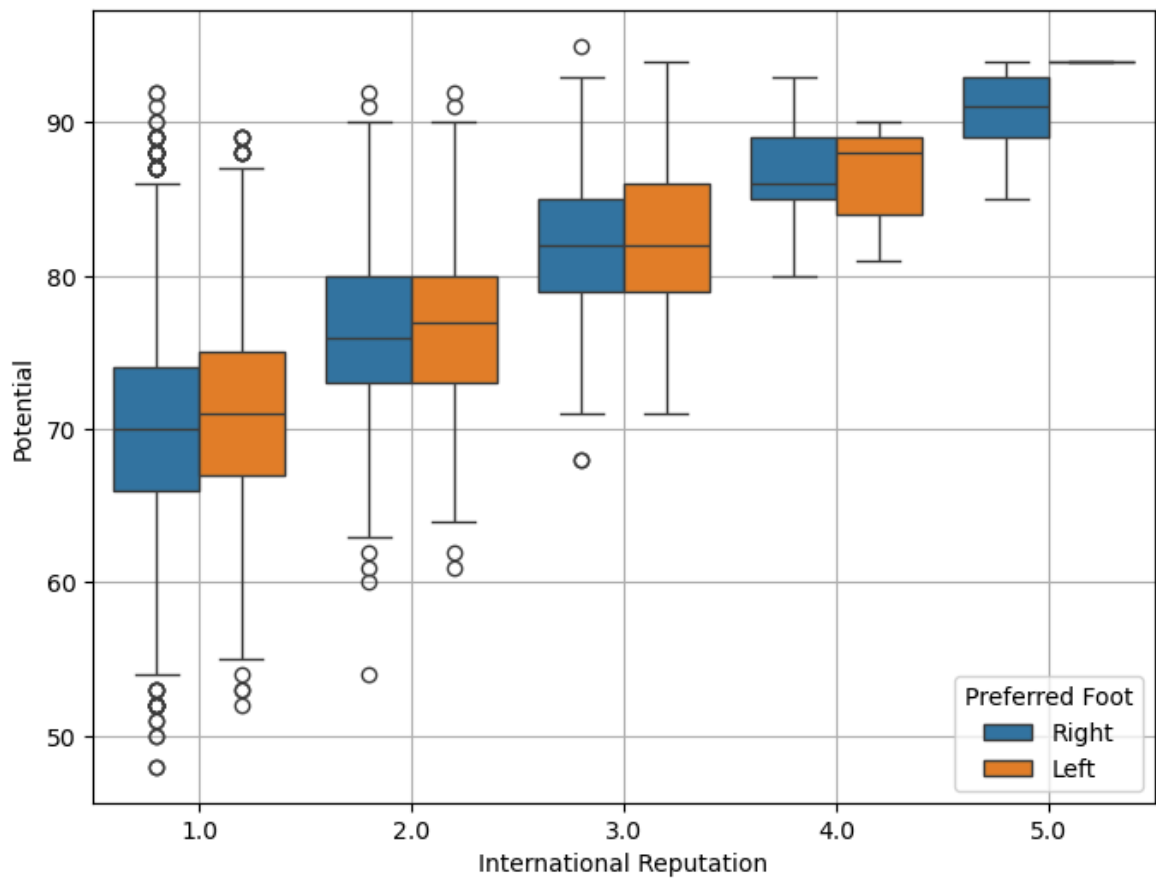
```
In [31]: f,ax=plt.subplots(figsize=(8,6))
sns.boxplot(x=fifa19['Potential'])
plt.grid()
plt.show()
```



```
In [32]: f,ax=plt.subplots(figsize=(8,6))
sns.boxplot(data=fifa19,x='International Reputation',y='Potential',palette='dark')
plt.grid()
plt.show()
```

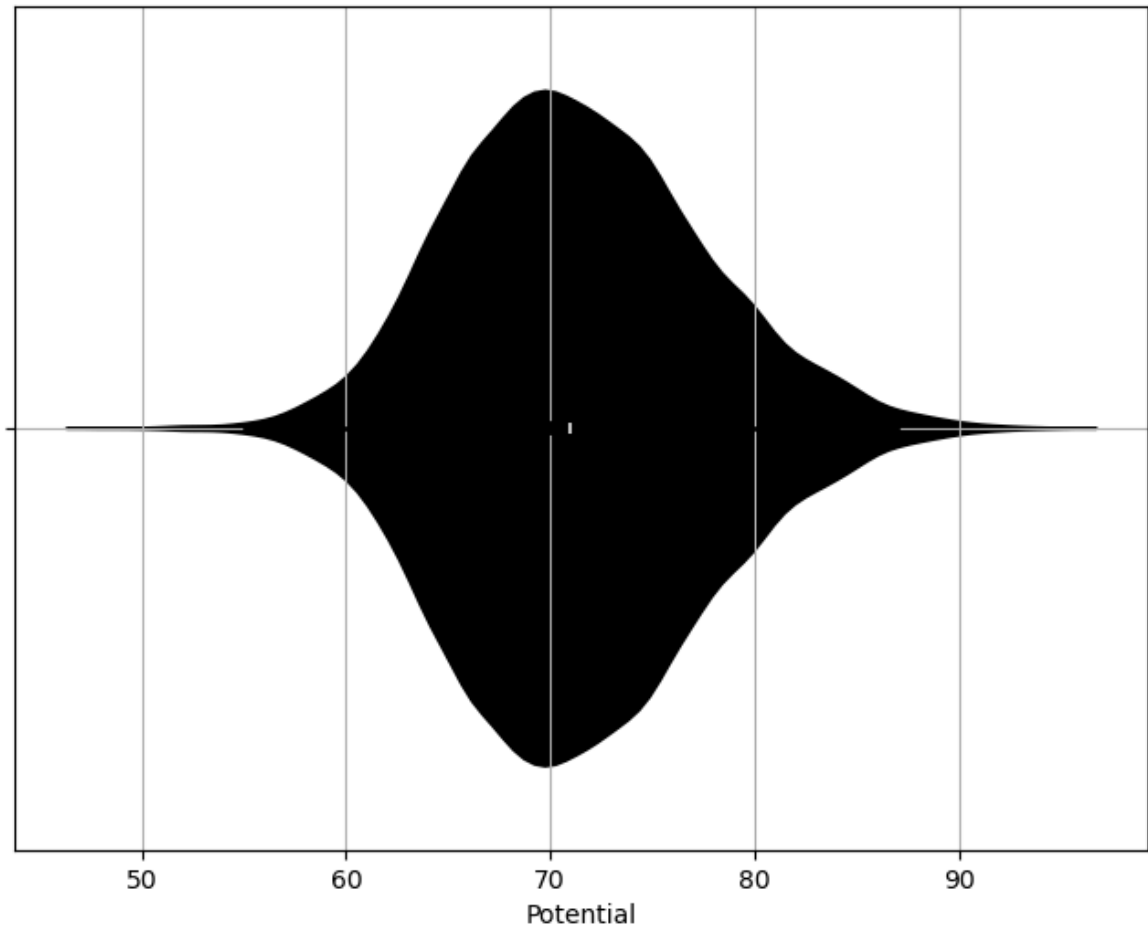


```
In [33]: f,ax=plt.subplots(figsize=(8,6))
sns.boxplot(data=fifa19,x='International Reputation',y='Potential',hue='Preferred Foot')
plt.grid()
plt.show()
```

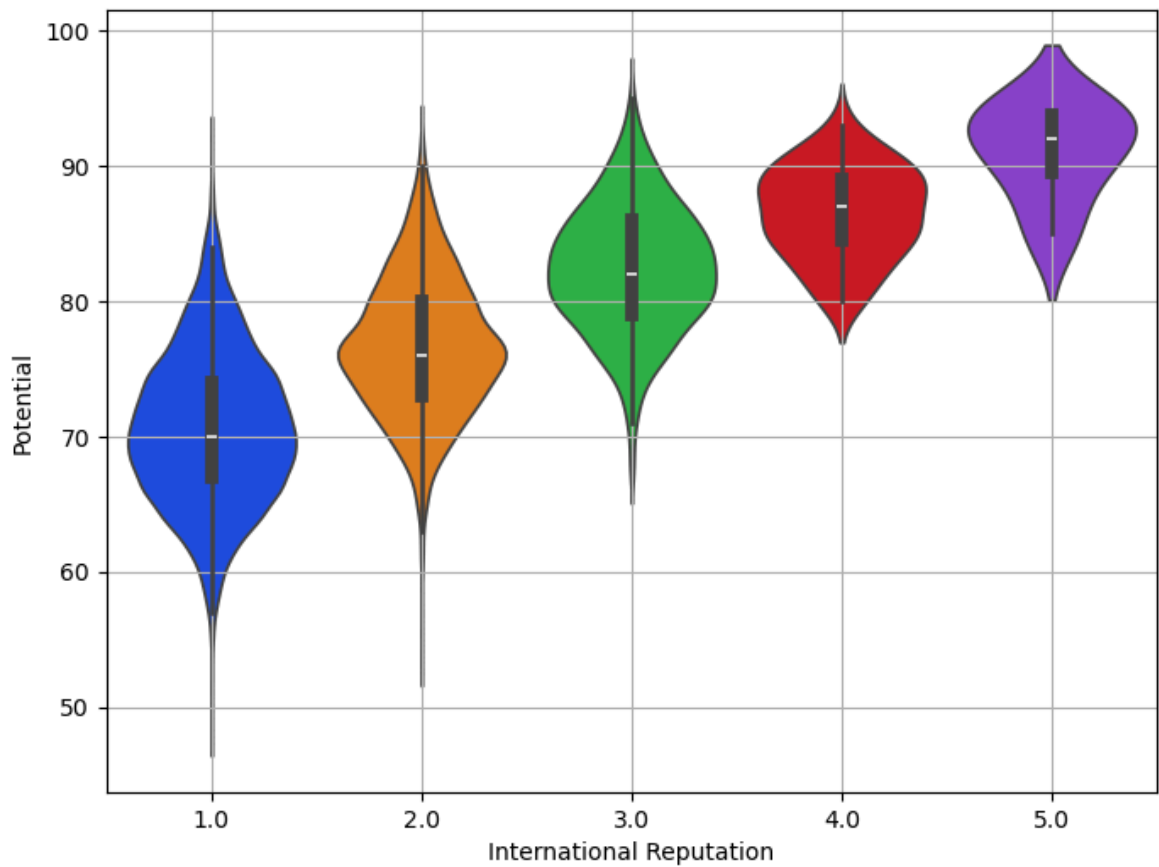


using violinplot() function

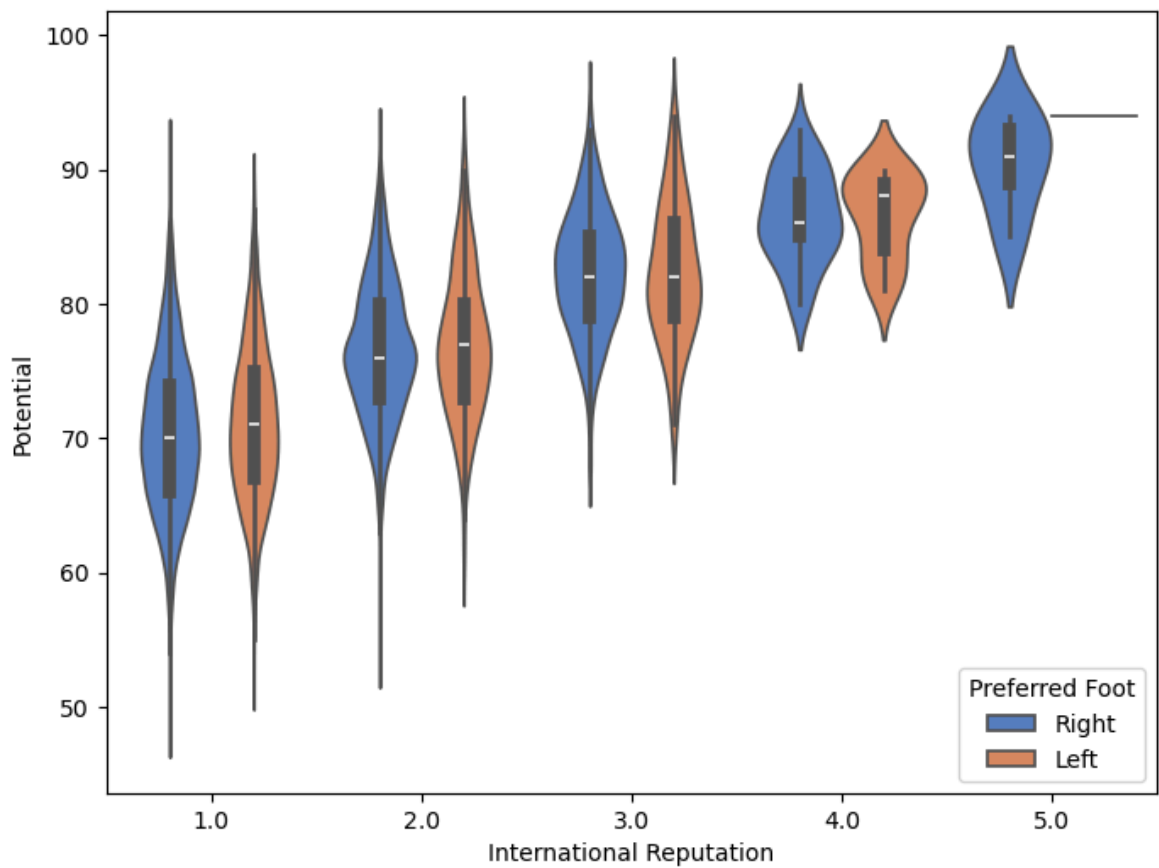
```
In [34]: f,ax=plt.subplots(figsize=(8,6))
sns.violinplot(x=fifa19['Potential'],color='black')
plt.grid()
plt.show()
```



```
In [35]: f,ax=plt.subplots(figsize=(8,6))
sns.violinplot(data=fifa19,x='International Reputation',y='Potential',palette='b')
plt.grid()
plt.show()
```

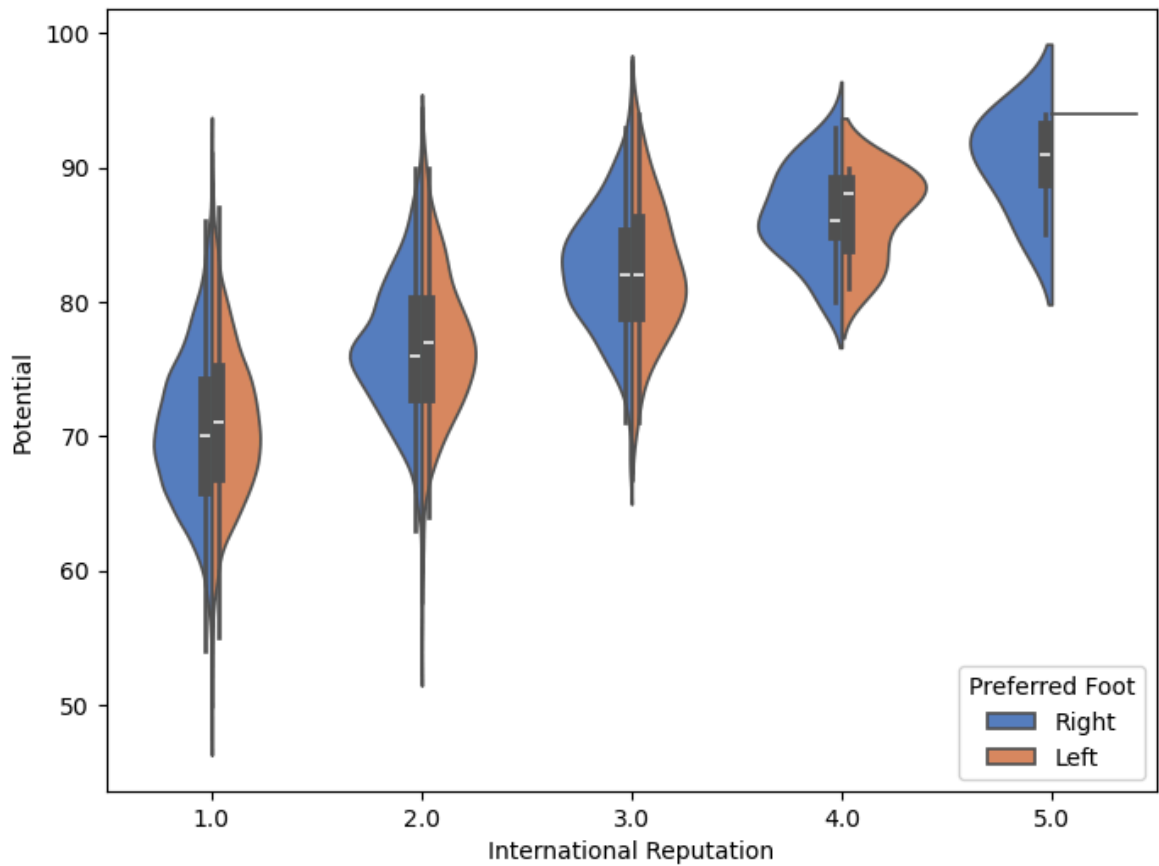


```
In [36]: f, ax = plt.subplots(figsize=(8, 6))
sns.violinplot(x="International Reputation", y="Potential", hue="Preferred Foot")
plt.show()
```



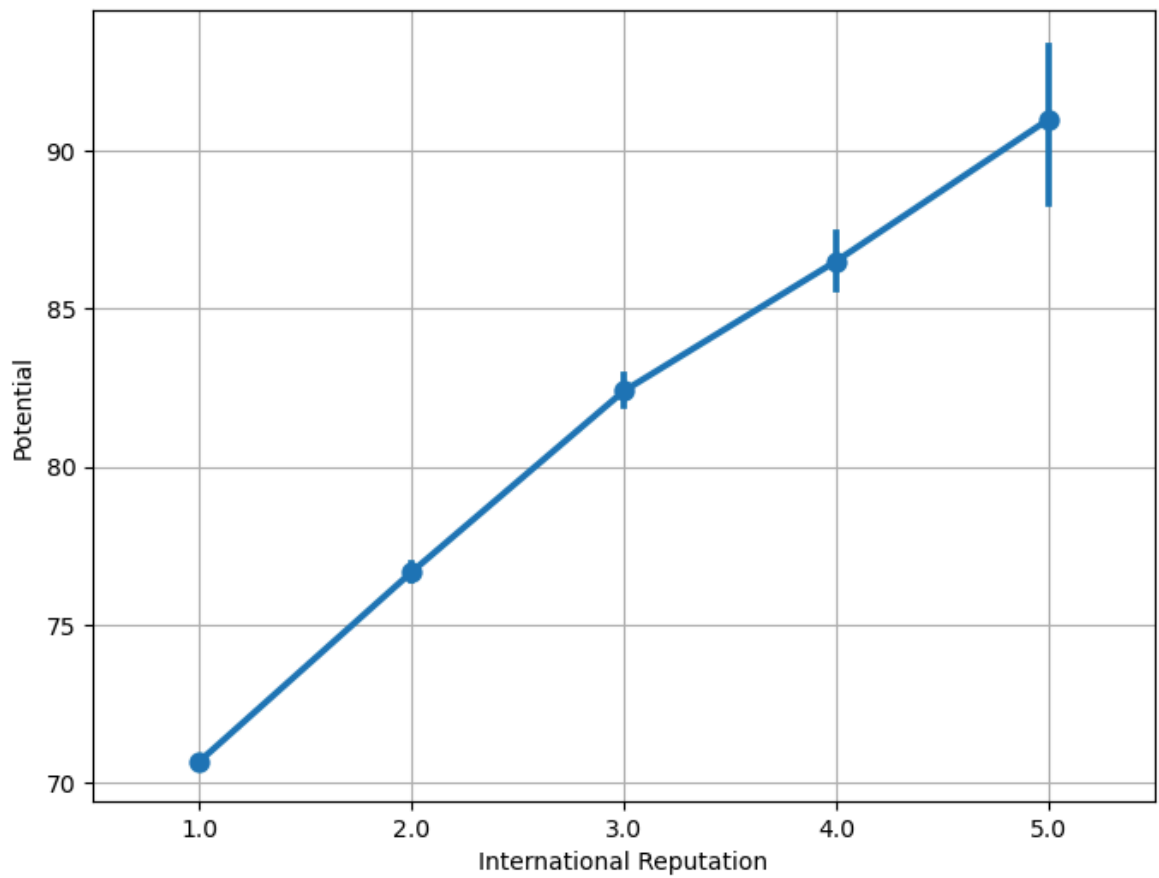
```
In [37]: #we can also draw split violine to compare the hue var as follows using split:
f, ax = plt.subplots(figsize=(8, 6))
```

```
sns.violinplot(x="International Reputation", y="Potential", hue="Preferred Foot"  
plt.show()
```

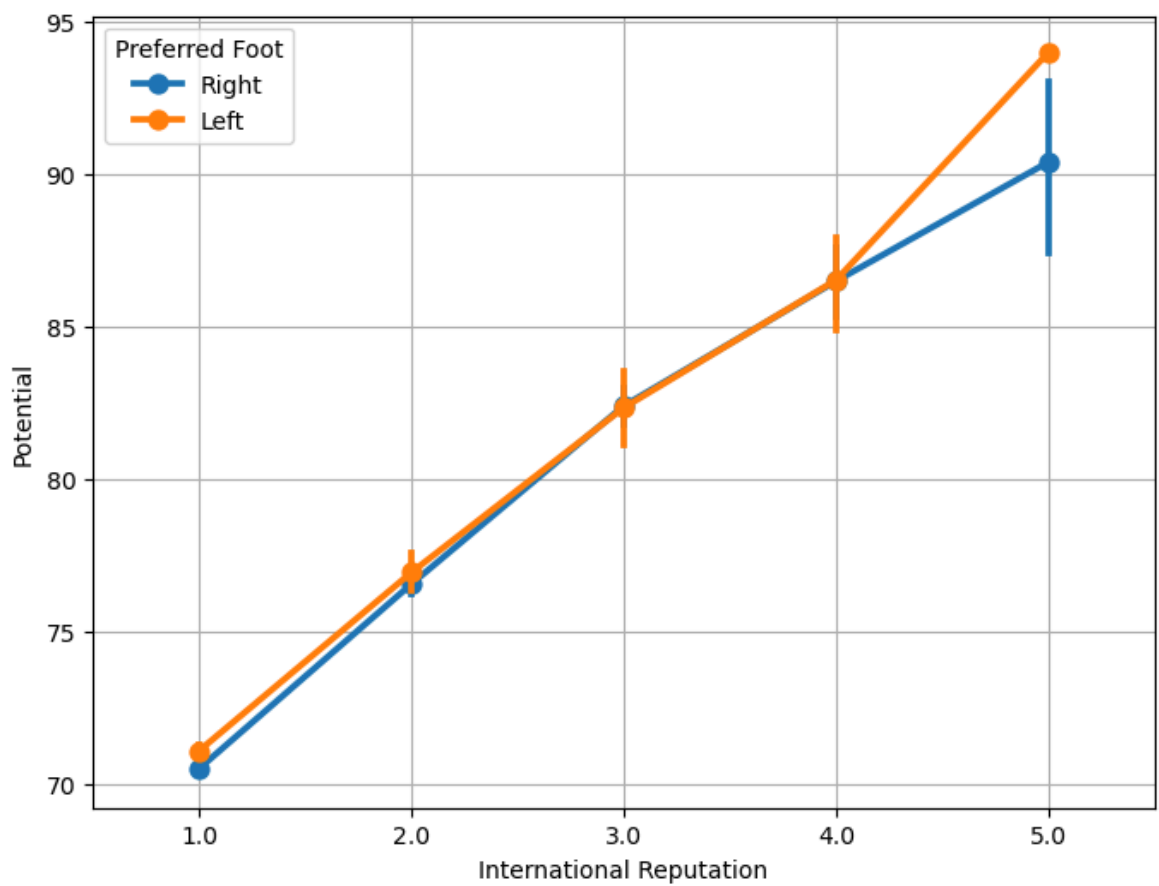


pointplot() function

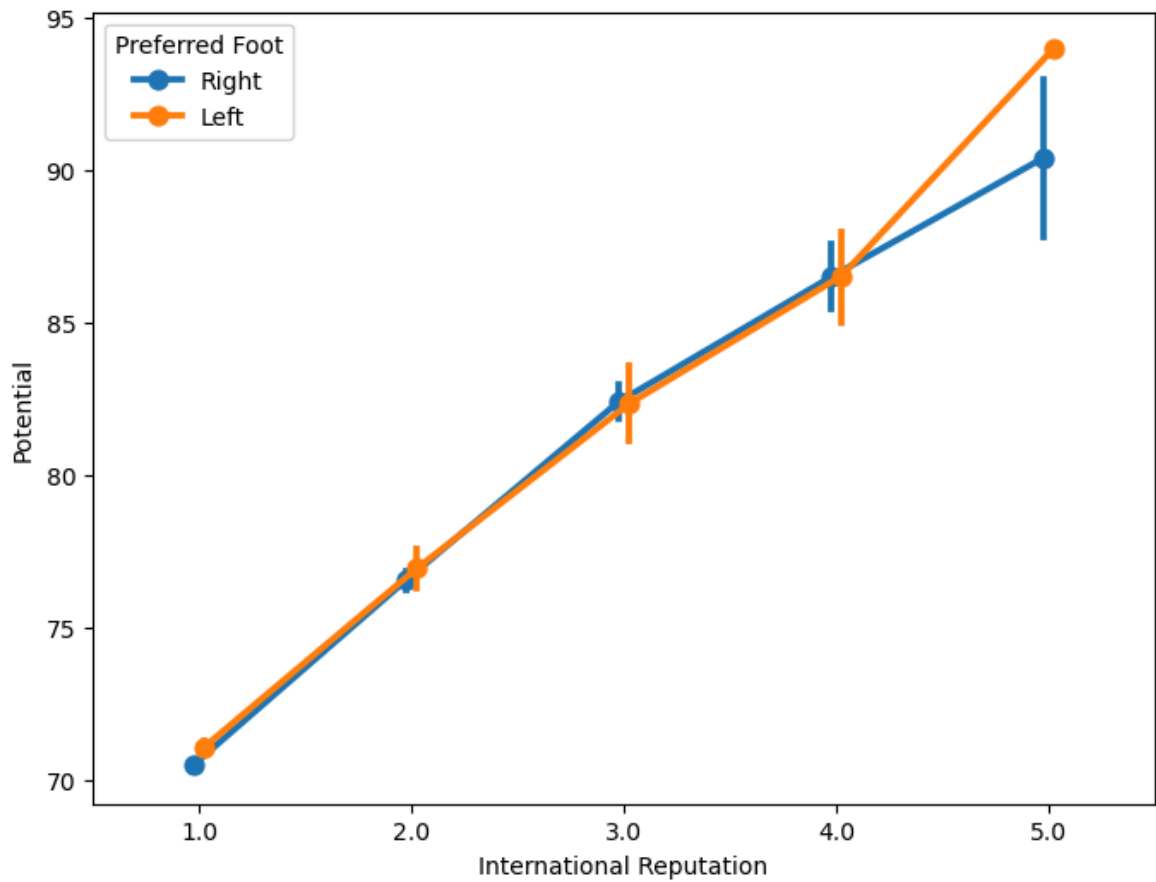
```
In [38]: f, ax = plt.subplots(figsize=(8, 6))  
sns.pointplot(x="International Reputation", y="Potential", data=fifa19)  
plt.grid()  
plt.show()
```

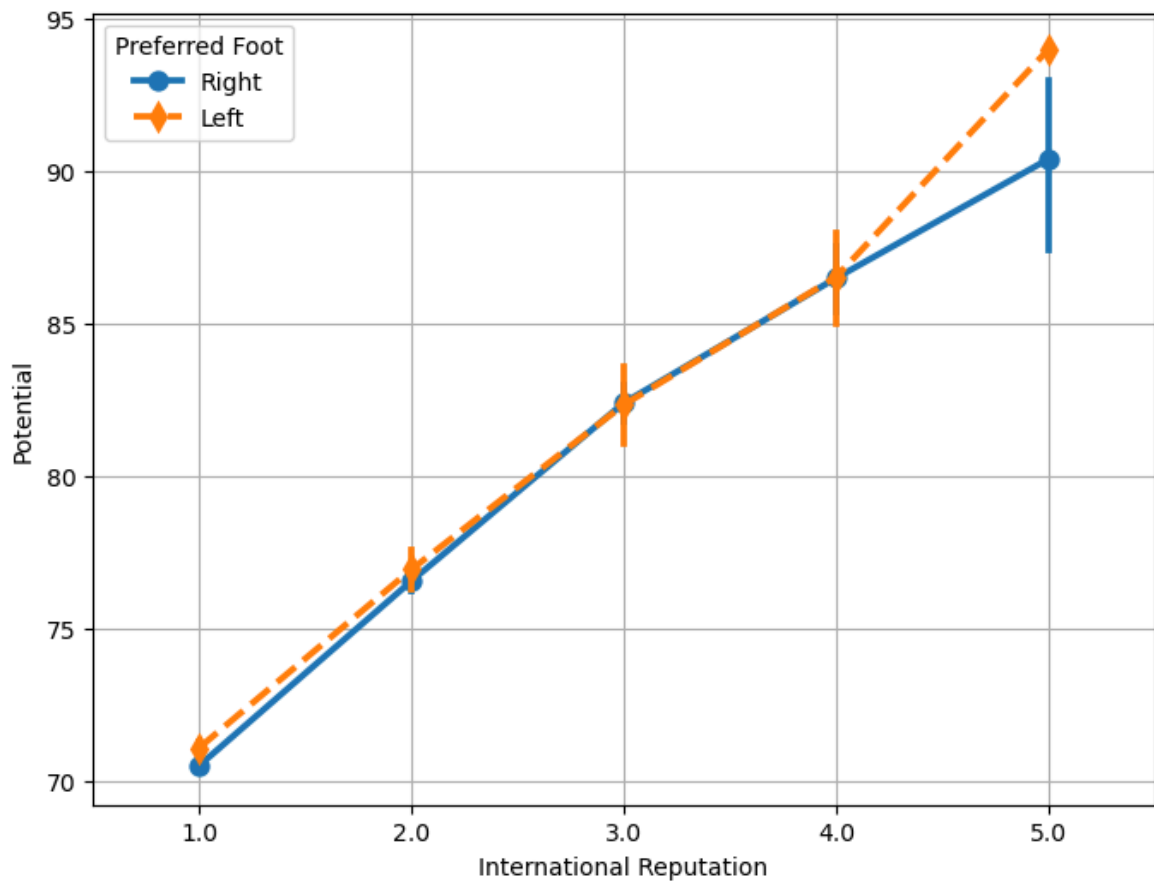
```
In [39]: f, ax = plt.subplots(figsize=(8, 6))
sns.pointplot(x="International Reputation", y="Potential", hue="Preferred Foot",
plt.grid()
plt.show()
```



```
In [40]: #we can separate the points of diff hue lvls along the catgorical axis as follow
f, ax = plt.subplots(figsize=(8, 6))
sns.pointplot(x="International Reputation", y="Potential", hue="Preferred Foot",
plt.show())
```

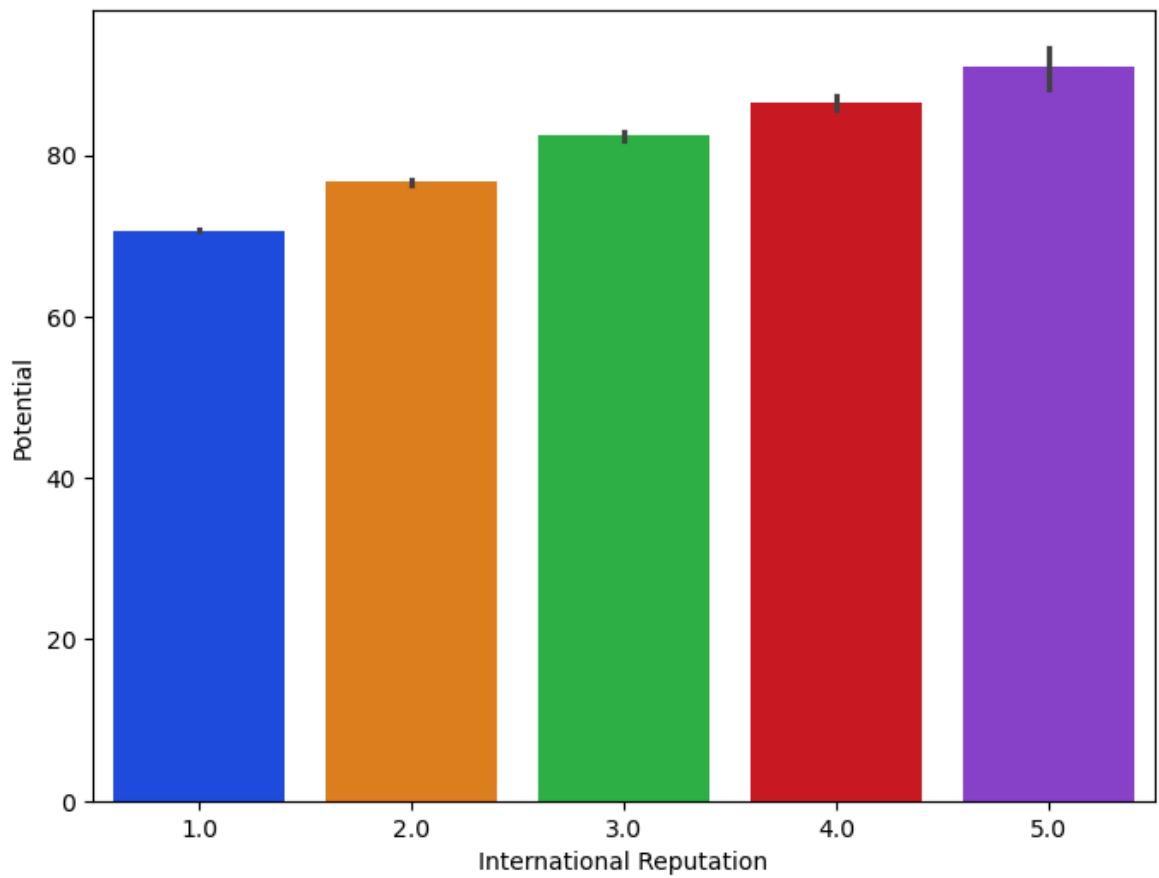


```
In [41]: f, ax = plt.subplots(figsize=(8, 6))
sns.pointplot(x="International Reputation", y="Potential", hue="Preferred Foot",
plt.grid()
plt.show())
```

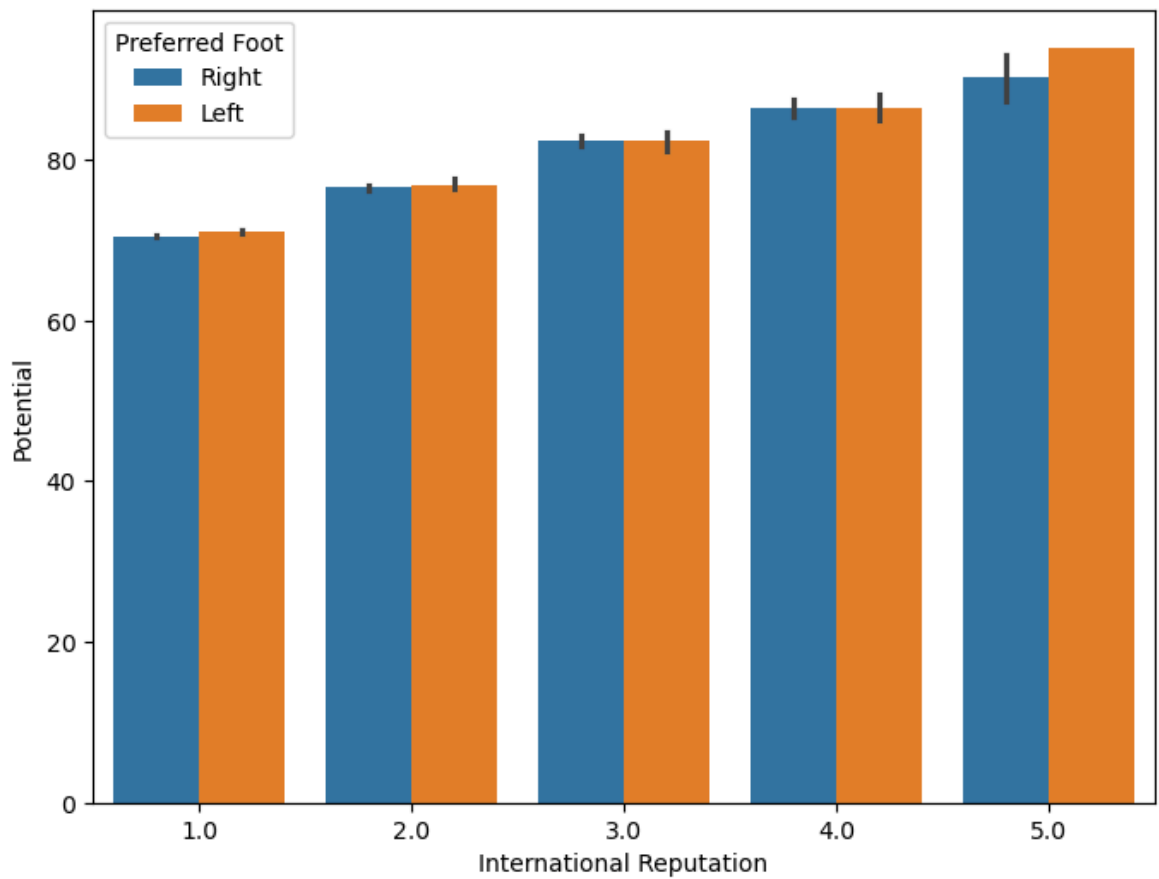


Barplot() function

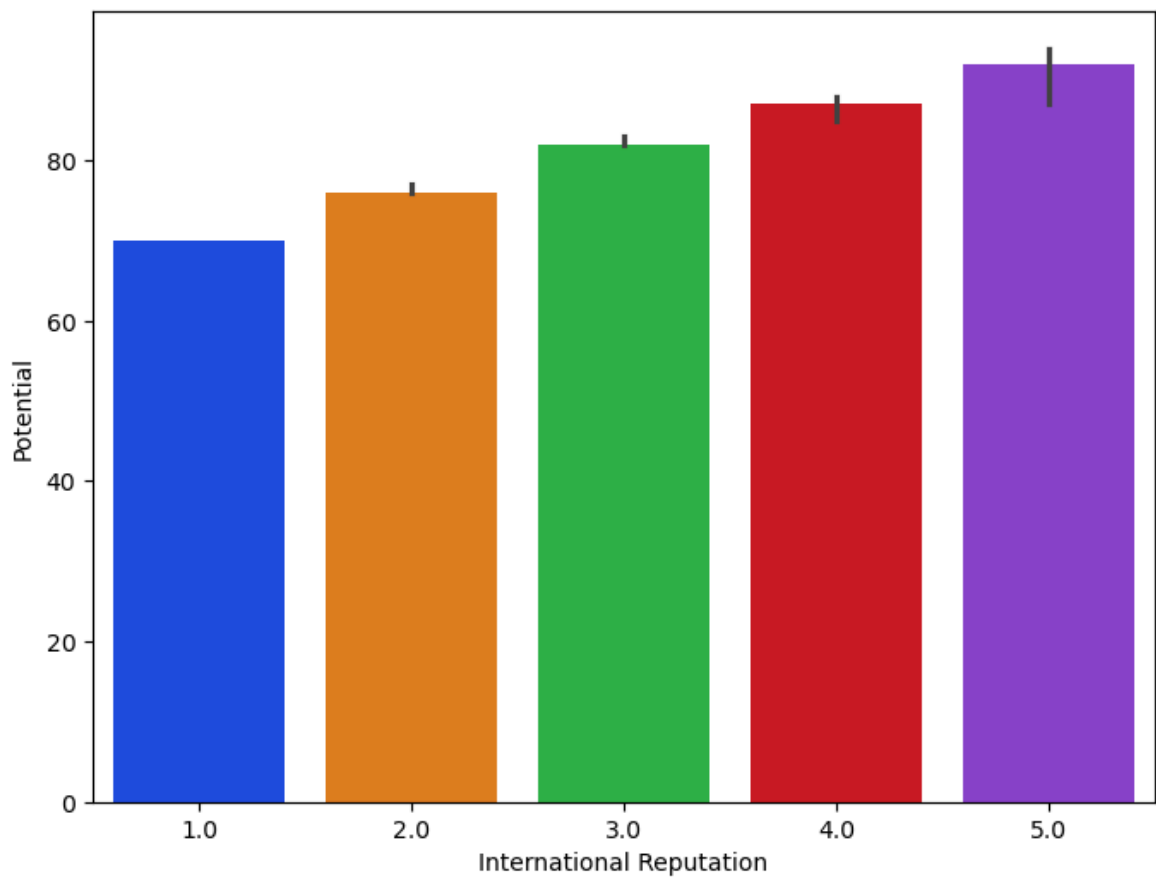
```
In [42]: f, ax = plt.subplots(figsize=(8, 6))  
sns.barplot(x="International Reputation", y="Potential", data=fifa19, palette='br')  
plt.show()
```



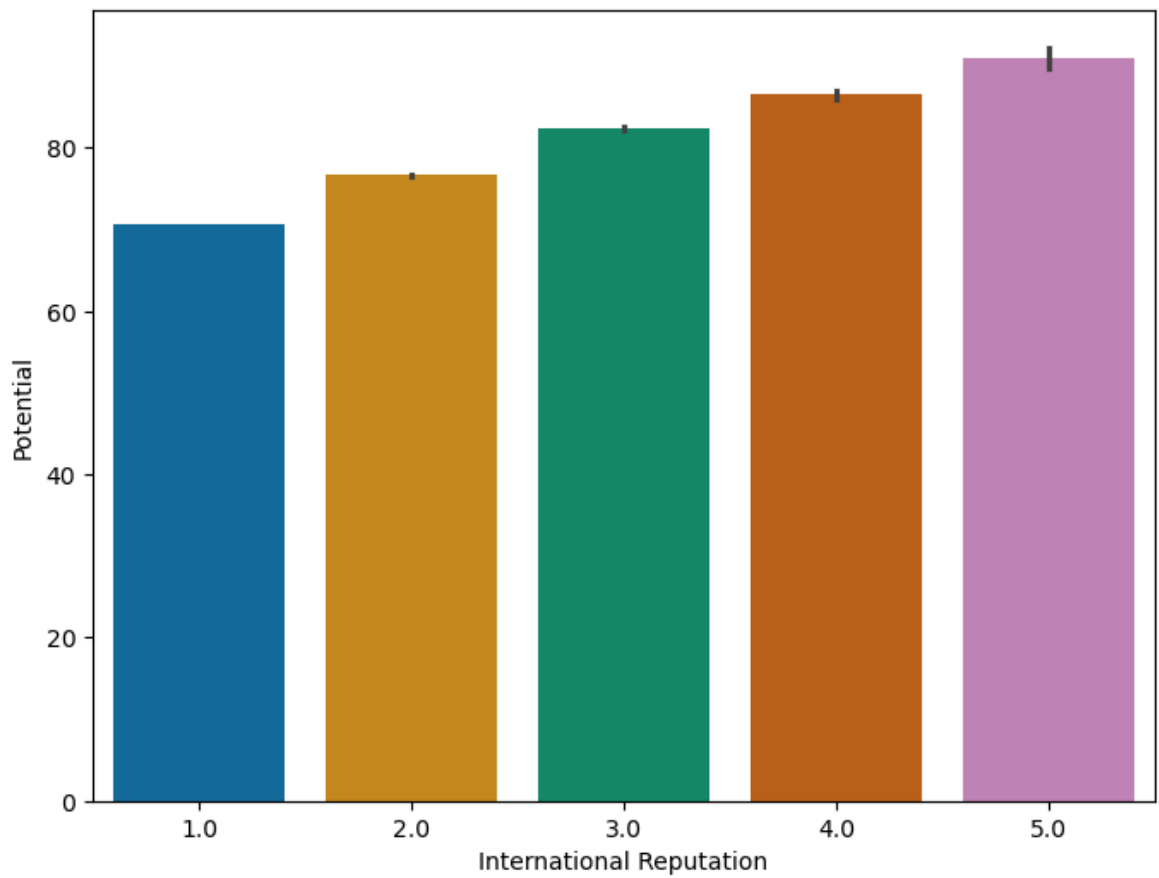
```
In [43]: f, ax = plt.subplots(figsize=(8, 6))
sns.barplot(x="International Reputation", y="Potential", hue='Preferred Foot', d
plt.show()
```



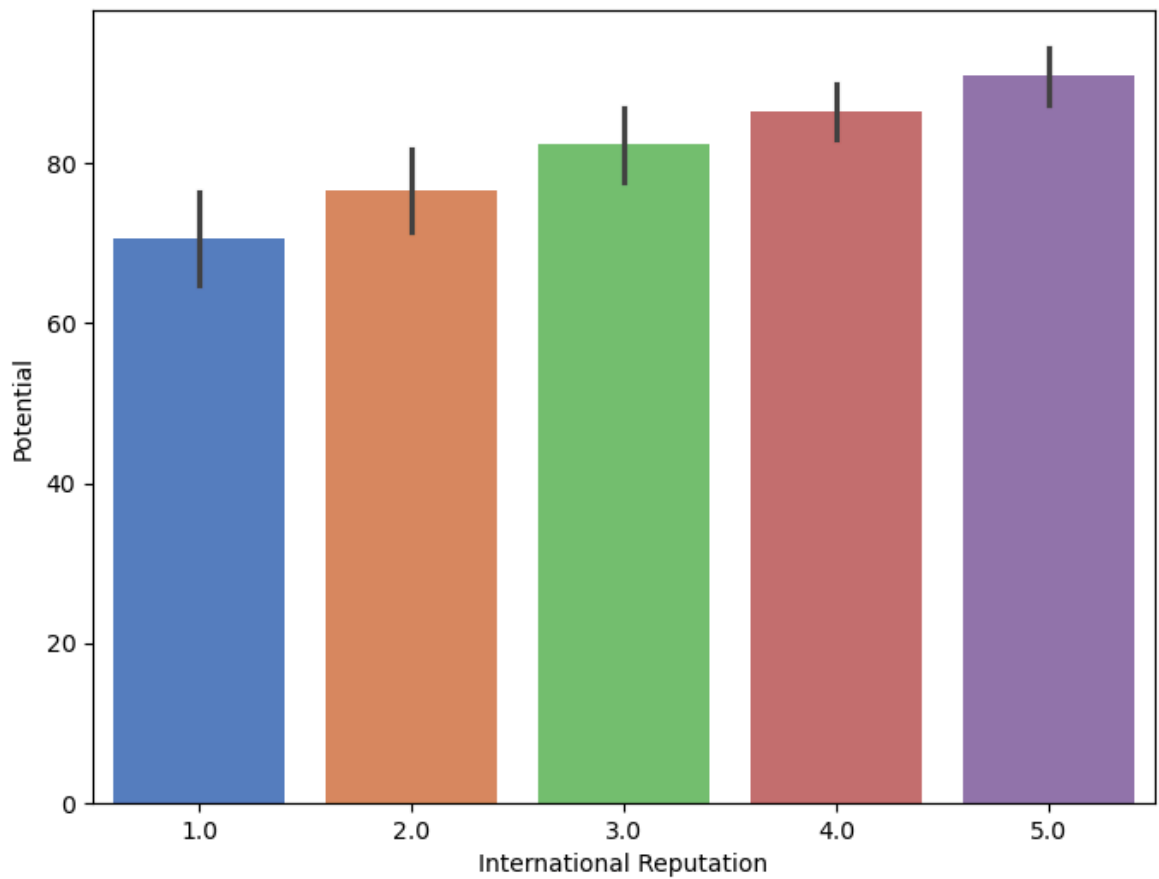
```
In [44]: f, ax = plt.subplots(figsize=(8, 6))
sns.barplot(x="International Reputation", y="Potential", data=fifa19,palette='br
plt.show()
```



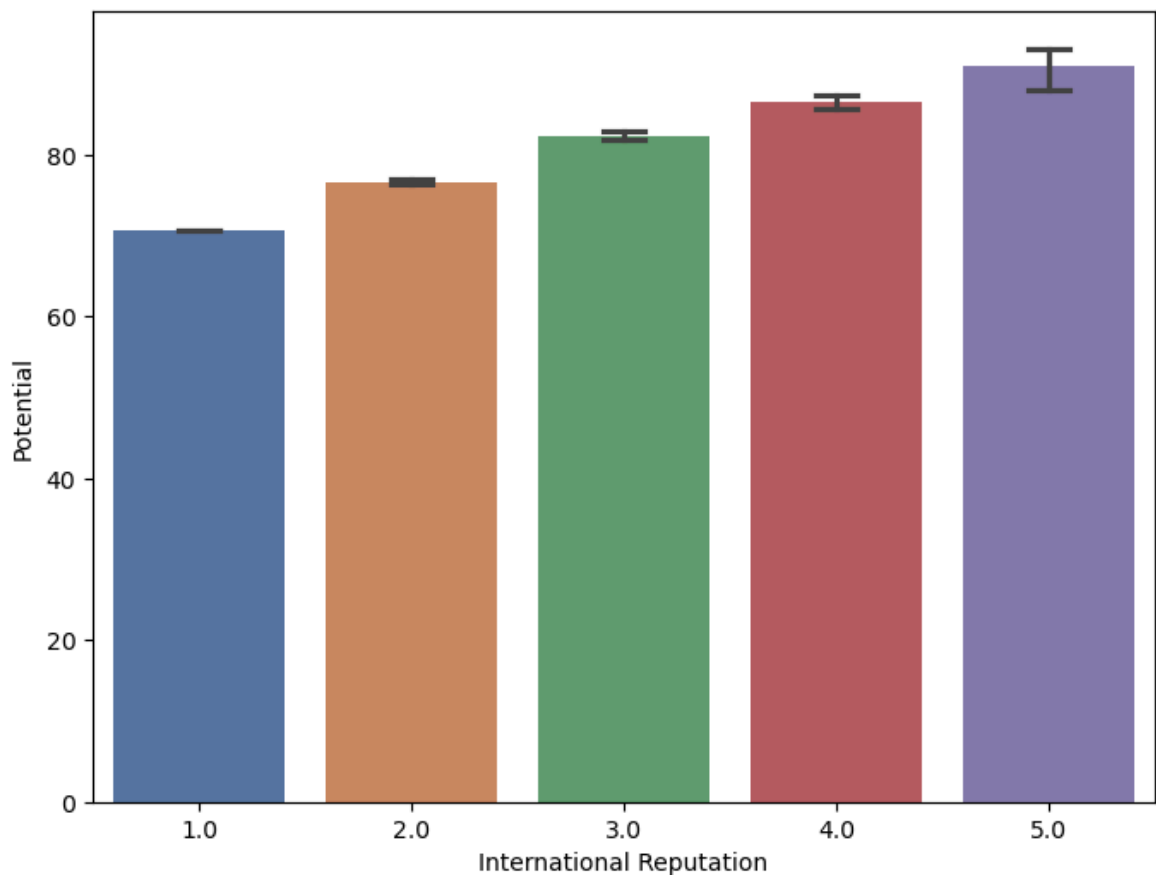
```
In [45]: f, ax = plt.subplots(figsize=(8, 6))
sns.barplot(x="International Reputation", y="Potential", data=fifa19,ci=68,palet
plt.show()
```



```
In [46]: f, ax = plt.subplots(figsize=(8, 6))
sns.barplot(x="International Reputation", y="Potential", data=fifa19, ci="sd", pa
plt.show()
```



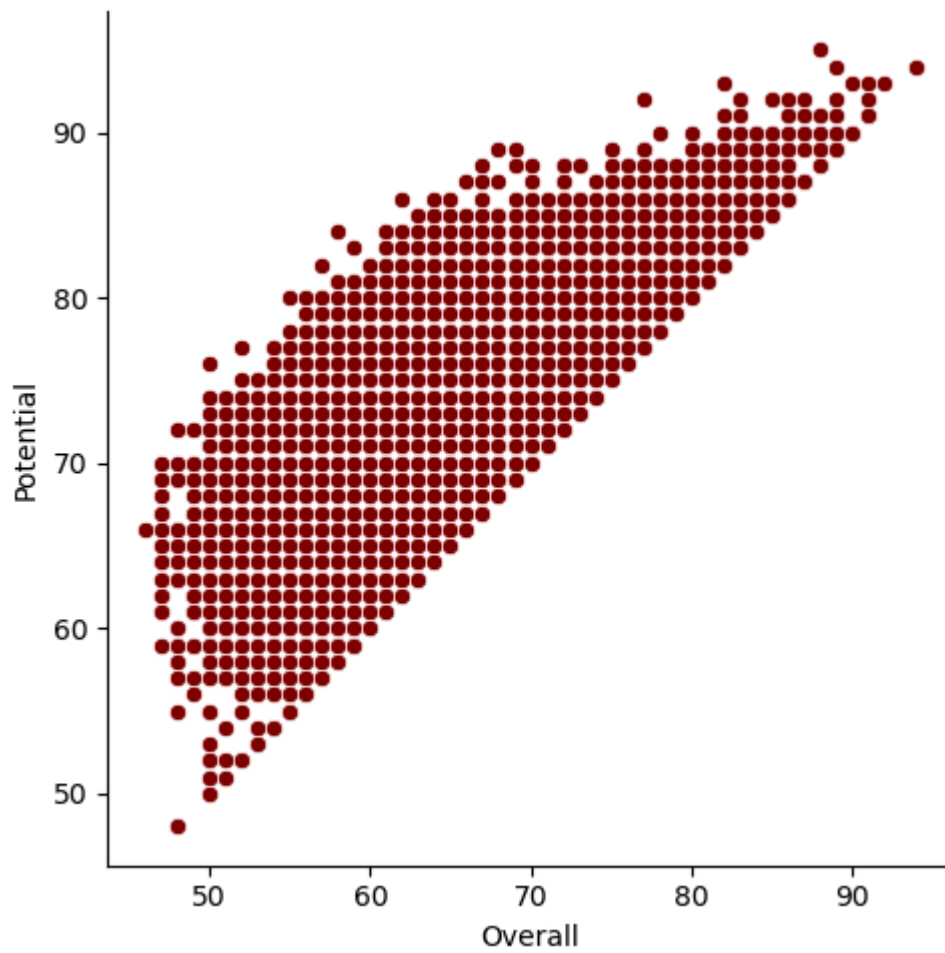
```
In [47]: f, ax = plt.subplots(figsize=(8, 6))
sns.barplot(x="International Reputation", y="Potential", data=fifa19, palette='d
plt.show()
```



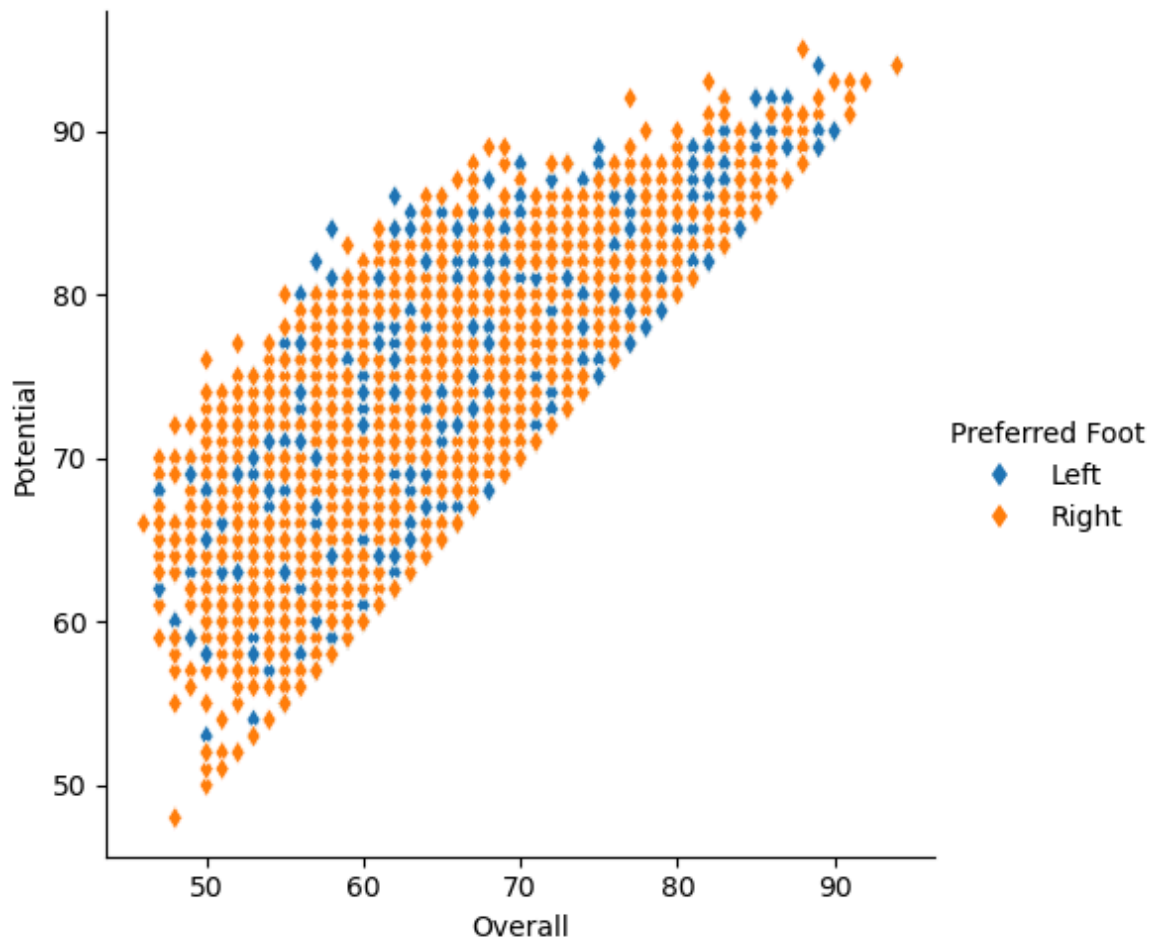
relplot() function

Seaborn 'relplot()' function - Seaborn 'relplot()' function helps us to draw figure-level interface for drawing relational plots onto a FacetGrid. - This function provides access to several different axes-level functions that show the relationship between two variables with semantic mappings of subsets. - The 'kind' parameter selects the underlying axes-level function to use - scatterplot() (with kind="scatter"; the default) - lineplot() (with kind="line") ---we can plot a scatterplot with height, weight variables using relplot() as follows---

```
In [48]: g = sns.relplot(x='Overall', y='Potential', data=fifa19, color='Maroon')
plt.show()
```

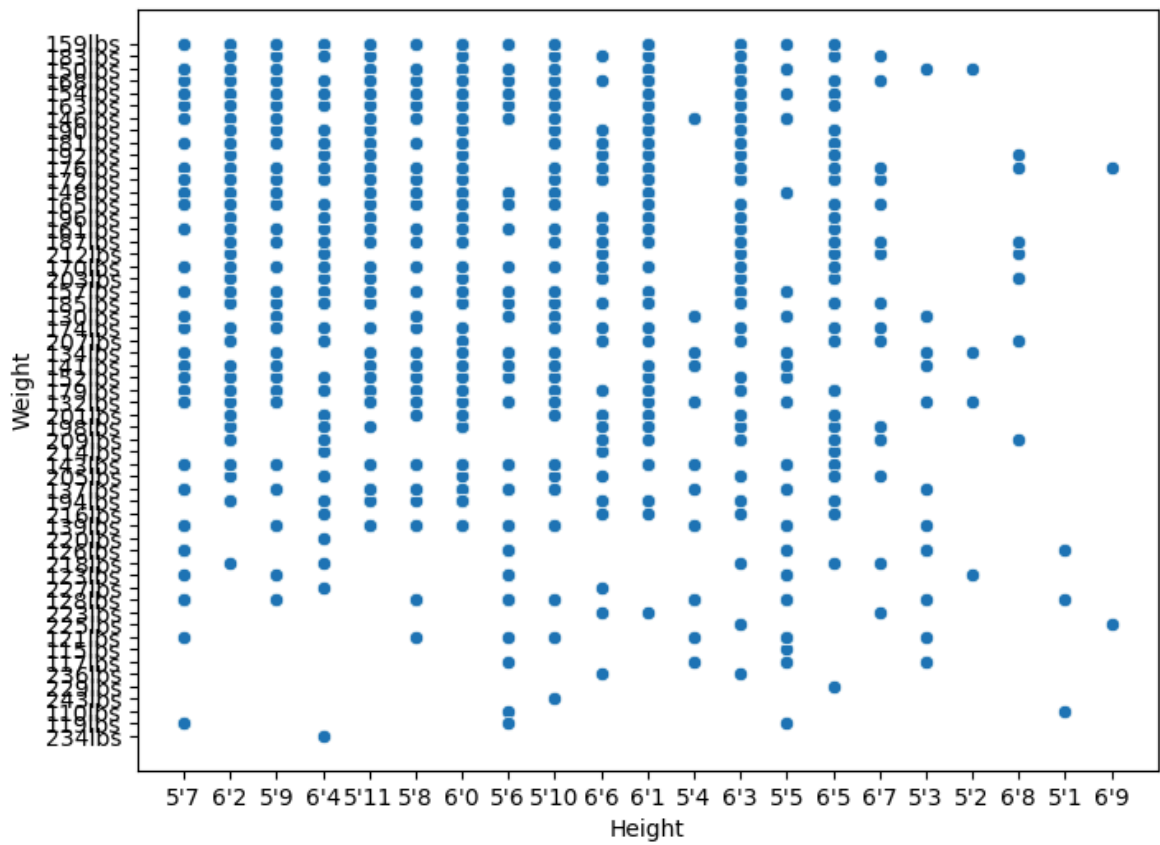


```
In [49]: g= sns.relplot(x='Overall',y='Potential',hue='Preferred Foot',marker='d',data=fi
plt.show()
```

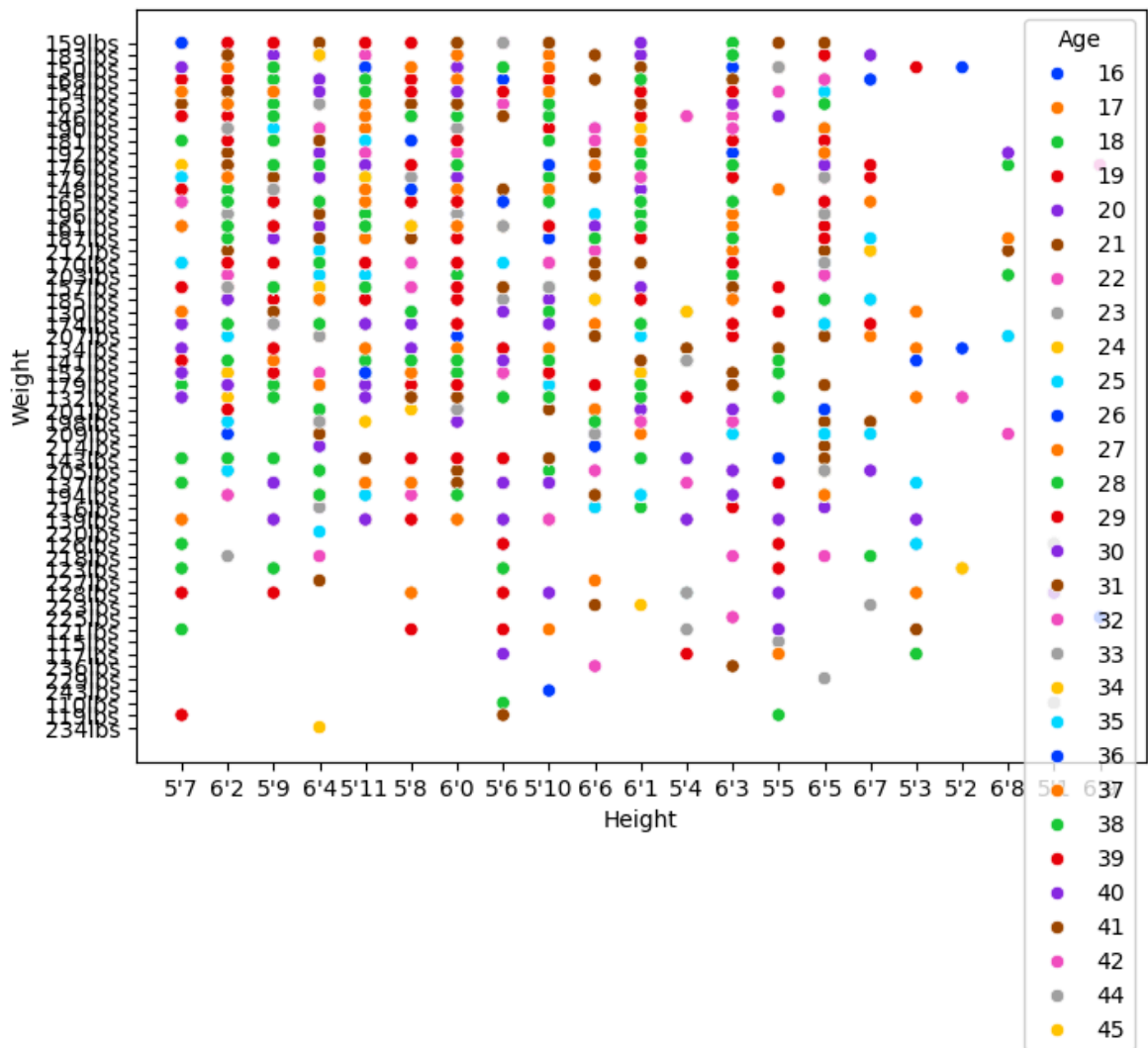



scatter plot

```
In [50]: f, ax = plt.subplots(figsize=(8, 6))
sns.scatterplot(x="Height", y="Weight", data=fifa19)
plt.show()
```

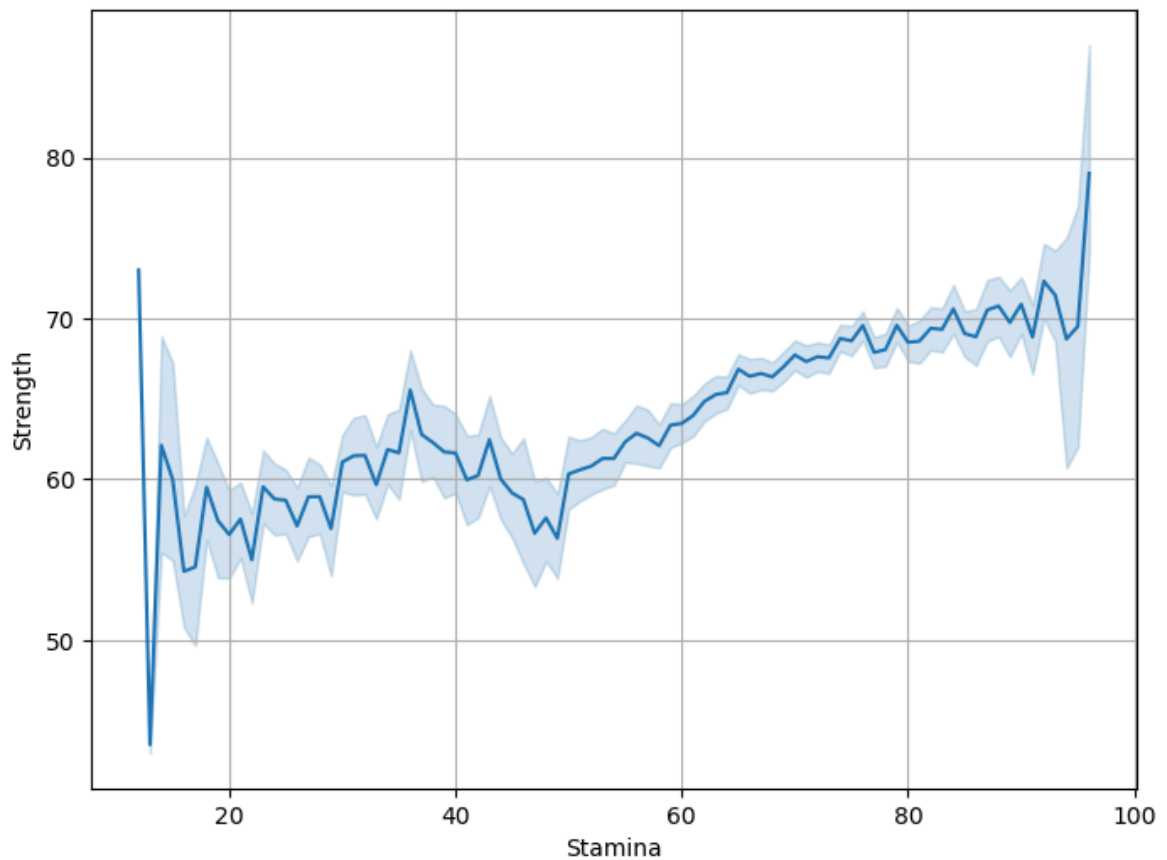


```
In [51]: f, ax = plt.subplots(figsize=(8, 6))
sns.scatterplot(x="Height", y="Weight", hue='Age', data=fifa19,palette='bright')
plt.show()
```



lineplot

```
In [52]: f,ax=plt.subplots(figsize=(8,6))
sns.lineplot(x='Stamina',y='Strength',data=fifa19)
plt.grid()
plt.show()
```

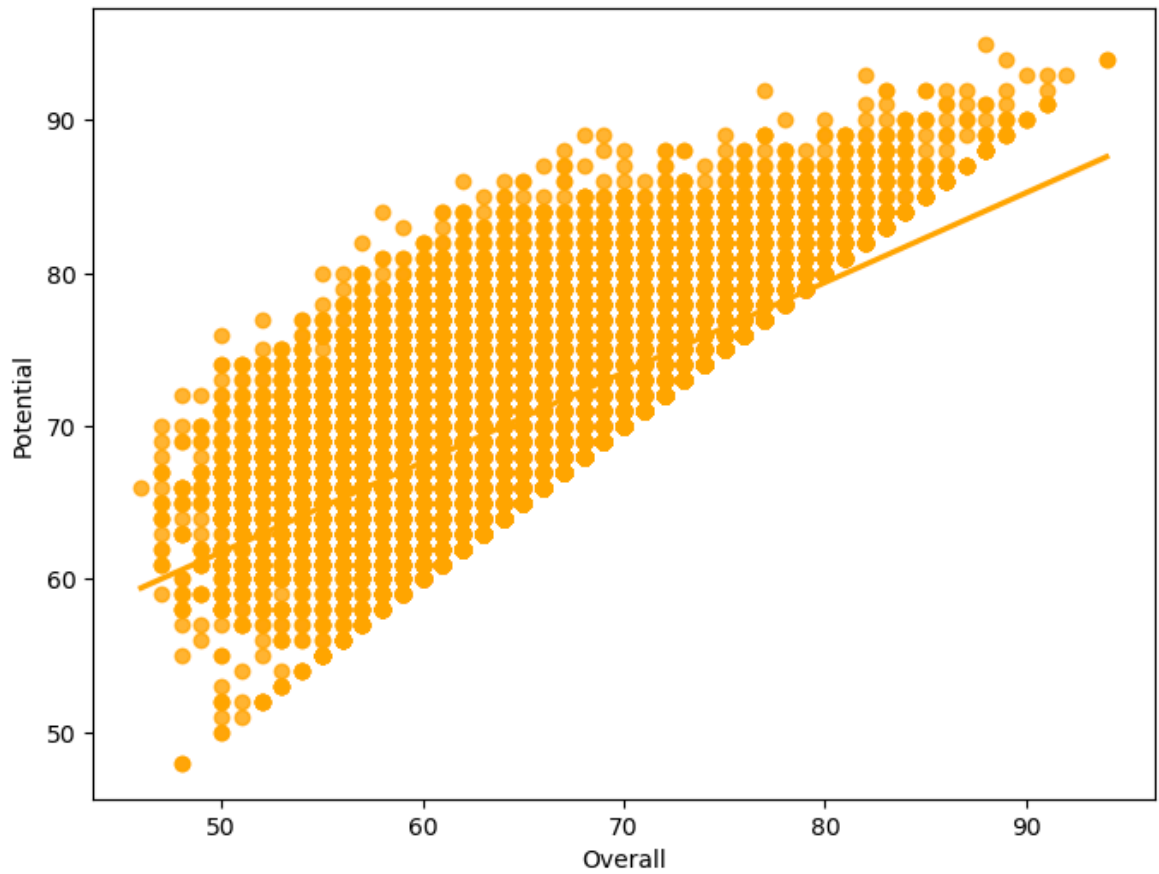


```
In [53]: #f,ax=plt.subplots(figsize=(8,6))
#sns.Lineplot(x='Stamina',y='Strength',hue='Age',data=fifa19)
#plt.grid()
#plt.show()
```

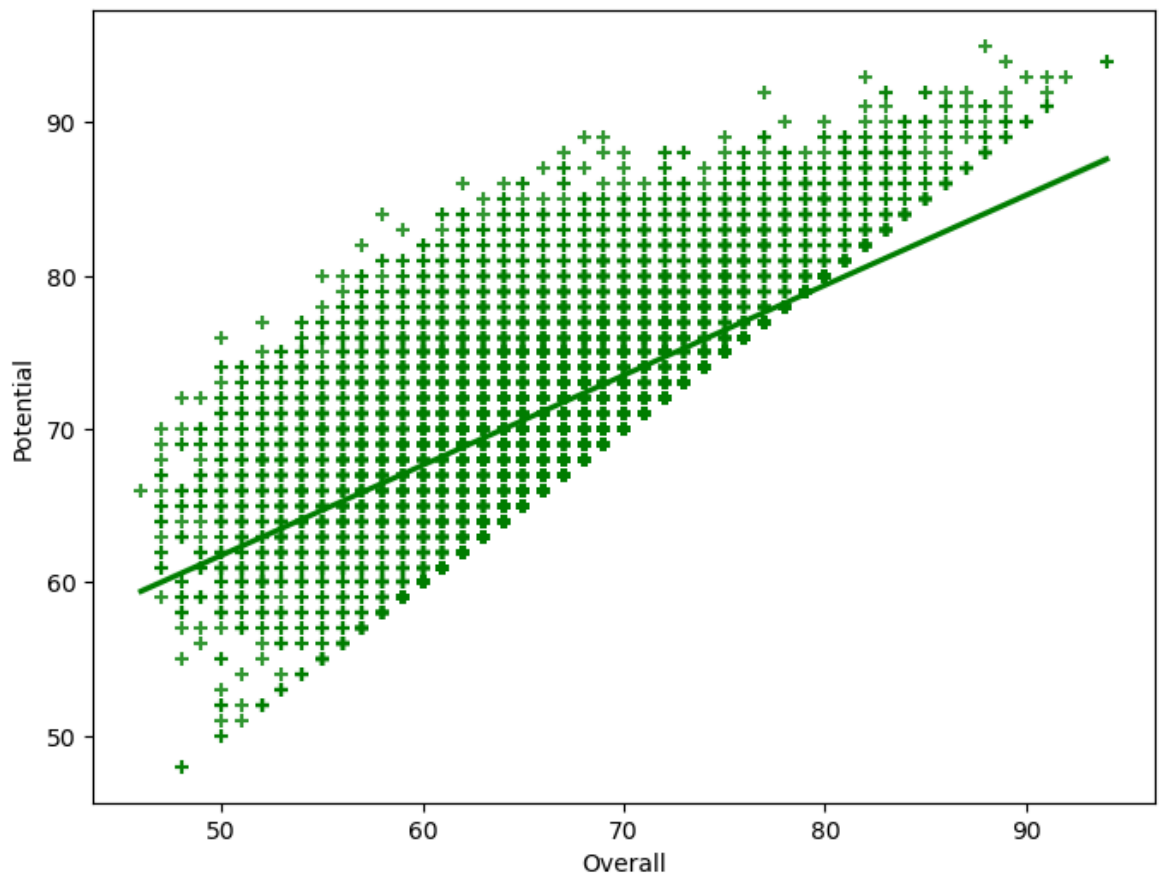
Regression plot

we can plot or model between Overall and Potential var with regplot() as follows-

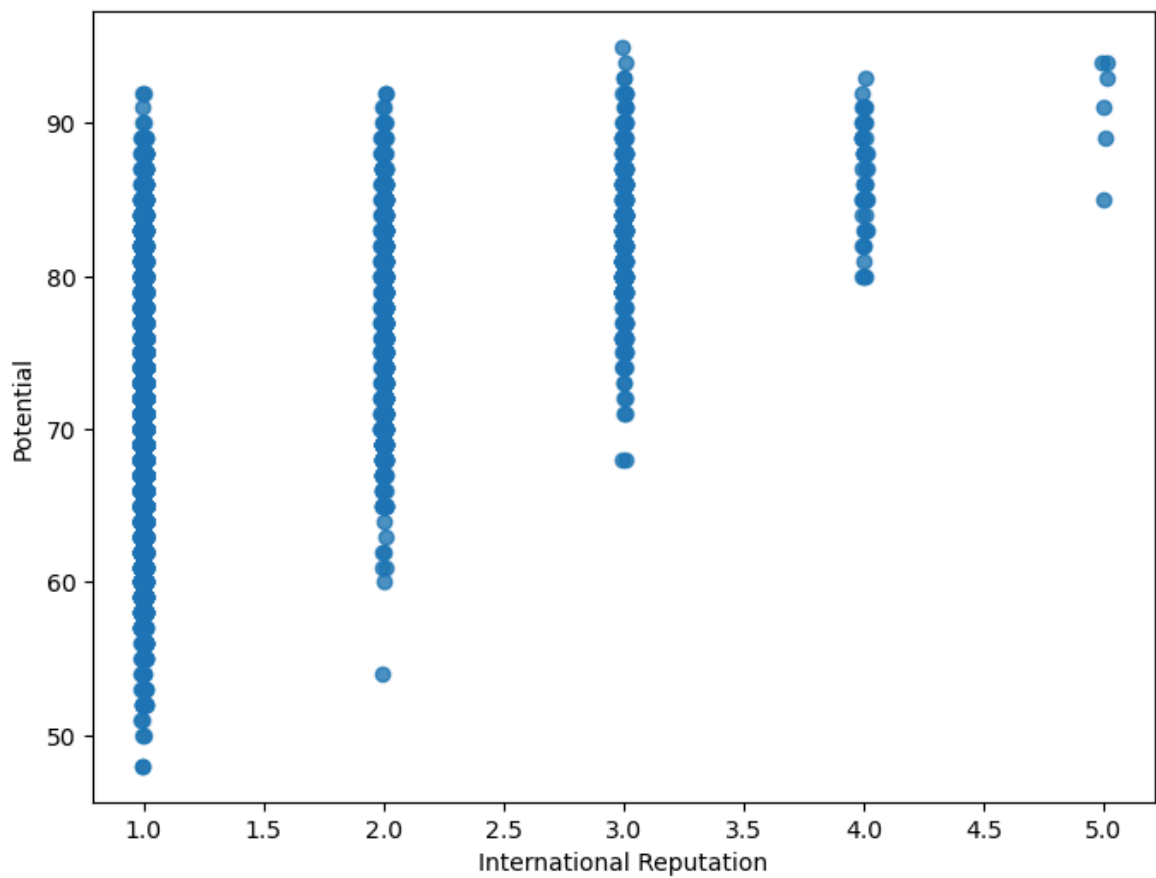
```
In [54]: f, ax = plt.subplots(figsize=(8, 6))
ax = sns.regplot(x="Overall", y="Potential", data=fifa19,color='orange')
plt.show()
```



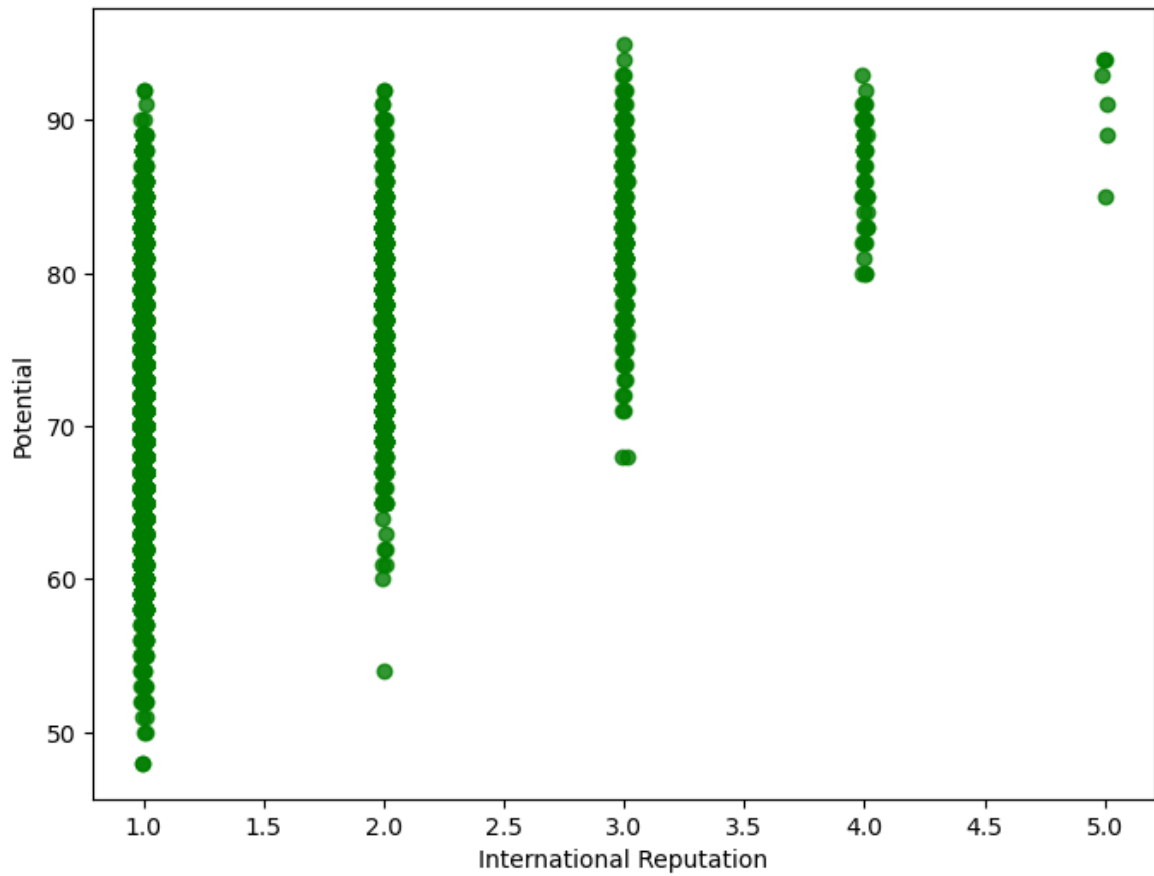
```
In [55]: f, ax = plt.subplots(figsize=(8, 6))
ax = sns.regplot(x="Overall", y="Potential", data=fifa19, color="g", marker="+")
plt.show()
```



```
In [56]: f, ax = plt.subplots(figsize=(8, 6))
sns.regplot(x="International Reputation", y="Potential", data=fifa19, x_jitter=.
plt.show()
```



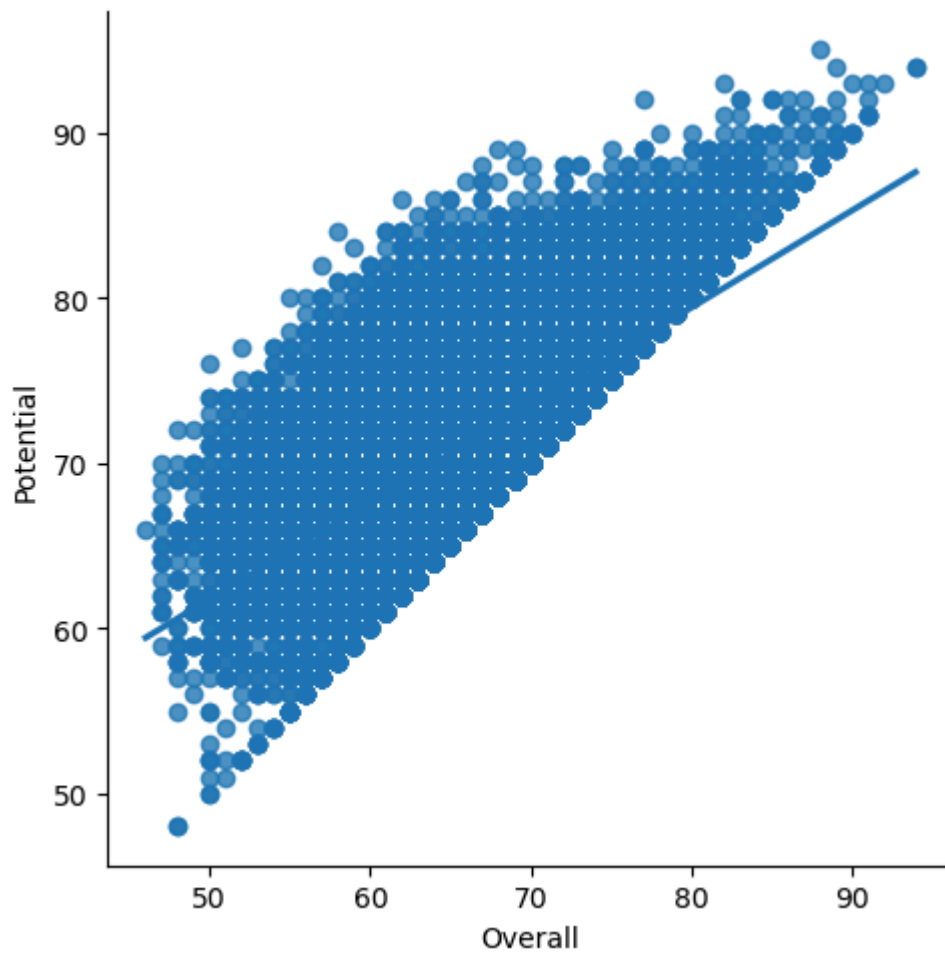
```
In [57]: f, ax = plt.subplots(figsize=(8, 6))
sns.regplot(x="International Reputation", y="Potential", data=fifa19, x_jitter=.
plt.show()
```



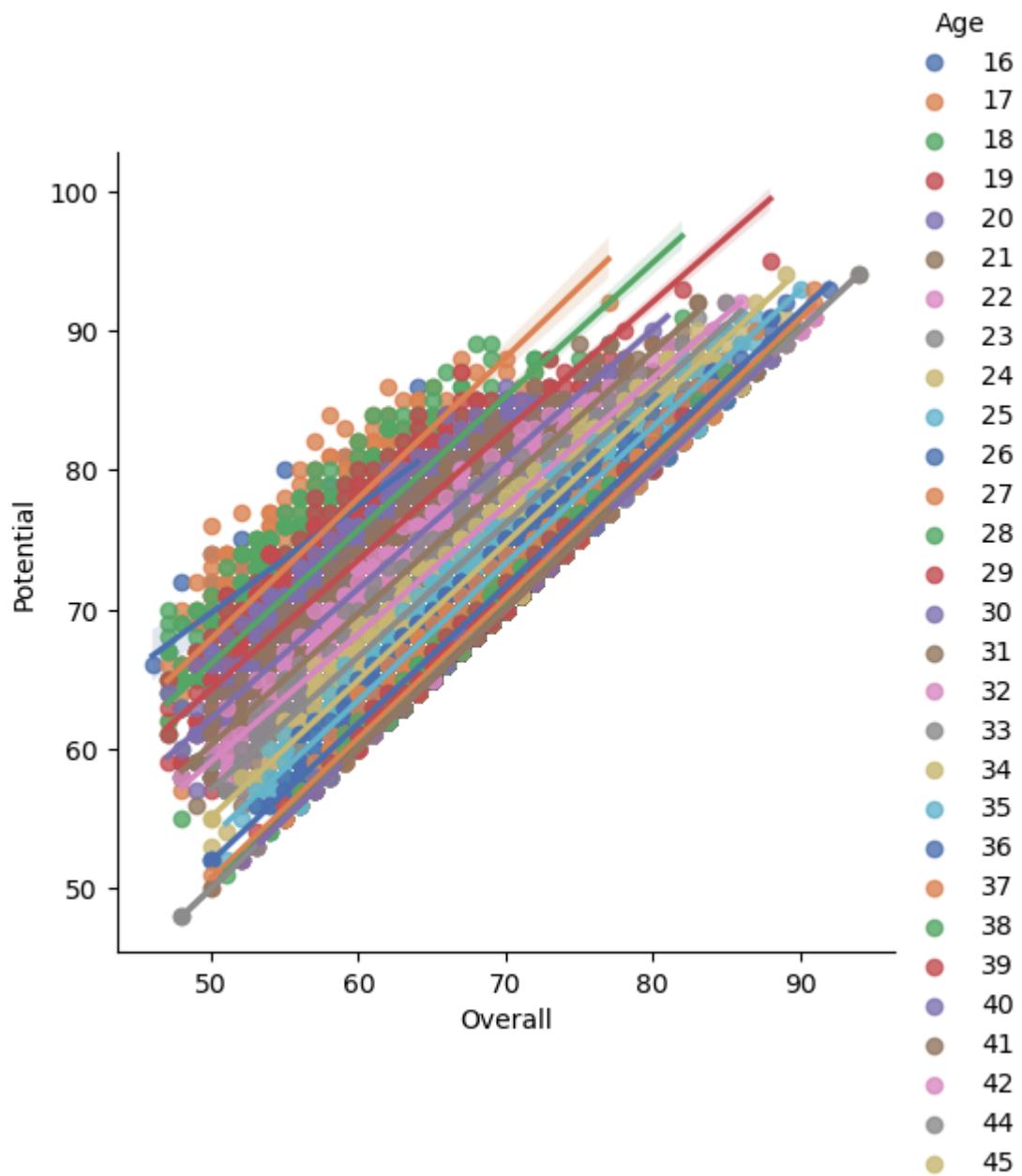
Implot() funvction

---this function combines regplot() and facetgird---

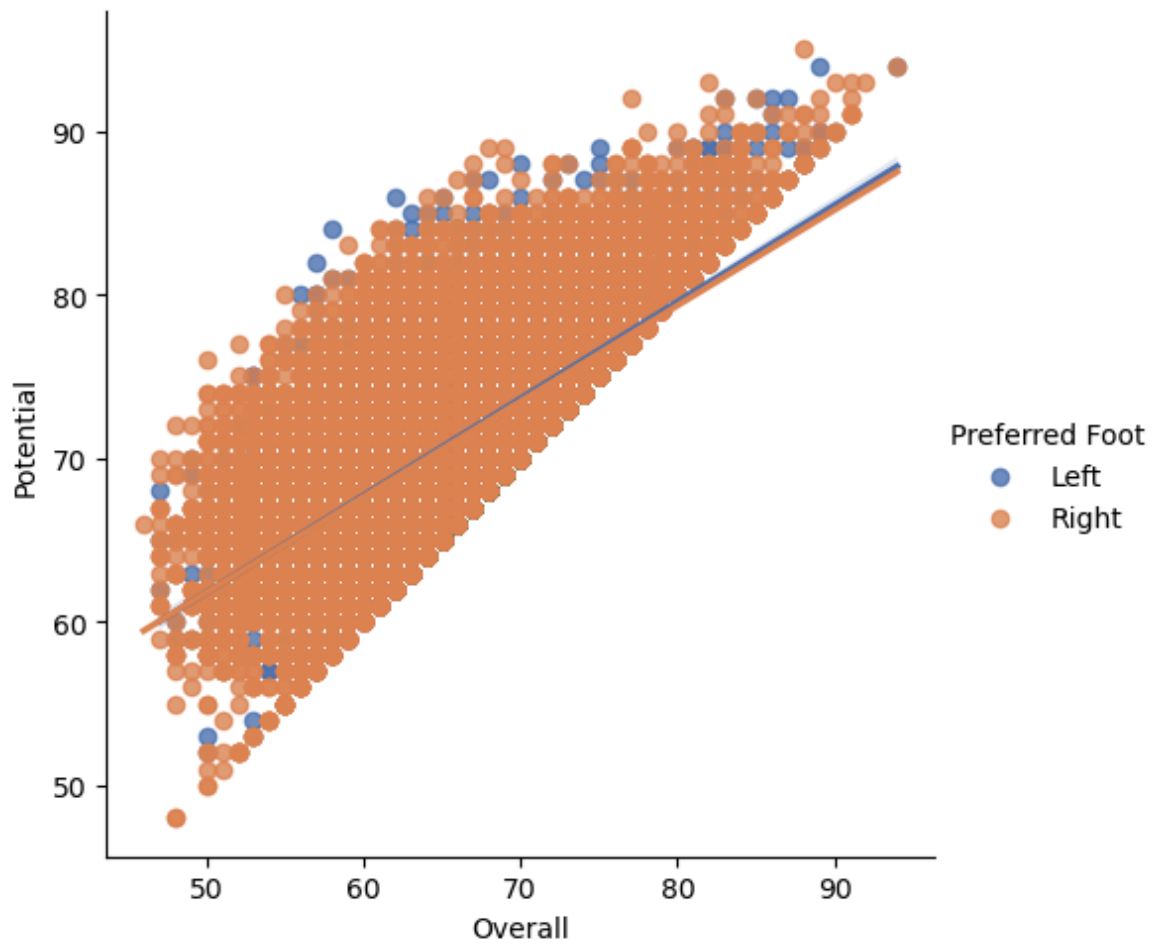
```
In [58]: g=sns.lmplot(x='Overall',y='Potential',data=fifa19,palette='deep')
plt.show()
```



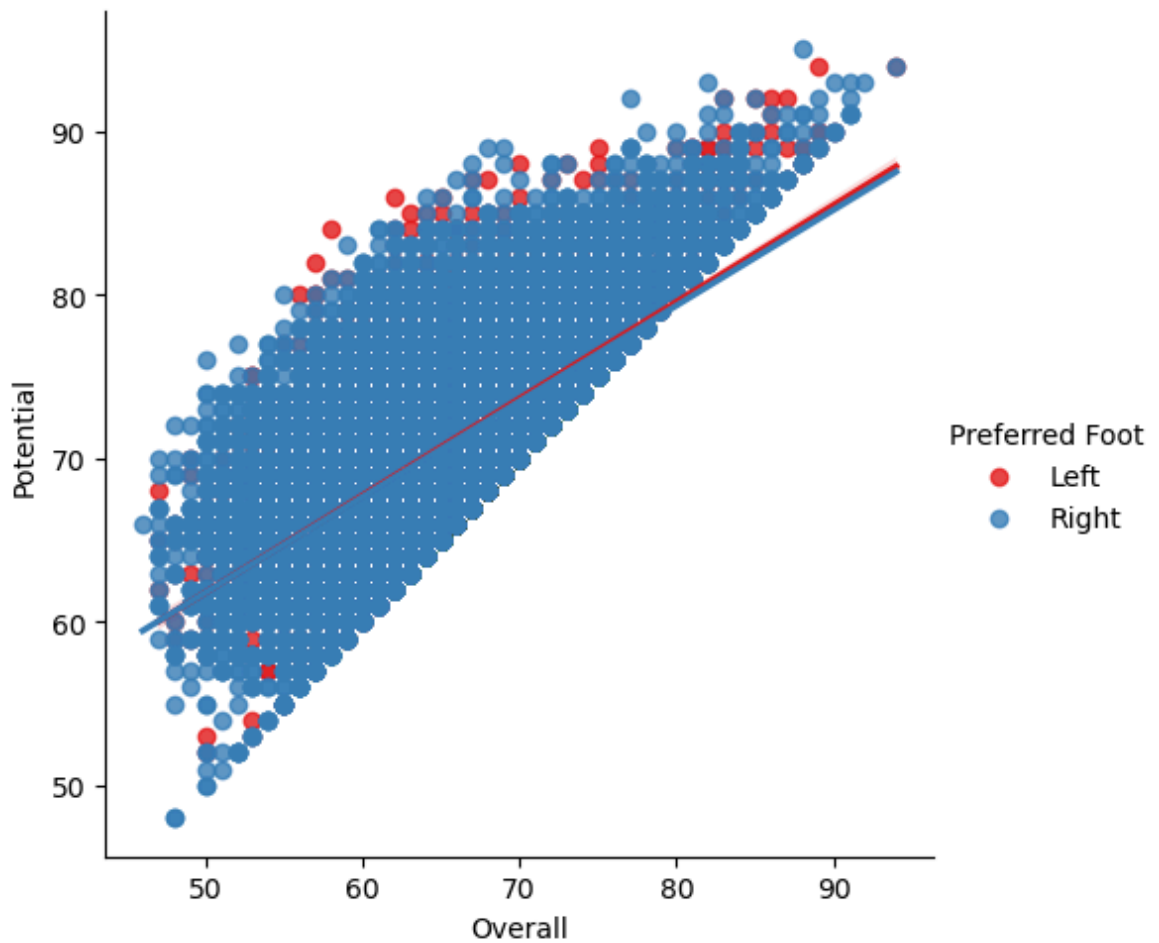
```
In [59]: g1=sns.lmplot(x='Overall',y='Potential',data=fifa19,hue='Age',palette='deep')  
plt.show()
```

```
In [60]: g2=sns.lmplot(x='Overall',y='Potential',data=fifa19,hue='Preferred Foot',palette
plt.show()
```



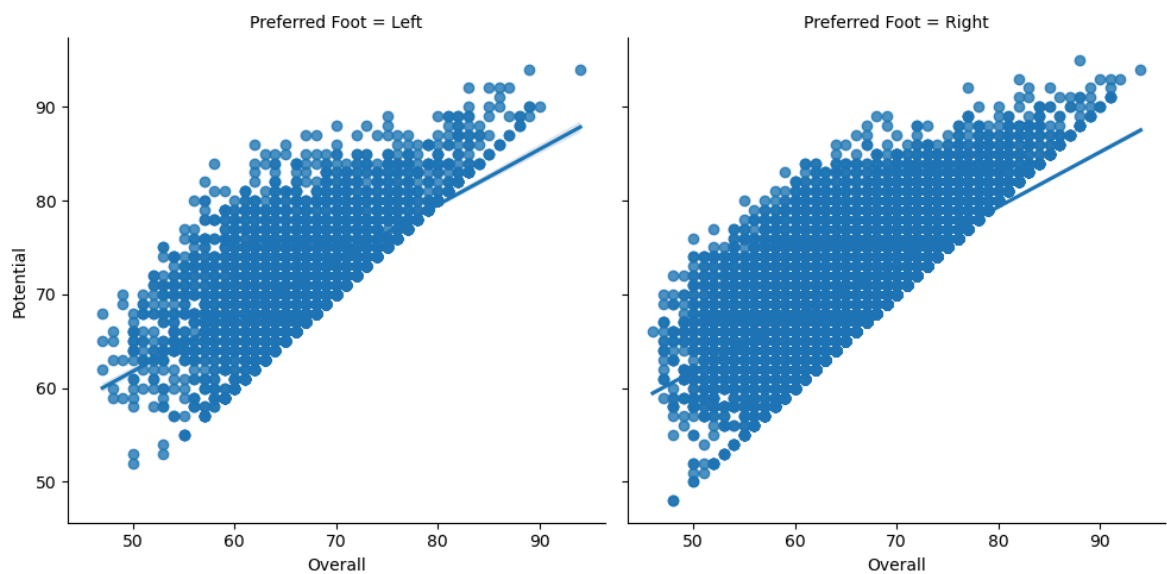
```
In [61]: g = sns.lmplot(x="Overall", y="Potential", hue="Preferred Foot", data=fifa19, pal
plt.show()
```



In []:

In [62]: *#we can plot the lvls of the third var across diff cols as follows--*

In [63]: `g= sns.lmplot(x="Overall", y="Potential", col="Preferred Foot", data=fifa19,pale
plt.show()`

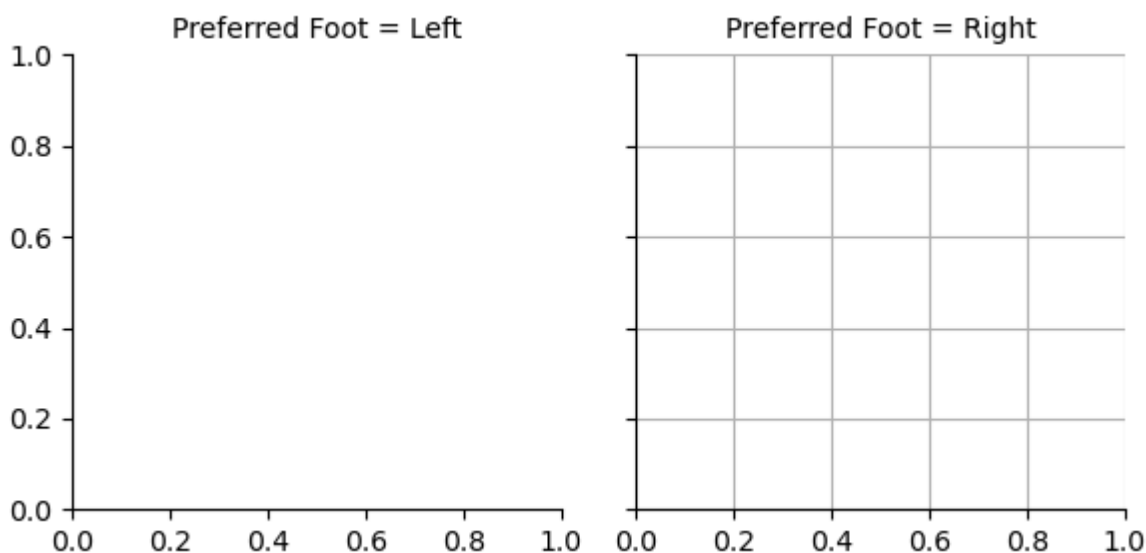


Multi-plot grids

Seaborn `FacetGrid()` function

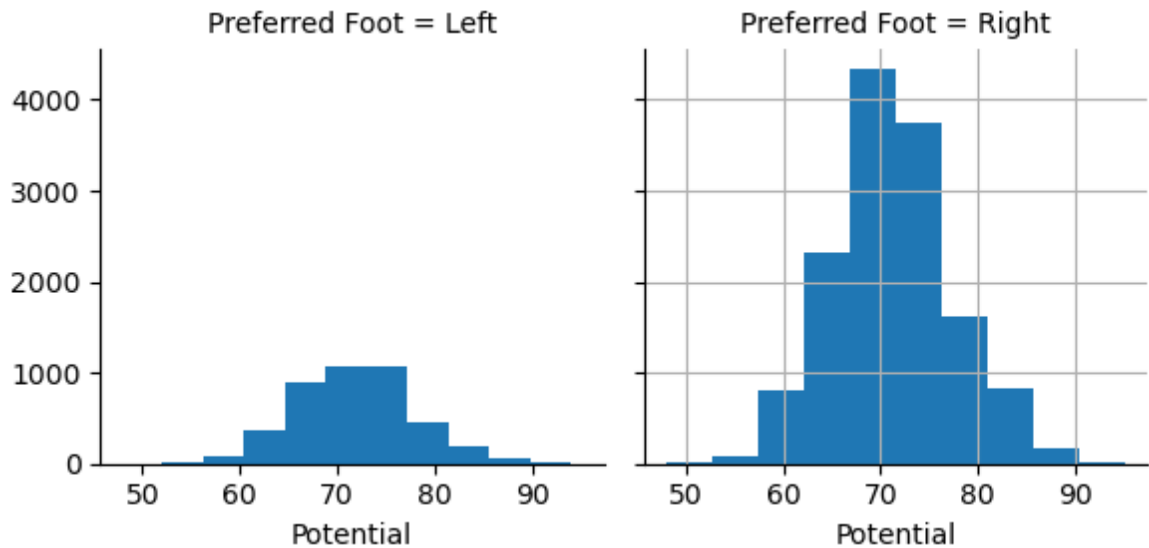
- The FacetGrid class is useful when you want to visualize the distribution of a variable or the relationship between multiple variables separately within subsets of your dataset.
- A FacetGrid can be drawn with up to three dimensions - `row`, `col` and `hue`. The first two have obvious correspondence with the resulting array of axes - the `hue` variable is a third dimension along a depth axis, where different levels are plotted with different colors.
- The class is used by initializing a FacetGrid object with a dataframe and the names of the variables that will form the `row`, `column` or `hue` dimensions of the grid.
- These variables should be categorical or discrete, and then the data at each level of the variable will be used for a facet along that axis.

```
In [64]: g=sns.FacetGrid(fifa19,col='Preferred Foot')
plt.grid(g)
plt.show()
```

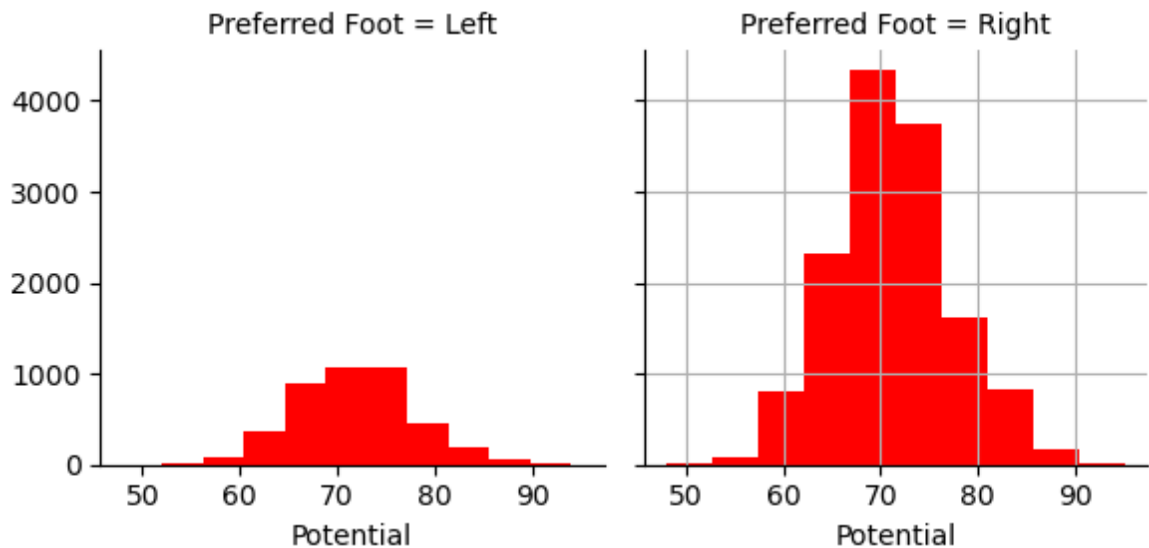


```
In [65]: # we can also draw a univariate plot of potential var on each facet as follows-
```

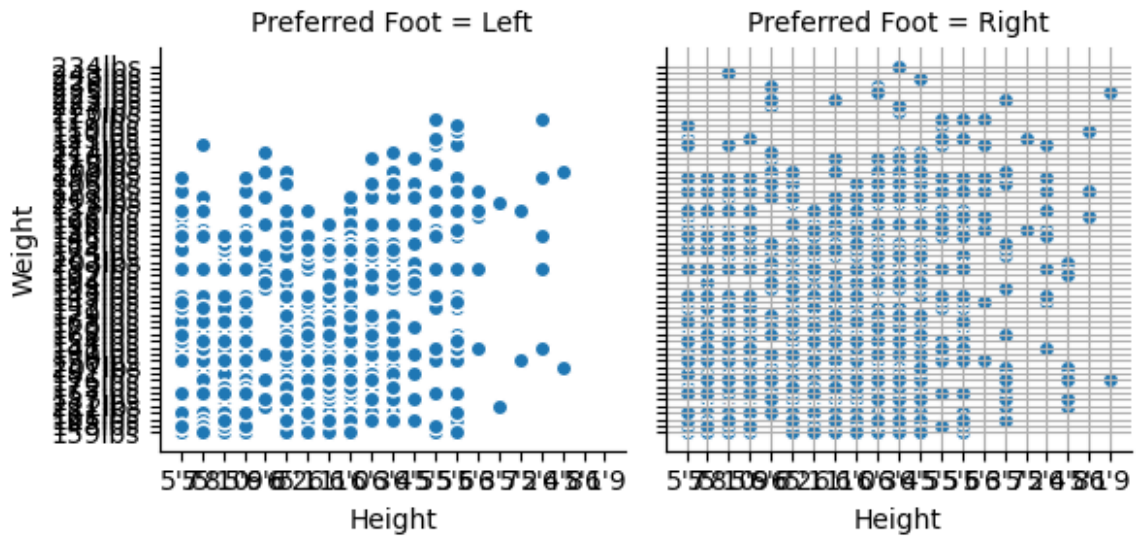
```
In [66]: g = sns.FacetGrid(fifa19, col="Preferred Foot")
g = g.map(plt.hist, "Potential")
plt.grid(g)
plt.show()
```



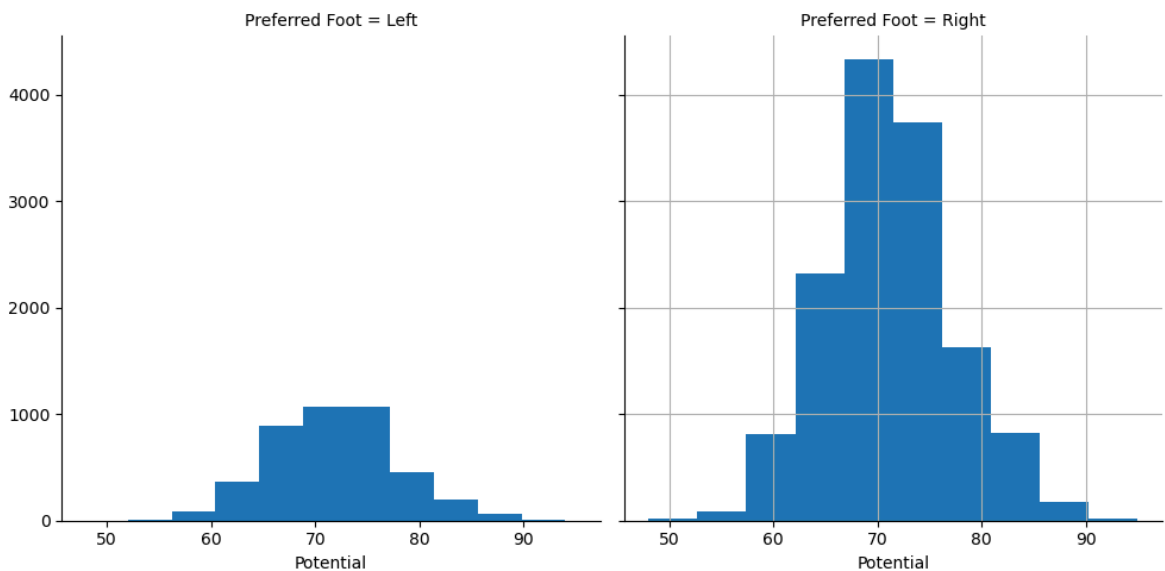
```
In [67]: g = sns.FacetGrid(fifa19, col="Preferred Foot")
g = g.map(plt.hist, "Potential", bins=10, color='r')
plt.grid(g)
plt.show()
```



```
In [68]: # here we have plotted bivariate function on each facet
g = sns.FacetGrid(fifa19, col="Preferred Foot")
g = g.map(plt.scatter, 'Height', 'Weight', edgecolor='w').add_legend()
plt.grid(g)
plt.show()
```



```
In [69]: g = sns.FacetGrid(fifa19, col="Preferred Foot", height=5, aspect=1)
g = g.map(plt.hist, "Potential")
plt.grid(g)
plt.show()
```



Seaborn Pairgrid() function

- This function plots subplot grid for plotting pairwise relationships in a dataset.
- This class maps each variable in a dataset onto a column and row in a grid of multiple axes.
- Different axes-level plotting functions can be used to draw bivariate plots in the upper and lower triangles, and the the marginal distribution of each variable can be shown on the diagonal.
- It can also represent an additional level of conditionalization with the hue parameter, which plots different subsets of data in different colors.
- This uses color to resolve elements on a third dimension, but only draws subsets on top of each other and will not tailor the hue parameter for the specific visualization

the way that axes-level functions that accept hue will.

```
In [70]: fifa19_new = fifa19[['Age', 'Potential', 'Strength', 'Stamina', 'Preferred Foot']]
```

```
In [71]: fifa19_new
```

```
Out[71]:
```

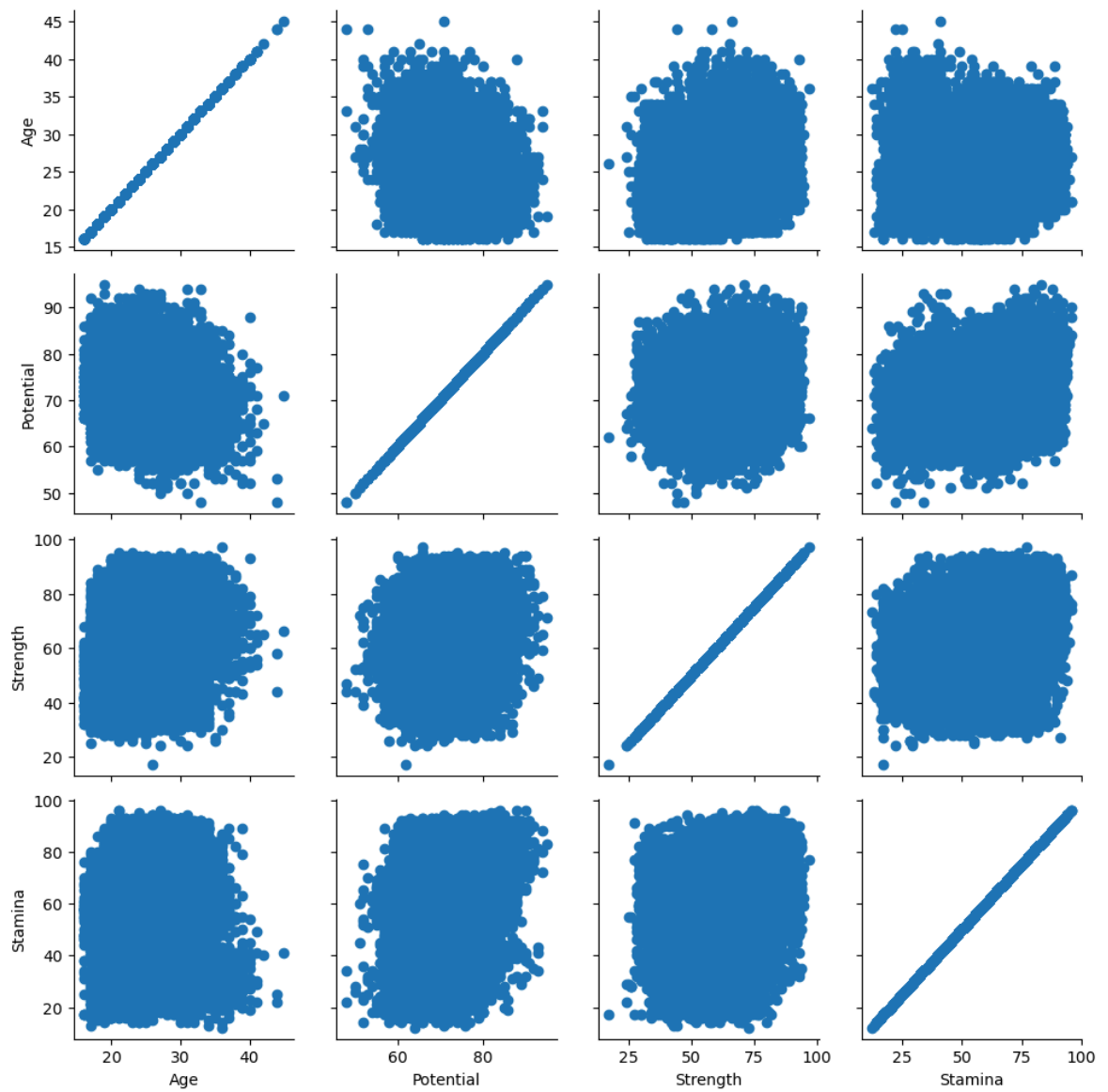
	Age	Potential	Strength	Stamina	Preferred Foot
0	31	94	59.0	72.0	Left
1	33	94	79.0	88.0	Right
2	26	93	49.0	81.0	Right
3	27	93	64.0	43.0	Right
4	27	92	75.0	90.0	Right
...
18202	19	65	47.0	40.0	Right
18203	19	63	67.0	43.0	Right
18204	16	67	32.0	55.0	Right
18205	17	66	48.0	40.0	Right
18206	16	66	60.0	47.0	Right

18207 rows × 5 columns

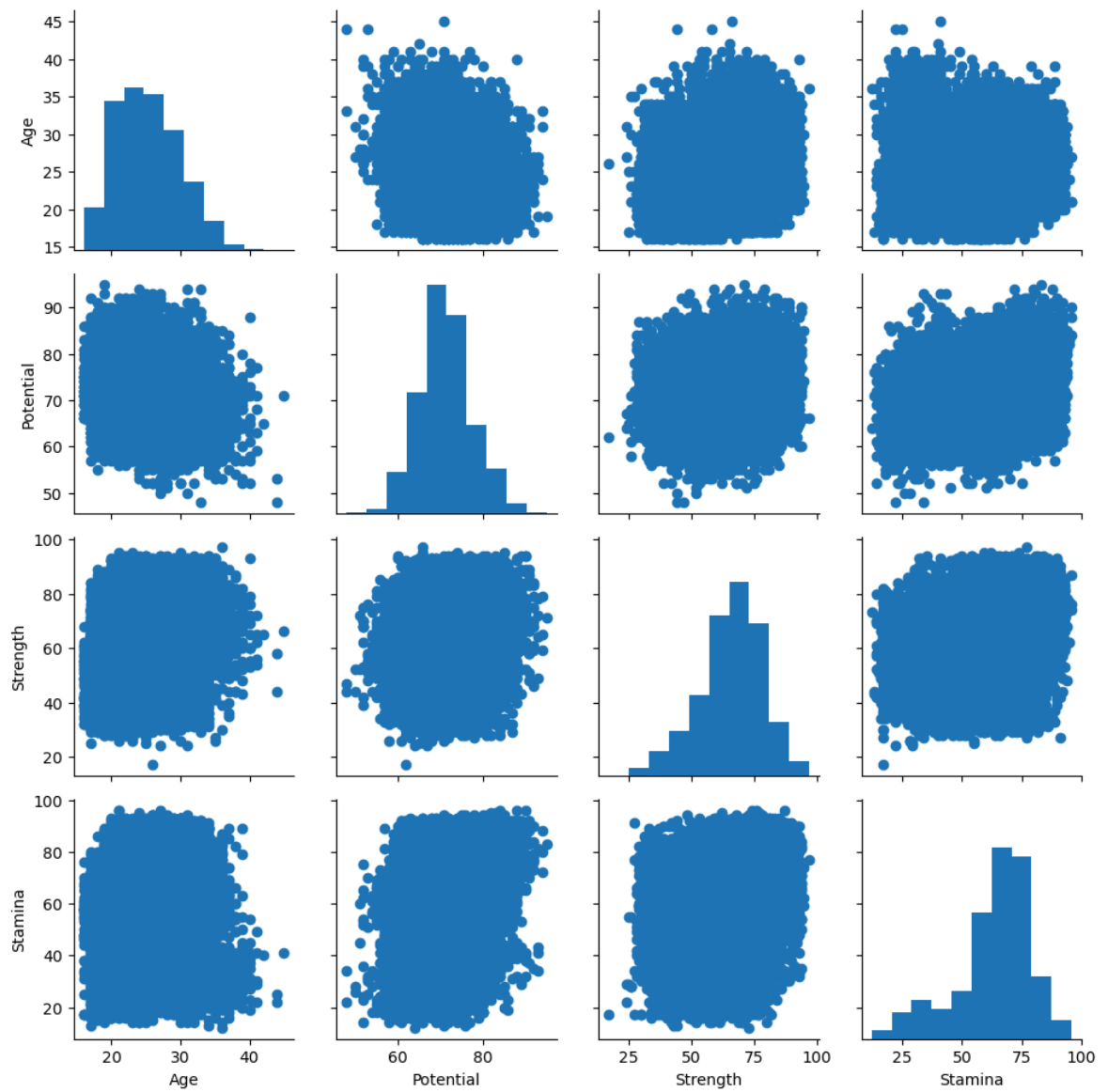
```
In [72]: type(fifa19_new)
```

```
Out[72]: pandas.core.frame.DataFrame
```

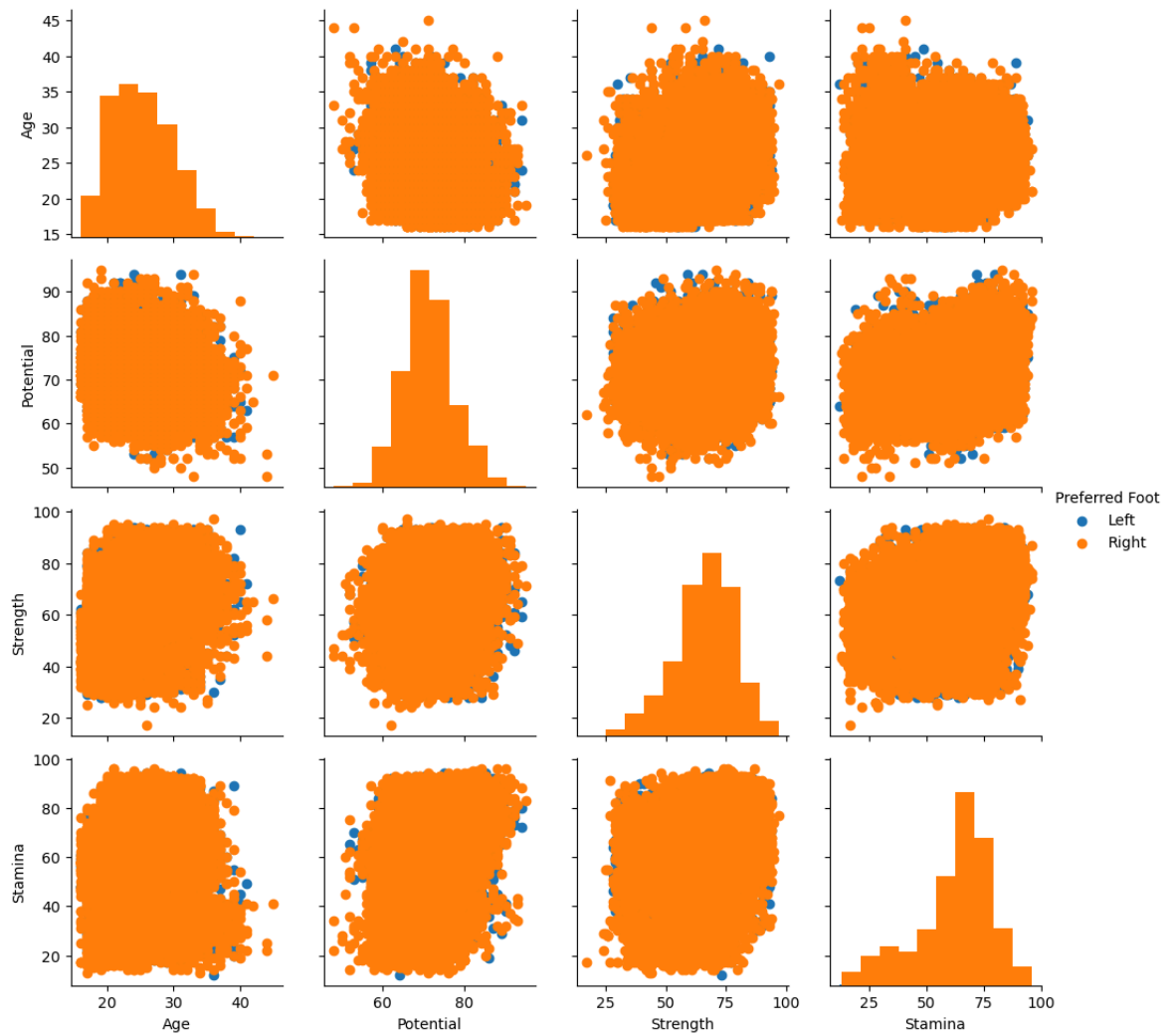
```
In [73]: g = sns.PairGrid(fifa19_new)
g = g.map(plt.scatter)
plt.show()
```



```
In [74]: g = sns.PairGrid(fifa19_new)
g = g.map_diag(plt.hist)
g = g.map_offdiag(plt.scatter)
plt.show()
```

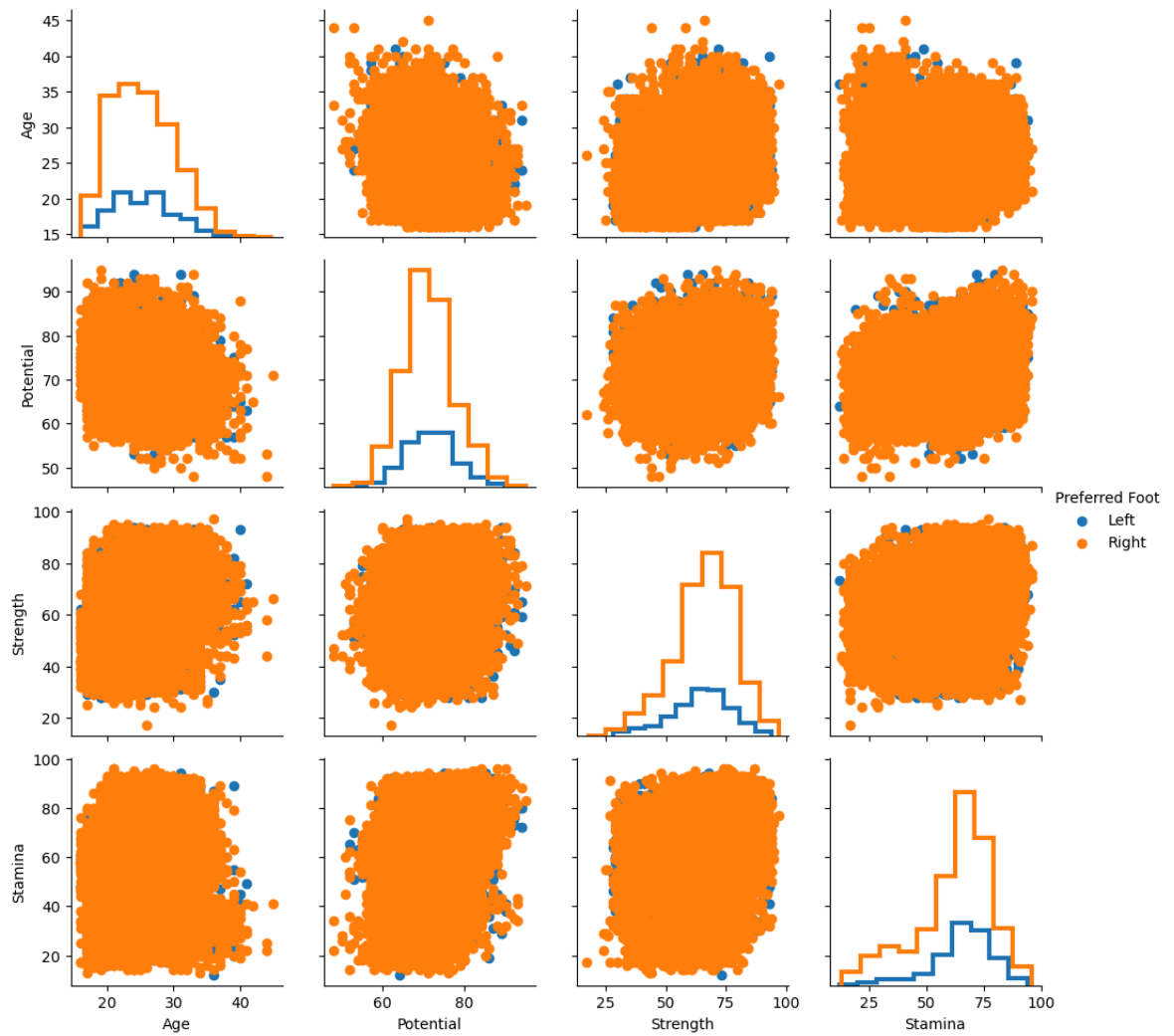



```
In [75]: g = sns.PairGrid(fifa19_new, hue="Preferred Foot")
g = g.map_diag(plt.hist)
g = g.map_offdiag(plt.scatter)
g = g.add_legend()
plt.show()
```

