world-data

November 24, 2024

0.1 Importing Libraries

[2]:
```
import pandas as pd
import numpy as np
import seaborn as sns
import matplotlib.pyplot as plt
```

0.2 Load and Check the DataSet

[4]:
```
df = pd.read_csv(r"D:\ANURAG KUMAR\Project\Data Analysis\Projeact 4\WorldData.
↪csv")
df.head()
```

[4]:

	Code	Name	Continent	Region	SurfaceArea
0	ABW	Aruba	North America	Caribbean	193.0
1	AFG	Afghanistan	Asia	Southern and Central Asia	652090.0
2	AGO	Angola	Africa	Central Africa	1246700.0
3	AIA	Anguilla	North America	Caribbean	96.0
4	ALB	Albania	Europe	Southern Europe	28748.0

	IndepYear	Population	LifeExpectancy	GNP	GNPOld
0	NaN	103000	78.4	828.0	793.0
1	1919.0	22720000	45.9	5976.0	NaN
2	1975.0	12878000	38.3	6648.0	7984.0
3	NaN	8000	76.1	63.2	NaN
4	1912.0	3401200	71.6	3205.0	2500.0

	LocalName	GovernmentForm
0	Aruba	Nonmetropolitan Territory of The Netherlands
1	Afganistan/Afqanestan	Islamic Emirate
2	Angola	Republic
3	Anguilla	Dependent Territory of the UK
4	Shqipëria	Republic

	HeadOfState	Capital	Code2
0	Beatrix	129.0	AW
1	Mohammad Omar	1.0	AF
2	José Eduardo dos Santos	56.0	AO

```
3             Elisabeth II     62.0    AI
4           Rexhep Mejdani     34.0    AL
```

0.3 Get all the Information About the DataSet

[6]:
```
df. info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 239 entries, 0 to 238
Data columns (total 15 columns):
 #   Column          Non-Null Count  Dtype
---  ------          --------------  -----
 0   Code            239 non-null    object
 1   Name            239 non-null    object
 2   Continent       239 non-null    object
 3   Region          239 non-null    object
 4   SurfaceArea     239 non-null    float64
 5   IndepYear       192 non-null    float64
 6   Population      239 non-null    int64
 7   LifeExpectancy  222 non-null    float64
 8   GNP             239 non-null    float64
 9   GNPOld          178 non-null    float64
 10  LocalName       239 non-null    object
 11  GovernmentForm  239 non-null    object
 12  HeadOfState     236 non-null    object
 13  Capital         232 non-null    float64
 14  Code2           238 non-null    object
dtypes: float64(6), int64(1), object(8)
memory usage: 28.1+ KB
```

0.4 Get Statistical Data on DataSet

[8]:
```
df.describe()
```

[8]:
```
         SurfaceArea    IndepYear    Population  LifeExpectancy           GNP  \
count   2.390000e+02   192.000000  2.390000e+02      222.000000  2.390000e+02
mean    6.232481e+05  1847.260417  2.543410e+07       66.486036  1.228239e+05
std     1.924140e+06   420.831370  1.093398e+08       11.519267  6.379976e+05
min     4.000000e-01 -1523.000000  0.000000e+00       37.200000  0.000000e+00
25%     2.275000e+03  1906.750000  2.380000e+05       60.300000  6.400000e+02
50%     7.174000e+04  1960.000000  3.869000e+06       70.150000  4.787000e+03
75%     3.987545e+05  1974.000000  1.493550e+07       75.500000  2.994450e+04
max     1.707540e+07  1994.000000  1.277558e+09       83.500000  8.510700e+06

              GNPOld      Capital
count   1.780000e+02   232.000000
mean    1.655343e+05  2071.306034
```

```
std    7.204689e+05  1184.095609
min    1.570000e+02     1.000000
25%    2.187000e+03   915.750000
50%    8.421000e+03  2449.500000
75%    7.114550e+04  3065.250000
max    8.110900e+06  4074.000000
```

0.5 Check For Missing Values

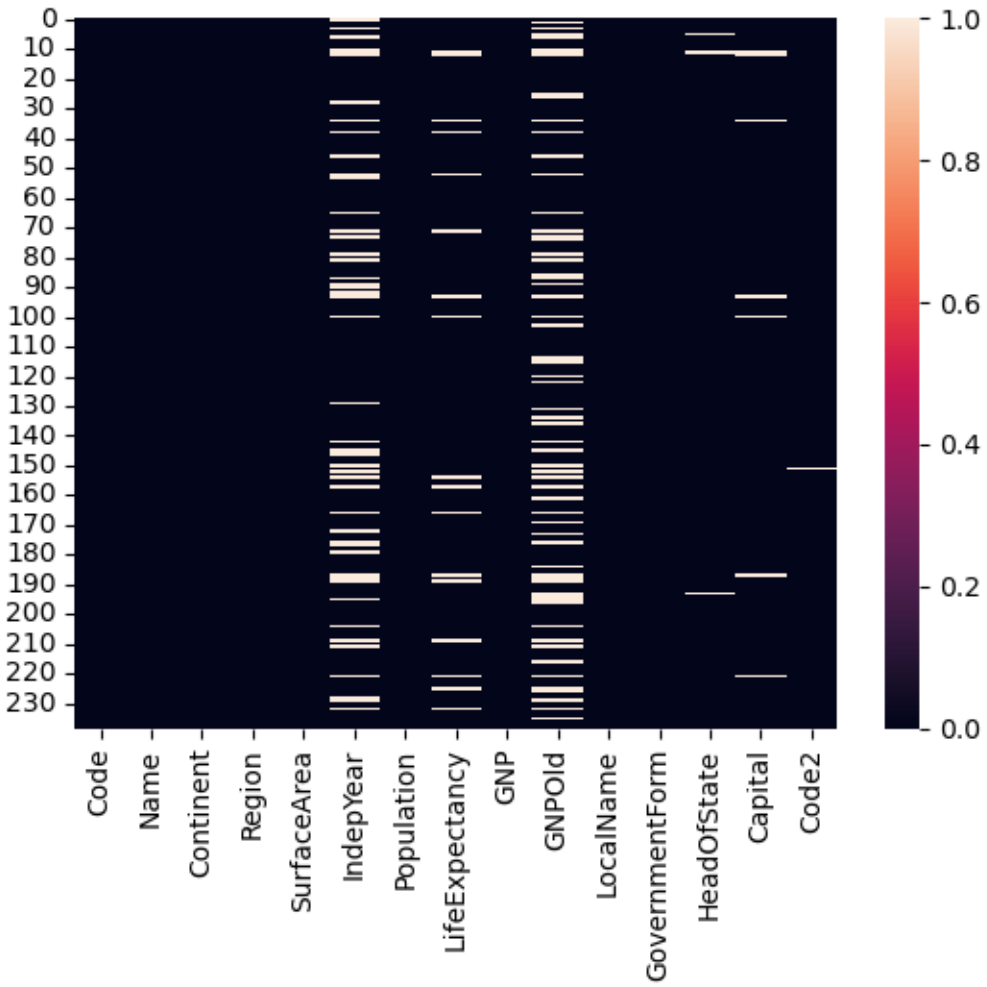
[10]:
```
df.isnull().sum()
# OR
df.isna().sum()
# Both Gives Same result
```

[10]:
```
Code              0
Name              0
Continent         0
Region            0
SurfaceArea       0
IndepYear        47
Population        0
LifeExpectancy   17
GNP               0
GNPOld           61
LocalName         0
GovernmentForm    0
HeadOfState       3
Capital           7
Code2             1
dtype: int64
```

0.6 Visualizing Missing values

[12]:
```
sns.heatmap(df.isnull())
```

[12]:
```
<Axes: >
```



0.7 Remove Unwanted Columns

[14]:
```
df.drop(columns='GNP', inplace=True)
df.drop(columns='GNPOld', inplace=True)
df.head()
```

[14]:

	Code	Name	Continent	Region	SurfaceArea	IndepYear	Population	LifeExpectancy	LocalName
0	ABW	Aruba	North America	Caribbean	193.0	NaN	103000	78.4	Aruba
1	AFG	Afghanistan	Asia	Southern and Central Asia	652090.0	1919.0	22720000	45.9	Afganistan/Afqanestan
2	AGO	Angola	Africa	Central Africa	1246700.0				
3	AIA	Anguilla	North America	Caribbean	96.0				
4	ALB	Albania	Europe	Southern Europe	28748.0				

```
2     1975.0    12878000            38.3                Angola
3        NaN        8000            76.1             Anguilla
4     1912.0     3401200            71.6              Shqipëria

                                GovernmentForm              HeadOfState  \
0  Nonmetropolitan Territory of The Netherlands                  Beatrix
1                               Islamic Emirate            Mohammad Omar
2                                     Republic  José Eduardo dos Santos
3                 Dependent Territory of the UK              Elisabeth II
4                                     Republic            Rexhep Mejdani

   Capital Code2
0    129.0    AW
1      1.0    AF
2     56.0    AO
3     62.0    AI
4     34.0    AL
```

0.8 Renaming a Column

[17]:
```
df.rename(columns = {'IndepYear' : 'IndependentYear'}, inplace = True)
```

0.9 Extract the names of all the columns in the Dataset

[20]:
```
df.columns
```

[20]:
```
Index(['Code', 'Name', 'Continent', 'Region', 'SurfaceArea', 'IndependentYear',
       'Population', 'LifeExpectancy', 'LocalName', 'GovernmentForm',
       'HeadOfState', 'Capital', 'Code2'],
      dtype='object')
```

0.10 Check Datatypes of all the Columns

[23]:
```
df.dtypes
```

[23]:
```
Code               object
Name               object
Continent          object
Region             object
SurfaceArea       float64
IndependentYear   float64
Population          int64
LifeExpectancy    float64
LocalName          object
GovernmentForm     object
HeadOfState        object
```

```
Capital            float64
Code2               object
dtype: object
```

0.11 Handling Missing or NULL Values

[26]:
```
df['IndependentYear'] = df['IndependentYear'].fillna(0) #Replacing With 0

df['LifeExpectancy'] = df['LifeExpectancy'].fillna(df['LifeExpectancy'].mean())␣
 ↪ # Replace with mean
df['Capital'] = df['Capital'].fillna(df['Capital'].mean())  # Replace with mean

df['HeadOfState'] = df['HeadOfState'].fillna('Unknown')   # Replace with␣
 ↪"Unknown"
df['Code2'] = df['Code2'].fillna('N/A')   # Replace with "N/A"
```

[28]:
```
df.isnull().sum()
```

[28]:
```
Code               0
Name               0
Continent          0
Region             0
SurfaceArea        0
IndependentYear    0
Population         0
LifeExpectancy     0
LocalName          0
GovernmentForm     0
HeadOfState        0
Capital            0
Code2              0
dtype: int64
```

0.12 Count Number of Unique Values in Each Column

[31]:
```
df.nunique()
```

[31]:
```
Code               239
Name               239
Continent            7
Region              25
SurfaceArea        238
IndependentYear     89
Population         226
LifeExpectancy     161
LocalName          239
```

```
GovernmentForm      35
HeadOfState        179
Capital            233
Code2              239
dtype: int64
```

0.13 Name All the Unique Values in 'Continent' Column

[34]:
```
df['Continent'].value_counts()
      # OR
df.Continent.value_counts()    # Both gives same result
```

[34]:
```
Continent
Africa           58
Asia             51
Europe           46
North America    37
Oceania          28
South America    14
Antarctica        5
Name: count, dtype: int64
```

0.14 Get all the values of 'Antarctica' Continent

[37]:
```
# Using Filter Method
df[df['Continent'] == 'Antarctica']

# Using groupby() method
df.groupby('Continent').get_group('Antarctica')
```

[37]:
```
     Code                                        Name   Continent  \
11    ATA                                  Antarctica  Antarctica
12    ATF                 French Southern territories  Antarctica
34    BVT                               Bouvet Island  Antarctica
93    HMD           Heard Island and McDonald Islands  Antarctica
187   SGS  South Georgia and the South Sandwich Islands  Antarctica

          Region  SurfaceArea  IndependentYear  Population  LifeExpectancy  \
11    Antarctica   13120000.0              0.0           0       66.486036
12    Antarctica       7780.0              0.0           0       66.486036
34    Antarctica         59.0              0.0           0       66.486036
93    Antarctica        359.0              0.0           0       66.486036
187   Antarctica       3903.0              0.0           0       66.486036

                                     LocalName  \
11                                           -
```

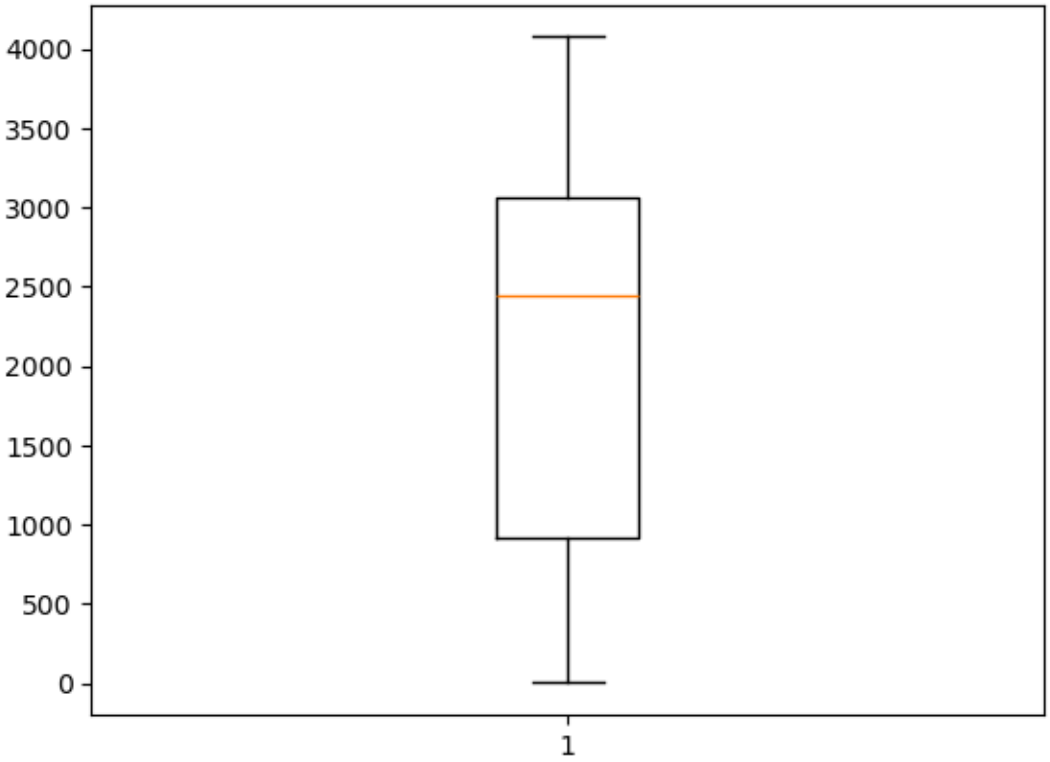
```
12                    Terres australes françaises
34                                     Bouvetøya
93                     Heard and McDonald Islands
187  South Georgia and the South Sandwich Islands

                          GovernmentForm     HeadOfState      Capital Code2
11                       Co-administrated          Unknown  2071.306034    AQ
12   Nonmetropolitan Territory of France  Jacques Chirac  2071.306034    TF
34         Dependent Territory of Norway         Harald V  2071.306034    BV
93                Territory of Australia     Elisabeth II  2071.306034    HM
187        Dependent Territory of the UK     Elisabeth II  2071.306034    GS
```

0.15 Visualize and analyse the "Capital" Column

[40]:
```
y = list(df.Capital)
 plt.boxplot(y)
 plt.show
```

[40]:
```
<function matplotlib.pyplot.show(close=None, block=None)>
```



[42]:
```
df['Capital'].max()  # Maximum Value
```

[42]: 4074.0

[44]:
```
df['Capital'].min()  # Minimum Value
```

[44]: 1.0

[46]:
```
df['Capital'].mean()  # Mean Value
```

[46]: 2071.3060344827586

[50]:
```
df['Capital'].sum()  # Sum of all values
```

[50]: 495042.1422413793

0.16 Find Out the Maximum Capital Of Each Region According to their Continents

[53]:
```
df.groupby(['Continent','Region']).max('Region')
```

[53]:

Continent	Region	SurfaceArea	IndependentYear
Africa	Central Africa	2344858.0	1975.0
	Eastern Africa	1104300.0	1993.0
	Northern Africa	2505813.0	1962.0
	Southern Africa	1221037.0	1990.0
	Western Africa	1267000.0	1975.0
Antarctica	Antarctica	13120000.0	0.0
Asia	Eastern Asia	9572900.0	1948.0
	Middle East	2149690.0	1991.0
	Southeast Asia	1904569.0	1984.0
	Southern and Central Asia	3287263.0	1991.0
Europe	Baltic Countries	65301.0	1991.0
	British Islands	242900.0	1921.0
	Eastern Europe	17075400.0	1993.0
	Nordic Countries	449964.0	1944.0
	Southern Europe	505992.0	1992.0
	Western Europe	551500.0	1955.0
North America	Caribbean	110861.0	1983.0
	Central America	1958201.0	1981.0
	North America	9970610.0	1867.0
Oceania	Australia and New Zealand	7741220.0	1907.0
	Melanesia	462840.0	1980.0
	Micronesia	726.0	1994.0
	Micronesia/Caribbean	16.0	0.0
	Polynesia	4000.0	1978.0
South America	South America	8547403.0	1975.0

		Population	LifeExpectancy	\
Continent	Region			
Africa	Central Africa	51654000	65.300000	
	Eastern Africa	62565000	72.700000	
	Northern Africa	68470000	75.500000	
	Southern Africa	40377000	51.100000	
	Western Africa	111506000	76.800000	
Antarctica	Antarctica	0	66.486036	
Asia	Eastern Asia	1277558000	81.600000	
	Middle East	66591000	78.600000	
	Southeast Asia	212107000	80.100000	
	Southern and Central Asia	1013662000	71.800000	
Europe	Baltic Countries	3698500	69.500000	
	British Islands	59623400	77.700000	
	Eastern Europe	146934000	74.500000	
	Nordic Countries	8861400	79.600000	
	Southern Europe	57680000	83.500000	
	Western Europe	82164700	79.600000	
North America	Caribbean	11201000	78.900000	
	Central America	98881000	75.800000	
	North America	278357000	79.400000	
Oceania	Australia and New Zealand	18886000	79.800000	
	Melanesia	4807000	72.800000	
	Micronesia	168000	77.800000	
	Micronesia/Caribbean	0	66.486036	
	Polynesia	235000	75.100000	
South America	South America	170115000	76.100000	

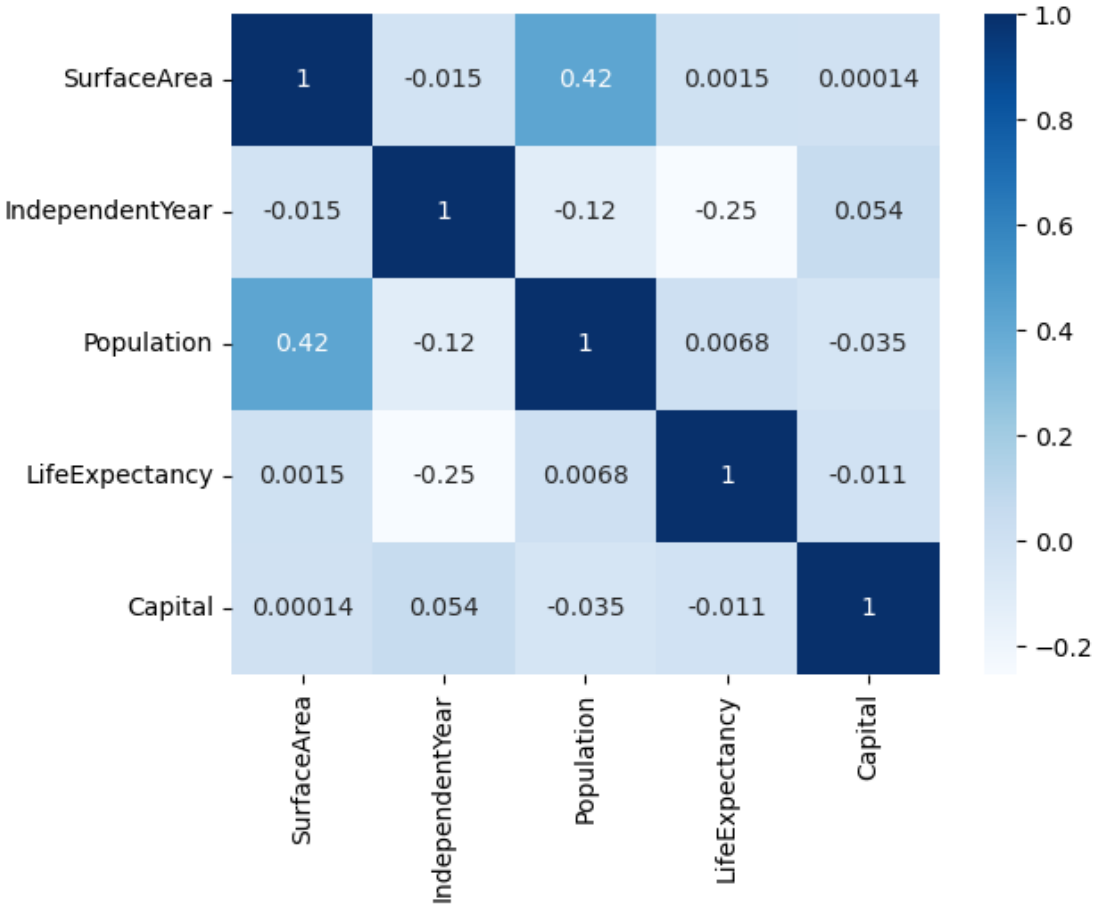
		Capital
Continent	Region	
Africa	Central Africa	3337.000000
	Eastern Africa	4068.000000
	Northern Africa	3349.000000
	Southern Africa	3244.000000
	Western Africa	3332.000000
Antarctica	Antarctica	2071.306034
Asia	Eastern Asia	3263.000000
	Middle East	4074.000000
	Southeast Asia	3770.000000
	Southern and Central Asia	3503.000000
Europe	Baltic Countries	3791.000000
	British Islands	1447.000000
	Eastern Europe	3580.000000
	Nordic Countries	3315.000000
	Southern Europe	3538.000000
	Western Europe	3248.000000
North America	Caribbean	4067.000000

```
                Central America            2882.000000
                North America              3813.000000
Oceania         Australia and New Zealand  3499.000000
                Melanesia                  3537.000000
                Micronesia                 2913.000000
                Micronesia/Caribbean       2071.306034
                Polynesia                  3536.000000
South America   South America              3539.000000
```

0.17 Find out the correlation between different variables in the given dataset

[56]:
```
numeric_df = df.select_dtypes(include=['number'])
sns.heatmap(numeric_df.corr(), cbar=True,annot=True,cmap='Blues')
```

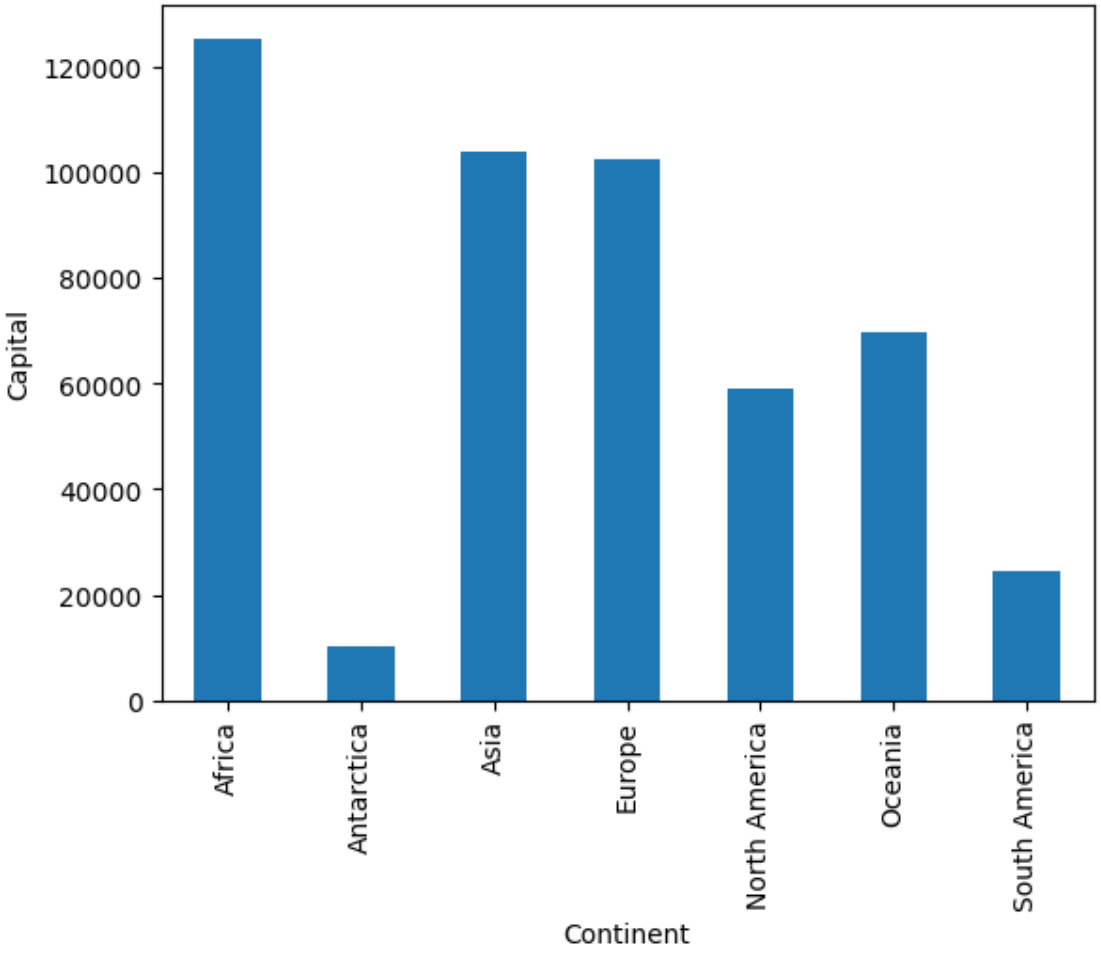
[56]: <Axes: >



0.18 Analyse how the capital of a region affects the Continent

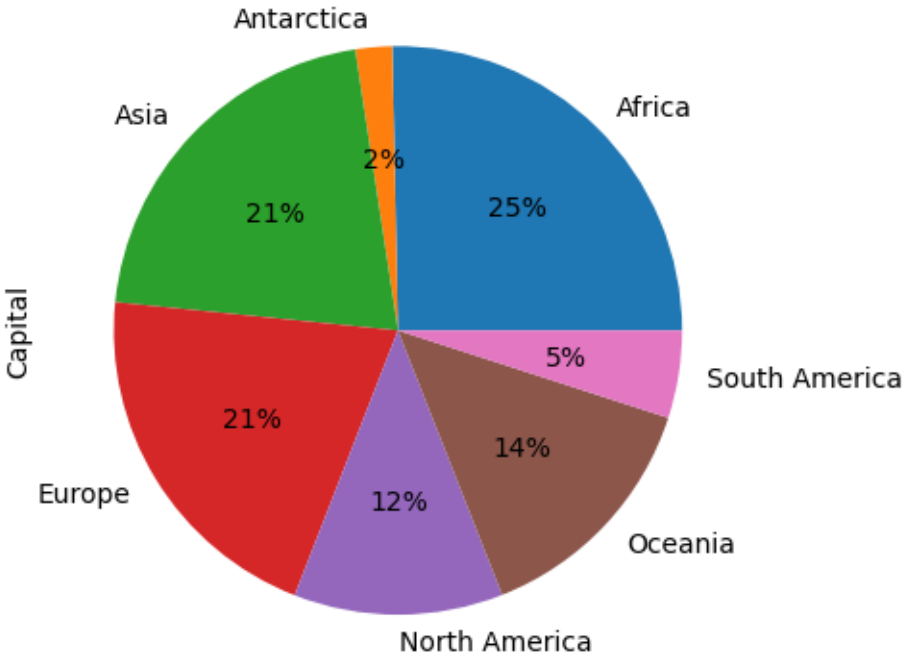
[60]:
```
df.groupby("Continent")['Capital'].sum().plot.bar()
plt.xlabel('Continent')
plt.ylabel('Capital')
```

[60]:
```
Text(0, 0.5, 'Capital')
```



[62]:
```
df.groupby("Continent")['Capital'].sum().plot.pie(autopct="%1.0f%%")
```

[62]:
```
<Axes: ylabel='Capital'>
```


Antarctica
Asia
Africa
2%
21%
25%
Capital
5%
South America
21%
14%
Europe
12%
Oceania
North America