# Weekly Assignment 5a

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Q.1 We'll We'll be working with the 120 years of Olympic History dataset. Download the dataset "athlete\_events.csv" and perform the following:

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

    olympic_data = pd.read_csv("../dataFiles/athlete_events.csv")
    olympic_data
```

ut[1]:		ID	Name	Sex	Age	Height	Weight	Team	NOC	Games	Year	Season
	0	1	A Dijiang	М	24.0	180.0	80.0	China	CHN	1992 Summer	1992	Summer
	1	2	A Lamusi	М	23.0	170.0	60.0	China	CHN	2012 Summer	2012	Summer
	2	3	Gunnar Nielsen Aaby	М	24.0	NaN	NaN	Denmark	DEN	1920 Summer	1920	Summer
	3	4	Edgar Lindenau Aabye	М	34.0	NaN	NaN	Denmark/Sweden	DEN	1900 Summer	1900	Summer
	4	5	Christine Jacoba Aaftink	F	21.0	185.0	82.0	Netherlands	NED	1988 Winter	1988	Winter
	•••											
	271111	135569	Andrzej ya	М	29.0	179.0	89.0	Poland-1	POL	1976 Winter	1976	Winter
	271112	135570	Piotr ya	М	27.0	176.0	59.0	Poland	POL	2014 Winter	2014	Winter
	271113	135570	Piotr ya	М	27.0	176.0	59.0	Poland	POL	2014 Winter	2014	Winter

	ID	Name	Sex	Age	Height	Weight	Team	NOC	Games	Year	Season
271114	135571	Tomasz Ireneusz ya	М	30.0	185.0	96.0	Poland	POL	1998 Winter	1998	Winter
271115	135571	Tomasz Ireneusz ya	М	34.0	185.0	96.0	Poland	POL	2002 Winter	2002	Winter

271116 rows × 15 columns

In [2]:

#Q.1.a Filter the DataFrame to only include the rows corresponding to medal winners fro
olympic\_filtered = olympic\_data[olympic\_data['Year'] >= 2016]
olympic\_filtered = olympic\_filtered[olympic\_filtered['Medal'].notna()]
olympic\_filtered

Out[2]: ID Height Weight NOC Games Year Name Sex Age **Team** Season Giovanni 2016 21.0 158 62 M 198.0 90.0 Italy ITA 2016 Summer Summer Abagnale **Patimat** 2016 161 65 21.0 165.0 2016 49.0 Azerbaijan AZE Summer Summer Abakarova 2016 175 73 Luc Abalo 31.0 182.0 86.0 FRA 2016 Summer M France Summer Saeid Morad 2016 2016 Summer 450 250 26.0 170.0 0.08 Iran Summer Abdevali Denis 2016 794 455 Mikhaylovich 24.0 161.0 62.0 **RUS** 2016 Summer M Russia Summer Ablyazin 2016 **269511** 134857 Zhu Ting 21.0 198.0 78.0 China 2016 Summer CHN Summer 2016 **270111** 135132 28.0 186.0 72.0 Serbia SRB 2016 Summer Bojana ivkovi Summer Shakhobiddin 2016 **270281** 135205 Shokirovich 23.0 UZB M 169.0 52.0 Uzbekistan 2016 Summer Summer Zoirov 2016 73.0 SRB 2016 Summer **270370** 135245 Milenko Zori 27.0 179.0 Serbia M Summer

	ID	Name	Sex	Age	Height	Weight	Team	NOC	Games	Year	Season	
271019	135525	Martin Zwicker	М	29.0	175.0	64.0	Germany	GER	2016 Summer	2016	Summer	l Ji

2023 rows × 15 columns

```
#Q.1.b Find out the number of medals awarded in 2016 for each sport.
awards_per_sport = olympic_filtered[['Sport','Year','Medal']]
awards_per_sport.groupby(['Sport','Year']).count()
```

Out[3]: Medal

Sport	Year	
Archery	2016	24
Athletics	2016	192
Badminton	2016	24
Basketball	2016	72
Beach Volleyball	2016	12
Boxing	2016	51
Canoeing	2016	82
Cycling	2016	84
Diving	2016	36
Equestrianism	2016	45
Fencing	2016	65
Football	2016	106
Golf	2016	6
Gymnastics	2016	66
Handball	2016	89
Hockey	2016	99
Judo	2016	56
<b>Modern Pentathlon</b>	2016	6
Rhythmic Gymnastics	2016	18
Rowing	2016	144
Rugby Sevens	2016	74
Sailing	2016	45
Shooting	2016	45
Swimming	2016	191

#### Medal

Sport	Year	
Synchronized Swimming	2016	32
Table Tennis	2016	24
Taekwondo	2016	32
Tennis	2016	24
Trampolining	2016	6
Triathlon	2016	6
Volleyball	2016	72
Water Polo	2016	78
Weightlifting	2016	45
Wrestling	2016	72

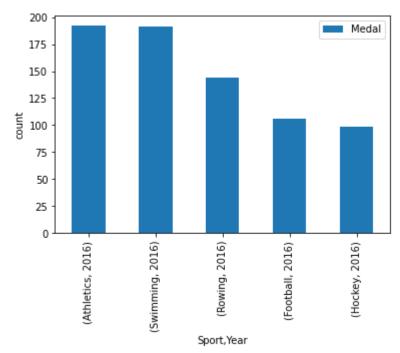
```
#Q.1.c Filter the DataFrame one more time to only include the records for the top five awards_count = awards_per_sport.groupby(['Sport', 'Year']).count() awards_per_sport_sorted = awards_count.sort_values('Medal', ascending=False) awards_per_sport_sorted.head(5)
```

### Out[4]: Medal

Sport	Year	
Athletics	2016	192
Swimming	2016	191
Rowing	2016	144
Football	2016	106
Hockey	2016	99

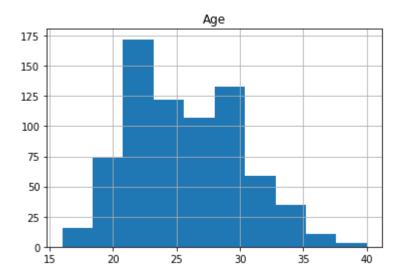
```
#Q.1.d Generate a bar plot of record counts corresponding to each of the top five sport
plot = awards_per_sport_sorted.head(5)
plot.plot.bar(ylabel= 'count')
```

Out[5]: <AxesSubplot:xlabel='Sport,Year', ylabel='count'>



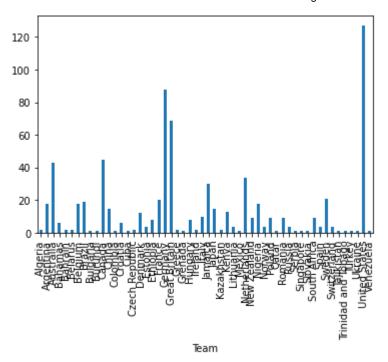
```
In [6]:
#Q.1.e Generate a histogram for the "Age" of all medal winners in the top five sports (
    top_five = olympic_filtered.loc[olympic_filtered['Sport'].isin(['Athletics', 'Swimming'
    top_five = top_five[top_five['Year'] >= 2016]
    medals_age = top_five[['Sport','Age','Medal']]
    medals = ['Gold','Silver','Bronze']
    medals_age.hist()
```

Out[6]: array([[<AxesSubplot:title={'center':'Age'}>]], dtype=object)

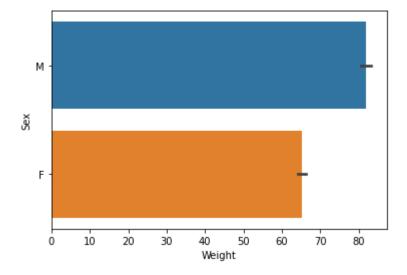


```
In [7]: country_by_medal = top_five.groupby(['Team'])['Medal'].count()
    country_by_medal.plot(kind='bar')
```

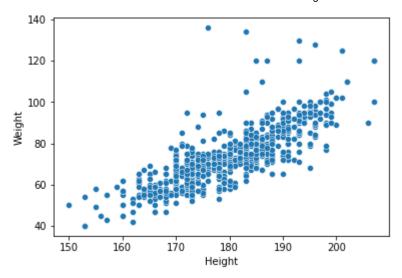
Out[7]: <AxesSubplot:xlabel='Team'>



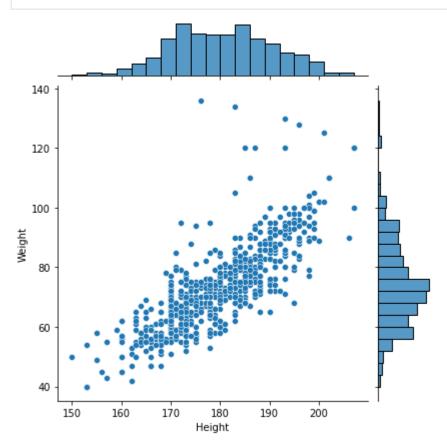
#Q.1.g Generate a bar plot indicating the average weight of players, categorized based
weight\_by\_gender = top\_five[['Weight','Sex']]
ax = sns.barplot(x='Weight', y="Sex", data=weight\_by\_gender)



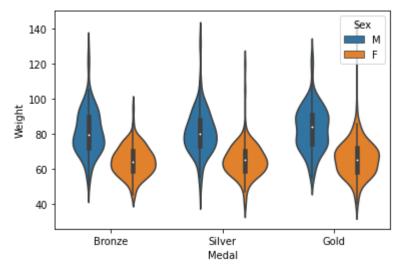
```
In [9]:
#Q.1.h Create a scatter plot with x=height and y=weight.
height_by_weight = top_five[['Weight', 'Height']]
ax = sns.scatterplot(x='Height', y='Weight', data=height_by_weight)
```



In [10]: #Q.1.i Create a joint plot with x=height and y=weight.
ax = sns.jointplot(x='Height', y='Weight', data=height\_by\_weight)



#Q.1.j Create two violin plots: 1 for distribution of weight by genders & class of meda
ax = sns.violinplot(x=top\_five['Medal'], y=top\_five['Weight'], hue=top\_five['Sex'], dat



Q.2 We will work with the HPI dataset. The objective is to draw a bar plot depicting the number of countries in each region and a heatmap indicating the number of countries in various ranges of wellbeing and life-expectancy.

In [12]: #Q.2.a Read from the dataset "hpi\_data\_countries.txt" (it is txt file separated by \t).
hpi\_df = pd.read\_csv('..\dataFiles\hpi\_data\_countries.txt', delimiter = "\t")
hpi\_df

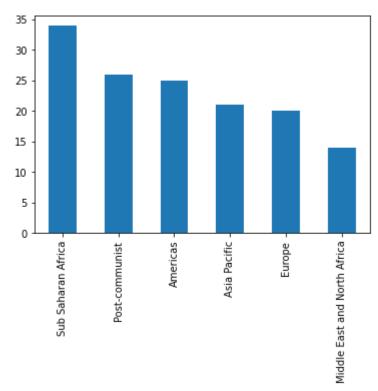
Out[12]:

	HPI Rank	Country	Region	Life Expectancy (years)	Wellbeing (0-10)	Inequality of outcomes	Ecological Footprint (gha/capita)	Happy Planet Index
0	1	Costa Rica	Americas	79.1	7.3	15%	2.8	44.7
1	2	Mexico	Americas	76.4	7.3	19%	2.9	40.7
2	3	Colombia	Americas	73.7	6.4	24%	1.9	40.7
3	4	Vanuatu	Asia Pacific	71.3	6.5	22%	1.9	40.6
4	5	Vietnam	Asia Pacific	75.5	5.5	19%	1.7	40.3
•••								
135	136	Mongolia	Asia Pacific	68.6	4.9	22%	6.1	14.3
136	137	Benin	Sub Saharan Africa	59.2	3.2	44%	1.4	13.4
137	138	Togo	Sub Saharan Africa	58.6	2.9	43%	1.1	13.2
138	139	Luxembourg	Europe	81.1	7.0	7%	15.8	13.2
139	140	Chad	Sub Saharan Africa	50.8	4.0	51%	1.5	12.8

140 rows × 8 columns

```
#Q.2.b Count the number of rows for each region to generate a bar chart.
region_count = hpi_df['Region'].value_counts()
region_count.plot(kind='bar')
```

## Out[13]: <AxesSubplot:>



In [14]:
#Q.2.c Use pd.cut to bin the values into different categories.
#Then generate a heatmap using cmap="Greens"
hpi\_df["Wellbeing (0-10)(binned)"] = pd.cut(hpi\_df["Wellbeing (0-10)"], bins=np.linspac hpi\_df["Life Expectancy (years)(binned)"] = pd.cut(hpi\_df["Life Expectancy (years)"], b heat\_df = hpi\_df.groupby(["Wellbeing (0-10)(binned)", "Life Expectancy (years)(binned)" heat\_df = heat\_df.sort\_values(ascending=False, by="Life Expectancy (years)(binned)")
sns.heatmap(heat\_df, cmap="Greens")