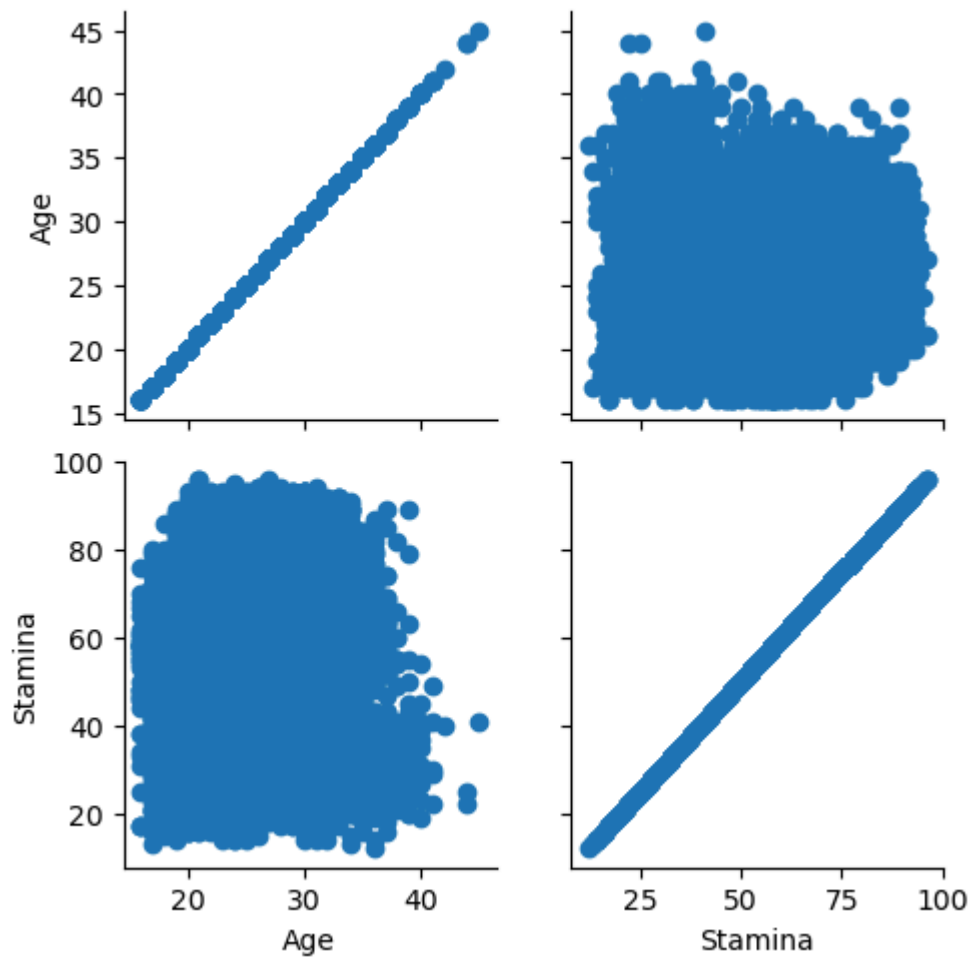


In [76]: *#we can use diff style to show mul histograms as follows:*

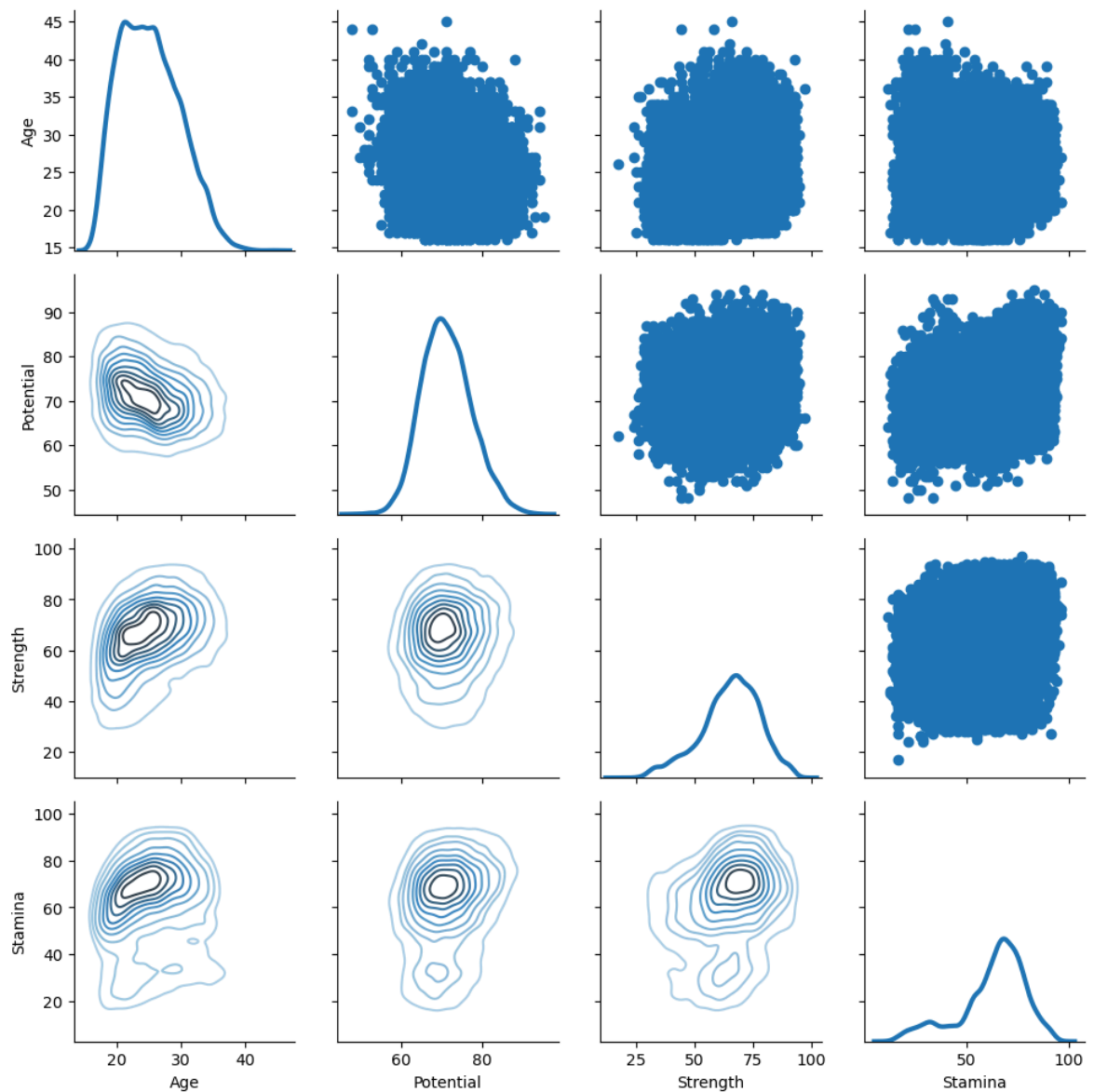
```
g= sns.PairGrid(fifa19_new,hue='Preferred Foot')
g=g.map_diag(plt.hist, histtype='step',linewidth=3)
g=g.map_offdiag(plt.scatter)
g=g.add_legend()
plt.show()
```



```
In [77]: #plotting subset of vars
g=sns.PairGrid(fifa19_new,vars=['Age','Stamina'])
g=g.map(plt.scatter)
plt.show()
```



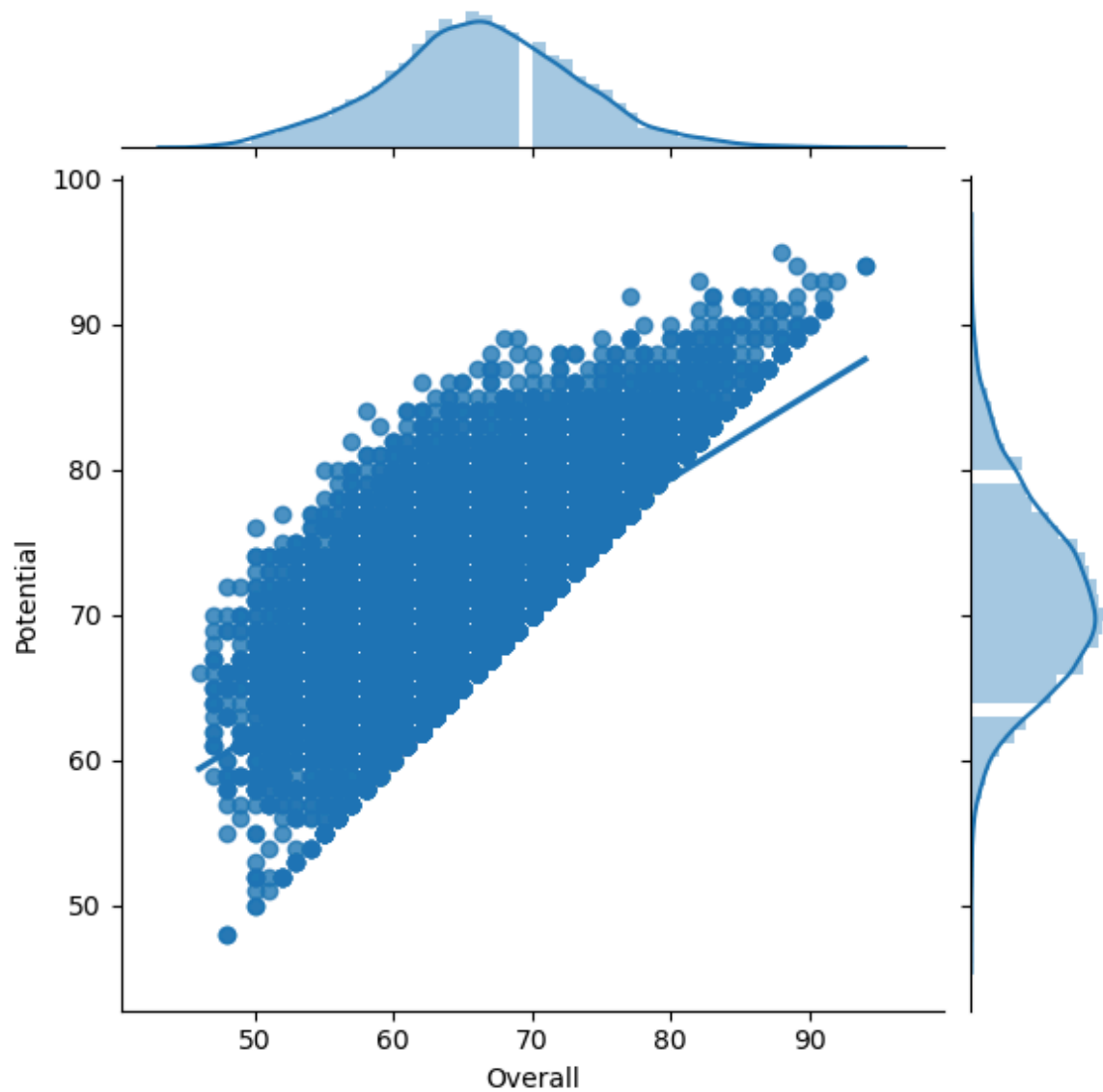
```
In [78]: g=sns.PairGrid(fifa19_new)
g=g.map_upper(plt.scatter)
g=g.map_lower(sns.kdeplot,cmap='Blues_d')
g=g.map_diag(sns.kdeplot,lw=3,legend=False)
plt.show()
```



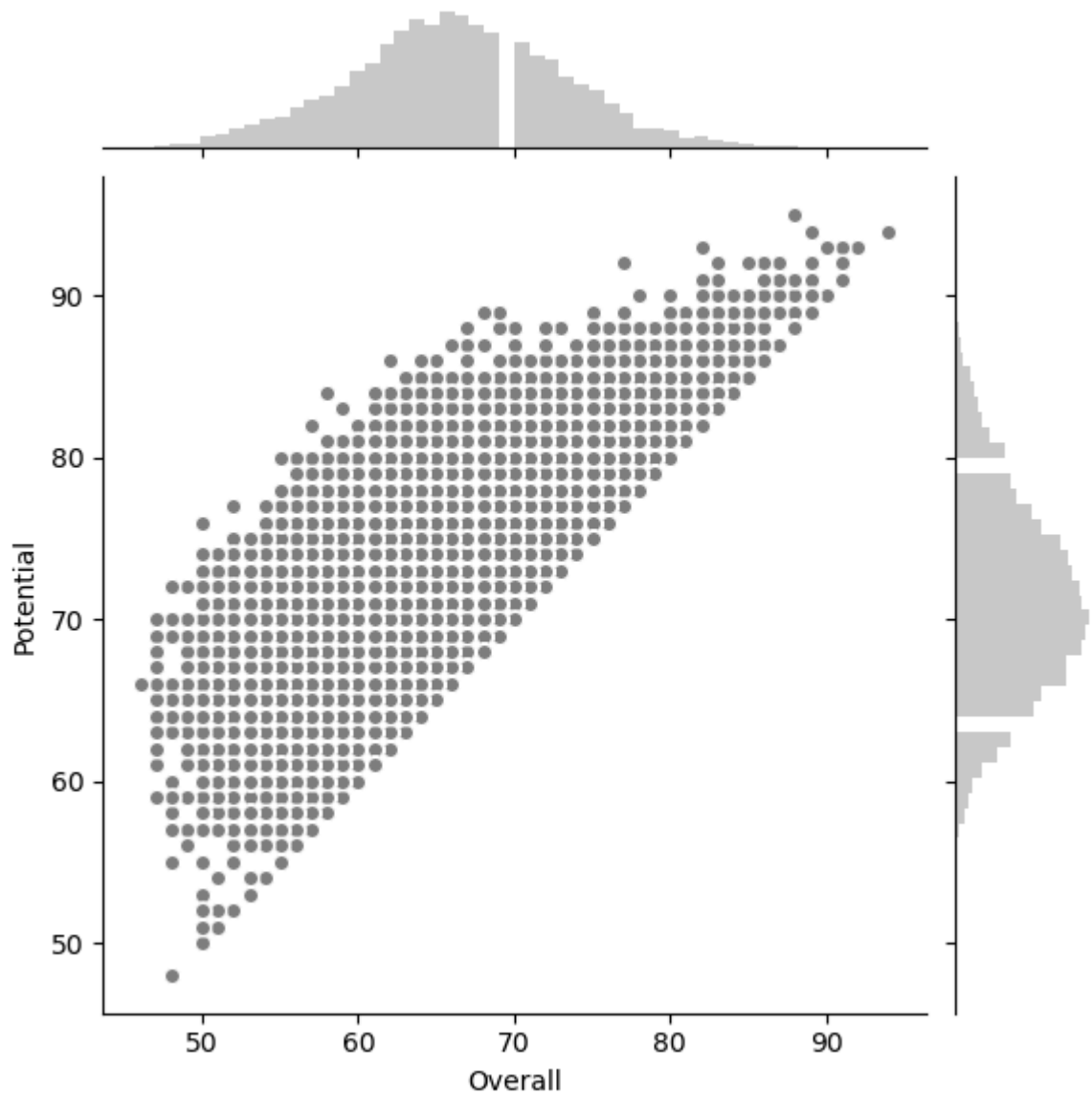
jointgrid() function

-- this func allows to plot bivariate graph with marginal univariate plot, it setup the grid of sublots

```
In [79]: g = sns.JointGrid(x="Overall", y="Potential", data=fifa19)
g = g.plot(sns.regplot, sns.distplot)
plt.show()
```

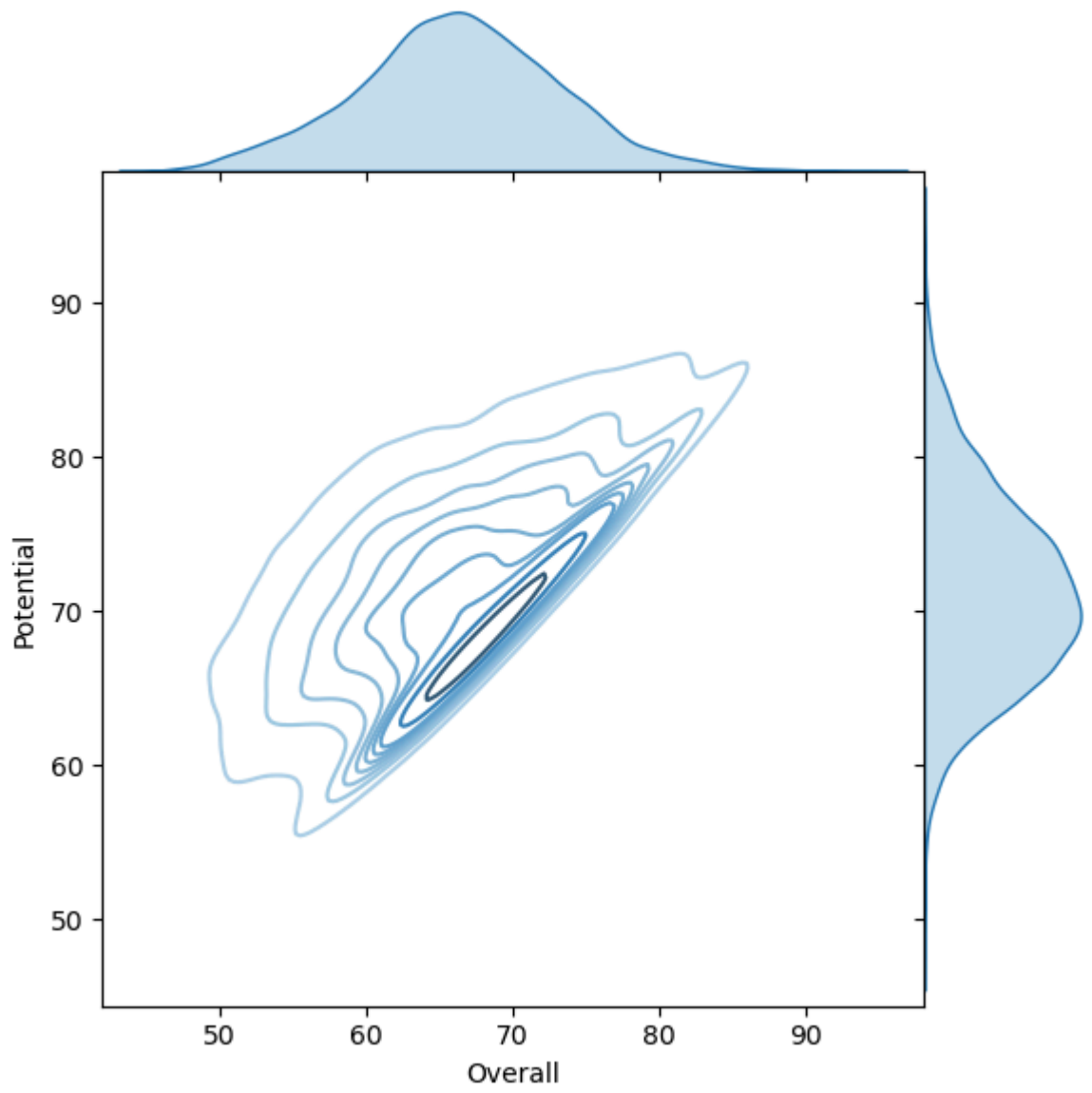


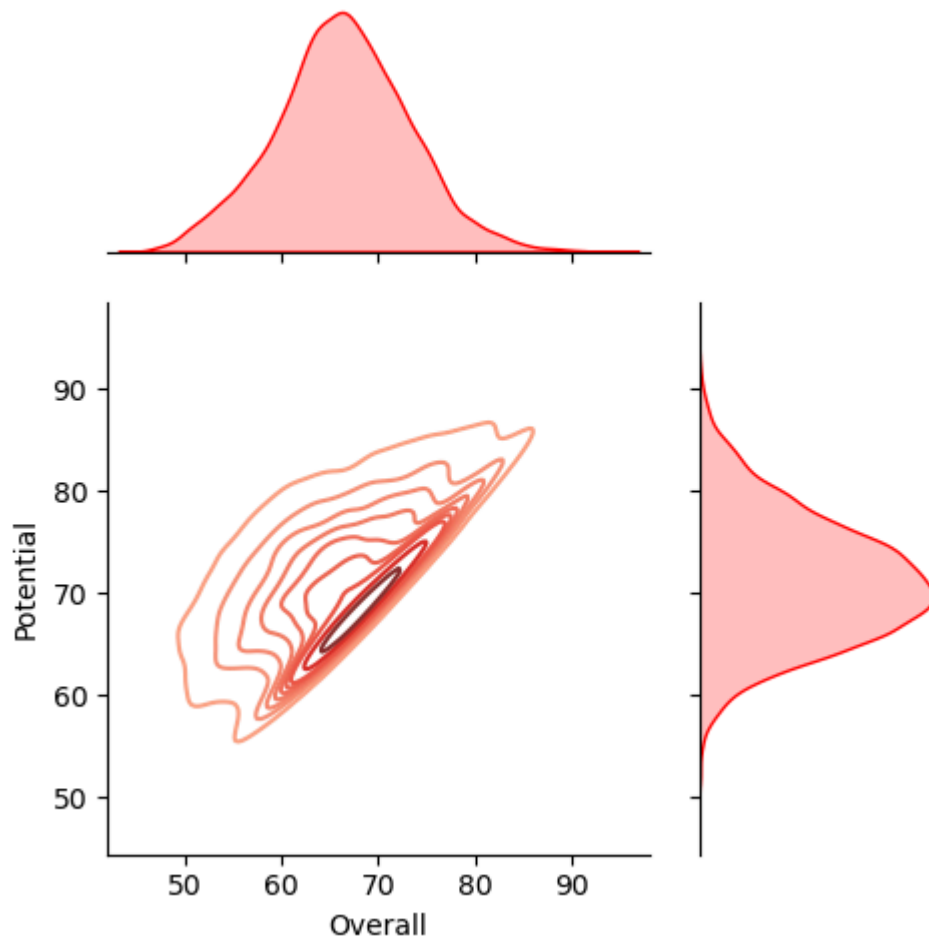
```
In [80]: g = sns.JointGrid(x="Overall", y="Potential", data=fifa19)
g = g.plot_joint(plt.scatter, color=".5", edgecolor="white")
g = g.plot_marginals(sns.distplot, kde=False, color=".5")
plt.show()
```



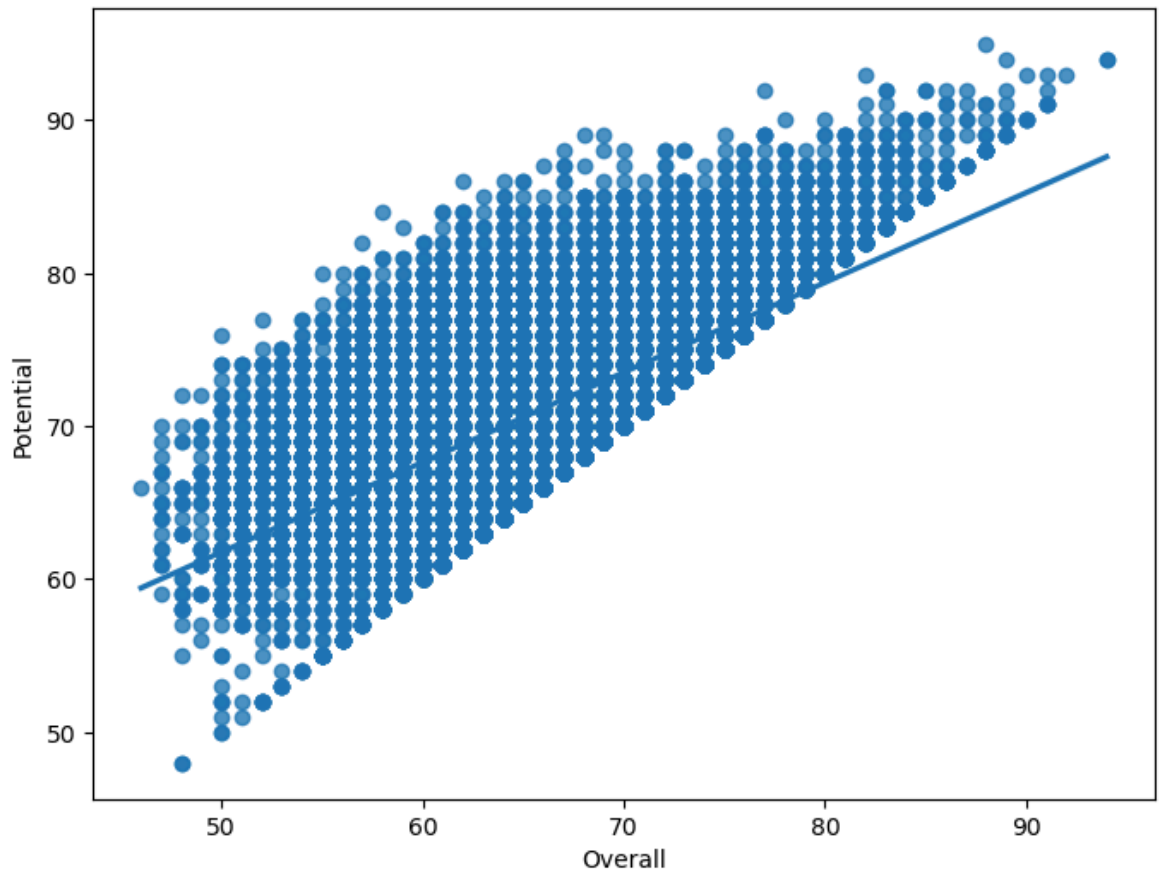
```
In [81]: g = sns.JointGrid(x="Overall", y="Potential", data=fifa19, space=0)
g = g.plot_joint(sns.kdeplot, cmap="Blues_d")
g = g.plot_marginals(sns.kdeplot, shade=True)
plt.show()
```

```
In [82]: g = sns.JointGrid(x="Overall", y="Potential", data=fifa19, height=5, ratio=2)
g = g.plot_joint(sns.kdeplot, cmap="Reds_d")
g = g.plot_marginals(sns.kdeplot, color="r", shade=True)
plt.show()
```

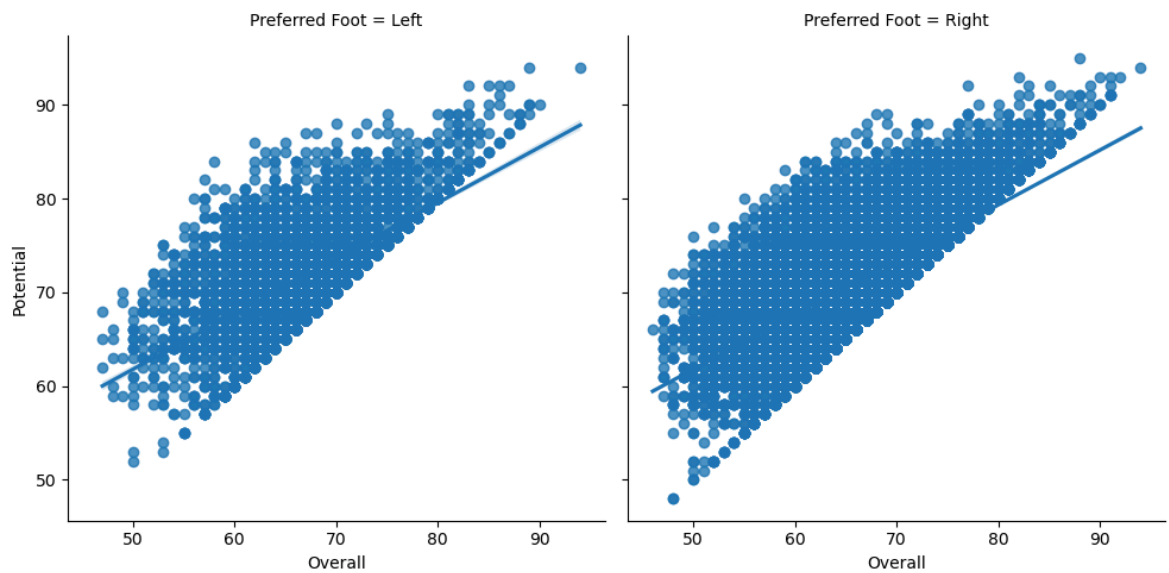


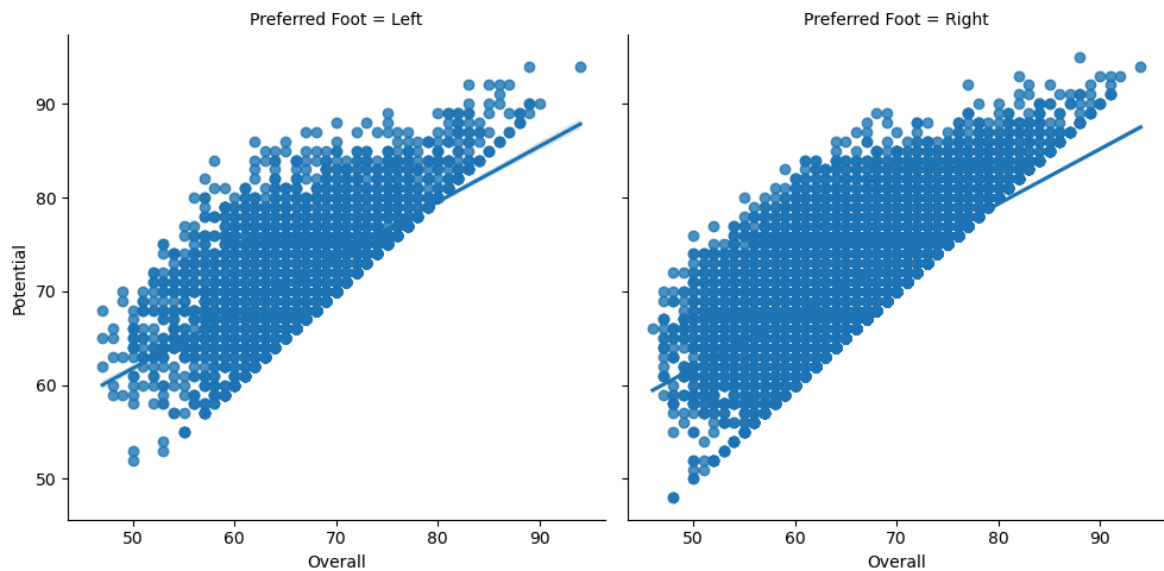


```
In [83]: f, ax = plt.subplots(figsize=(8, 6))
ax = sns.regplot(x="Overall", y="Potential", data=fifa19);
plt.show()
```



```
In [85]: sns.lmplot(x="Overall", y="Potential", col="Preferred Foot", data=fifa19, col_wrap=2,
plt.show())
```

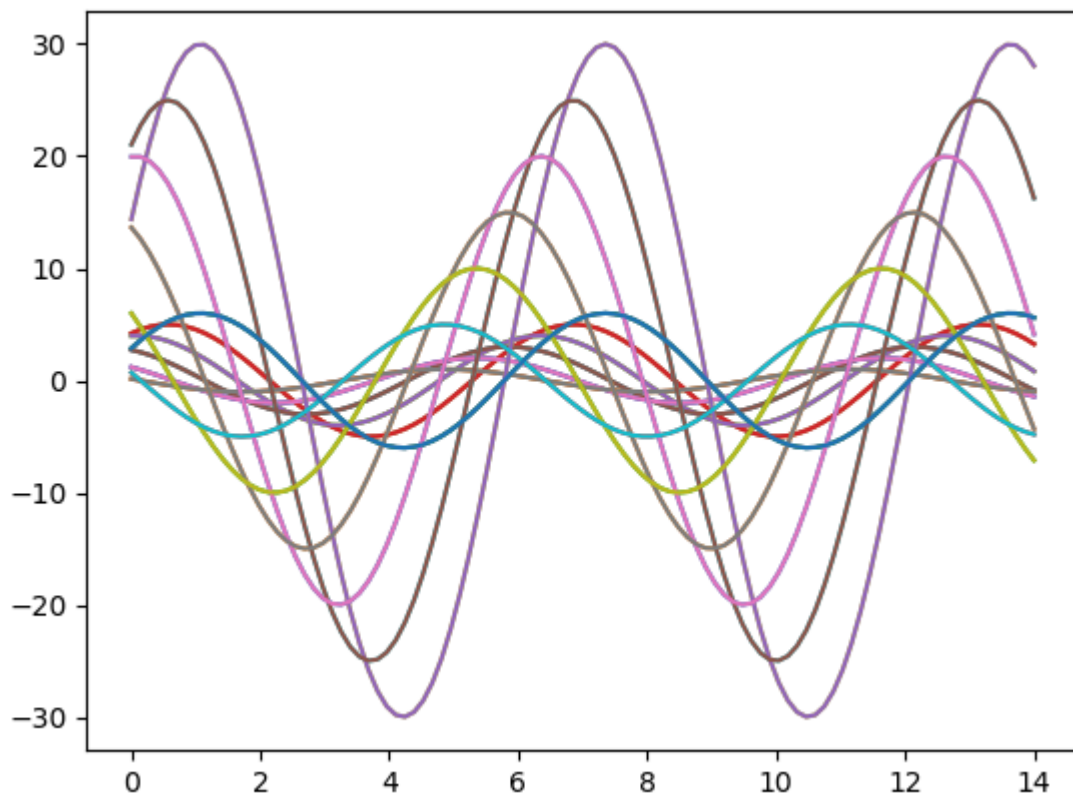


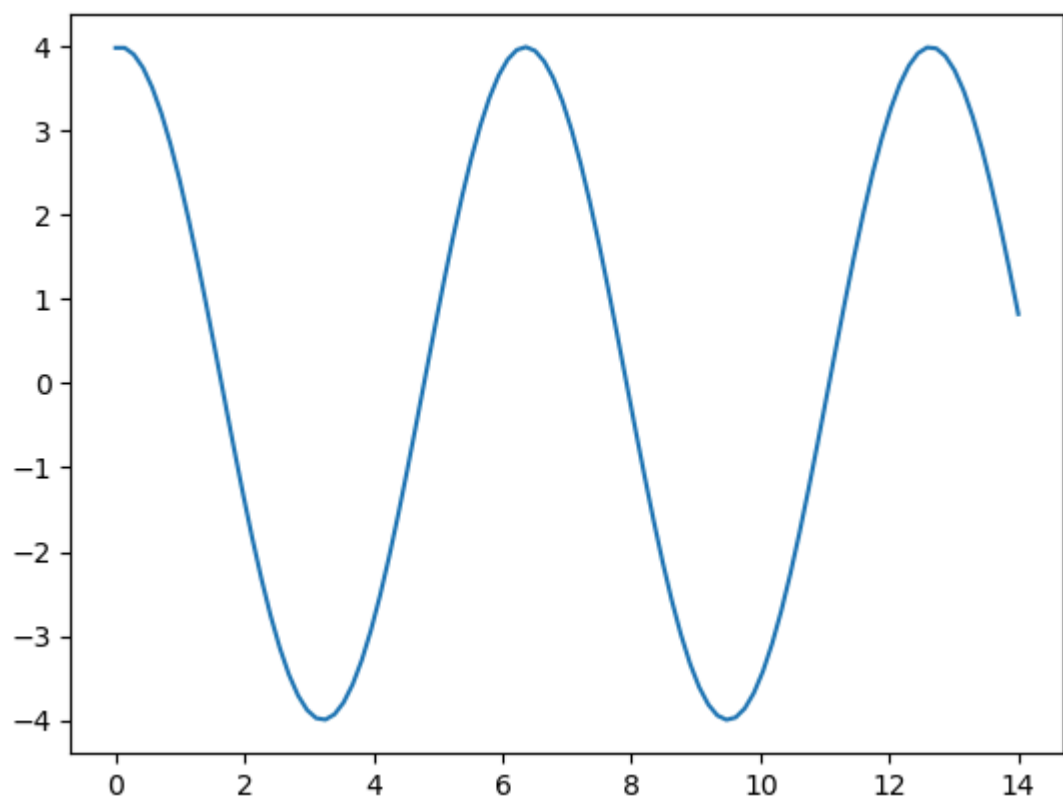
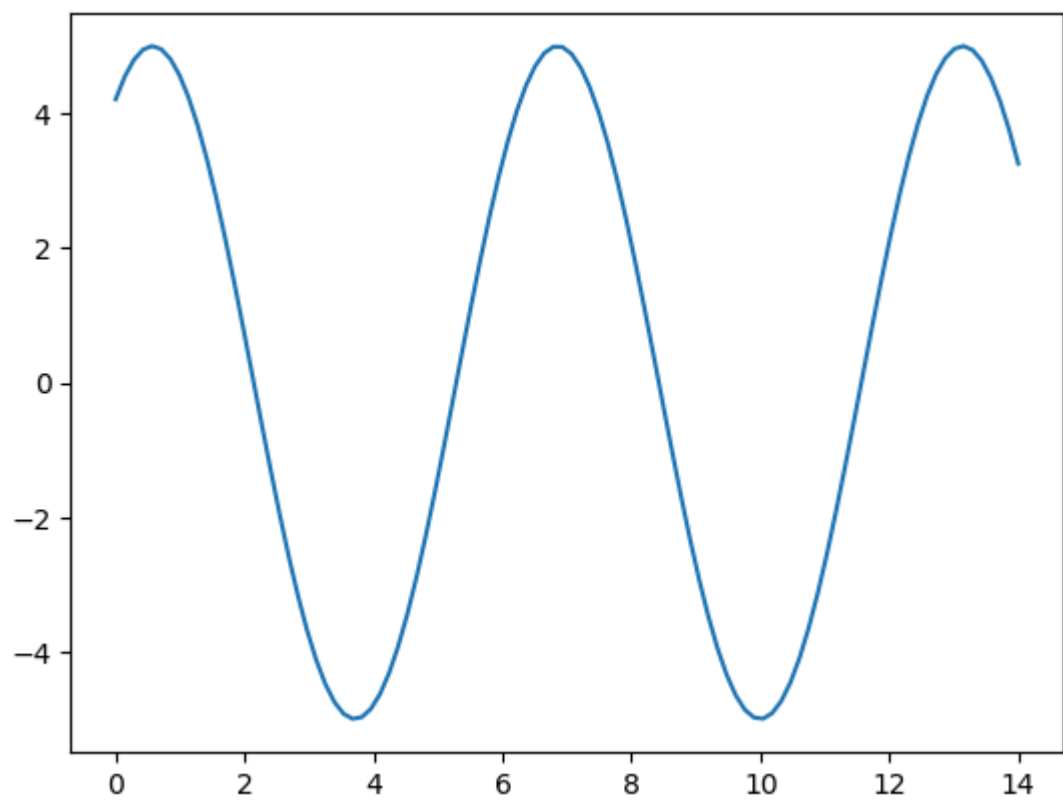


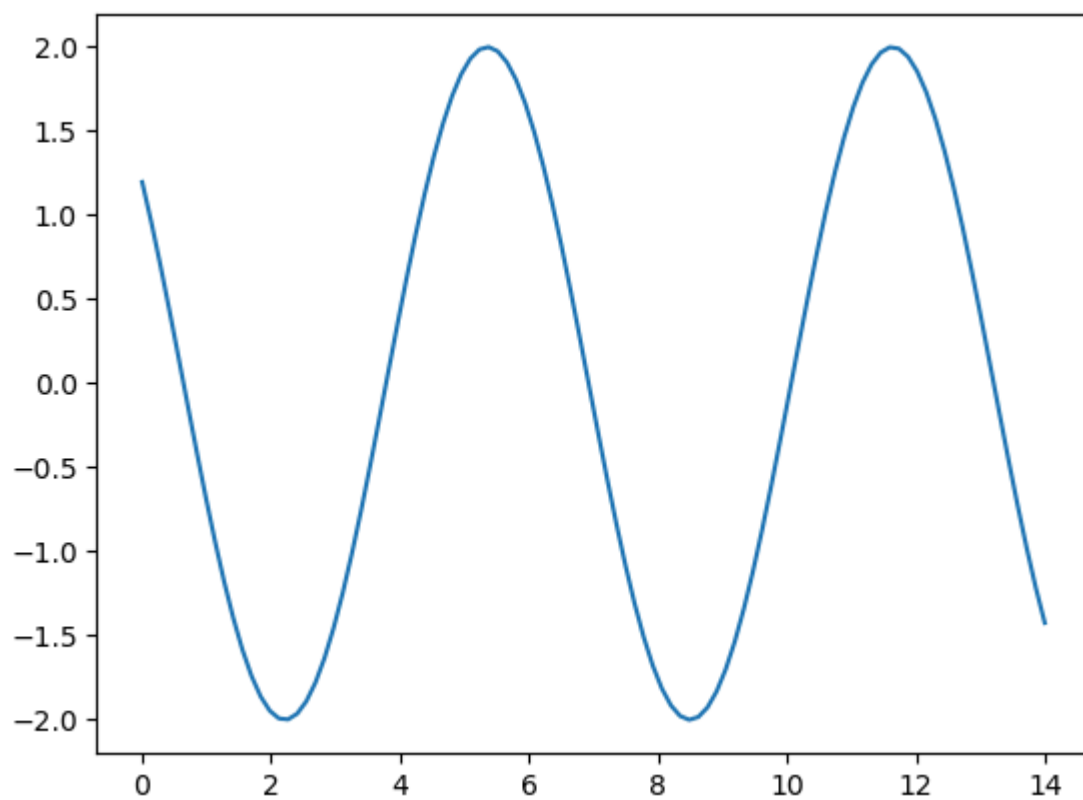
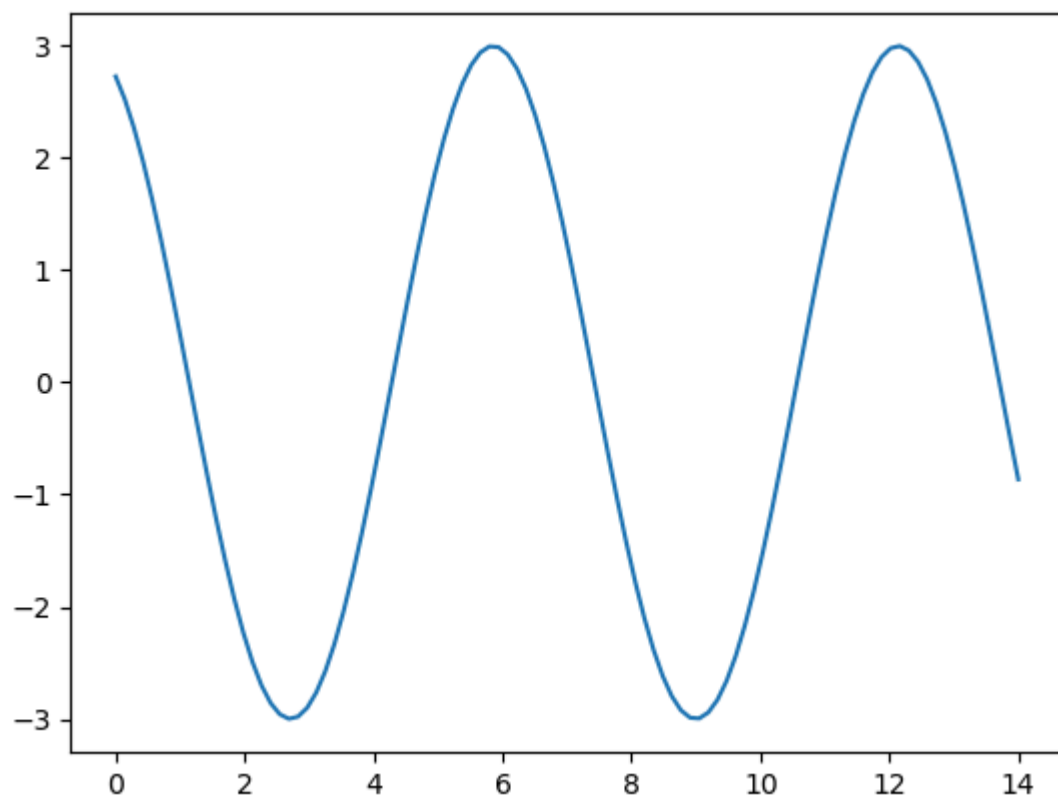
I will define a simple function to plot some offset sine waves, which will help us see the different stylistic parameters as follows -

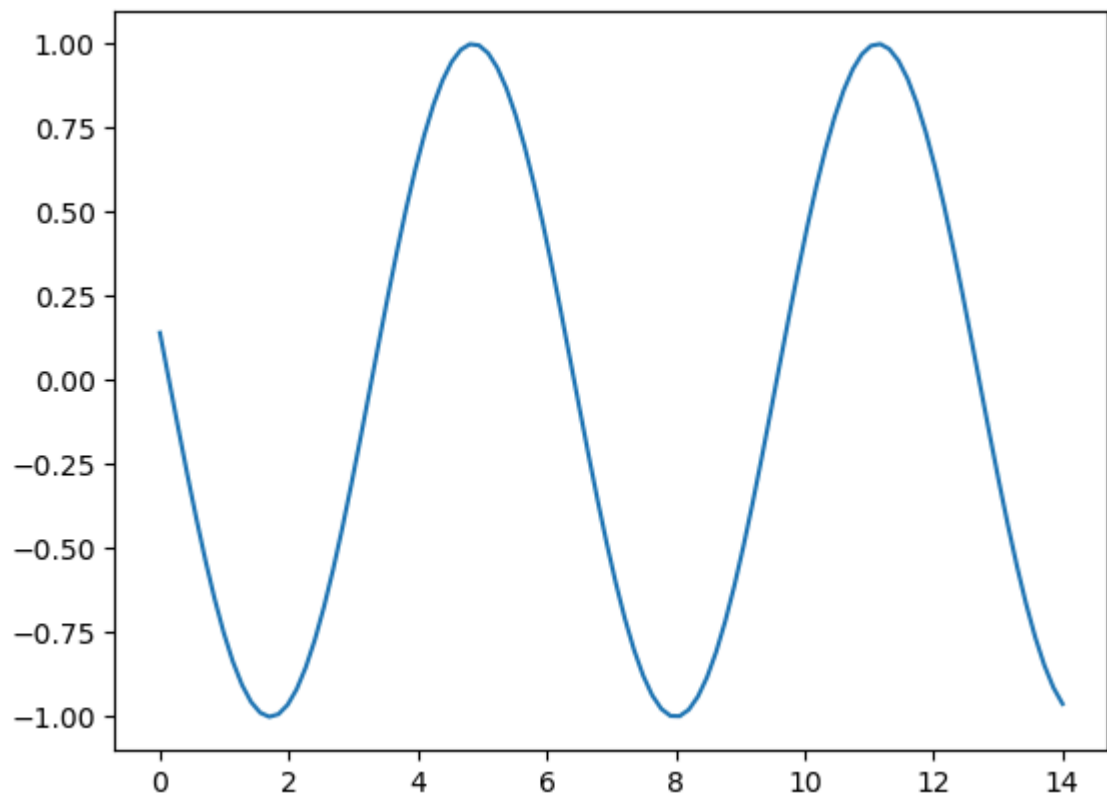
```
In [96]: def sinplot(flip=1):
x=np.linspace(0,14,100)
for i in range(1,7):
    plt.plot(x,np.sin(x+i*.5)*(7-i)*flip)
plt.show()
```

```
In [97]: sinplot() #this is what plot with default matplotlib parameters looks like
```



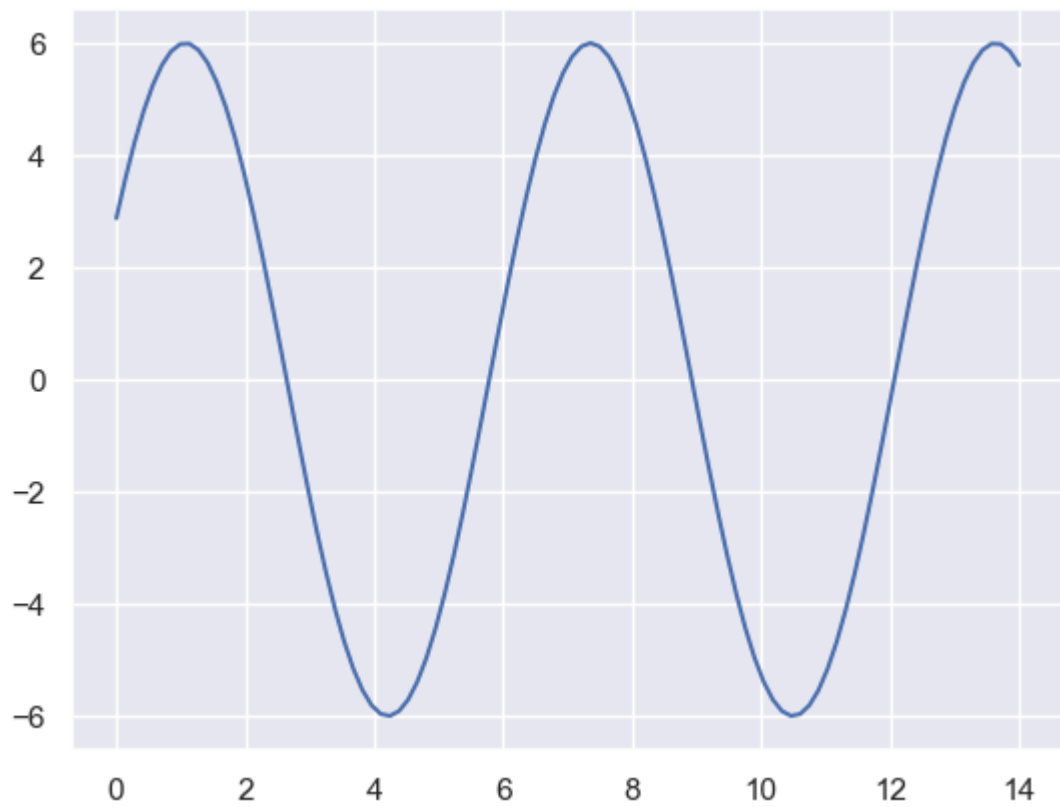


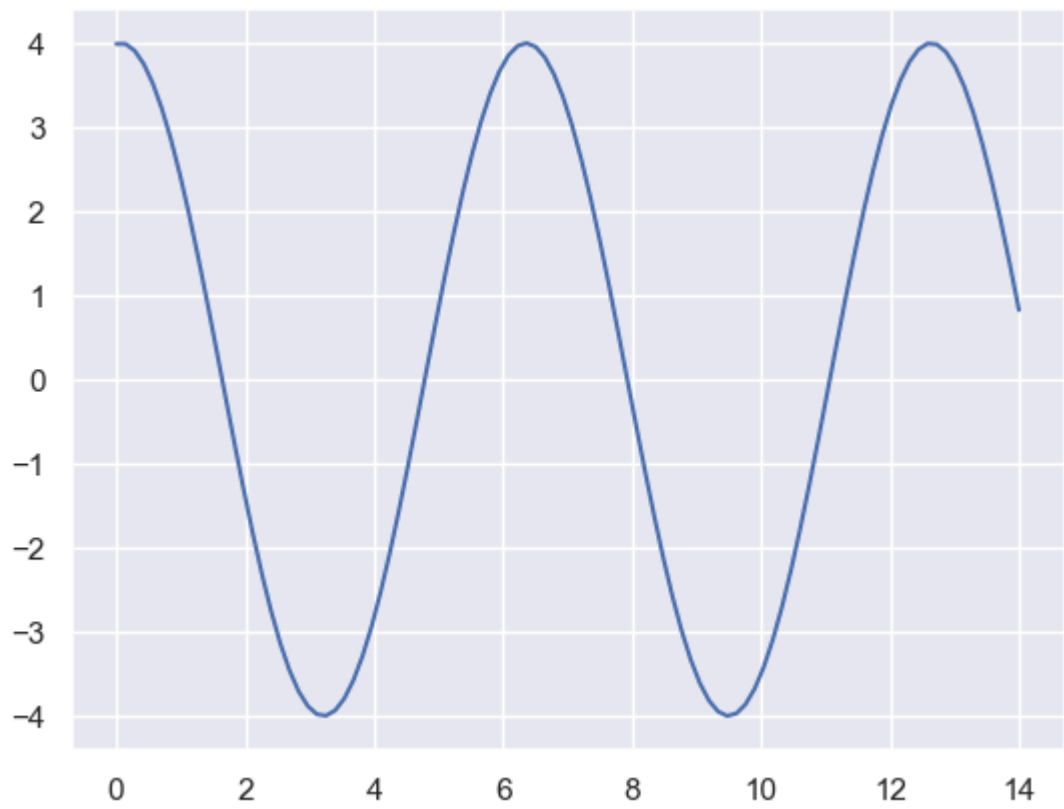
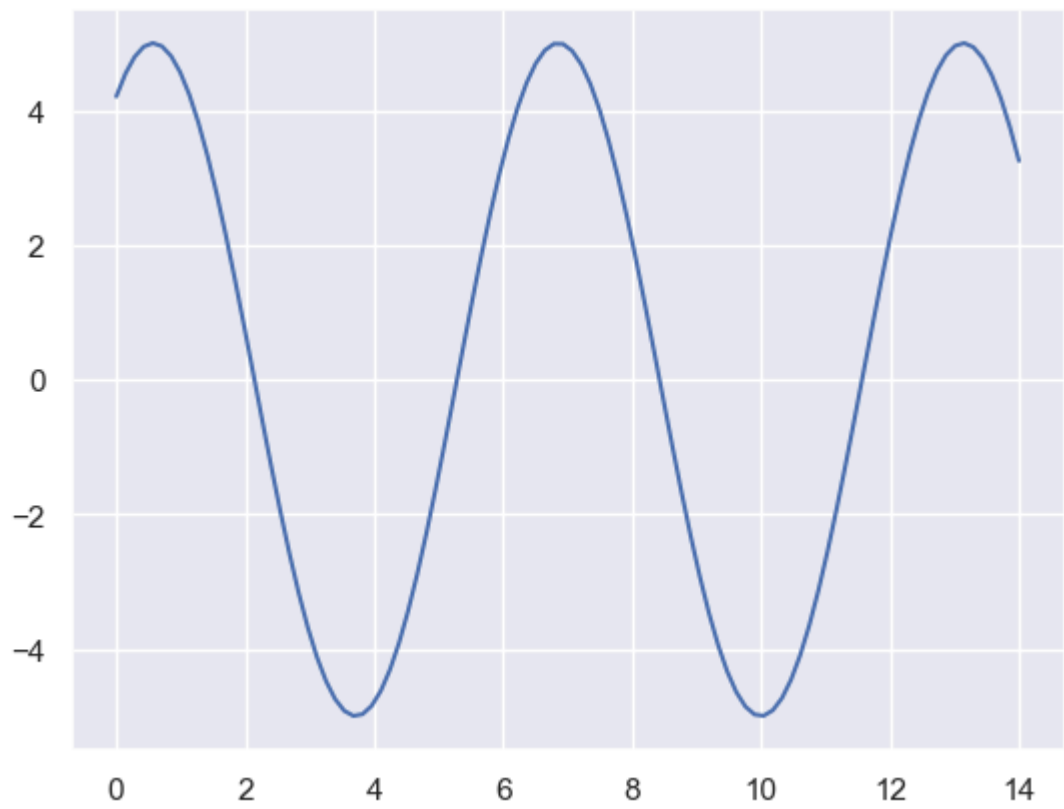


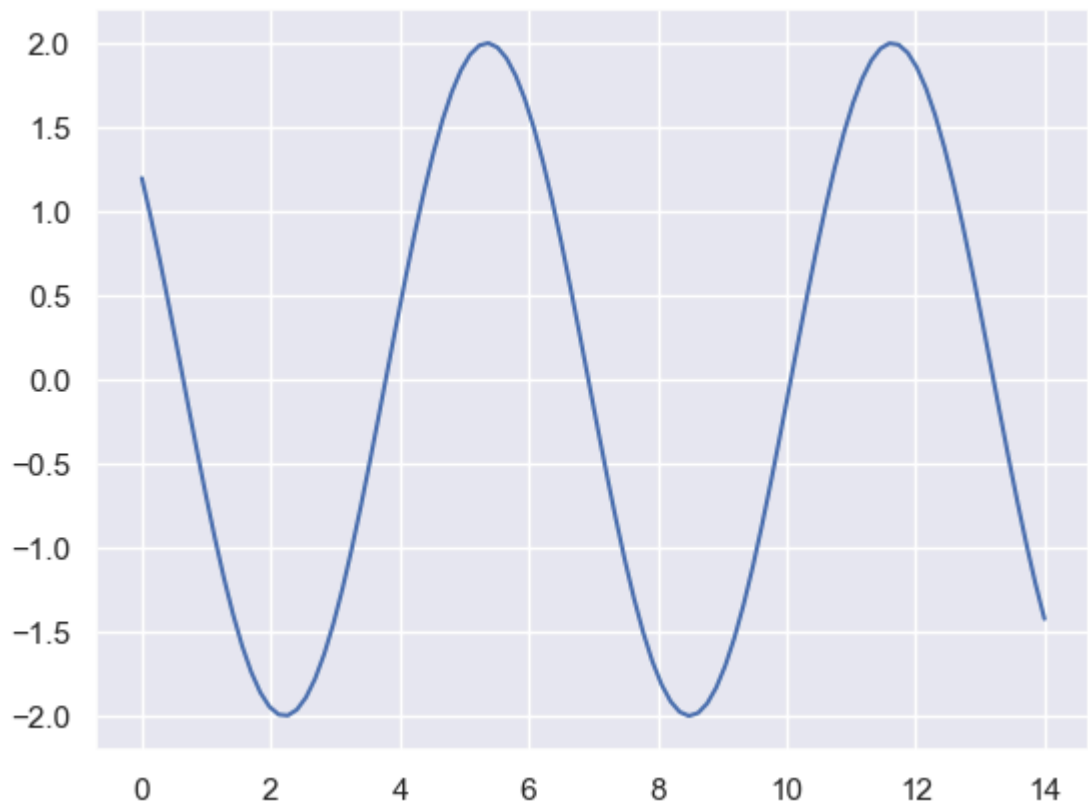
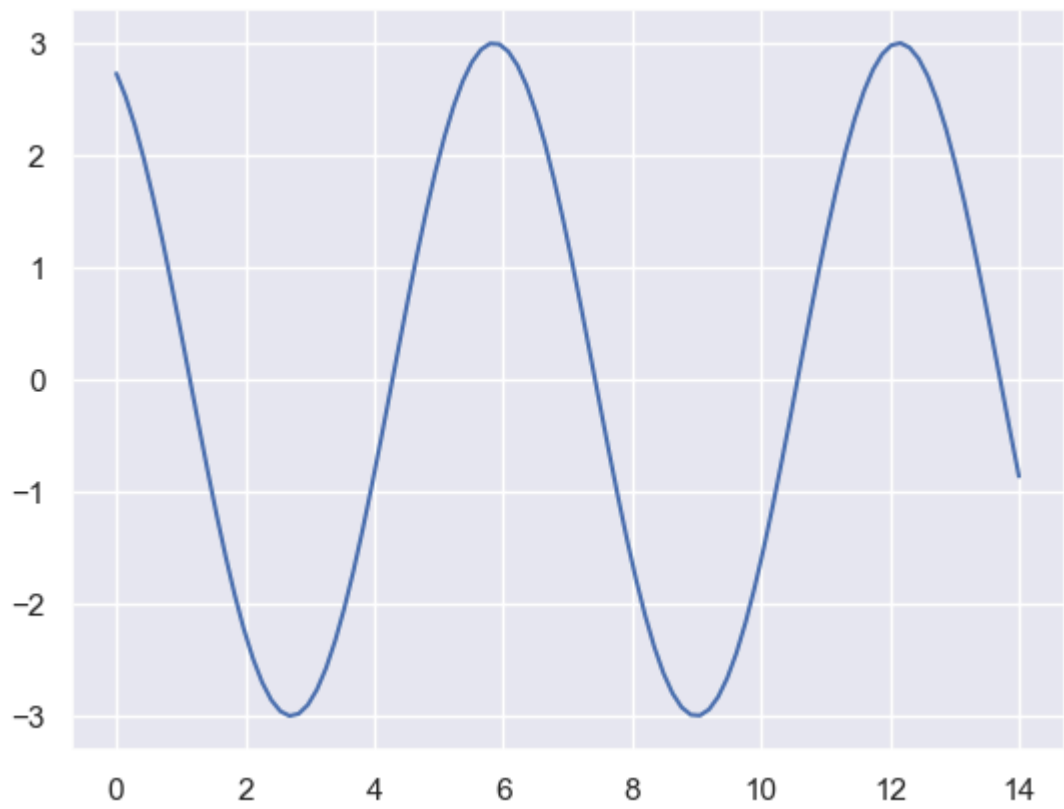


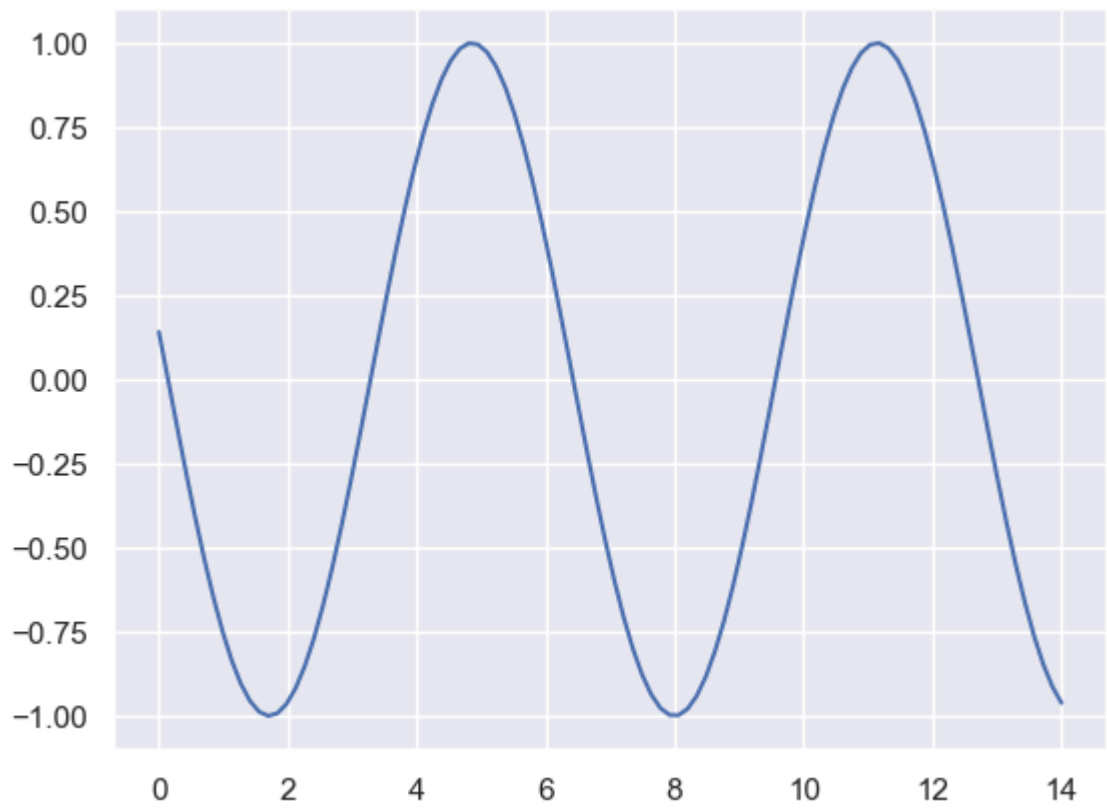
In [100...

```
sns.set()  
sinplot()
```

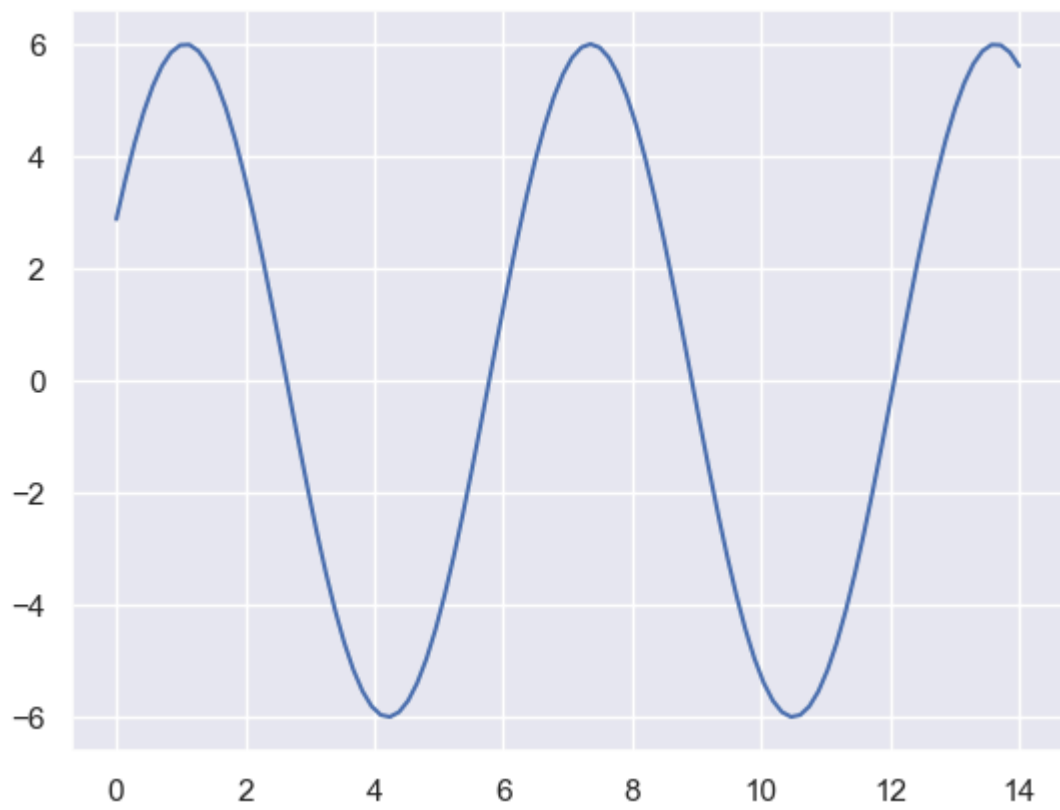


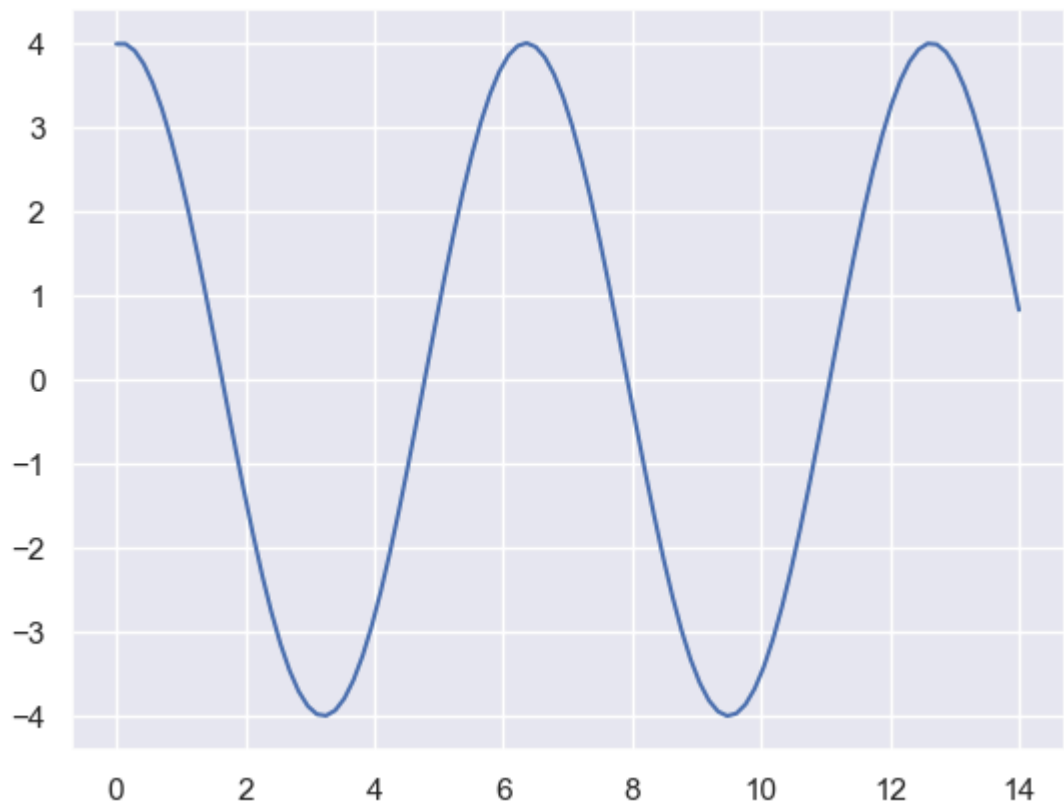
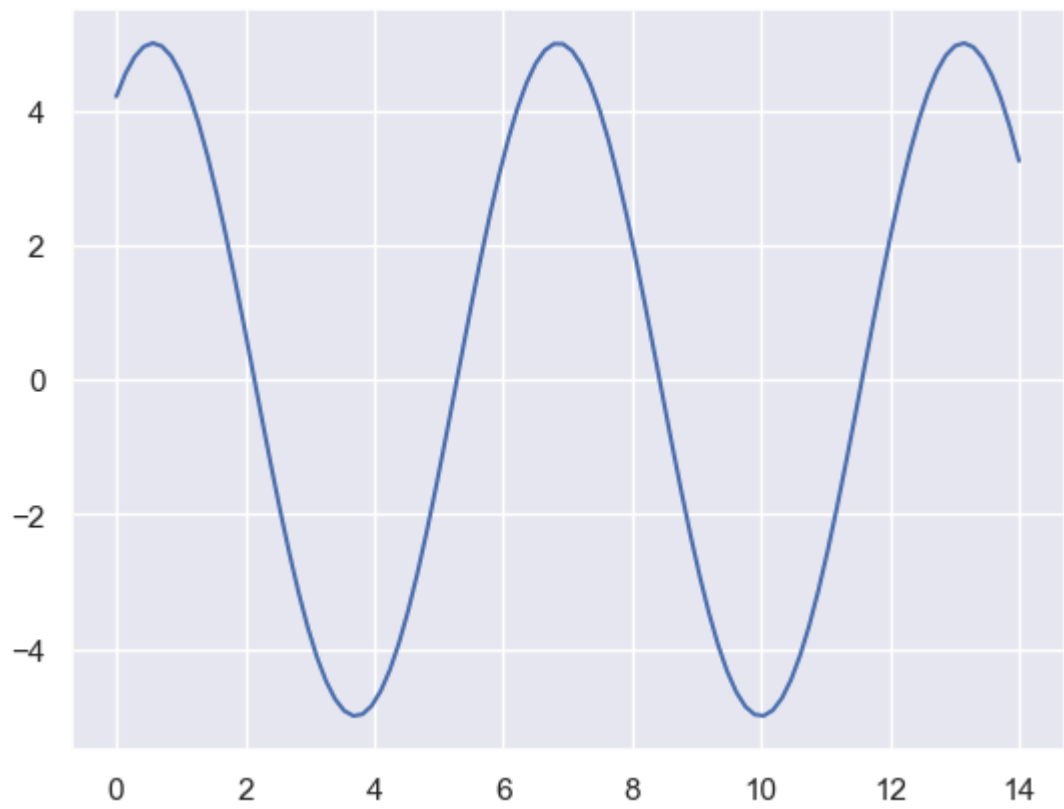


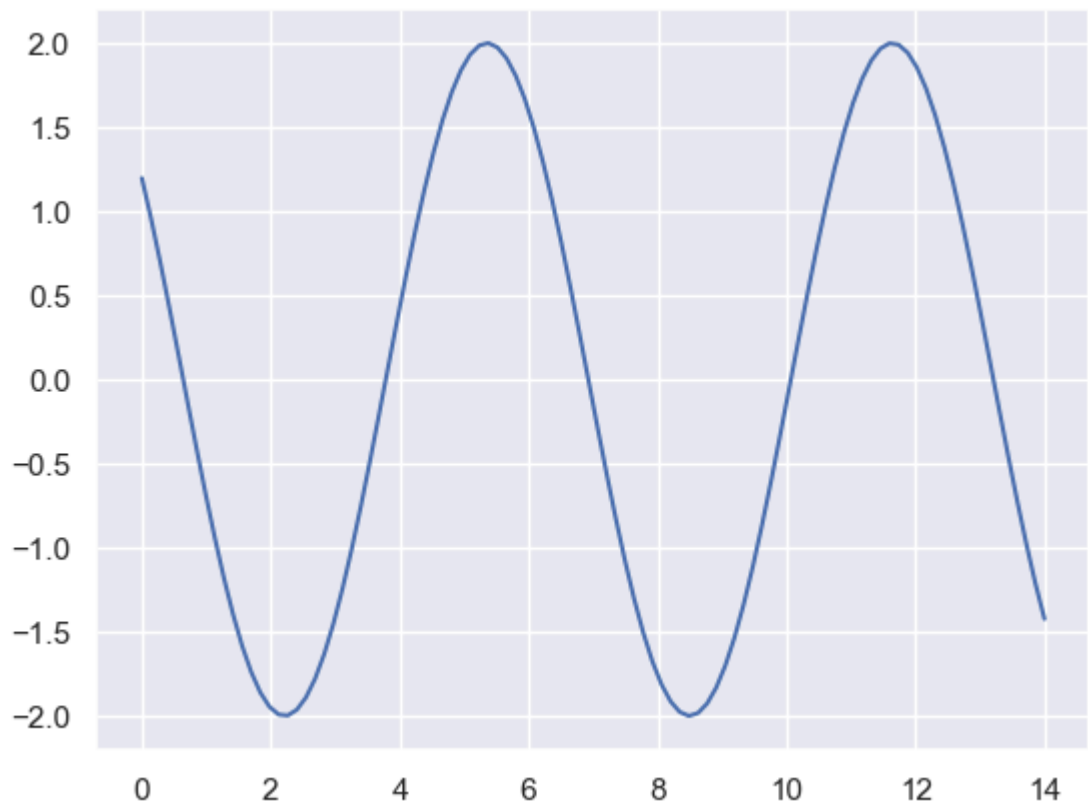
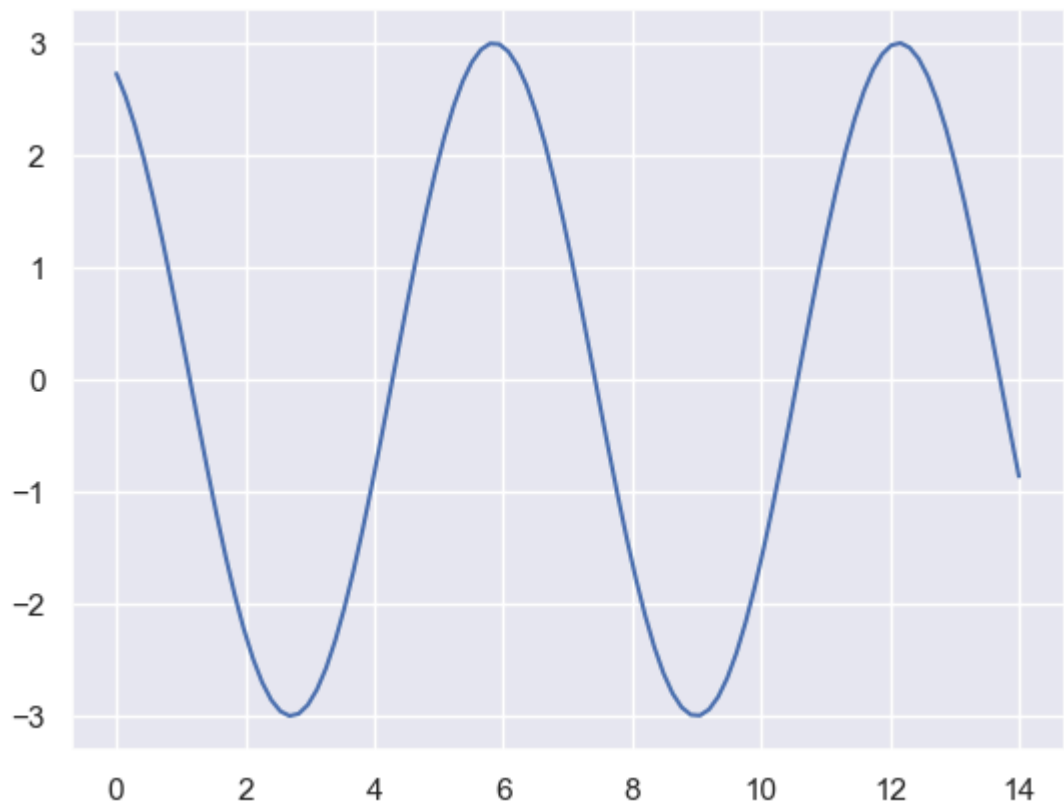


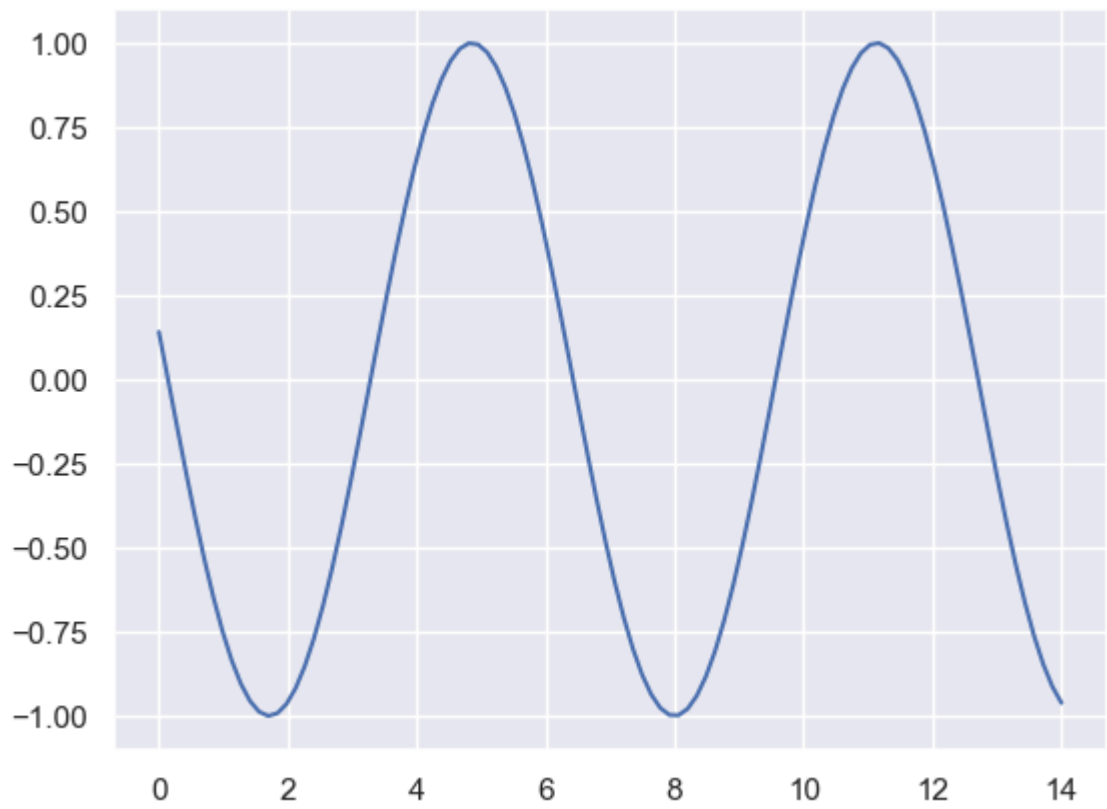


```
In [107... sns.set_style("whitegrid")  
sinplot()
```

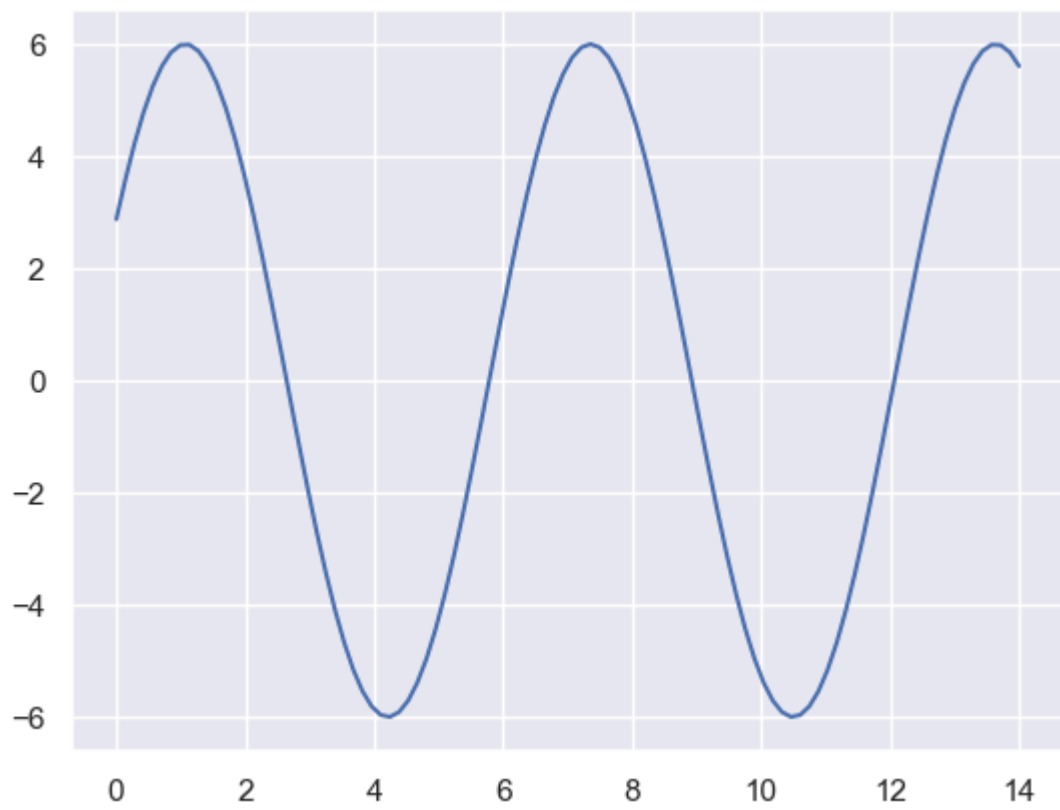


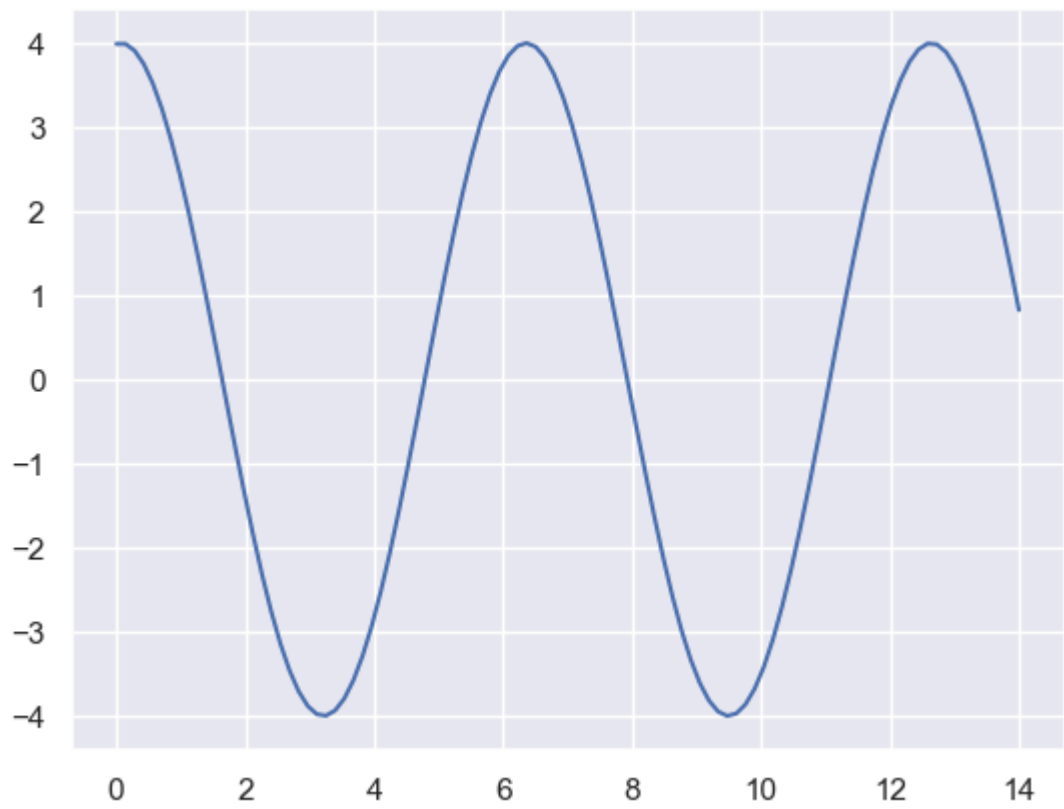
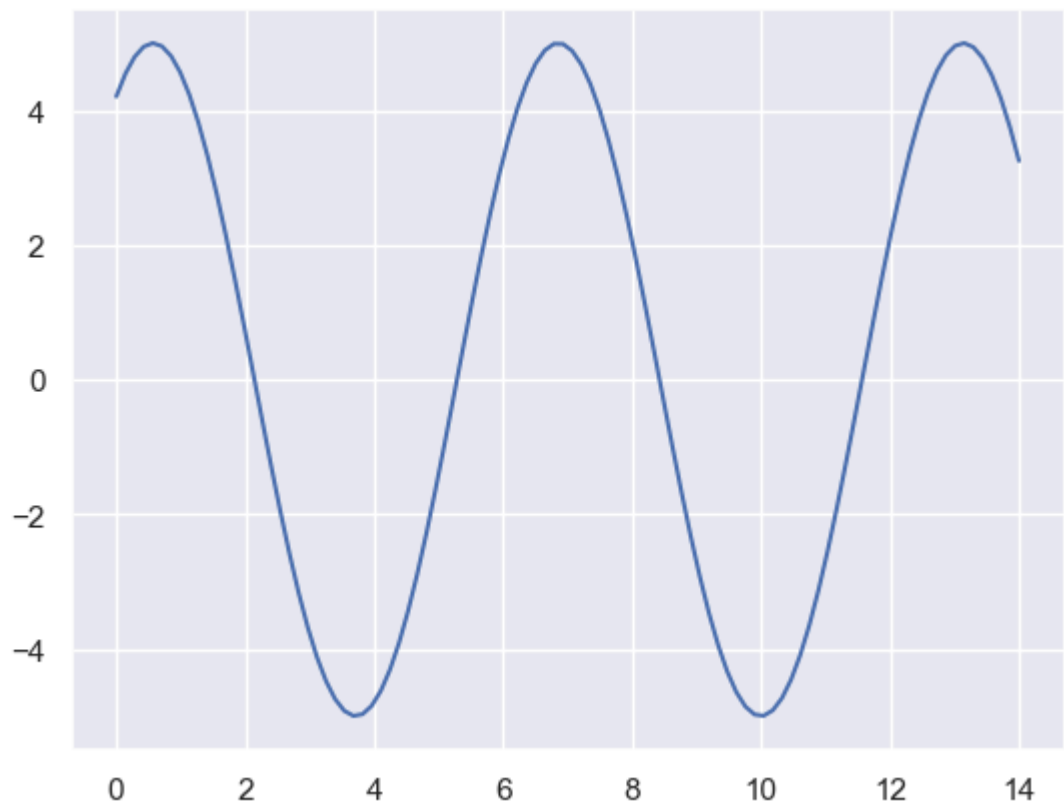


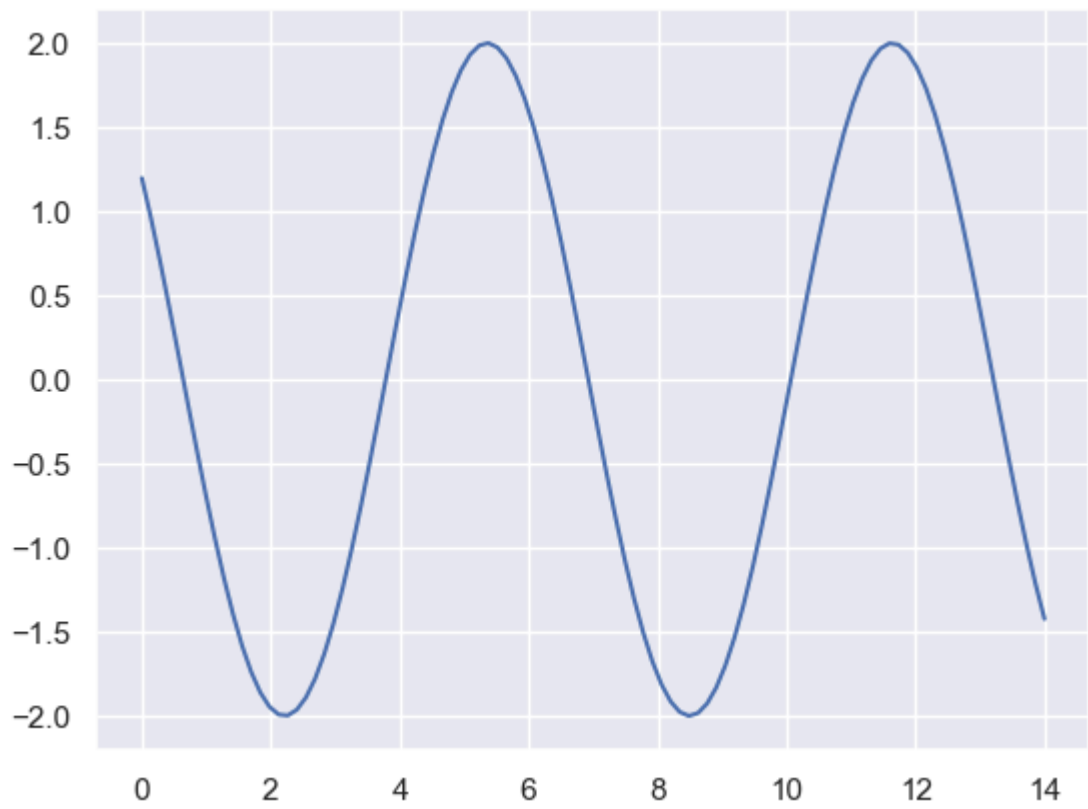
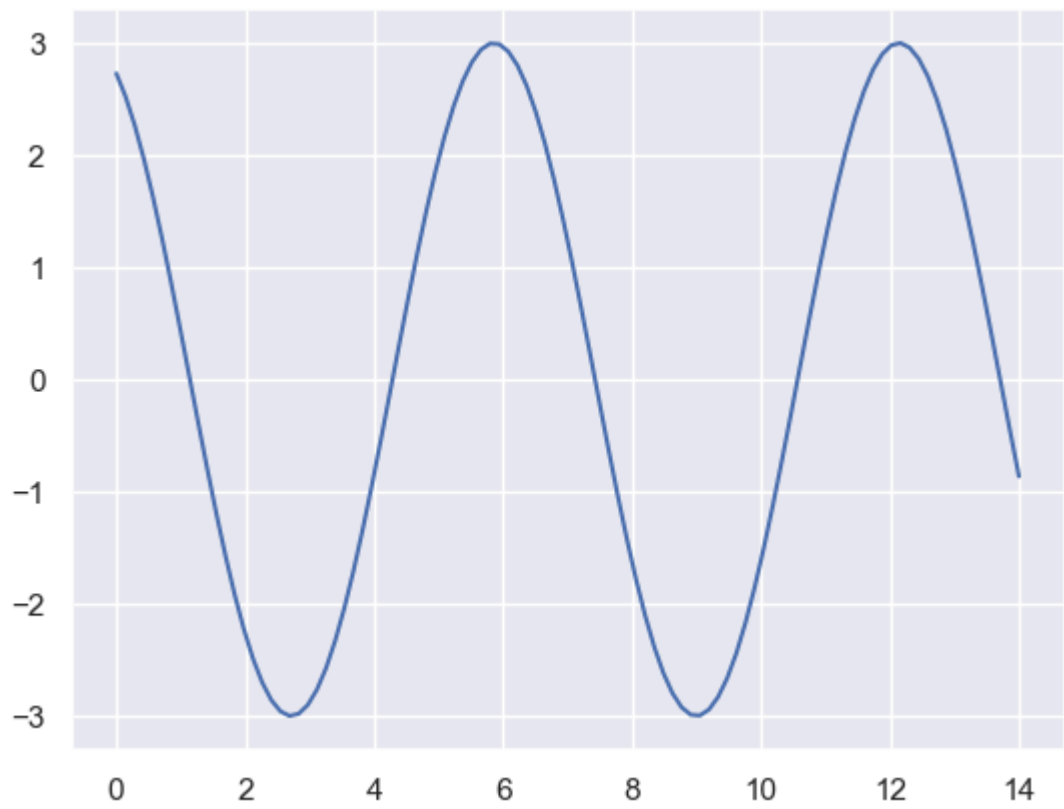


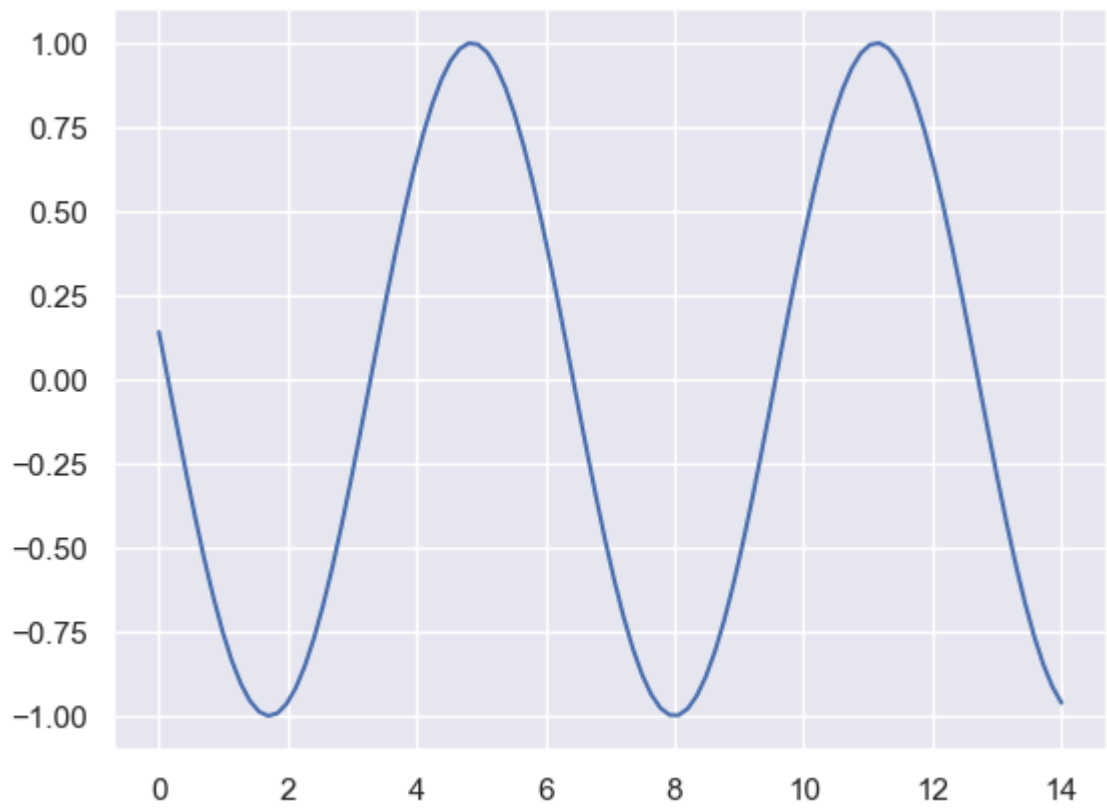


```
In [109... sns.set_style='dark'  
sinplot()
```

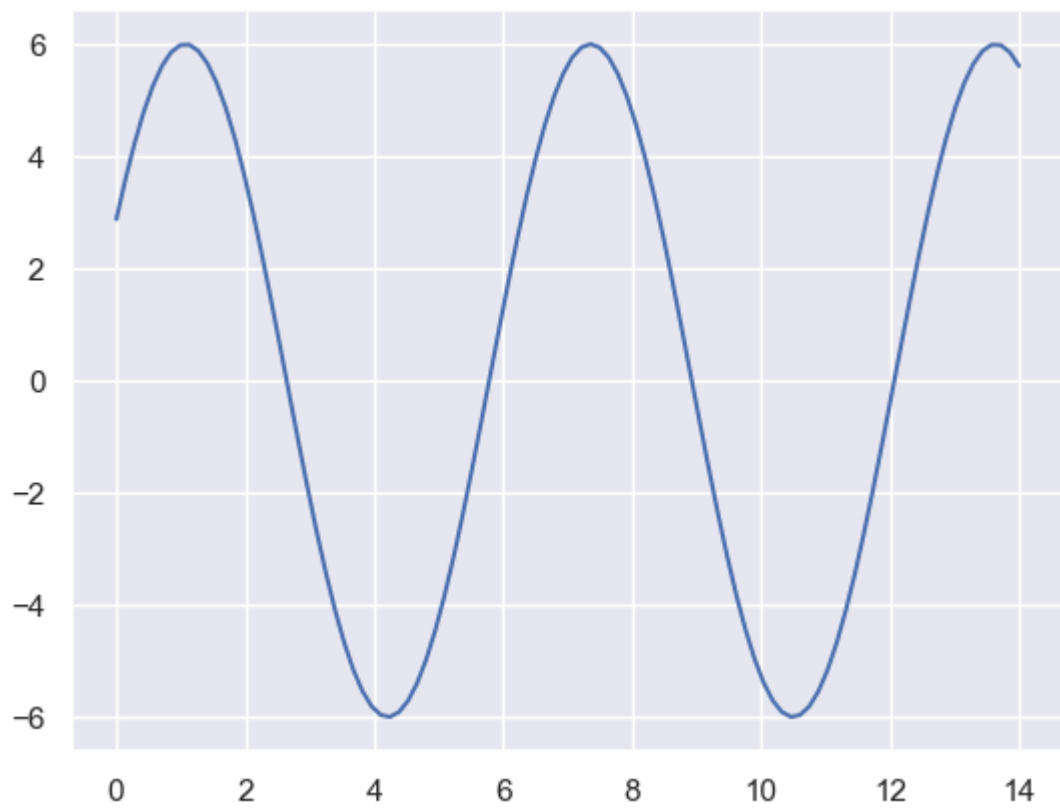


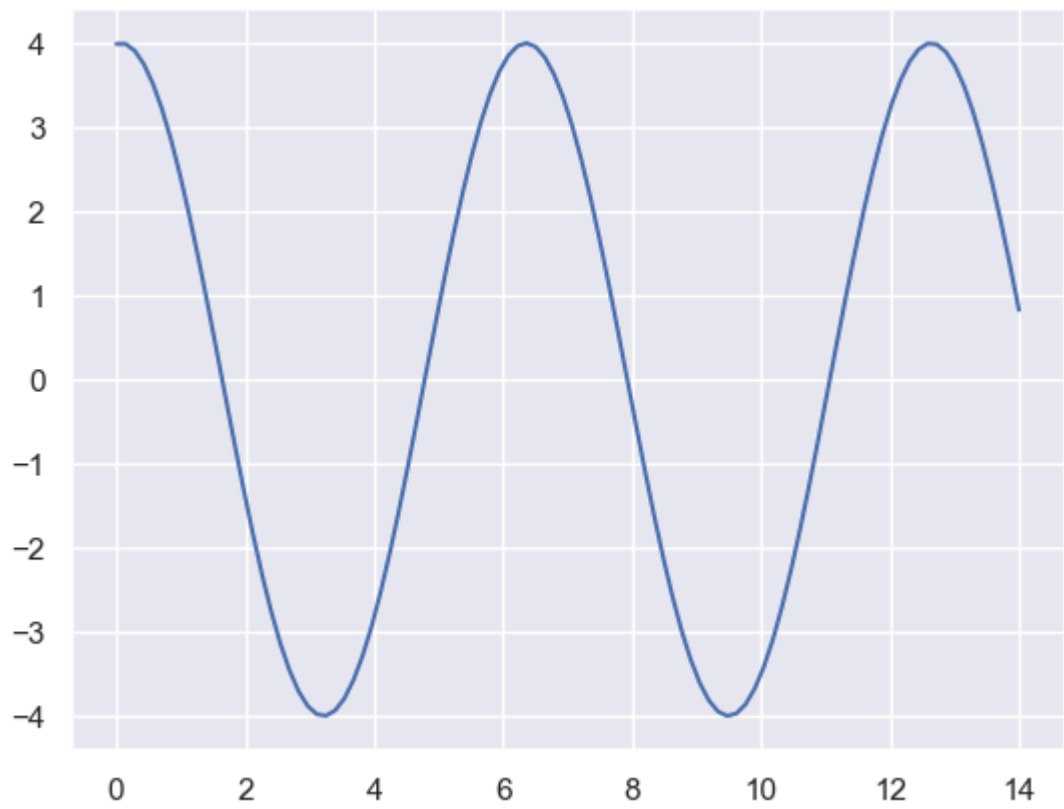
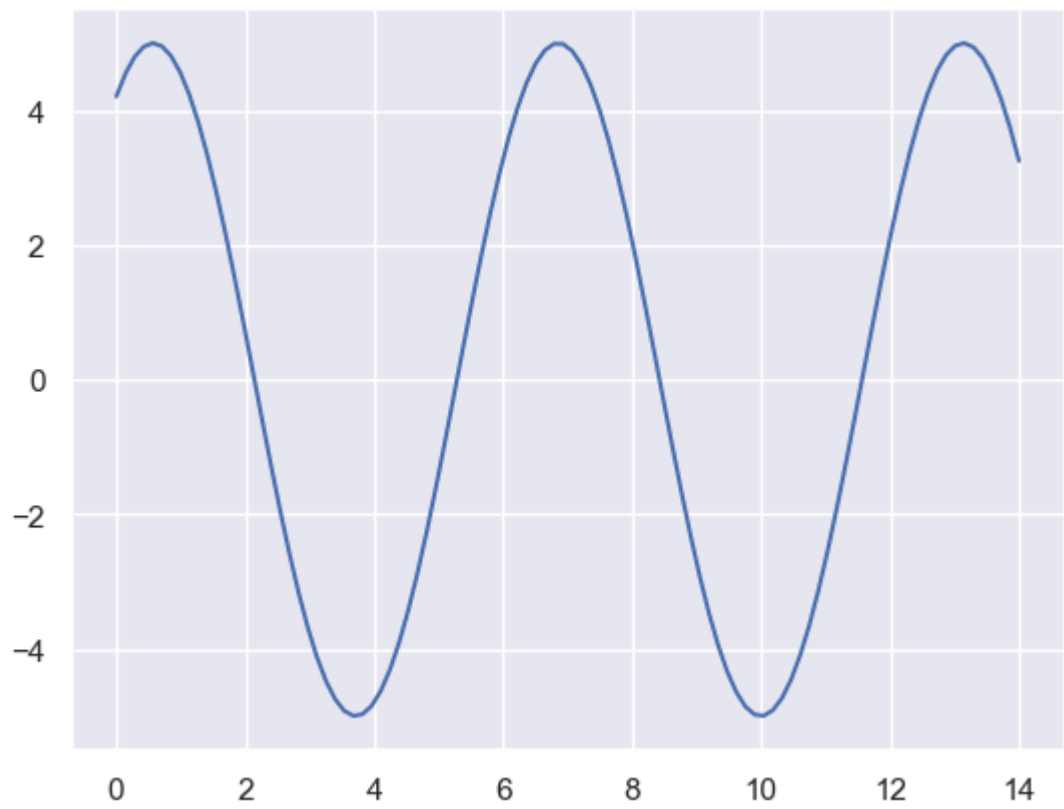


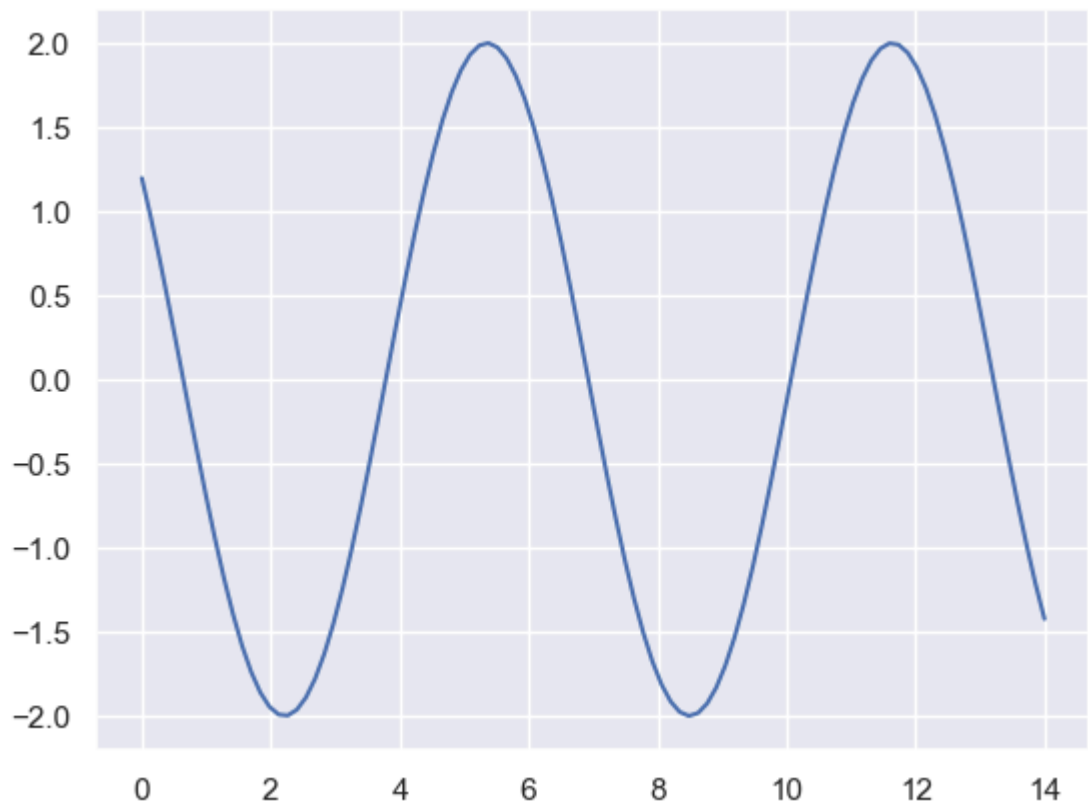
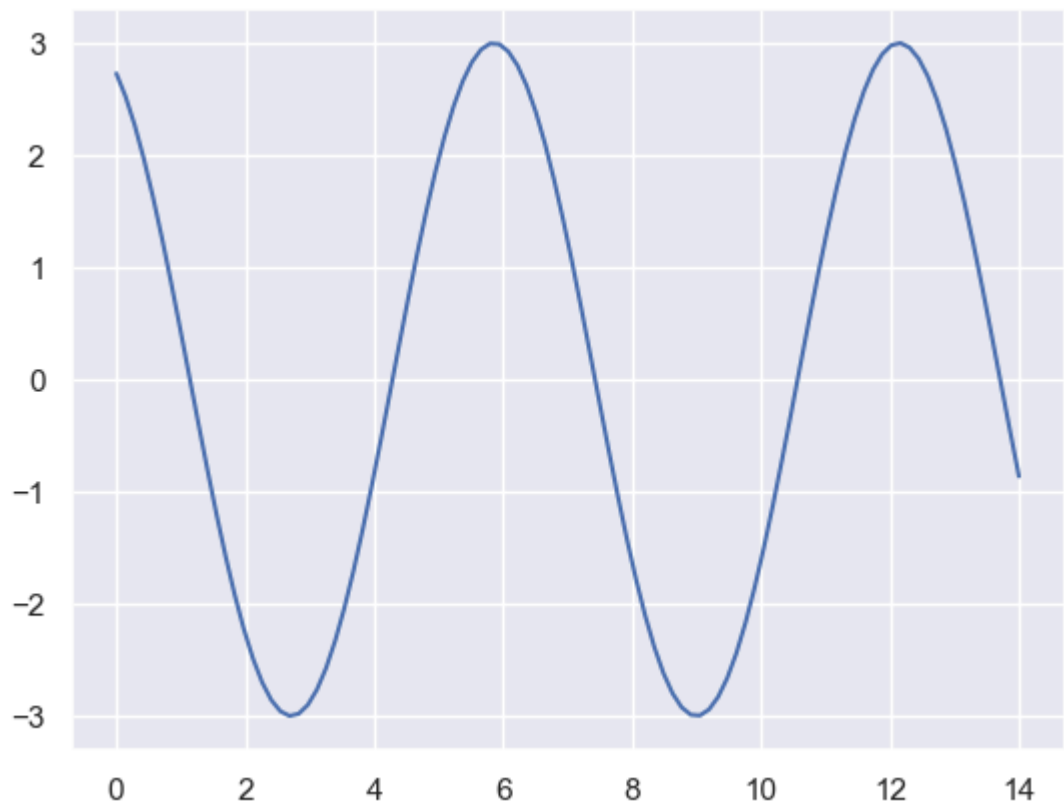


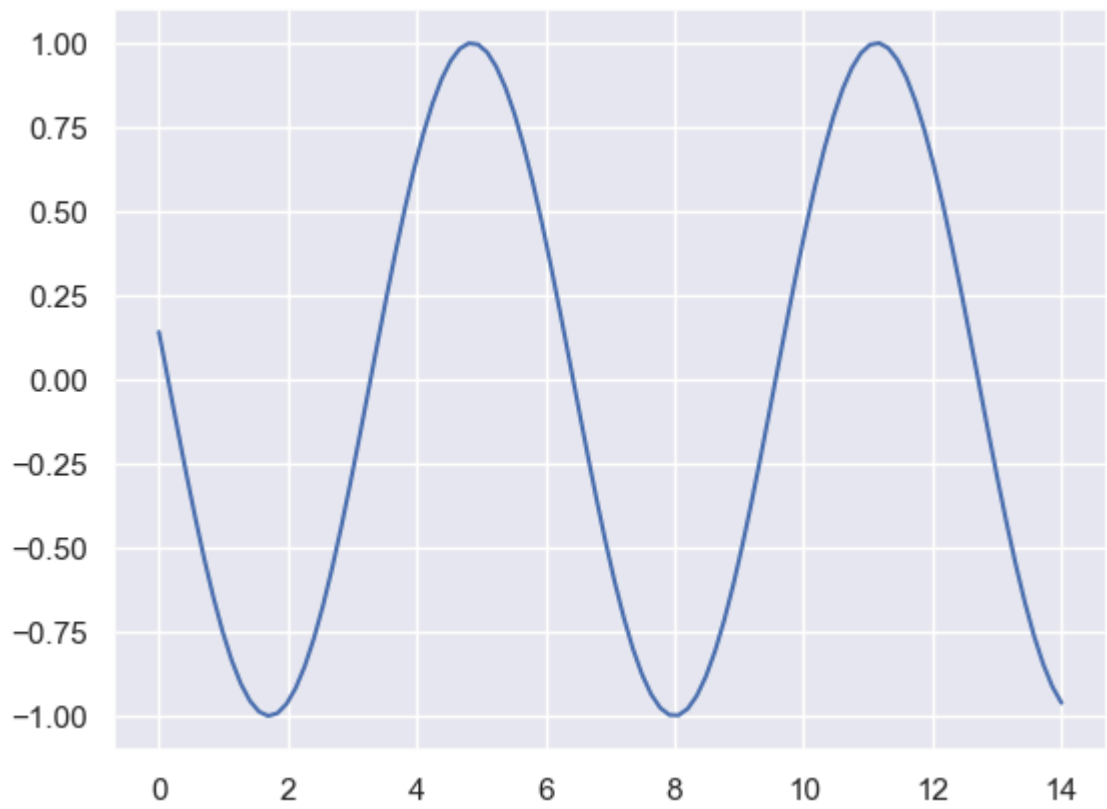


```
In [110... sns.set_style='white'  
sinplot()
```

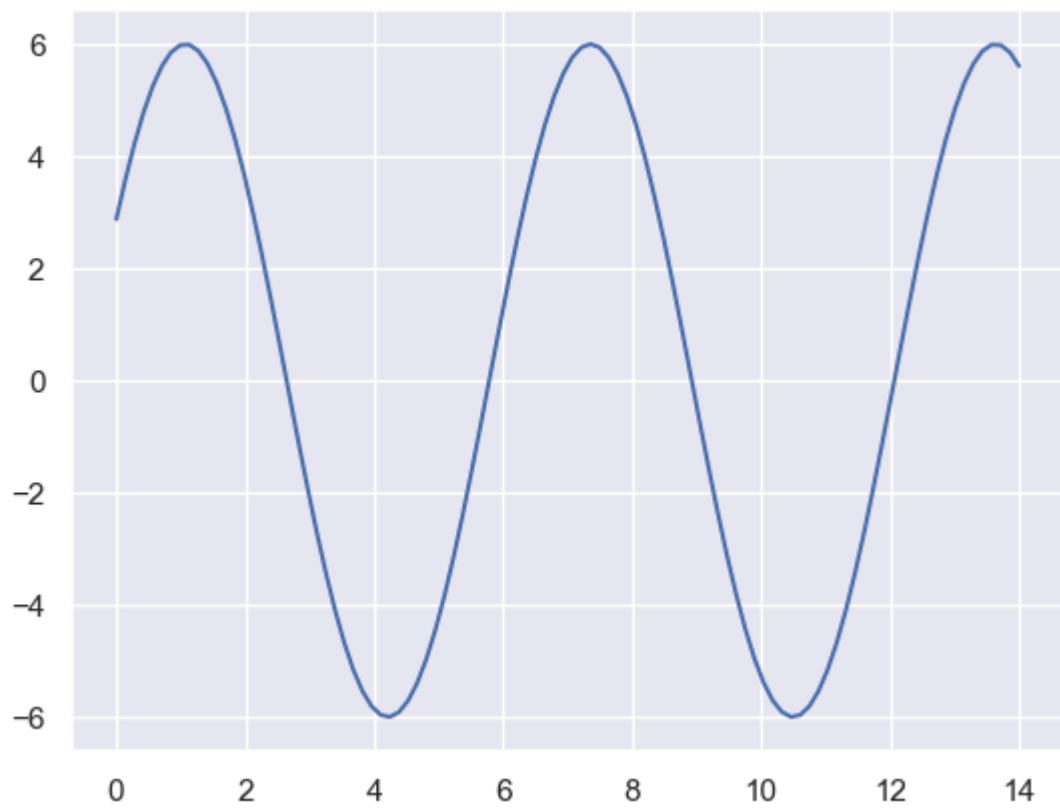


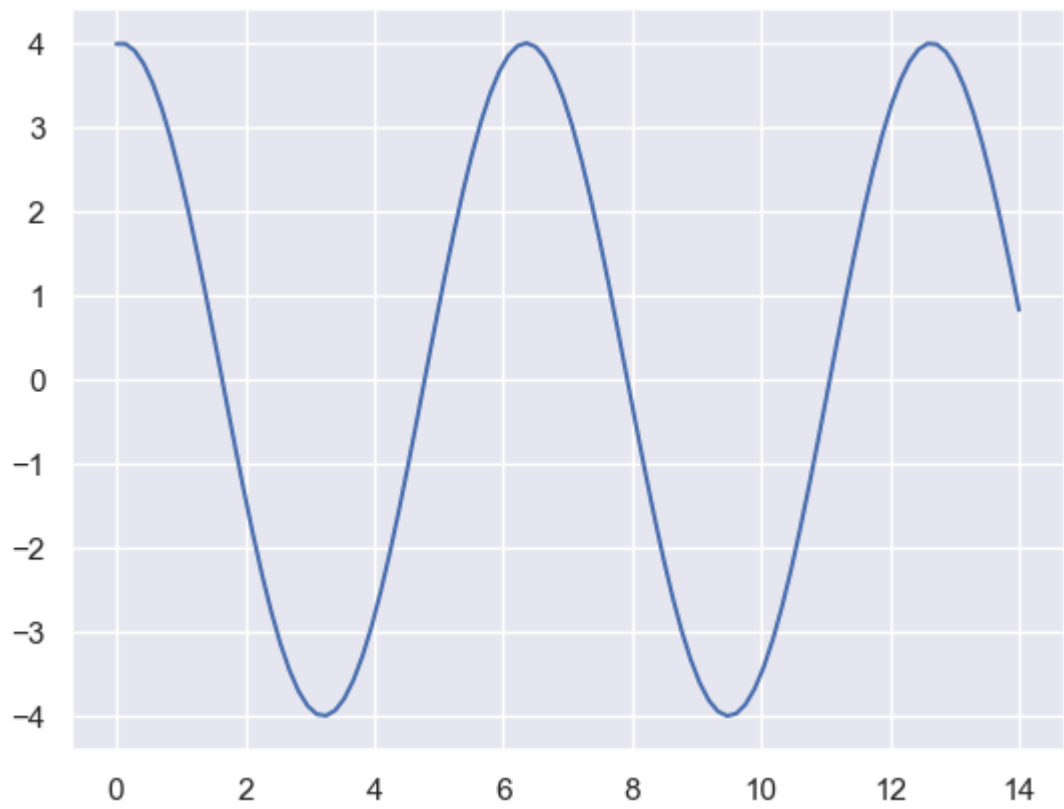
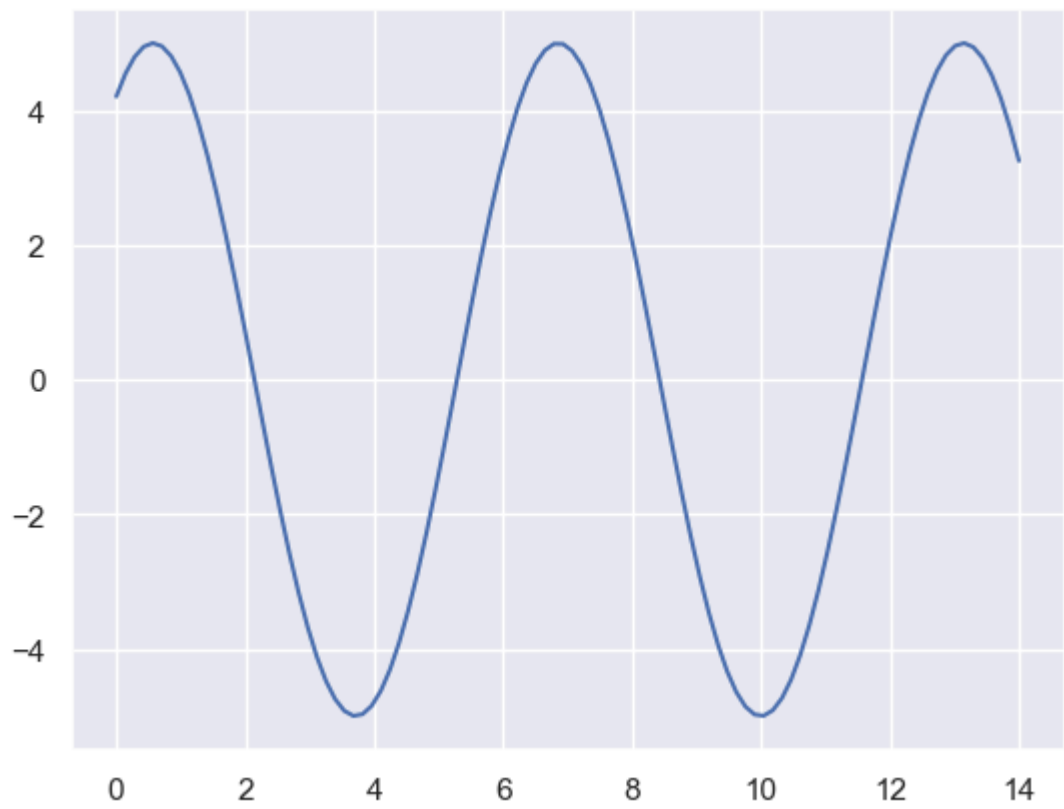


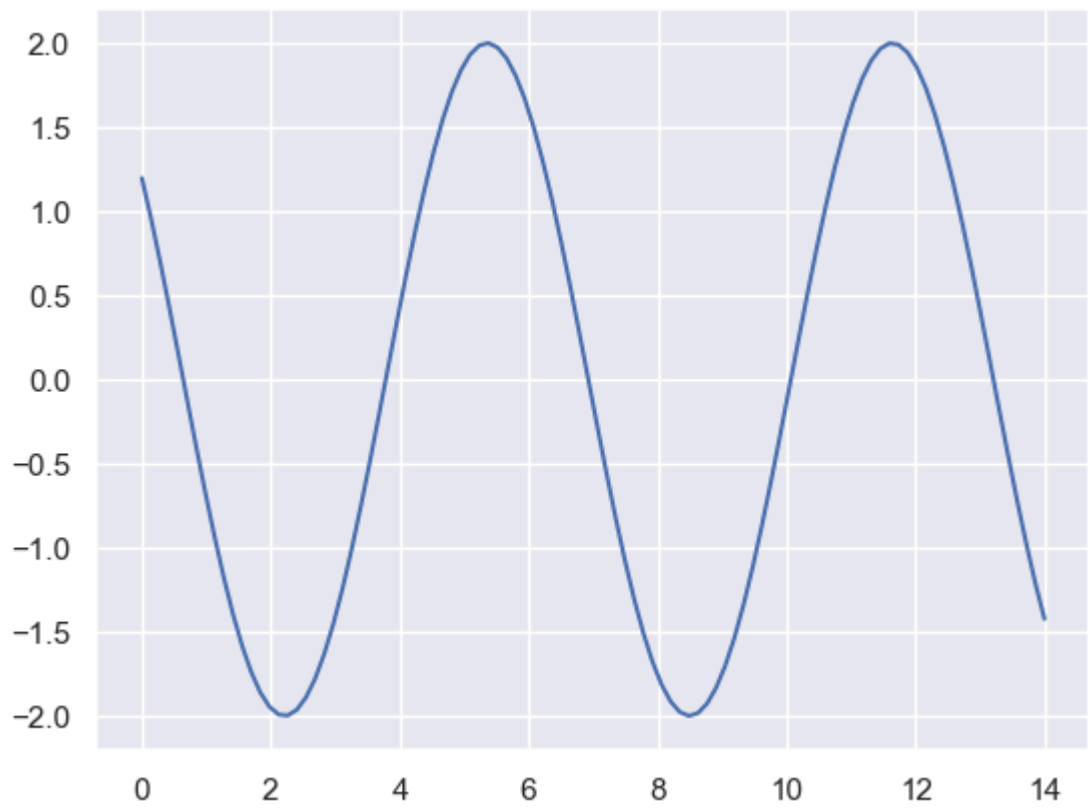
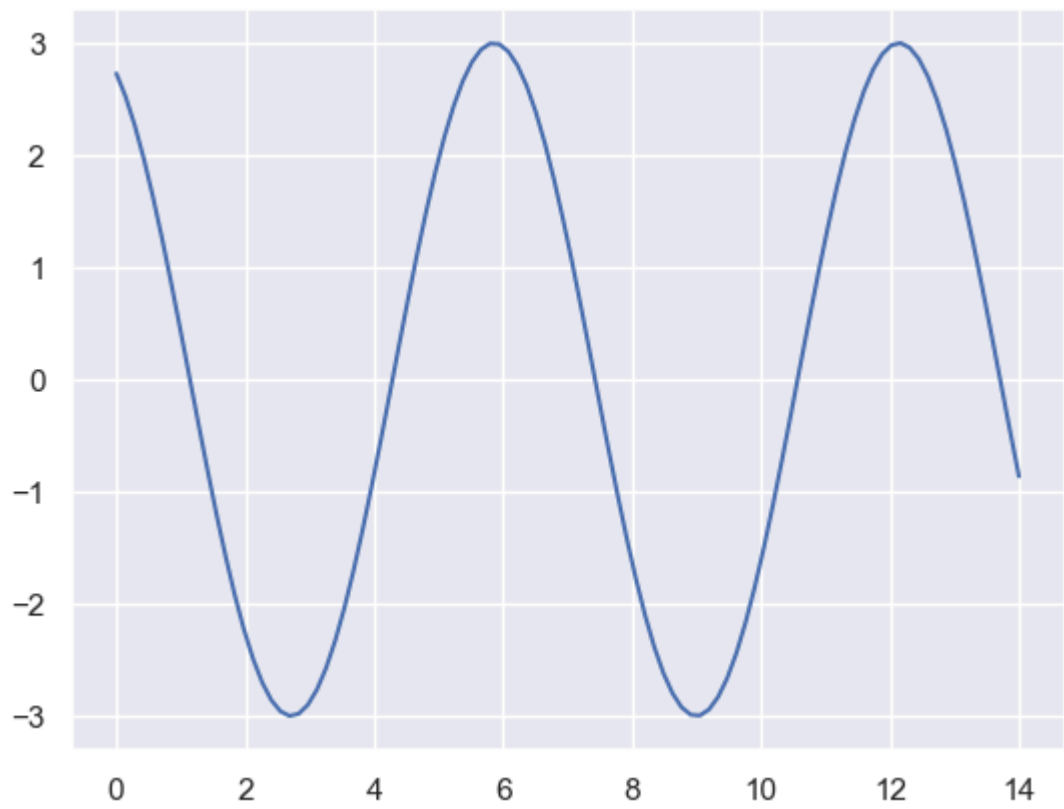


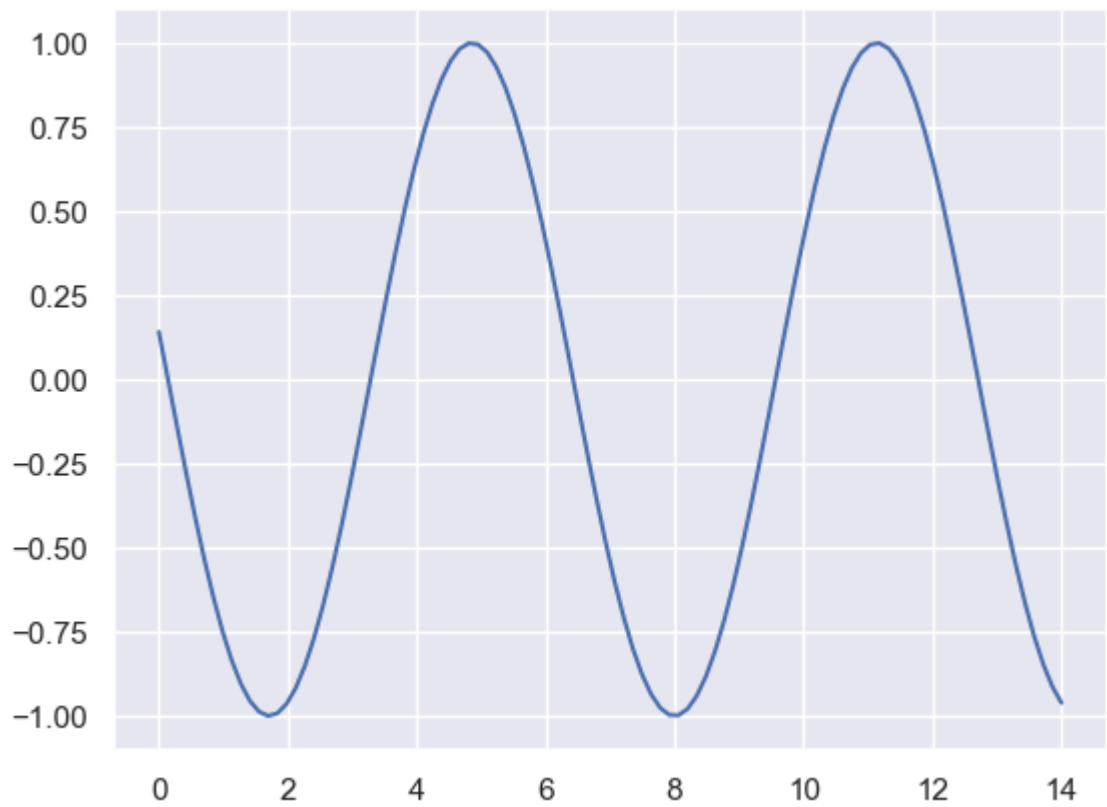


```
In [111... sns.set_style='ticks'  
sinplot()
```









**we have finally completed analysis on
FIFA Sports**

In []: