

L3-BCU_KBP-Part1

April 8, 2021

1 TP 3 : Scientific operations tools

1.1 Part I. - Socio-Eco. database

1.1.1 I.1) Import module gapminder and use the method ; head(), describe() & info() to familiarize with it.

```
[38]: from gapminder import gapminder as gap
import pandas as pd
import numpy as np
from matplotlib import pyplot as plt

pd.DataFrame.head(gap)
```

```
[38]:
```

	country	continent	year	lifeExp	pop	gdpPercap
0	Afghanistan	Asia	1952	28.801	8425333	779.445314
1	Afghanistan	Asia	1957	30.332	9240934	820.853030
2	Afghanistan	Asia	1962	31.997	10267083	853.100710
3	Afghanistan	Asia	1967	34.020	11537966	836.197138
4	Afghanistan	Asia	1972	36.088	13079460	739.981106

```
[39]: pd.DataFrame.describe(gap)
```

```
[39]:
```

	year	lifeExp	pop	gdpPercap
count	1704.00000	1704.000000	1.704000e+03	1704.000000
mean	1979.50000	59.474439	2.960121e+07	7215.327081
std	17.26533	12.917107	1.061579e+08	9857.454543
min	1952.00000	23.599000	6.001100e+04	241.165876
25%	1965.75000	48.198000	2.793664e+06	1202.060309
50%	1979.50000	60.712500	7.023596e+06	3531.846988
75%	1993.25000	70.845500	1.958522e+07	9325.462346
max	2007.00000	82.603000	1.318683e+09	113523.132900

```
[40]: pd.DataFrame.info(gap)
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1704 entries, 0 to 1703
Data columns (total 6 columns):
#   Column      Non-Null Count  Dtype
#
```

```

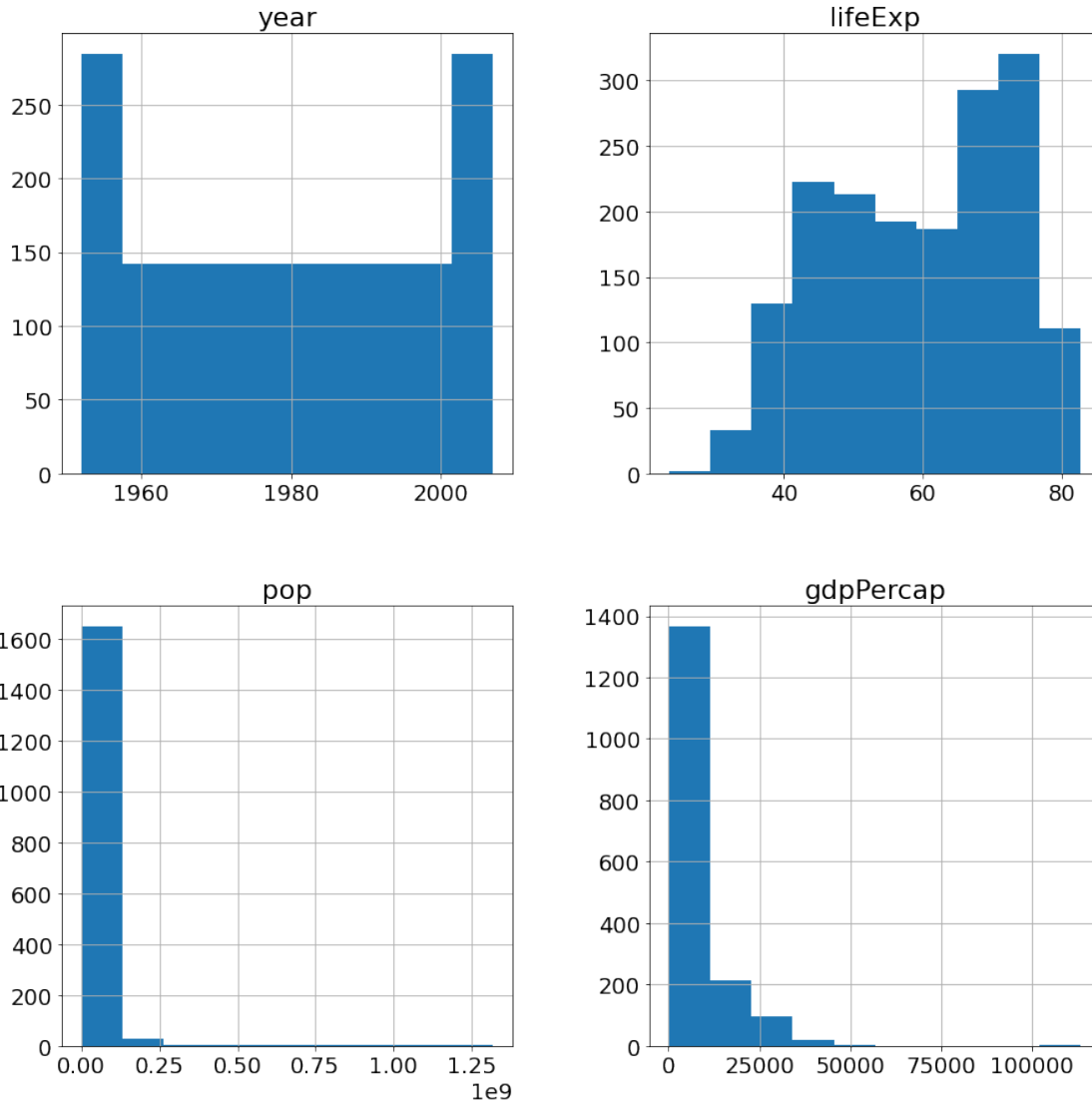
---  -----  -----  -----
0   country    1704 non-null  object
1   continent  1704 non-null  object
2   year       1704 non-null  int64
3   lifeExp    1704 non-null  float64
4   pop        1704 non-null  int64
5   gdpPercap  1704 non-null  float64
dtypes: float64(2), int64(2), object(2)
memory usage: 80.0+ KB

```

1.1.2 I.2) Use `pandas.DataFrame.hist(gapminder)` to have an idea of what looks like the database.

```
[41]: pd.DataFrame.hist(gap, figsize=(15,15))
```

```
[41]: array([[<AxesSubplot:title={'center':'year'}>,
             <AxesSubplot:title={'center':'lifeExp'}>],
             [<AxesSubplot:title={'center':'pop'}>,
              <AxesSubplot:title={'center':'gdpPercap'}>]], dtype=object)
```



Observations “**Q1 : Which conclusion can you make about this histogram ?**” : - With the first plot, we can say that measures haven’t been done every year. - Life expectancy is mostly around 70 years old, as shown by the second plot, but there are also some countries that have a lower life expectancy than 40 years old.

1.1.3 I.3) Find how many measures have been made for each country & if there is missing values.

```
[42]: ## Missing values (defined by <NA>)
print("lifeExp  : ", sum(pd.isna(gap["lifeExp"])))
print("year     : ", sum(pd.isna(gap["year"])))
print("gdpPerCap : ", sum(pd.isna(gap["gdpPerCap"])))
```

```
lifeExp  : 0
```

```
year      : 0
gdpPerCap : 0
```

With this, we can see that there is no missing values.

```
[43]: ## Number of analysis :
print("Nbr. of COUNTRY analysed      : ", len(gap['country'].unique()))
print("Nbr. of YEAR when there was analysis : ", len(gap['year'].unique()))
print()
print("Total analysis                : ", len(gap['country'].unique()) *
      ↪len(gap['year'].unique()))
```

```
Nbr. of COUNTRY analysed      : 142
Nbr. of YEAR when there was analysis : 12

Total analysis                : 1704
```

In conclusion, is there any missing values?

1.1.4 I.4) List unique values into ; CONTINENT, COUNTRY & YEAR columns.

```
[44]: print(gap['continent'].unique())
```

```
['Asia' 'Europe' 'Africa' 'Americas' 'Oceania']
```

```
[45]: print(gap['country'].unique())
```

```
['Afghanistan' 'Albania' 'Algeria' 'Angola' 'Argentina' 'Australia'
'Austria' 'Bahrain' 'Bangladesh' 'Belgium' 'Benin' 'Bolivia'
'Bosnia and Herzegovina' 'Botswana' 'Brazil' 'Bulgaria' 'Burkina Faso'
'Burundi' 'Cambodia' 'Cameroon' 'Canada' 'Central African Republic'
'Chad' 'Chile' 'China' 'Colombia' 'Comoros' 'Congo, Dem. Rep.'
'Congo, Rep.' 'Costa Rica' 'Cote d'Ivoire' 'Croatia' 'Cuba'
'Czech Republic' 'Denmark' 'Djibouti' 'Dominican Republic' 'Ecuador'
'Egypt' 'El Salvador' 'Equatorial Guinea' 'Eritrea' 'Ethiopia' 'Finland'
'France' 'Gabon' 'Gambia' 'Germany' 'Ghana' 'Greece' 'Guatemala' 'Guinea'
'Guinea-Bissau' 'Haiti' 'Honduras' 'Hong Kong, China' 'Hungary' 'Iceland'
'India' 'Indonesia' 'Iran' 'Iraq' 'Ireland' 'Israel' 'Italy' 'Jamaica'
'Japan' 'Jordan' 'Kenya' 'Korea, Dem. Rep.' 'Korea, Rep.' 'Kuwait'
'Lebanon' 'Lesotho' 'Liberia' 'Libya' 'Madagascar' 'Malawi' 'Malaysia'
'Mali' 'Mauritania' 'Mauritius' 'Mexico' 'Mongolia' 'Montenegro'
'Morocco' 'Mozambique' 'Myanmar' 'Namibia' 'Nepal' 'Netherlands'
'New Zealand' 'Nicaragua' 'Niger' 'Nigeria' 'Norway' 'Oman' 'Pakistan'
'Panama' 'Paraguay' 'Peru' 'Philippines' 'Poland' 'Portugal'
'Puerto Rico' 'Reunion' 'Romania' 'Rwanda' 'Sao Tome and Principe'
'Saudi Arabia' 'Senegal' 'Serbia' 'Sierra Leone' 'Singapore'
'Slovak Republic' 'Slovenia' 'Somalia' 'South Africa' 'Spain' 'Sri Lanka'
'Sudan' 'Swaziland' 'Sweden' 'Switzerland' 'Syria' 'Taiwan' 'Tanzania'
'Thailand' 'Togo' 'Trinidad and Tobago' 'Tunisia' 'Turkey' 'Uganda'
'United Kingdom' 'United States' 'Uruguay' 'Venezuela' 'Vietnam'
'West Bank and Gaza' 'Yemen, Rep.' 'Zambia' 'Zimbabwe']
```

```
[46]: print(gap['year'].unique())
```

```
[1952 1957 1962 1967 1972 1977 1982 1987 1992 1997 2002 2007]
```

1.1.5 I.5) Reckon the mean value of the life expectancy in 1952 & 2007, then make a bar chart out of it.

```
[47]: ## List for mean value of ; 1952 & 2007 :
avg_1952 = pd.DataFrame.mean(gap[gap["year"] == 1952]["lifeExp"])
avg_2007 = pd.DataFrame.mean(gap[gap["year"] == 2007]["lifeExp"])

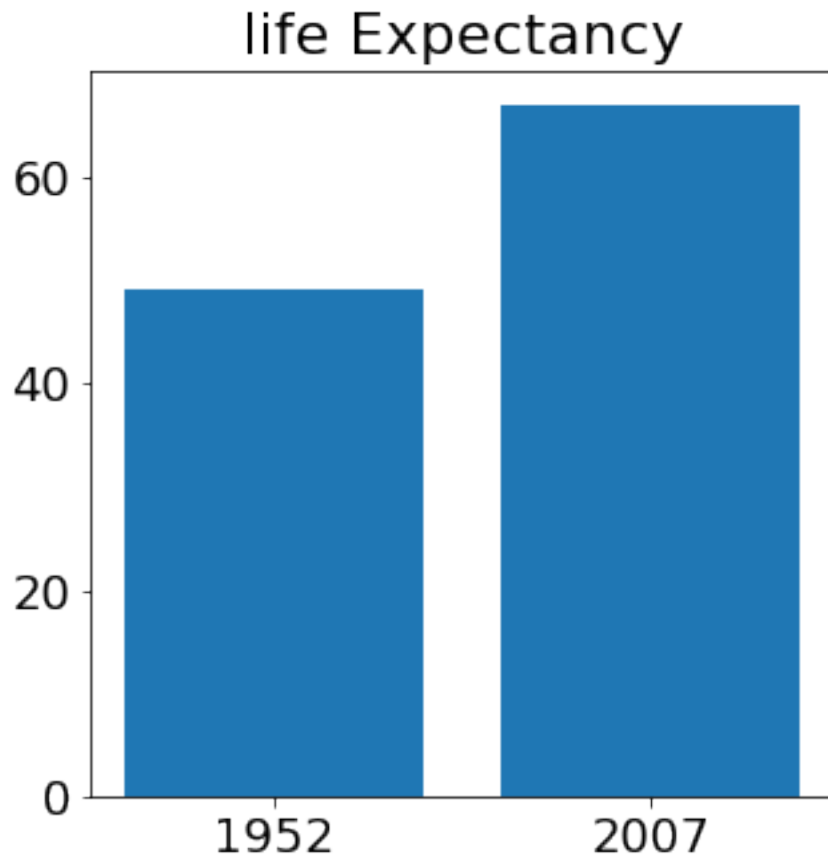
## Set plot size
plt.figure(figsize=(5,5))

## Set font size around graph
plt.rcParams.update({'font.size': 18})

## Fill bars
plt.bar(["1952", "2007"], [avg_1952, avg_2007])

## plot Title
plt.title("life Expectancy")

## Show plot
plt.show()
```



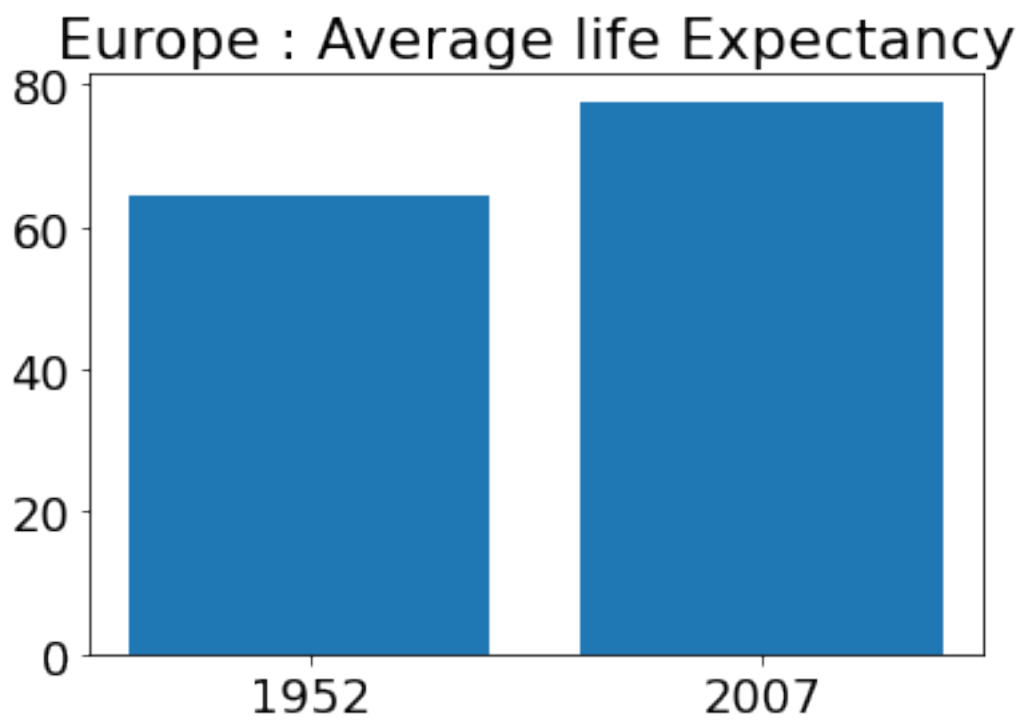
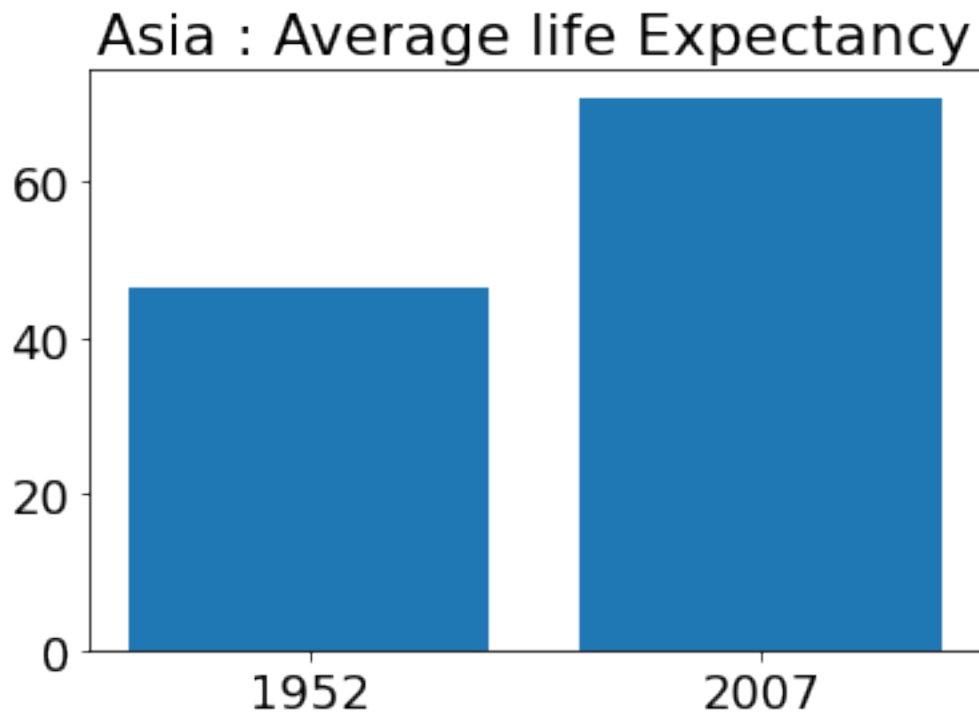
1.1.6 I.6) Reckon the mean value of the life expectancy for each CONTINENT in 1952 & 2007, then generate a bar chart to compare the different values.

```
[48]: ## For unique CONTINENT ..
for cont in pd.unique(gap["continent"]):
    ## Recover infos depending of the CONTINENT and the YEAR :
    ls_cont_1952 = gap[(gap["year"] == 1952) & (gap["continent"] == cont)]
    ls_cont_2007 = gap[(gap["year"] == 2007) & (gap["continent"] == cont)]

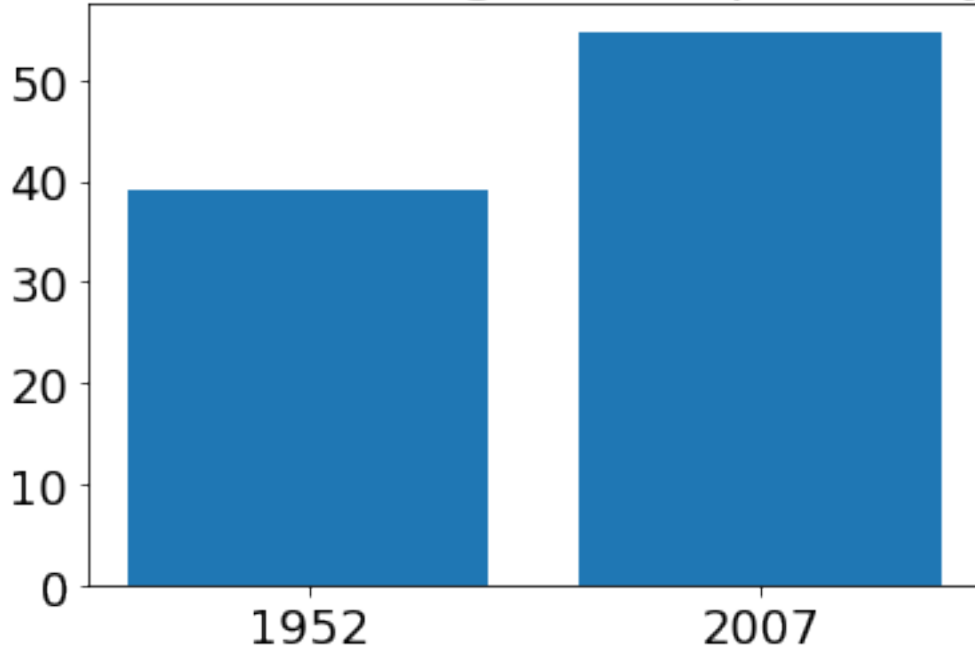
    ## Make the mean value :
    avg_1952 = pd.DataFrame.mean(ls_cont_1952["lifeExp"])
    avg_2007 = pd.DataFrame.mean(ls_cont_2007["lifeExp"])

    ## Show plot with ; BAR's infos & graph title
    plt.bar(["1952", "2007"], [avg_1952, avg_2007])
    plt.title(f"{cont} : Average life Expectancy")
    plt.show()
```

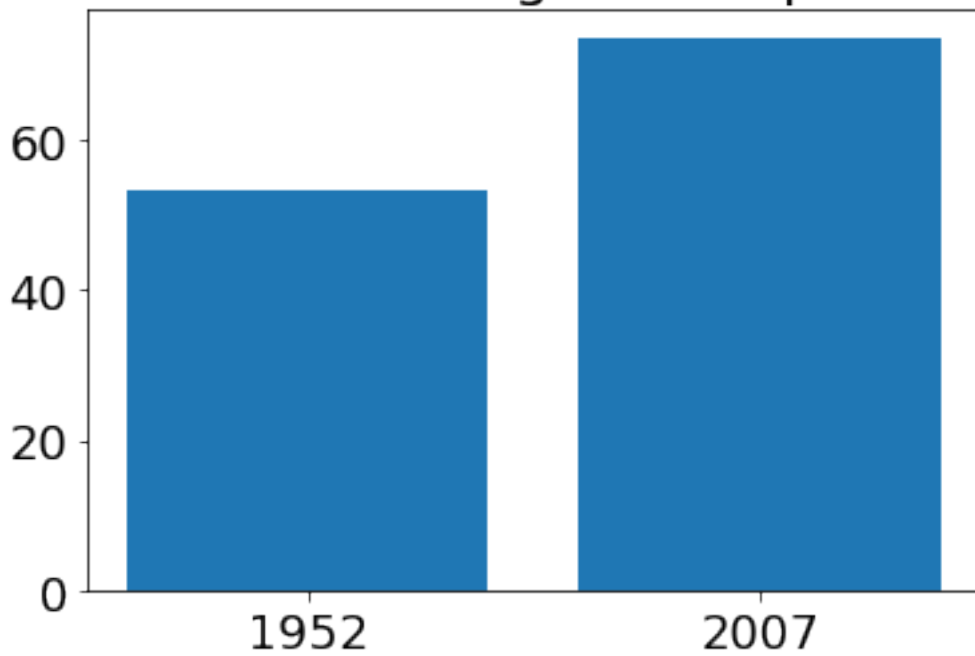
Only ONE bar chart.

