```
In [1]: import pandas as pd
          import numpy as np
          import matplotlib.pyplot as plt
          import seaborn as sns
          %matplotlib inline
 In [2]: from google.colab import files
         uploaded = files.upload()
          Choose files No file chosen
                                              Upload widget is only available when the cell has been executed in the current browser
         session. Please rerun this cell to enable.
         Saving Teams.xlsx to Teams.xlsx
         Saving Medals.xlsx to Medals.xlsx
         Saving events.csv to events.csv
         Saving EntriesGender.xlsx to EntriesGender.xlsx
         Saving Athletes.xlsx to Athletes.xlsx
In [44]: read_file = pd.read_excel ("Athletes.xlsx")
         read sfile = pd.read excel ("Medals.xlsx")
         read_gfile = pd.read_excel ("EntriesGender.xlsx")
         read_tfile = pd.read_excel ("Teams.xlsx")
          read file.to csv ("Athletes.csv",
                            index = None,
                            header=True)
          read_sfile.to_csv ("Medals.csv",
                             index = None.
                            header=True )
          read_gfile.to_csv ("EntriesGender.csv",
                            index = None,
                            header=True)
          read_tfile.to_csv ("Teams.csv",
                            index = None,
                            header=True)
         athletes = pd.read_csv('Athletes.csv')
         medals = pd.read_csv('Medals.csv')
          gender = pd.read csv('EntriesGender.csv')
         teams = pd.read csv('Teams.csv')
```

/usr/local/lib/python3.10/dist-packages/openpyxl/styles/stylesheet.py:237: UserWarning: Workbook contains no default style, apply openpyxl's default warn("Workbook contains no default style, apply openpyxl's default")

```
In [37]: athletes_df = pd.DataFrame(athletes)
athletes_df.head()
```

| Out[37]: |   | Name              | NOC    | Discipline          |
|----------|---|-------------------|--------|---------------------|
|          | 0 | AALERUD Katrine   | Norway | Cycling Road        |
|          | 1 | ABAD Nestor       | Spain  | Artistic Gymnastics |
|          | 2 | ABAGNALE Giovanni | Italy  | Rowing              |
|          | 3 | ABALDE Alberto    | Spain  | Basketball          |
|          | 4 | ABALDE Tamara     | Spain  | Basketball          |

```
In [38]: athletes_df.rename(columns = {'Discipline':'Sport'}, inplace = True)
```

## In [40]: athletes\_df.head()

| Out[40]: |   | Name              | NOC    | Sport               |
|----------|---|-------------------|--------|---------------------|
|          | 0 | AALERUD Katrine   | Norway | Cycling Road        |
|          | 1 | ABAD Nestor       | Spain  | Artistic Gymnastics |
|          | 2 | ABAGNALE Giovanni | Italy  | Rowing              |
|          | 3 | ABALDE Alberto    | Spain  | Basketball          |
|          | 4 | ABALDE Tamara     | Spain  | Basketball          |

```
In [41]: athletes_df.describe()
```

```
NOC
Out[41]:
                      Name
                                                    Sport
                      11085
                                                    11085
           count
                                            11085
                                                       46
          unique
                      11062
                                             206
             top CHEN Yang United States of America Athletics
                          2
                                             615
                                                     2068
            freq
          medals_df = pd.DataFrame(medals)
In [45]:
          gender_df = pd.DataFrame(gender)
          teams_df = pd.DataFrame(teams)
          medals_df.rename(columns = {'Team/NOC':'NOC'}, inplace = True)
In [47]:
In [48]: medals_df.head()
                                    NOC Gold Silver Bronze Total Rank by Total
Out[48]:
             Rank
                    United States of America
                                                               113
                                                                               1
          0
                                            39
                                                  41
                                                          33
                2 People's Republic of China
                                            38
                                                          18
                                                                88
                                                                               2
          1
                                                  32
          2
                                            27
                                                                58
                                                                               5
                                   Japan
                                                  14
                                                          17
          3
                4
                              Great Britain
                                            22
                                                  21
                                                          22
                                                                65
                                                                               4
                                    ROC
                                            20
                                                  28
                                                                71
          4
                5
                                                          23
                                                                               3
```

aathletes = athletes\_df.merge(medals\_df, how='left', on='NOC')

In [49]:

aathletes.head()

```
Out[49]:
                                   NOC
                                                     Sport Rank Gold Silver Bronze Total Rank by Total
                          Name
                                                                                         8.0
                AALERUD Katrine Norway
                                                                                   2.0
          0
                                               Cycling Road
                                                             20.0
                                                                    4.0
                                                                           2.0
                                                                                                       29.0
          1
                    ABAD Nestor
                                   Spain Artistic Gymnastics
                                                             22.0
                                                                    3.0
                                                                           0.8
                                                                                   6.0
                                                                                         17.0
                                                                                                       17.0
           2 ABAGNALE Giovanni
                                                            10.0
                                                                   10.0
                                                                          10.0
                                                                                  20.0
                                                                                         40.0
                                                                                                        7.0
                                    Italy
                                                   Rowing
                                                             22.0
                                                                    3.0
                                                                           8.0
                                                                                        17.0
          3
                  ABALDE Alberto
                                   Spain
                                                 Basketball
                                                                                   6.0
                                                                                                       17.0
           4
                  ABALDE Tamara
                                   Spain
                                                 Basketball
                                                             22.0
                                                                    3.0
                                                                           8.0
                                                                                   6.0
                                                                                        17.0
                                                                                                       17.0
           aathletes.rename(columns = {'Total':'TotalMedals'}, inplace = True)
In [92]:
           aathletes.drop(columns=['Rank by Total'], inplace = True)
           aathletes.head()
                                   NOC
                                                     Sport Rank Gold Silver Bronze TotalMedals
Out[92]:
                          Name
                AALERUD Katrine Norway
          0
                                               Cycling Road
                                                             20.0
                                                                    4.0
                                                                           2.0
                                                                                   2.0
                                                                                                8.0
          1
                                                                    3.0
                    ABAD Nestor
                                   Spain Artistic Gymnastics
                                                             22.0
                                                                           0.8
                                                                                   6.0
                                                                                               17.0
           2 ABAGNALE Giovanni
                                                             10.0
                                                                          10.0
                                                                                  20.0
                                    Italy
                                                   Rowing
                                                                   10.0
                                                                                               40.0
          3
                  ABALDE Alberto
                                   Spain
                                                 Basketball
                                                             22.0
                                                                    3.0
                                                                           8.0
                                                                                   6.0
                                                                                               17.0
           4
                  ABALDE Tamara
                                   Spain
                                                 Basketball
                                                             22.0
                                                                    3.0
                                                                           0.8
                                                                                   6.0
                                                                                               17.0
          gender_df.head()
In [93]:
Out[93]:
                         Sport Female Male Total
                  3x3 Basketball
          0
                                    32
                                           32
                                                 64
                       Archery
          1
                                    64
                                           64
                                                128
           2 Artistic Gymnastics
                                    98
                                          98
                                                196
              Artistic Swimming
                                   105
                                            0
                                                105
           4
                      Athletics
                                   969 1072 2041
          gender_df.rename(columns = {'Discipline':'Sport', 'Total' : 'TotalPlayers'}, inplace = True)
```

```
gender_df.head()
 In [96]:
                         Sport Female Male TotalPlayers
 Out[96]:
           0
                   3x3 Basketball
                                           32
                                                        64
                                     32
                        Archery
                                           64
           1
                                     64
                                                       128
            2 Artistic Gymnastics
                                           98
                                     98
                                                       196
            3 Artistic Swimming
                                    105
                                            0
                                                       105
            4
                       Athletics
                                    969
                                        1072
                                                      2041
            aathletes_gender = aathletes.merge(gender_df, how='left', on='Sport')
            aathletes gender.head()
                                                     Sport Rank Gold Silver Bronze TotalMedals Female Male TotalPlayers
 Out[97]:
                           Name
                                    NOC
                                               Cycling Road
                 AALERUD Katrine Norway
           0
                                                             20.0
                                                                    4.0
                                                                            2.0
                                                                                    2.0
                                                                                                 8.0
                                                                                                         70
                                                                                                               131
                                                                                                                            201
           1
                                                                    3.0
                                                                                                                98
                     ABAD Nestor
                                    Spain Artistic Gymnastics
                                                             22.0
                                                                            8.0
                                                                                    6.0
                                                                                                17.0
                                                                                                         98
                                                                                                                            196
            2 ABAGNALE Giovanni
                                     Italy
                                                    Rowing
                                                             10.0
                                                                   10.0
                                                                           10.0
                                                                                   20.0
                                                                                                40.0
                                                                                                        257
                                                                                                               265
                                                                                                                            522
                  ABALDE Alberto
                                                             22.0
                                                                    3.0
                                                                            8.0
                                                                                    6.0
                                                                                                               144
           3
                                    Spain
                                                  Basketball
                                                                                                17.0
                                                                                                        144
                                                                                                                            288
            4
                  ABALDE Tamara
                                                                                    6.0
                                                                                                                            288
                                    Spain
                                                  Basketball
                                                             22.0
                                                                     3.0
                                                                            8.0
                                                                                                17.0
                                                                                                        144
                                                                                                               144
           bathletes_df = pd.DataFrame(aathletes_gender)
In [118...
            bathletes df.head()
                                    NOC
                                                     Sport Rank Gold Silver Bronze TotalMedals Female Male TotalPlayers
Out[118]:
                           Name
                                                                                                              131
                                                                    4.0
                                                                            2.0
                                                                                    2.0
                                                                                                 8.0
                                                                                                                            201
            0
                 AALERUD Katrine Norway
                                                Cycling Road
                                                             20.0
                                                                                                         70
                                    Spain Artistic Gymnastics
                                                                    3.0
           1
                     ABAD Nestor
                                                             22.0
                                                                            0.8
                                                                                    6.0
                                                                                                17.0
                                                                                                         98
                                                                                                                98
                                                                                                                            196
                                                    Rowing
            2 ABAGNALE Giovanni
                                     Italy
                                                             10.0
                                                                    10.0
                                                                           10.0
                                                                                   20.0
                                                                                                40.0
                                                                                                        257
                                                                                                               265
                                                                                                                            522
```

Basketball

Basketball

22.0

22.0

3.0

3.0

8.0

8.0

6.0

6.0

17.0

17.0

144

144

144

144

288

288

3

4

**ABALDE Alberto** 

**ABALDE Tamara** 

Spain

Spain

```
bathletes_df.head()
In [119...
Out[119]:
                                    NOC
                                                      Sport Rank Gold Silver Bronze TotalMedals Female Male TotalPlayers
                           Name
                  AALERUD Katrine Norway
                                                Cycling Road
                                                             20.0
                                                                     4.0
                                                                            2.0
                                                                                    2.0
                                                                                                 8.0
                                                                                                          70
                                                                                                               131
            0
                                                                                                                            201
                     ABAD Nestor
                                    Spain Artistic Gymnastics
                                                                     3.0
                                                                            8.0
                                                                                    6.0
                                                                                                17.0
                                                                                                          98
                                                                                                                98
            1
                                                              22.0
                                                                                                                            196
            2 ABAGNALE Giovanni
                                     Italy
                                                              10.0
                                                                    10.0
                                                                                   20.0
                                                                                                                            522
                                                    Rowing
                                                                           10.0
                                                                                                40.0
                                                                                                         257
                                                                                                               265
                                                                                    6.0
            3
                   ABALDE Alberto
                                    Spain
                                                  Basketball
                                                              22.0
                                                                     3.0
                                                                            8.0
                                                                                                17.0
                                                                                                         144
                                                                                                               144
                                                                                                                            288
                                                                                    6.0
                                                  Basketball
            4
                   ABALDE Tamara
                                    Spain
                                                              22.0
                                                                     3.0
                                                                            8.0
                                                                                                17.0
                                                                                                         144
                                                                                                               144
                                                                                                                            288
            teams_df = pd.DataFrame(teams)
In [120...
            teams_df.head()
                                                        NOC
Out[120]:
                 Name
                           Discipline
                                                                Event
            0 Belgium 3x3 Basketball
                                                     Belgium
                                                                 Men
                 China 3x3 Basketball People's Republic of China
                                                                 Men
            2
                 China 3x3 Basketball People's Republic of China Women
                France 3x3 Basketball
                                                      France Women
            4
                  Italy 3x3 Basketball
                                                        Italy Women
            ateams_df = pd.DataFrame(teams_df)
In [122...
```

ateams\_df.rename(columns = {'Name' : 'Country', 'Discipline' : 'Sport'}, inplace = True)

ateams df.head()

ateams\_df.head()

| Out[122]: |   | Country | Sport          | NOC                        | Event |
|-----------|---|---------|----------------|----------------------------|-------|
|           | 0 | Belgium | 3x3 Basketball | Belgium                    | Men   |
|           | 1 | China   | 3x3 Basketball | People's Republic of China | Men   |
|           | 2 | China   | 3x3 Basketball | People's Republic of China | Women |
|           | 3 | France  | 3x3 Basketball | France                     | Women |
|           | 4 | Italy   | 3x3 Basketball | Italy                      | Women |

ABALDE

Tamara

Spain

22.0

3.0

80

In [138... olympics\_df.drop\_duplicates(subset=['Name'], keep='first', inplace=True)
 olympics\_df.head()

Out[138]: NOC Rank Gold Silver Bronze TotalMedals Female Male TotalPlayers **Sports** Name Country **Event** Mol Cycling Road\nBeach AALERUD 8.0 0 Norway 20.0 4.0 2.0 2.0 70 131 A./Sorum 201 Men Katrine Volleyball C. ABAD Artistic Mixed 8.0 3 Spain 22.0 6.0 17.0 98 98 196 Spain 3.0 Nestor Team Gymnastics\nArchery Rowing\n3x3 **ABAGNALE** 23 10.0 10.0 10.0 40.0 522 20.0 257 265 Italy Women Italy Basketball Giovanni **ABALDE** Mixed 60 Spain 22.0 3.0 8.0 17.0 288 Spain Basketball\nArchery 6.0 144 144 Alberto Team

17.0

144

144

Mixed

Team

Basketball\nArchery

Spain

288

8.0

6.0

| Out[139]: |   | Name               | NOC    | Sport_x                | Rank | Gold | Silver | Bronze | TotalMedals | Female | Male | TotalPlayers | Country               | Sport_y                | Event         |
|-----------|---|--------------------|--------|------------------------|------|------|--------|--------|-------------|--------|------|--------------|-----------------------|------------------------|---------------|
|           | 0 | AALERUD<br>Katrine | Norway | Cycling<br>Road        | 20.0 | 4.0  | 2.0    | 2.0    | 8.0         | 70     | 131  | 201          | Mol<br>A./Sorum<br>C. | Beach<br>Volleyball    | Men           |
|           | 1 | AALERUD<br>Katrine | Norway | Cycling<br>Road        | 20.0 | 4.0  | 2.0    | 2.0    | 8.0         | 70     | 131  | 201          | Norway                | Handball               | Men           |
|           | 2 | AALERUD<br>Katrine | Norway | Cycling<br>Road        | 20.0 | 4.0  | 2.0    | 2.0    | 8.0         | 70     | 131  | 201          | Norway                | Handball               | Women         |
|           | 3 | ABAD<br>Nestor     | Spain  | Artistic<br>Gymnastics | 22.0 | 3.0  | 8.0    | 6.0    | 17.0        | 98     | 98   | 196          | Spain                 | Archery                | Mixed<br>Team |
|           | 4 | ABAD<br>Nestor     | Spain  | Artistic<br>Gymnastics | 22.0 | 3.0  | 8.0    | 6.0    | 17.0        | 98     | 98   | 196          | Spain                 | Artistic<br>Gymnastics | Men's<br>Team |

| Out[140]: |    | Name                 | NOC    | Sport_x                | Rank | Gold | Silver | Bronze | TotalMedals | Female | Male | TotalPlayers | Country               | Sport_y             | Event         |
|-----------|----|----------------------|--------|------------------------|------|------|--------|--------|-------------|--------|------|--------------|-----------------------|---------------------|---------------|
|           | 0  | AALERUD<br>Katrine   | Norway | Cycling<br>Road        | 20.0 | 4.0  | 2.0    | 2.0    | 8.0         | 70     | 131  | 201          | Mol<br>A./Sorum<br>C. | Beach<br>Volleyball | Men           |
|           | 3  | ABAD<br>Nestor       | Spain  | Artistic<br>Gymnastics | 22.0 | 3.0  | 8.0    | 6.0    | 17.0        | 98     | 98   | 196          | Spain                 | Archery             | Mixed<br>Team |
|           | 23 | ABAGNALE<br>Giovanni | Italy  | Rowing                 | 10.0 | 10.0 | 10.0   | 20.0   | 40.0        | 257    | 265  | 522          | Italy                 | 3x3<br>Basketball   | Women         |
|           | 60 | ABALDE<br>Alberto    | Spain  | Basketball             | 22.0 | 3.0  | 8.0    | 6.0    | 17.0        | 144    | 144  | 288          | Spain                 | Archery             | Mixed<br>Team |
|           | 80 | ABALDE<br>Tamara     | Spain  | Basketball             | 22.0 | 3.0  | 8.0    | 6.0    | 17.0        | 144    | 144  | 288          | Spain                 | Archery             | Mixed<br>Team |

```
In [162... #pd.get_dummies(olympics_df['Sport_x'])
    df = pd.DataFrame(olympics_df['Sport_x'])
    df.head()
    #pd.get_dummies(olympics_df['Sport_y'])
```

```
df1 = pd.DataFrame(olympics_df['Sport_y'])
df1.head()
dolympics_df= pd.concat([df, df1], axis=0, ignore_index=True)
dolympics_df['Sports'] = pd.concat([olympics_df['Sport_x'], olympics_df['Sport_y']], ignore_index=True)
dolympics_df.head()
```

### Out[162]:

| /            | Sport_y | Sport_x             |   |
|--------------|---------|---------------------|---|
| l Cy         | NaN     | Cycling Road        | 0 |
| N Artistic G | NaN     | Artistic Gymnastics | 1 |
| J            | NaN     | Rowing              | 2 |
| J            | NaN     | Basketball          | 3 |
| J            | NaN     | Basketball          | 4 |

In [167...

z = olympics\_df.merge(dolympics\_df, left\_index=True, right\_index=True)
z.head()

### Out[167]:

| : |    | Name                 | NOC    | Rank | Gold | Silver | Bronze | TotalMedals | Female | Male | TotalPlayers | Country               | Event         | Sport_x         | Sport_y | Spor          |
|---|----|----------------------|--------|------|------|--------|--------|-------------|--------|------|--------------|-----------------------|---------------|-----------------|---------|---------------|
|   | 0  | AALERUD<br>Katrine   | Norway | 20.0 | 4.0  | 2.0    | 2.0    | 8.0         | 70     | 131  | 201          | Mol<br>A./Sorum<br>C. | Men           | Cycling<br>Road | NaN     | Cyclir<br>Roa |
|   | 3  | ABAD<br>Nestor       | Spain  | 22.0 | 3.0  | 8.0    | 6.0    | 17.0        | 98     | 98   | 196          | Spain                 | Mixed<br>Team | Basketball      | NaN     | Basketb       |
|   | 23 | ABAGNALE<br>Giovanni | Italy  | 10.0 | 10.0 | 10.0   | 20.0   | 40.0        | 257    | 265  | 522          | Italy                 | Women         | Wrestling       | NaN     | Wrestlir      |
|   | 60 | ABALDE<br>Alberto    | Spain  | 22.0 | 3.0  | 8.0    | 6.0    | 17.0        | 144    | 144  | 288          | Spain                 | Mixed<br>Team | Athletics       | NaN     | Athleti       |
|   | 80 | ABALDE<br>Tamara     | Spain  | 22.0 | 3.0  | 8.0    | 6.0    | 17.0        | 144    | 144  | 288          | Spain                 | Mixed<br>Team | Triathlon       | NaN     | Triathlo      |

4

In [168... z.drop(columns=['Sport\_x', 'Sport\_y'], inplace = True)
z.head()

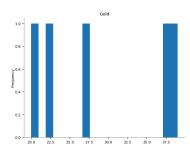
| ut[168]:             |   | Name                                    | NOC   | Rank     | Gold   | Silver | Bronze | TotalMedals           | Female | Male | TotalPlayers | Country            | Event         | Sports          |
|----------------------|---|---|---|----------|--------|--------|--------|-----------------------|--------|------|--------------|--------------------|---------------|-----------------|
|                      | 0   | AALERUD<br>Katrine                      |   | 20.0     | 4.0    | 2.0    | 2.0    | 8.0                   | 70     | 131  | 201          | Mol<br>A./Sorum C. | Men           | Cycling<br>Road |
|                      | 3   | ABAD Nestor                             | Spain   | 22.0     | 3.0    | 8.0    | 6.0    | 17.0                  | 98     | 98   | 196          | Spain              | Mixed<br>Team | Basketball      |
|                      | 23  | ABAGNALE<br>Giovann                     | IT2I\/  | 10.0     | 10.0   | 10.0   | 20.0   | 40.0                  | 257    | 265  | 522          | Italy              | Women         | Wrestling       |
|                      | 60  | ABALDE Alberto                          | Spain   | 22.0     | 3.0    | 8.0    | 6.0    | 17.0                  | 144    | 144  | 288          | Spain              | Mixed<br>Team | Athletics       |
|                      | 80  | ABALDE Tamara                           | Spain   | 22.0     | 3.0    | 8.0    | 6.0    | 17.0                  | 144    | 144  | 288          | Spain              | Mixed<br>Team | Triathlon       |
| In [169<br>Out[169]: | Nam<br>NOC<br>Ran<br>Gol<br>Sil<br>Bro<br>Tot<br>Fem<br>Mal | k<br>d<br>ver<br>nze<br>alMedals<br>ale | 0<br>0<br>866<br>866<br>866<br>866<br>866<br>0<br>0 | 0()      |        |        |        |                       |        |      |              |                    |               |                 |
|                      | Cou<br>Eve<br>dty   | ntry<br>nt<br>pe: int64                 | 938<br>938  | F.1      |        |        |        |                       |        |      |              |                    |               |                 |
| n [170               |   | rop_duplicate<br>ead()                  | s(subset  | :=[ 'NO0 | .'], k | eep='f | irst', | inplace= <b>Tru</b> e | e)     |      |              |                    |               |                 |

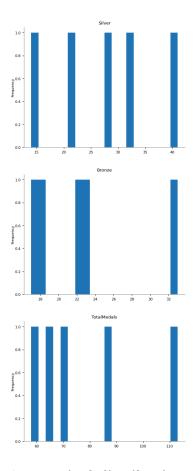
| Out[170]: |     | Name                 | NOC    | Rank | Gold | Silver | Bronze | TotalMedals | Female | Male | TotalPlayers | Country                  | Event         | Sports          |
|-----------|-----|----------------------|--------|------|------|--------|--------|-------------|--------|------|--------------|--------------------------|---------------|-----------------|
|           | 0   | AALERUD<br>Katrine   | Norway | 20.0 | 4.0  | 2.0    | 2.0    | 8.0         | 70     | 131  | 201          | Mol A./Sorum<br>C.       | Men           | Cycling<br>Road |
|           | 3   | ABAD Nestor          | Spain  | 22.0 | 3.0  | 8.0    | 6.0    | 17.0        | 98     | 98   | 196          | Spain                    | Mixed<br>Team | Basketball      |
|           | 23  | ABAGNALE<br>Giovanni | Italy  | 10.0 | 10.0 | 10.0   | 20.0   | 40.0        | 257    | 265  | 522          | Italy                    | Women         | Wrestling       |
|           | 100 | ABALO Luc            | France | 8.0  | 10.0 | 12.0   | 11.0   | 33.0        | 168    | 168  | 336          | France                   | Women         | Wrestling       |
|           | 133 | ABAROA Cesar         | Chile  | NaN  | NaN  | NaN    | NaN    | NaN         | 257    | 265  | 522          | Grimalt<br>M./Grimalt E. | Men           | Athletics       |

### Out[177]:

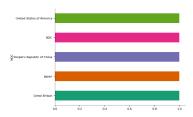
|   | NOC                        | Gold | Silver | Bronze | TotalMedals |
|---|----------------------------|------|--------|--------|-------------|
| 0 | United States of America   | 39.0 | 41.0   | 33.0   | 113.0       |
| 1 | People's Republic of China | 38.0 | 32.0   | 18.0   | 88.0        |
| 2 | Japan                      | 27.0 | 14.0   | 17.0   | 58.0        |
| 3 | Great Britain              | 22.0 | 21.0   | 22.0   | 65.0        |
| 4 | ROC                        | 20.0 | 28.0   | 23.0   | 71.0        |

# Distributions

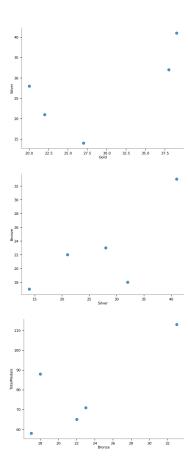




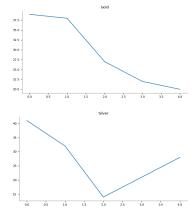
# Categorical distributions

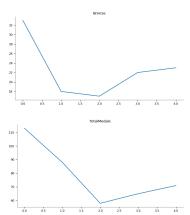


2-d distributions



# Values





### **Faceted distributions**

### <string>:5: FutureWarning:

Passing `palette` without assigning `hue` is deprecated and will be removed in v0.14.0. Assign the `y` variable to `hue` and set `legend=False` for the same effect.



#### <string>:5: FutureWarning:

Passing `palette` without assigning `hue` is deprecated and will be removed in v0.14.0. Assign the `y` variable to `hue` and set `legend=False` for the same effect.



### <string>:5: FutureWarning:

Passing `palette` without assigning `hue` is deprecated and will be removed in v0.14.0. Assign the `y` variable to `hue` and set `legend=False` for the same effect.

The state of the s

### <string>:5: FutureWarning:

Passing `palette` without assigning `hue` is deprecated and will be removed in v0.14.0. Assign the `y` variable to `hue` and set `legend=False` for the same effect.

Season for an an annual and an annual an annua

In [192... goldcountries = sorted\_df[sorted\_df['Gold'] > 0]
 goldcountries

Out[192]:

|     | NOC                        | Gold | Silver | Bronze | TotalMedals |
|-----|----------------------------|------|--------|--------|-------------|
| 0   | United States of America   | 39.0 | 41.0   | 33.0   | 113.0       |
| 1   | People's Republic of China | 38.0 | 32.0   | 18.0   | 88.0        |
| 2   | Japan                      | 27.0 | 14.0   | 17.0   | 58.0        |
| 3   | Great Britain              | 22.0 | 21.0   | 22.0   | 65.0        |
| 4   | ROC                        | 20.0 | 28.0   | 23.0   | 71.0        |
| ••• |                            |      |        |        |             |
| 57  | Slovakia                   | 1.0  | 2.0    | 1.0    | 4.0         |
| 58  | Fiji                       | 1.0  | 0.0    | 1.0    | 2.0         |
| 59  | Venezuela                  | 1.0  | 3.0    | 0.0    | 4.0         |
| 60  | Morocco                    | 1.0  | 0.0    | 0.0    | 1.0         |
| 61  | South Africa               | 1.0  | 2.0    | 0.0    | 3.0         |

62 rows × 5 columns

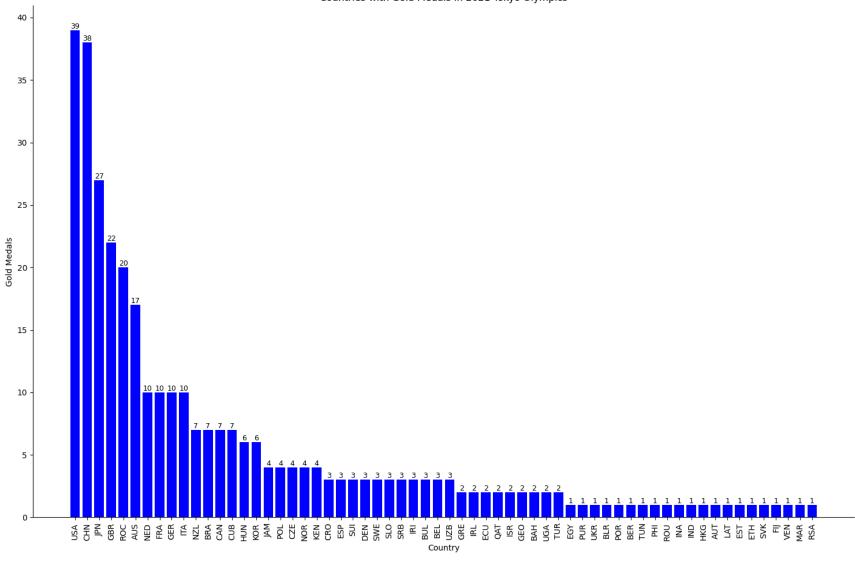
```
country_shortforms = {
In [193...
               'United States of America': 'USA',
               "People's Republic of China": 'CHN',
               'Japan': 'JPN',
               'Great Britain': 'GBR',
               'ROC': 'ROC', # Russian Olympic Committee
               'Australia': 'AUS',
               'Netherlands': 'NED',
               'France': 'FRA',
               'Germany': 'GER',
               'Italy': 'ITA',
               'Canada': 'CAN',
               'Brazil': 'BRA',
               'New Zealand': 'NZL',
               'Cuba': 'CUB',
               'Hungary': 'HUN',
               'Republic of Korea': 'KOR',
               'Poland': 'POL',
               'Czech Republic': 'CZE',
               'Kenya': 'KEN',
               'Norway': 'NOR',
               'Jamaica': 'JAM',
               'Spain': 'ESP',
               'Sweden': 'SWE',
               'Switzerland': 'SUI',
               'Denmark': 'DEN',
               'Croatia': 'CRO',
               'Islamic Republic of Iran': 'IRI',
               'Serbia': 'SRB',
               'Belgium': 'BEL',
               'Bulgaria': 'BUL',
               'Slovenia': 'SLO',
               'Uzbekistan': 'UZB',
               'Georgia': 'GEO',
               'Chinese Taipei': 'TPE',
               'Turkey': 'TUR',
               'Greece': 'GRE',
               'Uganda': 'UGA',
               'Ecuador': 'ECU',
               'Ireland': 'IRL',
               'Israel': 'ISR',
               'Qatar': 'QAT',
               'Bahamas': 'BAH',
               'Kosovo': 'KOS',
```

```
'Ukraine': 'UKR',
'Belarus': 'BLR',
'Romania': 'ROU',
'Venezuela': 'VEN',
'India': 'IND',
'Hong Kong, China': 'HKG',
'Philippines': 'PHI',
'Slovakia': 'SVK',
'South Africa': 'RSA',
'Austria': 'AUT',
'Egypt': 'EGY',
'Indonesia': 'INA',
'Ethiopia': 'ETH',
'Portugal': 'POR',
'Tunisia': 'TUN',
'Estonia': 'EST',
'Fiji': 'FIJ',
'Latvia': 'LAT',
'Thailand': 'THA',
'Bermuda': 'BER',
'Morocco': 'MAR',
'Puerto Rico': 'PUR'
```

```
import matplotlib.pyplot as plt
In [197...
          # Countries with zero gold medals
          sorted_df = sorted_df[sorted_df['Gold'] > 0]
          # Plotting with improved readability
          plt.figure(figsize=(15, 10))
          # Create bar plot with filtered country names on x-axis and gold medal counts on y-axis
          bars = plt.bar(sorted_df['NOC'], sorted_df['Gold'], color=['blue'])
          country shortforms = {
              'United States of America': 'USA',
              "People's Republic of China": 'CHN',
              'Japan': 'JPN',
              'Great Britain': 'GBR',
              'ROC': 'ROC',
              'Australia': 'AUS',
              'Netherlands': 'NED',
              'France': 'FRA',
               'Germany': 'GER',
```

```
'Italy': 'ITA',
'Canada': 'CAN',
'Brazil': 'BRA',
'New Zealand': 'NZL',
'Cuba': 'CUB',
'Hungary': 'HUN',
'Republic of Korea': 'KOR',
'Poland': 'POL',
'Czech Republic': 'CZE',
'Kenya': 'KEN',
'Norway': 'NOR',
'Jamaica': 'JAM',
'Spain': 'ESP',
'Sweden': 'SWE',
'Switzerland': 'SUI',
'Denmark': 'DEN',
'Croatia': 'CRO',
'Islamic Republic of Iran': 'IRI',
'Serbia': 'SRB',
'Belgium': 'BEL',
'Bulgaria': 'BUL',
'Slovenia': 'SLO',
'Uzbekistan': 'UZB',
'Georgia': 'GEO',
'Chinese Taipei': 'TPE',
'Turkey': 'TUR',
'Greece': 'GRE',
'Uganda': 'UGA',
'Ecuador': 'ECU',
'Ireland': 'IRL',
'Israel': 'ISR',
'Qatar': 'QAT',
'Bahamas': 'BAH',
'Kosovo': 'KOS',
'Ukraine': 'UKR',
'Belarus': 'BLR',
'Romania': 'ROU',
'Venezuela': 'VEN',
'India': 'IND',
'Hong Kong, China': 'HKG',
'Philippines': 'PHI',
'Slovakia': 'SVK',
'South Africa': 'RSA',
'Austria': 'AUT',
'Egypt': 'EGY',
```

```
'Indonesia': 'INA',
    'Ethiopia': 'ETH',
    'Portugal': 'POR',
    'Tunisia': 'TUN',
    'Estonia': 'EST',
    'Fiji': 'FIJ',
    'Latvia': 'LAT',
    'Thailand': 'THA',
    'Bermuda': 'BER',
    'Morocco': 'MAR',
    'Puerto Rico': 'PUR'
# Set plot title and labels
plt.title('Countries with Gold Medals in 2021 Tokyo Olympics')
plt.xlabel('Country')
plt.ylabel('Gold Medals')
# Rotate x-axis labels vertically
plt.xticks(sorted_df['NOC'], [country_shortforms[c] for c in sorted_df['NOC']], rotation='vertical', fontsize=10)
# Hide top and right spines
plt.gca().spines['top'].set_visible(False)
plt.gca().spines['right'].set_visible(False)
# Add text annotations on top of each bar
for bar in bars:
    plt.text(bar.get_x() + bar.get_width()/2, bar.get_height(), str(int(bar.get_height())), ha='center', va='bottom', f
# Show the plot
plt.tight_layout()
plt.show()
```



In [208... countries\_with\_silver\_medals = sorted\_df[sorted\_df['Silver'] > 0]
 countries\_with\_silver\_medals

Out[208]:

|    | NOC                        | Gold | Silver | Bronze | TotalMedals |
|----|----------------------------|------|--------|--------|-------------|
| 0  | United States of America   | 39.0 | 41.0   | 33.0   | 113.0       |
| 1  | People's Republic of China | 38.0 | 32.0   | 18.0   | 88.0        |
| 2  | Japan                      | 27.0 | 14.0   | 17.0   | 58.0        |
| 3  | Great Britain              | 22.0 | 21.0   | 22.0   | 65.0        |
| 4  | ROC                        | 20.0 | 28.0   | 23.0   | 71.0        |
| 5  | Australia                  | 17.0 | 7.0    | 22.0   | 46.0        |
| 6  | Netherlands                | 10.0 | 12.0   | 14.0   | 36.0        |
| 7  | France                     | 10.0 | 12.0   | 11.0   | 33.0        |
| 8  | Germany                    | 10.0 | 11.0   | 16.0   | 37.0        |
| 9  | Italy                      | 10.0 | 10.0   | 20.0   | 40.0        |
| 10 | New Zealand                | 7.0  | 6.0    | 7.0    | 20.0        |
| 11 | Brazil                     | 7.0  | 6.0    | 8.0    | 21.0        |
| 12 | Canada                     | 7.0  | 6.0    | 11.0   | 24.0        |
| 13 | Cuba                       | 7.0  | 3.0    | 5.0    | 15.0        |
| 14 | Hungary                    | 6.0  | 7.0    | 7.0    | 20.0        |
| 15 | Republic of Korea          | 6.0  | 4.0    | 10.0   | 20.0        |
| 16 | Jamaica                    | 4.0  | 1.0    | 4.0    | 9.0         |
| 17 | Poland                     | 4.0  | 5.0    | 5.0    | 14.0        |
| 18 | Czech Republic             | 4.0  | 4.0    | 3.0    | 11.0        |
| 19 | Norway                     | 4.0  | 2.0    | 2.0    | 8.0         |
| 20 | Kenya                      | 4.0  | 4.0    | 2.0    | 10.0        |
| 21 | Croatia                    | 3.0  | 3.0    | 2.0    | 8.0         |
| 22 | Spain                      | 3.0  | 8.0    | 6.0    | 17.0        |
| 23 | Switzerland                | 3.0  | 4.0    | 6.0    | 13.0        |
| 24 | Denmark                    | 3.0  | 4.0    | 4.0    | 11.0        |

|    | NOC                      | Gold | Silver | Bronze | TotalMedals |
|----|--------------------------|------|--------|--------|-------------|
| 25 | Sweden                   | 3.0  | 6.0    | 0.0    | 9.0         |
| 26 | Slovenia                 | 3.0  | 1.0    | 1.0    | 5.0         |
| 27 | Serbia                   | 3.0  | 1.0    | 5.0    | 9.0         |
| 28 | Islamic Republic of Iran | 3.0  | 2.0    | 2.0    | 7.0         |
| 29 | Bulgaria                 | 3.0  | 1.0    | 2.0    | 6.0         |
| 30 | Belgium                  | 3.0  | 1.0    | 3.0    | 7.0         |
| 32 | Greece                   | 2.0  | 1.0    | 1.0    | 4.0         |
| 34 | Ecuador                  | 2.0  | 1.0    | 0.0    | 3.0         |
| 37 | Georgia                  | 2.0  | 5.0    | 1.0    | 8.0         |
| 39 | Uganda                   | 2.0  | 1.0    | 1.0    | 4.0         |
| 40 | Turkey                   | 2.0  | 2.0    | 9.0    | 13.0        |
| 41 | Egypt                    | 1.0  | 1.0    | 4.0    | 6.0         |
| 43 | Ukraine                  | 1.0  | 6.0    | 12.0   | 19.0        |
| 44 | Belarus                  | 1.0  | 3.0    | 3.0    | 7.0         |
| 45 | Portugal                 | 1.0  | 1.0    | 2.0    | 4.0         |
| 47 | Tunisia                  | 1.0  | 1.0    | 0.0    | 2.0         |
| 48 | Philippines              | 1.0  | 2.0    | 1.0    | 4.0         |
| 49 | Romania                  | 1.0  | 3.0    | 0.0    | 4.0         |
| 50 | Indonesia                | 1.0  | 1.0    | 3.0    | 5.0         |
| 51 | India                    | 1.0  | 2.0    | 4.0    | 7.0         |
| 52 | Hong Kong, China         | 1.0  | 2.0    | 3.0    | 6.0         |
| 53 | Austria                  | 1.0  | 1.0    | 5.0    | 7.0         |
| 56 | Ethiopia                 | 1.0  | 1.0    | 2.0    | 4.0         |
| 57 | Slovakia                 | 1.0  | 2.0    | 1.0    | 4.0         |
| 59 | Venezuela                | 1.0  | 3.0    | 0.0    | 4.0         |

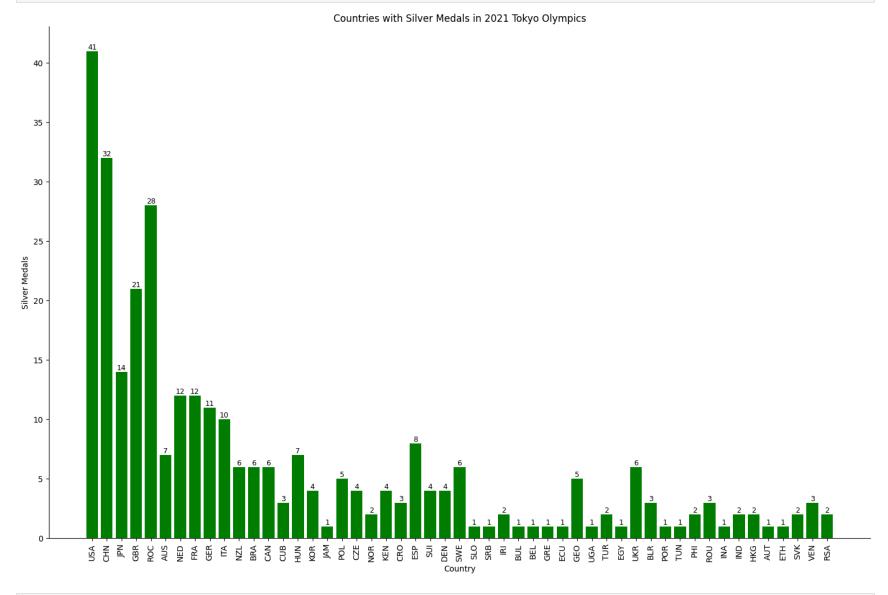
**61** South Africa 1.0 2.0 0.0

```
import matplotlib.pyplot as plt
In [211...
          # Countries with silver medals
          sorted_df = sorted_df[sorted_df['Silver'] > 0]
          # Plotting with improved readability
          plt.figure(figsize=(15, 10))
          #Bar plot with filtered country names on x-axis and silver medal counts on y-axis
          bars = plt.bar(sorted_df['NOC'], sorted_df['Silver'], color=['green'])
          # Short forms of countries
          country short forms = {
               'United States of America': 'USA',
              "People's Republic of China": 'CHN',
               'Japan': 'JPN',
               'Great Britain': 'GBR',
               'ROC': 'ROC',
               'Australia': 'AUS',
               'Netherlands': 'NED',
               'France': 'FRA',
               'Germany': 'GER',
               'Italy': 'ITA',
               'New Zealand': 'NZL',
               'Brazil': 'BRA',
               'Canada': 'CAN',
               'Cuba': 'CUB',
               'Hungary': 'HUN',
               'Republic of Korea': 'KOR',
               'Jamaica': 'JAM',
               'Poland': 'POL',
               'Czech Republic': 'CZE',
               'Norway': 'NOR',
              'Kenya': 'KEN',
               'Croatia': 'CRO',
               'Spain': 'ESP',
               'Switzerland': 'SUI',
              'Denmark': 'DEN',
               'Sweden': 'SWE',
```

3.0

```
'Slovenia': 'SLO',
    'Serbia': 'SRB',
    'Islamic Republic of Iran': 'IRI',
    'Bulgaria': 'BUL',
    'Belgium': 'BEL',
    'Greece': 'GRE',
    'Ecuador': 'ECU',
    'Georgia': 'GEO',
    'Uganda': 'UGA',
    'Turkey': 'TUR',
    'Egypt': 'EGY',
    'Ukraine': 'UKR',
    'Belarus': 'BLR',
    'Portugal': 'POR',
    'Tunisia': 'TUN',
    'Philippines': 'PHI',
    'Romania': 'ROU',
    'Indonesia': 'INA',
    'India': 'IND',
    'Hong Kong, China': 'HKG',
    'Austria': 'AUT',
    'Ethiopia': 'ETH',
    'Slovakia': 'SVK',
    'Venezuela': 'VEN',
    'South Africa': 'RSA'
# Set plot title and labels
plt.title('Countries with Silver Medals in 2021 Tokyo Olympics')
plt.xlabel('Country')
plt.ylabel('Silver Medals')
plt.xticks(sorted_df['NOC'], [country_shortforms[c] for c in sorted_df['NOC']], rotation='vertical', fontsize=10)
# Hide top and right spines
plt.gca().spines['top'].set_visible(False)
plt.gca().spines['right'].set_visible(False)
# Add text annotations on top of each bar
for i, bar in enumerate(bars):
    plt.text(bar.get_x() + bar.get_width()/2, bar.get_height(), str(int(bar.get_height())), ha='center', va='bottom', fe
# Show the plot
```





countries\_with\_bronze = sorted\_df[sorted\_df['Bronze'] > 0]
countries\_with\_bronze

Out[212]:

|    | NOC                        | Gold | Silver | Bronze | TotalMedals |
|----|----------------------------|------|--------|--------|-------------|
| 0  | United States of America   | 39.0 | 41.0   | 33.0   | 113.0       |
| 1  | People's Republic of China | 38.0 | 32.0   | 18.0   | 88.0        |
| 2  | Japan                      | 27.0 | 14.0   | 17.0   | 58.0        |
| 3  | Great Britain              | 22.0 | 21.0   | 22.0   | 65.0        |
| 4  | ROC                        | 20.0 | 28.0   | 23.0   | 71.0        |
| 5  | Australia                  | 17.0 | 7.0    | 22.0   | 46.0        |
| 6  | Netherlands                | 10.0 | 12.0   | 14.0   | 36.0        |
| 7  | France                     | 10.0 | 12.0   | 11.0   | 33.0        |
| 8  | Germany                    | 10.0 | 11.0   | 16.0   | 37.0        |
| 9  | Italy                      | 10.0 | 10.0   | 20.0   | 40.0        |
| 10 | New Zealand                | 7.0  | 6.0    | 7.0    | 20.0        |
| 11 | Brazil                     | 7.0  | 6.0    | 8.0    | 21.0        |
| 12 | Canada                     | 7.0  | 6.0    | 11.0   | 24.0        |
| 13 | Cuba                       | 7.0  | 3.0    | 5.0    | 15.0        |
| 14 | Hungary                    | 6.0  | 7.0    | 7.0    | 20.0        |
| 15 | Republic of Korea          | 6.0  | 4.0    | 10.0   | 20.0        |
| 16 | Jamaica                    | 4.0  | 1.0    | 4.0    | 9.0         |
| 17 | Poland                     | 4.0  | 5.0    | 5.0    | 14.0        |
| 18 | Czech Republic             | 4.0  | 4.0    | 3.0    | 11.0        |
| 19 | Norway                     | 4.0  | 2.0    | 2.0    | 8.0         |
| 20 | Kenya                      | 4.0  | 4.0    | 2.0    | 10.0        |
| 21 | Croatia                    | 3.0  | 3.0    | 2.0    | 8.0         |
| 22 | Spain                      | 3.0  | 8.0    | 6.0    | 17.0        |
| 23 | Switzerland                | 3.0  | 4.0    | 6.0    | 13.0        |
| 24 | Denmark                    | 3.0  | 4.0    | 4.0    | 11.0        |

|    | NOC                      | Gold | Silver | Bronze | TotalMedals |
|----|--------------------------|------|--------|--------|-------------|
| 26 | Slovenia                 | 3.0  | 1.0    | 1.0    | 5.0         |
| 27 | Serbia                   | 3.0  | 1.0    | 5.0    | 9.0         |
| 28 | Islamic Republic of Iran | 3.0  | 2.0    | 2.0    | 7.0         |
| 29 | Bulgaria                 | 3.0  | 1.0    | 2.0    | 6.0         |
| 30 | Belgium                  | 3.0  | 1.0    | 3.0    | 7.0         |
| 32 | Greece                   | 2.0  | 1.0    | 1.0    | 4.0         |
| 37 | Georgia                  | 2.0  | 5.0    | 1.0    | 8.0         |
| 39 | Uganda                   | 2.0  | 1.0    | 1.0    | 4.0         |
| 40 | Turkey                   | 2.0  | 2.0    | 9.0    | 13.0        |
| 41 | Egypt                    | 1.0  | 1.0    | 4.0    | 6.0         |
| 43 | Ukraine                  | 1.0  | 6.0    | 12.0   | 19.0        |
| 44 | Belarus                  | 1.0  | 3.0    | 3.0    | 7.0         |
| 45 | Portugal                 | 1.0  | 1.0    | 2.0    | 4.0         |
| 48 | Philippines              | 1.0  | 2.0    | 1.0    | 4.0         |
| 50 | Indonesia                | 1.0  | 1.0    | 3.0    | 5.0         |
| 51 | India                    | 1.0  | 2.0    | 4.0    | 7.0         |
| 52 | Hong Kong, China         | 1.0  | 2.0    | 3.0    | 6.0         |
| 53 | Austria                  | 1.0  | 1.0    | 5.0    | 7.0         |
| 56 | Ethiopia                 | 1.0  | 1.0    | 2.0    | 4.0         |
| 57 | Slovakia                 | 1.0  | 2.0    | 1.0    | 4.0         |

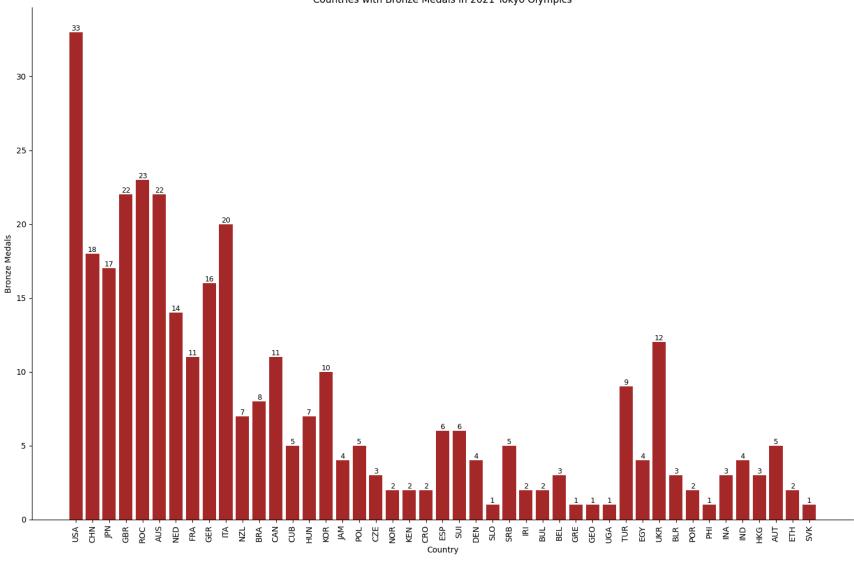
```
import matplotlib.pyplot as plt

# Plotting with improved readability
plt.figure(figsize=(15, 10))

sorted_df = sorted_df[sorted_df['Bronze'] > 0]
```

```
# Bar plot with filtered country names on x-axis and bronze medal counts on y-axis
bars = plt.bar(sorted_df['NOC'], sorted_df['Bronze'], color='brown')
# Short forms of countries
country short forms = {
    'United States of America': 'USA',
    "People's Republic of China": 'CHN',
    'Japan': 'JPN',
    'Great Britain': 'GBR',
    'ROC': 'ROC',
    'Australia': 'AUS',
    'Netherlands': 'NED',
    'France': 'FRA',
    'Germany': 'GER',
    'Italy': 'ITA',
    'New Zealand': 'NZL',
    'Brazil': 'BRA',
    'Canada': 'CAN',
    'Cuba': 'CUB',
    'Hungary': 'HUN',
    'Republic of Korea': 'KOR',
    'Jamaica': 'JAM',
    'Poland': 'POL',
    'Czech Republic': 'CZE',
    'Norway': 'NOR',
    'Kenya': 'KEN',
    'Croatia': 'CRO',
    'Spain': 'ESP',
    'Switzerland': 'SUI',
    'Denmark': 'DEN',
    'Slovenia': 'SLO',
    'Serbia': 'SRB',
    'Islamic Republic of Iran': 'IRI',
    'Bulgaria': 'BUL',
    'Belgium': 'BEL',
    'Greece': 'GRE',
    'Georgia': 'GEO',
    'Uganda': 'UGA',
    'Turkey': 'TUR',
    'Egypt': 'EGY',
    'Ukraine': 'UKR',
    'Belarus': 'BLR',
    'Portugal': 'POR',
    'Philippines': 'PHI',
    'Indonesia': 'INA',
```

```
'India': 'IND',
    'Hong Kong, China': 'HKG',
    'Austria': 'AUT',
    'Ethiopia': 'ETH',
    'Slovakia': 'SVK'
# Set plot title and labels
plt.title('Countries with Bronze Medals in 2021 Tokyo Olympics')
plt.xlabel('Country')
plt.ylabel('Bronze Medals')
plt.xticks(sorted_df['NOC'], [country_short_forms[c] for c in sorted_df['NOC']], rotation='vertical', fontsize=10)
# Hide top and right spines
plt.gca().spines['top'].set_visible(False)
plt.gca().spines['right'].set_visible(False)
# Add text annotations on top of each bar
for i, bar in enumerate(bars):
    plt.text(bar.get_x() + bar.get_width()/2, bar.get_height(), str(int(bar.get_height())), ha='center', va='bottom', f
# Show the plot
plt.tight_layout()
plt.show()
```



totalmedals= sorted\_df[sorted\_df['TotalMedals'] > 0]
totalmedals

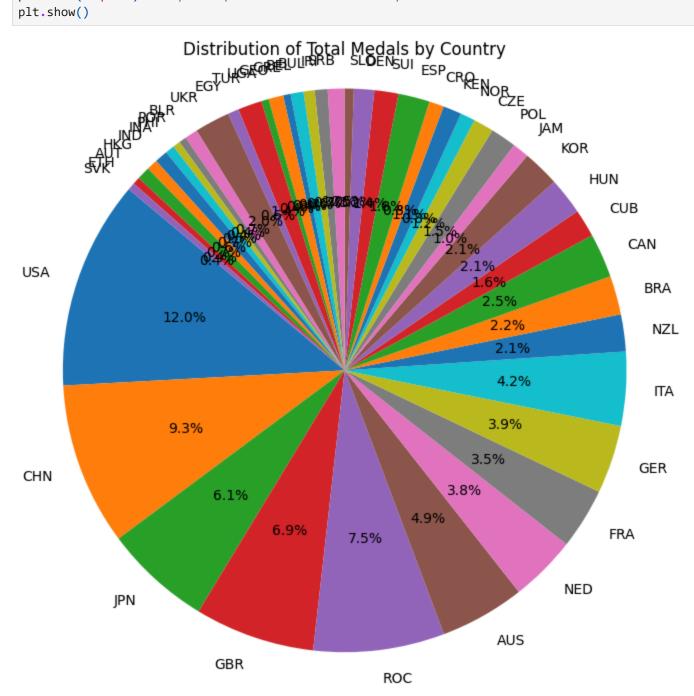
Out[217]:

|    | NOC                        | Gold | Silver | Bronze | TotalMedals |
|----|----------------------------|------|--------|--------|-------------|
| 0  | United States of America   | 39.0 | 41.0   | 33.0   | 113.0       |
| 1  | People's Republic of China | 38.0 | 32.0   | 18.0   | 88.0        |
| 2  | Japan                      | 27.0 | 14.0   | 17.0   | 58.0        |
| 3  | Great Britain              | 22.0 | 21.0   | 22.0   | 65.0        |
| 4  | ROC                        | 20.0 | 28.0   | 23.0   | 71.0        |
| 5  | Australia                  | 17.0 | 7.0    | 22.0   | 46.0        |
| 6  | Netherlands                | 10.0 | 12.0   | 14.0   | 36.0        |
| 7  | France                     | 10.0 | 12.0   | 11.0   | 33.0        |
| 8  | Germany                    | 10.0 | 11.0   | 16.0   | 37.0        |
| 9  | Italy                      | 10.0 | 10.0   | 20.0   | 40.0        |
| 10 | New Zealand                | 7.0  | 6.0    | 7.0    | 20.0        |
| 11 | Brazil                     | 7.0  | 6.0    | 8.0    | 21.0        |
| 12 | Canada                     | 7.0  | 6.0    | 11.0   | 24.0        |
| 13 | Cuba                       | 7.0  | 3.0    | 5.0    | 15.0        |
| 14 | Hungary                    | 6.0  | 7.0    | 7.0    | 20.0        |
| 15 | Republic of Korea          | 6.0  | 4.0    | 10.0   | 20.0        |
| 16 | Jamaica                    | 4.0  | 1.0    | 4.0    | 9.0         |
| 17 | Poland                     | 4.0  | 5.0    | 5.0    | 14.0        |
| 18 | Czech Republic             | 4.0  | 4.0    | 3.0    | 11.0        |
| 19 | Norway                     | 4.0  | 2.0    | 2.0    | 8.0         |
| 20 | Kenya                      | 4.0  | 4.0    | 2.0    | 10.0        |
| 21 | Croatia                    | 3.0  | 3.0    | 2.0    | 8.0         |
| 22 | Spain                      | 3.0  | 8.0    | 6.0    | 17.0        |
| 23 | Switzerland                | 3.0  | 4.0    | 6.0    | 13.0        |
| 24 | Denmark                    | 3.0  | 4.0    | 4.0    | 11.0        |

|    | NOC                      | Gold | Silver | Bronze | TotalMedals |
|----|--------------------------|------|--------|--------|-------------|
| 26 | Slovenia                 | 3.0  | 1.0    | 1.0    | 5.0         |
| 27 | Serbia                   | 3.0  | 1.0    | 5.0    | 9.0         |
| 28 | Islamic Republic of Iran | 3.0  | 2.0    | 2.0    | 7.0         |
| 29 | Bulgaria                 | 3.0  | 1.0    | 2.0    | 6.0         |
| 30 | Belgium                  | 3.0  | 1.0    | 3.0    | 7.0         |
| 32 | Greece                   | 2.0  | 1.0    | 1.0    | 4.0         |
| 37 | Georgia                  | 2.0  | 5.0    | 1.0    | 8.0         |
| 39 | Uganda                   | 2.0  | 1.0    | 1.0    | 4.0         |
| 40 | Turkey                   | 2.0  | 2.0    | 9.0    | 13.0        |
| 41 | Egypt                    | 1.0  | 1.0    | 4.0    | 6.0         |
| 43 | Ukraine                  | 1.0  | 6.0    | 12.0   | 19.0        |
| 44 | Belarus                  | 1.0  | 3.0    | 3.0    | 7.0         |
| 45 | Portugal                 | 1.0  | 1.0    | 2.0    | 4.0         |
| 48 | Philippines              | 1.0  | 2.0    | 1.0    | 4.0         |
| 50 | Indonesia                | 1.0  | 1.0    | 3.0    | 5.0         |
| 51 | India                    | 1.0  | 2.0    | 4.0    | 7.0         |
| 52 | Hong Kong, China         | 1.0  | 2.0    | 3.0    | 6.0         |
| 53 | Austria                  | 1.0  | 1.0    | 5.0    | 7.0         |
| 56 | Ethiopia                 | 1.0  | 1.0    | 2.0    | 4.0         |
| 57 | Slovakia                 | 1.0  | 2.0    | 1.0    | 4.0         |

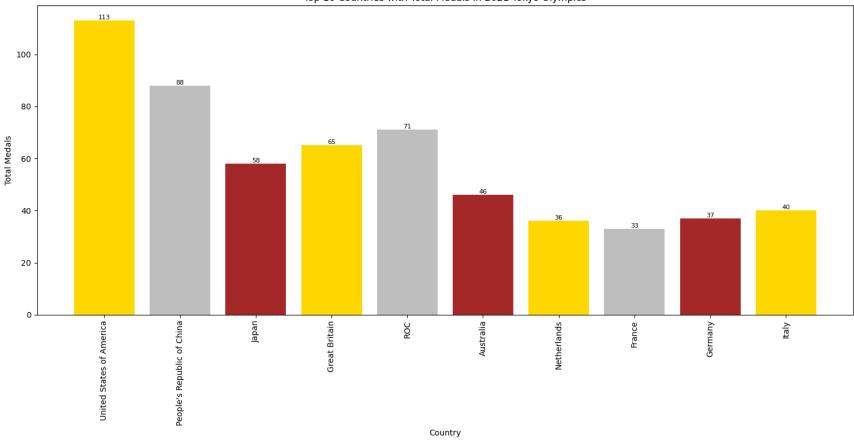
```
In [304...
#pie chart of total medals
country_short_forms = {
    'United States of America': 'USA',
    "People's Republic of China": 'CHN',
    'Japan': 'JPN',
    'Great Britain': 'GBR',
    'ROC': 'ROC',
```

```
'Australia': 'AUS',
    'Netherlands': 'NED',
    'France': 'FRA',
    'Germany': 'GER',
    'Italy': 'ITA',
    'New Zealand': 'NZL',
    'Brazil': 'BRA',
    'Canada': 'CAN',
    'Cuba': 'CUB',
    'Hungary': 'HUN',
    'Republic of Korea': 'KOR',
    'Jamaica': 'JAM',
    'Poland': 'POL',
    'Czech Republic': 'CZE',
    'Norway': 'NOR',
    'Kenya': 'KEN',
    'Croatia': 'CRO',
    'Spain': 'ESP',
    'Switzerland': 'SUI',
    'Denmark': 'DEN',
    'Slovenia': 'SLO',
    'Serbia': 'SRB',
    'Islamic Republic of Iran': 'IRI',
    'Bulgaria': 'BUL',
    'Belgium': 'BEL',
    'Greece': 'GRE',
    'Georgia': 'GEO',
    'Uganda': 'UGA',
    'Turkey': 'TUR',
    'Egypt': 'EGY',
    'Ukraine': 'UKR',
    'Belarus': 'BLR',
    'Portugal': 'POR',
    'Philippines': 'PHI',
    'Indonesia': 'INA',
    'India': 'IND',
    'Hong Kong, China': 'HKG',
    'Austria': 'AUT',
    'Ethiopia': 'ETH',
    'Slovakia': 'SVK'
# Create a pie chart for total medals won by each country
plt.figure(figsize=(8, 8))
plt.pie(sorted_df['TotalMedals'], labels=[country_short_forms[c] for c in sorted_df['NOC']], autopct='%1.1f%%', startan
```



```
# Selecting top 10 countries
In [223...
          top_10 = sorted_df.head(10)
          # Create a bar chart for top 10 countries with total medals
          plt.figure(figsize=(15, 8))
          bars = plt.bar(top_10['NOC'], top_10['TotalMedals'], color=['gold', 'silver', 'brown'])
          plt.xlabel('Country')
          plt.ylabel('Total Medals')
          plt.title('Top 10 Countries with Total Medals in 2021 Tokyo Olympics')
          plt.xticks(rotation='vertical')
          plt.tight_layout()
          # Add text annotations on top of each bar
          for bar in bars:
              yval = bar.get_height()
              plt.text(bar.get_x() + bar.get_width()/2, yval, int(yval), va='bottom', ha='center', fontsize=8)
          plt.show()
```

Top 10 Countries with Total Medals in 2021 Tokyo Olympics



```
In [237... from sklearn.linear_model import LinearRegression
import numpy as np

In [241... df = sorted_df
df.shape
df.insert(0, 'Year', [2021]*45)
df.head()
```

| • |   | Year | NOC                        | Gold | Silver | Bronze | TotalMedals |
|---|---|------|----------------------------|------|--------|--------|-------------|
|   | 0 | 2021 | United States of America   | 39.0 | 41.0   | 33.0   | 113.0       |
|   | 1 | 2021 | People's Republic of China | 38.0 | 32.0   | 18.0   | 88.0        |
|   | 2 | 2021 | Japan                      | 27.0 | 14.0   | 17.0   | 58.0        |
|   | 3 | 2021 | Great Britain              | 22.0 | 21.0   | 22.0   | 65.0        |
|   | 4 | 2021 | ROC                        | 20.0 | 28.0   | 23.0   | 71.0        |

Out[241]

```
# Assume a simple linear increase
In [248...
          years = np.array([2021]).reshape(-1, 1) # Current year
          future_years = np.array([2024]).reshape(-1, 1) # Year to predict
          # Prepare the linear regression model
          model = LinearRegression()
          # Predict each medal type
          predictions = {}
          for medal_type in ['Gold', 'Silver', 'Bronze', 'TotalMedals']:
              X = df['Year'].values.reshape(-1, 1)
              y = df[medal_type]
              # Fit the model
              model.fit(X, y)
              # Predict for 2024
              predictions[medal_type] = model.predict(future_years)[0]
          # Display the predictions
          for medal_type, prediction in predictions.items():
              print(f"Predicted {medal_type} in 2024: {prediction:.2f}")
          # Add predicted medals to the DataFrame for 2024
          df_pred = df.copy()
          df_pred['Gold'] = df_pred['Gold'] + (predictions['Gold'] - df_pred['Gold'].mean())
          df_pred['Silver'] = df_pred['Silver'] + (predictions['Silver'] - df_pred['Silver'].mean())
          df_pred['Bronze'] = df_pred['Bronze'] + (predictions['Bronze'] - df_pred['Bronze'].mean())
          df_pred['TotalMedals'] = df_pred['Gold'] + df_pred['Silver'] + df_pred['Bronze']
          # Sort DataFrame by predicted total medals
          df_pred = df_pred.sort_values(by='TotalMedals', ascending=False)
```

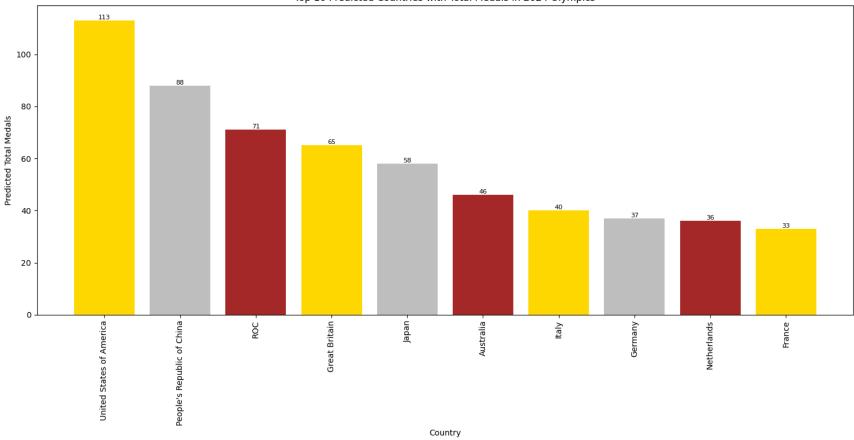
```
Predicted Gold in 2024: 6.87
Predicted Silver in 2024: 6.49
Predicted Bronze in 2024: 7.62
Predicted TotalMedals in 2024: 20.98
```

```
In [249... #Top 10 predicted countries with total medals
top_10_pred = df_pred.head(10)

plt.figure(figsize=(15, 8))
bars = plt.bar(top_10_pred['NOC'], top_10_pred['TotalMedals'], color=['gold', 'silver', 'brown'])
plt.xlabel('Country')
plt.ylabel('Predicted Total Medals')
plt.title('Top 10 Predicted Countries with Total Medals in 2024 Olympics')
plt.xticks(rotation='vertical')
plt.tight_layout()

# Add text annotations on top of each bar
for bar in bars:
    yval = bar.get_height()
    plt.text(bar.get_x() + bar.get_width()/2, yval, int(yval), va='bottom', ha='center', fontsize=8)

plt.show()
```



In [272... from google.colab import files
 uploaded = files.upload()

Choose files No file chosen

Upload widget is only available when the cell has been executed in the current browser

session. Please rerun this cell to enable.

Saving athlete\_events.csv to athlete\_events.csv

```
# Load the data
pred = pd.read_csv('athlete_events.csv')
pred.head()

# Create a DataFrame with counts of medals for each year
preddf = pred.groupby(['NOC','Year', 'Medal']).size().reset_index(name='Count')
```

```
# Display the first few rows of the new DataFrame preddf.head()
```

```
        Out[303]:
        NOC
        Year
        Medal
        Count

        0
        AFG
        2008
        Bronze
        1

        1
        AFG
        2012
        Bronze
        1

        2
        AHO
        1988
        Silver
        1

        3
        ALG
        1984
        Bronze
        2
```

**4** ALG 1992 Bronze

```
In [282... preddf.drop_duplicates(inplace=True)
    preddf.head()
```

Gold

201

1

## Vear Medal Count 0 1896 Bronze 38 1 1896 Gold 62 2 1896 Silver 43 3 1900 Bronze 175

**4** 1900

```
new = df
In [305...
          new = pd.melt(df, id_vars=['Year', 'NOC', 'TotalMedals'], value_vars=['Gold', 'Silver', 'Bronze'],
                            var_name='Medal', value_name='Count')
          new.drop(columns=['TotalMedals'], inplace=True)
          country_short_forms = {
              'United States of America': 'USA',
              "People's Republic of China": 'CHN',
              'Japan': 'JPN',
              'Great Britain': 'GBR',
              'ROC': 'ROC',
              'Australia': 'AUS',
              'Netherlands': 'NED',
              'France': 'FRA',
              'Germany': 'GER',
              'Italy': 'ITA',
```

```
'New Zealand': 'NZL',
    'Brazil': 'BRA',
    'Canada': 'CAN',
    'Cuba': 'CUB',
    'Hungary': 'HUN',
    'Republic of Korea': 'KOR',
    'Jamaica': 'JAM',
    'Poland': 'POL',
    'Czech Republic': 'CZE',
    'Norway': 'NOR',
    'Kenya': 'KEN',
    'Croatia': 'CRO',
    'Spain': 'ESP',
    'Switzerland': 'SUI',
    'Denmark': 'DEN',
    'Slovenia': 'SLO',
    'Serbia': 'SRB',
    'Islamic Republic of Iran': 'IRI',
    'Bulgaria': 'BUL',
    'Belgium': 'BEL',
    'Greece': 'GRE',
    'Georgia': 'GEO',
    'Uganda': 'UGA',
    'Turkey': 'TUR',
    'Egypt': 'EGY',
    'Ukraine': 'UKR',
    'Belarus': 'BLR',
    'Portugal': 'POR',
    'Philippines': 'PHI',
    'Indonesia': 'INA',
    'India': 'IND',
    'Hong Kong, China': 'HKG',
    'Austria': 'AUT',
    'Ethiopia': 'ETH',
    'Slovakia': 'SVK'
new['NOC'] = new['NOC'].map(country_short_forms)
new.head()
```

```
Out[305]:
             Year NOC Medal Count
          0 2021 USA
                        Gold
                               39.0
          1 2021 CHN
                        Gold
                               38.0
                  JPN
          2 2021
                        Gold
                               27.0
          3 2021
                  GBR
                        Gold
                               22.0
          4 2021 ROC
                        Gold
                               20.0
```

In [306... new.drop\_duplicates(inplace=True)
 new.head()

Out[306]:

|   | Year | NOC | Medal | Count |
|---|------|-----|-------|-------|
| 0 | 2021 | USA | Gold  | 39.0  |
| 1 | 2021 | CHN | Gold  | 38.0  |
| 2 | 2021 | JPN | Gold  | 27.0  |
| 3 | 2021 | GBR | Gold  | 22.0  |
| 4 | 2021 | ROC | Gold  | 20.0  |

```
In [313... preddf = pd.merge(preddf, new, on=['NOC','Year', 'Medal','Count'], how='left')
    preddf.head()
    preddf
```

```
Out[313]:
                NOC Year Medal Count
             0 AFG 2008 Bronze
             1 AFG 2012 Bronze
             2 AHO 1988
                           Silver
             3 ALG 1984 Bronze
                ALG 1992 Bronze
          3234 ZIM 2004 Bronze
          3235
                ZIM 2004
                            Gold
                ZIM 2004
          3236
                           Silver
          3237
                ZIM 2008
                            Gold
                                     1
                ZIM 2008
                                     3
          3238
                           Silver
```

3239 rows × 4 columns

```
In [308... model = LinearRegression()

# Assume a simple linear increase
future_years = np.array([2024]).reshape(-1, 1) # Year to predict

# Container for predictions
predictions = {}

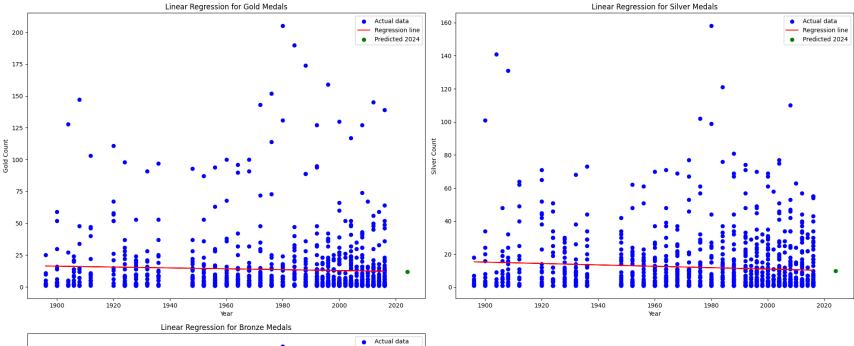
# Define medal types
medal_types = ['Gold', 'Silver', 'Bronze']

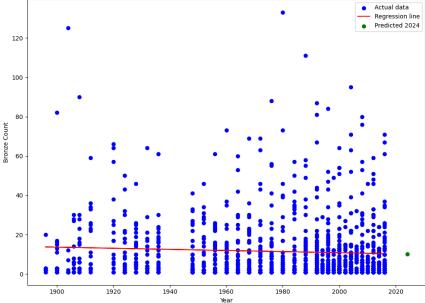
# Create a linear regression graph for each medal type
plt.figure(figsize=(20, 15))

for i, medal_type in enumerate(medal_types, start=1):
    plt.subplot(2, 2, i)

# Filter data for the current medal type
    medal_data = preddf[preddf['Medal'] == medal_type]
    X = medal_data['Year'].values.reshape(-1, 1)
```

```
y = medal_data['Count']
    # Fit the model
   model.fit(X, y)
   # Predict for 2024
    predictions[medal_type] = model.predict(future_years)[0]
   # Plot the actual data points
    plt.scatter(X, y, color='blue', label='Actual data')
    # Plot the regression line
    plt.plot(X, model.predict(X), color='red', label='Regression line')
    # Plot the prediction for 2024
    plt.scatter(future_years, predictions[medal_type], color='green', label='Predicted 2024')
    plt.xlabel('Year')
    plt.ylabel(f'{medal type} Count')
    plt.title(f'Linear Regression for {medal_type} Medals')
    plt.legend()
plt.tight_layout()
plt.show()
# Display the predictions
for medal_type, prediction in predictions.items():
    print(f"Predicted {medal type} count in 2024: {prediction:.2f}")
```





