Clustering with Countries_exercise.csv

Import the relevant libraries

```
In [4]: import pandas as pd
            import matplotlib.pyplot as plt
            {\tt import} \ {\tt seaborn} \ {\tt as} \ {\tt sns}
            sns.set()
            \textbf{from} \text{ sklearn.cluster } \textbf{import} \text{ KMeans}
```

Load the data

Out[5]:

In [6]:

```
# Load the data
In [5]:
        raw data = pd.read csv('Countries exercise.csv')
        # Check the data
        raw data
```

	name	Longitude	Latitude
0	Aruba	-69.982677	12.520880
1	Afghanistan	66.004734	33.835231
2	Angola	17.537368	-12.293361
3	Anguilla	-63.064989	18.223959
4	Albania	20.049834	41.142450
•••			
236	Samoa	-172.164851	-13.753243
237	Yemen	47.586762	15.909280
238	South Africa	25.083901	-29.000341
239	Zambia	27.774759	-13.458242
240	Zimbabwe	29.851441	-19.004204

241 rows × 3 columns

data = raw_data.copy()

Remove the duplicate index column from the dataset.

Plot the data

In [7]: plt.scatter(data['Longitude'], data['Latitude'])

Plot the 'Longtitude' and 'Latitude' columns.

```
plt.xlim(-180,180)
plt.ylim(-90, 90)
plt.show()
  80
  60
 40
 20
   0
-20
-40
```

0

50

100

150

In [8]: x = data.iloc[:,1:3]

Select the features

-150

-100

-50

-60

-80

Create a copy of that data and remove all parameters apart from *Longitude* and *Latitude*.

```
Out[8]:
                 Longitude
                               Latitude
             0
                 -69.982677
                             12.520880
                             33.835231
                 66.004734
                 17.537368
                            -12.293361
                 -63.064989
                             18.223959
                  20.049834
                             41.142450
```

-172.164851 -13.753243 236 15.909280 237 47.586762 25.083901 -29.000341 238 239 27.774759 -13.458242 240 29.851441 -19.004204 241 rows × 2 columns

kmeans = KMeans(7)

In [9]:

In [10]:

Clustering

kmeans.fit(x)

I'm changing index in kmeans() then running the remaining kernels.

```
KMeans(n_clusters=7)
Out[10]:
        Clustering Resutls
In [11]:
        identified clusters = kmeans.fit predict(x)
```

array([1, 0, 3, 1, 4, 4, 4, 0, 1, 0, 6, 3, 2, 3, 1, 2, 4, 0, 3, 4,

Aruba

1 Afghanistan

identified clusters

```
Out[11]:
                0, 4, 0, 1, 1, 4, 1, 4, 1, 1, 1, 1, 1, 2, 0, 3, 5, 4, 1, 0, 5, 5,
                3, 5, 6, 1, 3, 5, 1, 1, 1, 1, 4, 4, 4, 4, 3, 1, 4, 1, 5, 1, 4, 3,
                4, 4, 3, 4, 2, 1, 4, 4, 2, 5, 4, 0, 4, 5, 5, 5, 5, 5, 4, 1, 4, 1,
                2, 1, 2, 3, 1, 4, 1, 4, 2, 4, 0, 2, 3, 4, 0, 0, 4, 4, 4, 1, 4, 4,
```

```
3, 2, 5, 2, 5, 1, 6, 4, 4, 0, 2, 2, 0, 0, 1, 6, 1, 2, 2, 2, 4, 1,
                2, 4, 1, 4, 6, 0, 4, 0, 3, 5, 0, 5, 3, 5, 2, 1, 5, 2, 5, 1, 4, 3,
                3, 1, 4, 5, 1, 4, 4, 4, 3, 1, 3, 4, 1, 5, 5, 2, 0, 0, 2, 6, 1, 4,
                4, 2, 3, 3, 4, 1, 1, 0, 4, 1, 1, 1, 1, 2, 2, 6, 6, 0, 3, 3, 3])
In [12]: data_with_clusters = data.copy()
         data with clusters['Cluster'] = identified clusters
         data with clusters
Out[12]:
                        Longitude
                                   Latitude Cluster
                 name
```

0

12.520880

33.835231

2	Angola	17.537368	-12.293361	3	
3	Anguilla	-63.064989	18.223959	1	
4	Albania	20.049834	41.142450	4	
•••					
236	Samoa	-172.164851	-13.753243	6	
237	Yemen	47.586762	15.909280	0	
238	South Africa	25.083901	-29.000341	3	
239	Zambia	27.774759	-13.458242	3	
240	Zimbabwe	29.851441	-19.004204	3	
241 rows × 4 columns					
<pre>plt.scatter(data['Longitude'], data['Lat plt.xlim(-180,180) plt.ylim(-90, 90) plt.show()</pre>					

-69.982677

66.004734

```
60
```



In []:

In [13]: