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Not Trusted

Python 3 (ipykernel) O

Scenario

You are a sports data analyst and you have been tasked with summarizing data from the matches from a previous EuroCup. Your manager would like the following questions answered.

```
In [5]: ## Import the library
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
import plotly.express as px
from ydata_profiling import ProfileReport
%matplotlib inline
import warnings
warnings.filterwarnings("ignore")
plt.style.use('fivethirtyeight')
sns.set()

pd.options.display.float_format = '{:.2f}'.format
pd.options.display.max_rows = None
pd.options.display.max_columns = None
```

```
In [6]: data = pd.read_csv('https://raw.githubusercontent.com/guipsamora/pandas_exercises/master/02_Filtering_%26_Sorting/Euro12/Euro_201
```

```
In [7]: data.sample(10)
```

```
Out[7]:
```

	Team	Goals	Shots on target	Shots off target	Shooting Accuracy	% Goals-to-shots	Total shots (inc. Blocked)	Hit Woodwork	Penalty goals	Penalties not scored	Headed goals	Passes	Passes completed	Passing Accuracy	Touches	Crosses	D
3	England	5	11	18	50.0%	17.2%	40	0	0	0	3	1488	1200	80.6%	2440	58	
0	Croatia	4	13	12	51.9%	16.0%	32	0	0	0	2	1076	828	76.9%	1706	60	
4	France	3	22	24	37.9%	6.5%	65	1	0	0	0	2066	1803	87.2%	2909	55	
1	Czech Republic	4	13	18	41.9%	12.9%	39	0	0	0	0	1565	1223	78.1%	2358	46	
2	Denmark	4	10	10	50.0%	20.0%	27	1	0	0	3	1298	1082	83.3%	1873	43	
10	Portugal	6	22	42	34.3%	9.3%	82	6	0	0	2	1891	1461	77.2%	2958	91	
13	Spain	12	42	33	55.9%	16.0%	100	0	1	0	2	4317	3820	88.4%	5585	69	
12	Russia	5	9	31	22.5%	12.5%	59	2	0	0	1	1602	1345	83.9%	2278	40	
8	Netherlands	2	12	36	25.0%	4.1%	60	2	0	0	0	1556	1381	88.7%	2163	50	
14	Sweden	5	17	19	47.2%	13.8%	39	3	0	0	1	1192	965	80.9%	1806	44	

```
In [8]: data.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 16 entries, 0 to 15
Data columns (total 35 columns):
 #   Column           Non-Null Count  Dtype  
--- 
 0   Team             16 non-null    object 
 1   Goals            16 non-null    int64  
 2   Shots on target 16 non-null    int64  
 3   Shots off target 16 non-null    int64  
 4   Shooting Accuracy 16 non-null    object 
 5   % Goals-to-shots 16 non-null    object 
 6   Total shots (inc. Blocked) 16 non-null    int64  
 7   Hit Woodwork     16 non-null    int64  
 8   Penalty goals    16 non-null    int64  
 9   Penalties not scored 16 non-null    int64  
 10  Headed goals    16 non-null    int64  
 11  Passes           16 non-null    int64  
 12  Passes completed 16 non-null    int64  
 13  Passing Accuracy 16 non-null    object 
 14  Touches          16 non-null    int64  
 15  Crosses          16 non-null    int64  
 16  Dribbles         16 non-null    int64  
 17  Corners Taken   16 non-null    int64  
 18  Tackles          16 non-null    int64  
 19  Clearances       16 non-null    int64  
 20  Interceptions    16 non-null    int64  
 21  Clearances off line 15 non-null    float64 
 22  Clean Sheets     16 non-null    int64  
 23  Blocks            16 non-null    int64  
 24  Goals conceded   16 non-null    int64  
 25  Saves made        16 non-null    int64  
 26  Saves-to-shots ratio 16 non-null    object 
 27  Fouls Won         16 non-null    int64  
 28  Fouls Conceded   16 non-null    int64  
 29  Offside           16 non-null    int64
```

```
29  URTSUES      10 non-null    int64
30  Yellow Cards  16 non-null    int64
31  Red Cards     16 non-null    int64
32  Subs on       16 non-null    int64
33  Subs off      16 non-null    int64
34  Players Used 16 non-null    int64
dtypes: float64(1), int64(29), object(5)
memory usage: 4.5+ KB
```

```
In [9]: data.shape
```

```
Out[9]: (16, 35)
```

```
In [12]: data.duplicated().sum()
```

```
Out[12]: 0
```

How many teams participated in the Euro2012?

```
In [30]: data.Team.nunique()
```

```
Out[30]: 16
```

What is the number of columns in the dataset?

```
In [31]: data.shape[1]
```

```
Out[31]: 35
```

View only the columns Team, Yellow Cards and Red Cards and assign them to a dataframe called discipline.

```
In [35]: discipline=data[['Team','Yellow Cards','Red Cards']]
discipline
```

```
Out[35]:
```

	Team	Yellow Cards	Red Cards
0	Croatia	9	0
1	Czech Republic	7	0
2	Denmark	4	0
3	England	5	0
4	France	6	0
5	Germany	4	0
6	Greece	9	1
7	Italy	16	0
8	Netherlands	5	0
9	Poland	7	1
10	Portugal	12	0
11	Republic of Ireland	6	1
12	Russia	6	0
13	Spain	11	0
14	Sweden	7	0
15	Ukraine	5	0

Sort the teams by Red Cards, then to Yellow Cards.

```
In [41]: discipline = discipline.sort_values(by=["Red Cards", "Yellow Cards"], ascending=False)
discipline
```

```
Out[41]:
```

	Team	Yellow Cards	Red Cards
6	Greece	9	1
9	Poland	7	1
11	Republic of Ireland	6	1
7	Italy	16	0
10	Portugal	12	0
13	Spain	11	0
0	Croatia	9	0
1	Czech Republic	7	0
14	Sweden	7	0
4	France	6	0
12	Russia	6	0
3	England	5	0
8	Netherlands	5	0
15	Ukraine	5	0

2	Denmark	4	0
5	Germany	4	0

Calculate the mean Yellow Cards given per Team.

In [43]: `discipline.groupby("Team")['Yellow Cards'].mean().reset_index()`

Out[43]:

Team	Yellow Cards
0 Croatia	9.00
1 Czech Republic	7.00
2 Denmark	4.00
3 England	5.00
4 France	6.00
5 Germany	4.00
6 Greece	9.00
7 Italy	16.00
8 Netherlands	5.00
9 Poland	7.00
10 Portugal	12.00
11 Republic of Ireland	6.00
12 Russia	6.00
13 Spain	11.00
14 Sweden	7.00
15 Ukraine	5.00

Filter teams that scored more than 6 goals.

In [112]: `data.query('Goals > 6')`

Out[112]:

Team	Goals	Shots on target	Shots off target	Shooting Accuracy	% Goals-to-shots	Total shots (inc. Blocked)	Woodwork	Hit	Penalty goals	Penalties not scored	Headed goals	Passes	Passes completed	Passing Accuracy	Touches	Crosses	Drib
5 Germany	10	32	32	47.8%	15.6%	80	2	1	0	2	2774	2427	87.4%	3761	101		
13 Spain	12	42	33	55.9%	16.0%	100	0	1	0	2	4317	3820	88.4%	5585	69		

Select the teams that start with the letter G.

In [57]: `data[data.Team.str.startswith('G')]`

Out[57]:

Team	Goals	Shots on target	Shots off target	Shooting Accuracy	% Goals-to-shots	Total shots (inc. Blocked)	Woodwork	Hit	Penalty goals	Penalties not scored	Headed goals	Passes	Passes completed	Passing Accuracy	Touches	Crosses	Drib
5 Germany	10	32	32	47.8%	15.6%	80	2	1	0	2	2774	2427	87.4%	3761	101		
6 Greece	5	8	18	30.7%	19.2%	32	1	1	1	0	1187	911	76.7%	2016	52		

Select the first 7 columns.

In [61]: `data[data.columns[:7]]`

Out[61]:

Team	Goals	Shots on target	Shots off target	Shooting Accuracy	% Goals-to-shots	Total shots (inc. Blocked)
0 Croatia	4	13	12	51.9%	16.0%	32
1 Czech Republic	4	13	18	41.9%	12.9%	39
2 Denmark	4	10	10	50.0%	20.0%	27
3 England	5	11	18	50.0%	17.2%	40
4 France	3	22	24	37.9%	6.5%	65
5 Germany	10	32	32	47.8%	15.6%	80
6 Greece	5	8	18	30.7%	19.2%	32
7 Italy	6	34	45	43.0%	7.5%	110
8 Netherlands	2	12	36	25.0%	4.1%	60
9 Poland	2	15	23	39.4%	5.2%	48
10 Portugal	6	22	42	34.3%	9.3%	82
11 Republic of Ireland	1	7	12	36.8%	5.2%	28
12 Russia	5	9	31	22.5%	12.5%	59
13 Spain	12	42	33	55.9%	16.0%	100
14 Sweden	5	17	19	47.2%	13.8%	39

```
15      Ukraine    2       7      26     21.2%    6.0%      38
```

Select all columns except the last 3.

```
In [65]: data[data.columns[:-3]]
```

Out[65]:

	Team	Goals	Shots on target	Shots off target	Shooting Accuracy	% Goals-to-shots	Total shots (inc. Blocked)	Woodwork	Hit	Penalty goals	Penalties not scored	Headed goals	Passes	Passes completed	Passing Accuracy	Touches	Crosses	D
0	Croatia	4	13	12	51.9%	16.0%	32	0	0	0	0	2	1076	828	76.9%	1706	60	
1	Czech Republic	4	13	18	41.9%	12.9%	39	0	0	0	0	0	1565	1223	78.1%	2358	46	
2	Denmark	4	10	10	50.0%	20.0%	27	1	0	0	0	3	1298	1082	83.3%	1873	43	
3	England	5	11	18	50.0%	17.2%	40	0	0	0	0	3	1488	1200	80.6%	2440	58	
4	France	3	22	24	37.9%	6.5%	65	1	0	0	0	0	2066	1803	87.2%	2909	55	
5	Germany	10	32	32	47.8%	15.6%	80	2	1	0	2	2774	2427	87.4%	3761	101		
6	Greece	5	8	18	30.7%	19.2%	32	1	1	1	0	1	1187	911	76.7%	2016	52	
7	Italy	6	34	45	43.0%	7.5%	110	2	0	0	0	2	3016	2531	83.9%	4363	75	
8	Netherlands	2	12	36	25.0%	4.1%	60	2	0	0	0	0	1556	1381	88.7%	2163	50	
9	Poland	2	15	23	39.4%	5.2%	48	0	0	0	1	1	1059	852	80.4%	1724	55	
10	Portugal	6	22	42	34.3%	9.3%	82	6	0	0	0	2	1891	1461	77.2%	2958	91	
11	Republic of Ireland	1	7	12	36.8%	5.2%	28	0	0	0	1	1	851	606	71.2%	1433	43	
12	Russia	5	9	31	22.5%	12.5%	59	2	0	0	1	1	1602	1345	83.9%	2278	40	
13	Spain	12	42	33	55.9%	16.0%	100	0	1	0	2	4317	3820	88.4%	5585	69		
14	Sweden	5	17	19	47.2%	13.8%	39	3	0	0	1	1	1192	965	80.9%	1806	44	
15	Ukraine	2	7	26	21.2%	6.0%	38	0	0	0	2	1276	1043	81.7%	1894	33		

Present only the Shooting Accuracy from England, Italy and Russia.

```
In [106]: Shooting=data[['Team','Shooting Accuracy']]
```

```
In [105]: Shooting[Shooting['Team'].isin(['England', 'Italy', 'Russia'])]
```

Out[105]:

Team	Shooting Accuracy	
3	England	50.0%
7	Italy	43.0%
12	Russia	22.5%