Predicted Gold count in 2024: 12.02 Predicted Silver count in 2024: 9.98 Predicted Bronze count in 2024: 10.13

```
predictions = {
In [317...
              'Gold': {},
              'Silver': {},
              'Bronze': {}
          # Define medal types
          medal_types = ['Gold', 'Silver', 'Bronze']
          # Loop through each medal type and perform linear regression
          for medal type in medal types:
              # Filter data for the current medal type
              medal_data = preddf[preddf['Medal'] == medal_type]
              # Prepare X (Years) and y (Counts)
              X = medal data[['Year']]
              y = medal_data['Count']
              # Initialize the model
              model = LinearRegression()
              # Fit the model
              model.fit(X, y)
              # Predict for 2024
              future_years = np.array([[2024]])
              predictions[medal_type] = model.predict(future_years)[0]
          # Aggregate total predicted medals per country
          predicted_totals = {}
          for country in df['NOC'].unique():
              total_medals = sum(predictions[medal_type] for medal_type in medal_types)
              predicted_totals[country] = total_medals
          # Sort countries by predicted total medals in descending order
          sorted_countries = sorted(predicted_totals.items(), key=lambda x: x[1], reverse=True)
          # Display the sorted list of countries with predicted total medals
          print("Predicted Medal Totals for 2024 Olympics:")
          for country, total_medals in sorted_countries:
              print(f"{country}: {total_medals:.2f} medals")
          # The country with the highest predicted total medals
```

```
predicted_winner = sorted_countries[0][0]
print(f"\nPredicted winner of the 2024 Olympics: {predicted_winner}")
```

Predicted Medal Totals for 2024 Olympics: United States of America: 32.13 medals People's Republic of China: 32.13 medals

Japan: 32.13 medals

Great Britain: 32.13 medals

ROC: 32.13 medals

Australia: 32.13 medals Netherlands: 32.13 medals

France: 32.13 medals Germany: 32.13 medals Italy: 32.13 medals

New Zealand: 32.13 medals Brazil: 32.13 medals

Canada: 32.13 medals Cuba: 32.13 medals Hungary: 32.13 medals

Republic of Korea: 32.13 medals

Jamaica: 32.13 medals Poland: 32.13 medals

Czech Republic: 32.13 medals

Norway: 32.13 medals Kenya: 32.13 medals Croatia: 32.13 medals Spain: 32.13 medals

Switzerland: 32.13 medals Denmark: 32.13 medals Slovenia: 32.13 medals Serbia: 32.13 medals

Islamic Republic of Iran: 32.13 medals

Bulgaria: 32.13 medals
Belgium: 32.13 medals
Greece: 32.13 medals
Georgia: 32.13 medals
Uganda: 32.13 medals
Turkey: 32.13 medals
Egypt: 32.13 medals
Ukraine: 32.13 medals
Belarus: 32.13 medals
Portugal: 32.13 medals

Philippines: 32.13 medals Indonesia: 32.13 medals

India: 32.13 medals

Hong Kong, China: 32.13 medals

Austria: 32.13 medals Ethiopia: 32.13 medals

## Slovakia: 32.13 medals

Predicted winner of the 2024 Olympics: United States of America

/usr/local/lib/python3.10/dist-packages/sklearn/base.py:439: UserWarning: X does not have valid feature names, but Line arRegression was fitted with feature names
 warnings.warn(
/usr/local/lib/python3.10/dist-packages/sklearn/base.py:439: UserWarning: X does not have valid feature names, but Line arRegression was fitted with feature names
 warnings.warn(
/usr/local/lib/python3.10/dist-packages/sklearn/base.py:439: UserWarning: X does not have valid feature names, but Line arRegression was fitted with feature names
 warnings.warn(

In [319...

!jupyter nbconvert --to html

```
This application is used to convert notebook files (*.ipynb)
        to various other formats.
        WARNING: THE COMMANDLINE INTERFACE MAY CHANGE IN FUTURE RELEASES.
Options 0
======
The options below are convenience aliases to configurable class-options,
as listed in the "Equivalent to" description-line of the aliases.
To see all configurable class-options for some <cmd>, use:
    <cmd> --help-all
--debug
    set log level to logging.DEBUG (maximize logging output)
    Equivalent to: [--Application.log level=10]
--show-config
    Show the application's configuration (human-readable format)
    Equivalent to: [--Application.show_config=True]
--show-config-json
    Show the application's configuration (json format)
    Equivalent to: [--Application.show_config_json=True]
--generate-config
    generate default config file
    Equivalent to: [--JupyterApp.generate config=True]
- y
    Answer yes to any questions instead of prompting.
    Equivalent to: [--JupyterApp.answer yes=True]
--execute
    Execute the notebook prior to export.
    Equivalent to: [--ExecutePreprocessor.enabled=True]
--allow-errors
    Continue notebook execution even if one of the cells throws an error and include the error message in the cell outp
ut (the default behaviour is to abort conversion). This flag is only relevant if '--execute' was specified, too.
    Equivalent to: [--ExecutePreprocessor.allow_errors=True]
--stdin
    read a single notebook file from stdin. Write the resulting notebook with default basename 'notebook.*'
    Equivalent to: [--NbConvertApp.from_stdin=True]
--stdout
    Write notebook output to stdout instead of files.
    Equivalent to: [--NbConvertApp.writer_class=StdoutWriter]
--inplace
    Run nbconvert in place, overwriting the existing notebook (only
            relevant when converting to notebook format)
    Equivalent to: [--NbConvertApp.use output suffix=False --NbConvertApp.export format=notebook --FilesWriter.build di
rectory=]
```

```
--clear-output
       Clear output of current file and save in place,
                       overwriting the existing notebook.
        Equivalent to: [--NbConvertApp.use output suffix=False --NbConvertApp.export format=notebook --FilesWriter.build di
rectory= --ClearOutputPreprocessor.enabled=True]
--no-prompt
       Exclude input and output prompts from converted document.
       Equivalent to: [--TemplateExporter.exclude input prompt=True --TemplateExporter.exclude output prompt=True]
--no-input
       Exclude input cells and output prompts from converted document.
                       This mode is ideal for generating code-free reports.
        Equivalent to: [--TemplateExporter.exclude output prompt=True --TemplateExporter.exclude input=True --Templa
ter.exclude input prompt=True]
--allow-chromium-download
       Whether to allow downloading chromium if no suitable version is found on the system.
       Equivalent to: [--WebPDFExporter.allow chromium download=True]
--disable-chromium-sandbox
       Disable chromium security sandbox when converting to PDF...
       Equivalent to: [--WebPDFExporter.disable sandbox=True]
--show-input
       Shows code input. This flag is only useful for dejavu users.
       Equivalent to: [--TemplateExporter.exclude_input=False]
--embed-images
       Embed the images as base64 dataurls in the output. This flag is only useful for the HTML/WebPDF/Slides exports.
       Equivalent to: [--HTMLExporter.embed_images=True]
--sanitize-html
       Whether the HTML in Markdown cells and cell outputs should be sanitized..
       Equivalent to: [--HTMLExporter.sanitize html=True]
--log-level=<Enum>
       Set the log level by value or name.
       Choices: any of [0, 10, 20, 30, 40, 50, 'DEBUG', 'INFO', 'WARN', 'ERROR', 'CRITICAL']
       Default: 30
       Equivalent to: [--Application.log level]
--config=<Unicode>
       Full path of a config file.
       Default: ''
       Equivalent to: [--JupyterApp.config_file]
--to=<Unicode>
       The export format to be used, either one of the built-in formats
                       ['asciidoc', 'custom', 'html', 'latex', 'markdown', 'notebook', 'pdf', 'python', 'rst', 'script', 'slides',
'webpdf']
                       or a dotted object name that represents the import path for an
                       ``Exporter`` class
       Default: ''
       Equivalent to: [--NbConvertApp.export_format]
```

```
--template=<Unicode>
    Name of the template to use
    Default: ''
    Equivalent to: [--TemplateExporter.template name]
--template-file=<Unicode>
    Name of the template file to use
    Default: None
    Equivalent to: [--TemplateExporter.template file]
--theme=<Unicode>
    Template specific theme(e.g. the name of a JupyterLab CSS theme distributed
    as prebuilt extension for the lab template)
    Default: 'light'
    Equivalent to: [--HTMLExporter.theme]
--sanitize_html=<Bool>
    Whether the HTML in Markdown cells and cell outputs should be sanitized. This
    should be set to True by nbviewer or similar tools.
    Default: False
    Equivalent to: [--HTMLExporter.sanitize html]
--writer=<DottedObjectName>
    Writer class used to write the
                                        results of the conversion
    Default: 'FilesWriter'
    Equivalent to: [--NbConvertApp.writer_class]
--post=<DottedOrNone>
    PostProcessor class used to write the
                                        results of the conversion
    Default: ''
    Equivalent to: [--NbConvertApp.postprocessor class]
--output=<Unicode>
    overwrite base name use for output files.
                can only be used when converting one notebook at a time.
   Default: ''
    Equivalent to: [--NbConvertApp.output base]
--output-dir=<Unicode>
    Directory to write output(s) to. Defaults
                                  to output to the directory of each notebook. To recover
                                  previous default behaviour (outputting to the current
                                  working directory) use . as the flag value.
    Default: ''
    Equivalent to: [--FilesWriter.build_directory]
--reveal-prefix=<Unicode>
    The URL prefix for reveal.js (version 3.x).
            This defaults to the reveal CDN, but can be any url pointing to a copy
            of reveal.js.
            For speaker notes to work, this must be a relative path to a local
```

```
copy of reveal.js: e.g., "reveal.js".
            If a relative path is given, it must be a subdirectory of the
            current directory (from which the server is run).
            See the usage documentation
            (https://nbconvert.readthedocs.io/en/latest/usage.html#reveal-js-html-slideshow)
            for more details.
    Default: ''
    Equivalent to: [--SlidesExporter.reveal url prefix]
--nbformat=<Enum>
    The nbformat version to write.
            Use this to downgrade notebooks.
    Choices: any of [1, 2, 3, 4]
    Default: 4
    Equivalent to: [--NotebookExporter.nbformat_version]
Examples
    The simplest way to use nbconvert is
            > jupyter nbconvert mynotebook.ipynb --to html
            Options include ['asciidoc', 'custom', 'html', 'latex', 'markdown', 'notebook', 'pdf', 'python', 'rst', 'sc
ript', 'slides', 'webpdf'].
            > jupyter nbconvert --to latex mynotebook.ipynb
            Both HTML and LaTeX support multiple output templates. LaTeX includes
            'base', 'article' and 'report'. HTML includes 'basic', 'lab' and
            'classic'. You can specify the flavor of the format used.
            > jupyter nbconvert --to html --template lab mynotebook.ipynb
            You can also pipe the output to stdout, rather than a file
            > jupyter nbconvert mynotebook.ipynb --stdout
            PDF is generated via latex
            > jupyter nbconvert mynotebook.ipynb --to pdf
            You can get (and serve) a Reveal.js-powered slideshow
            > jupyter nbconvert myslides.ipynb --to slides --post serve
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