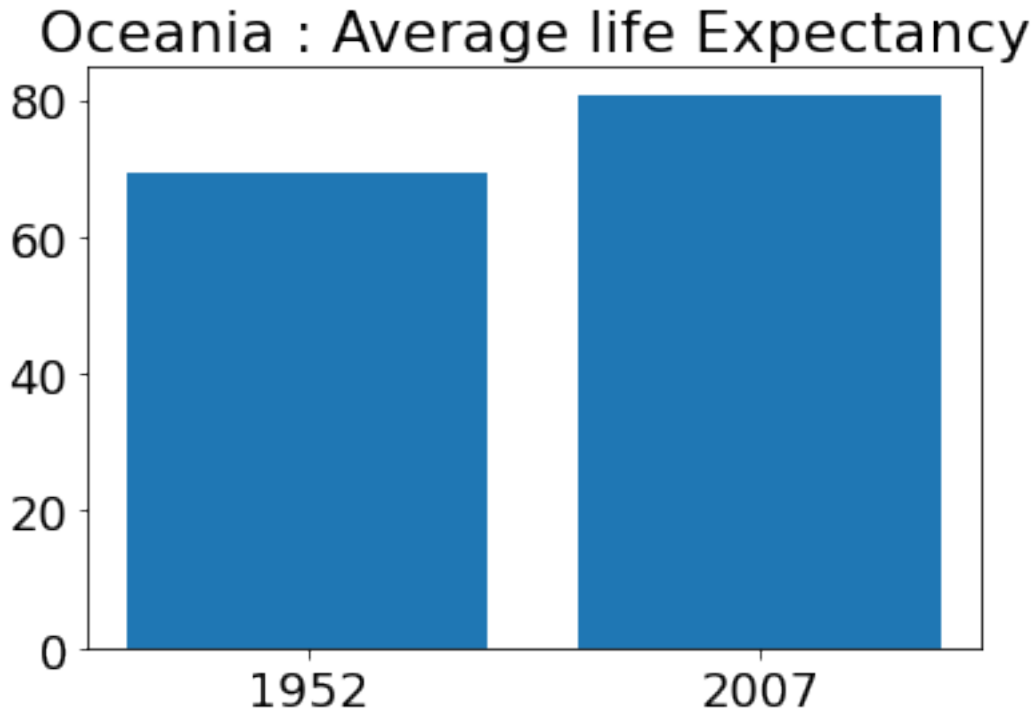


Africa : Average life Expectancy



Americas : Average life Expectancy





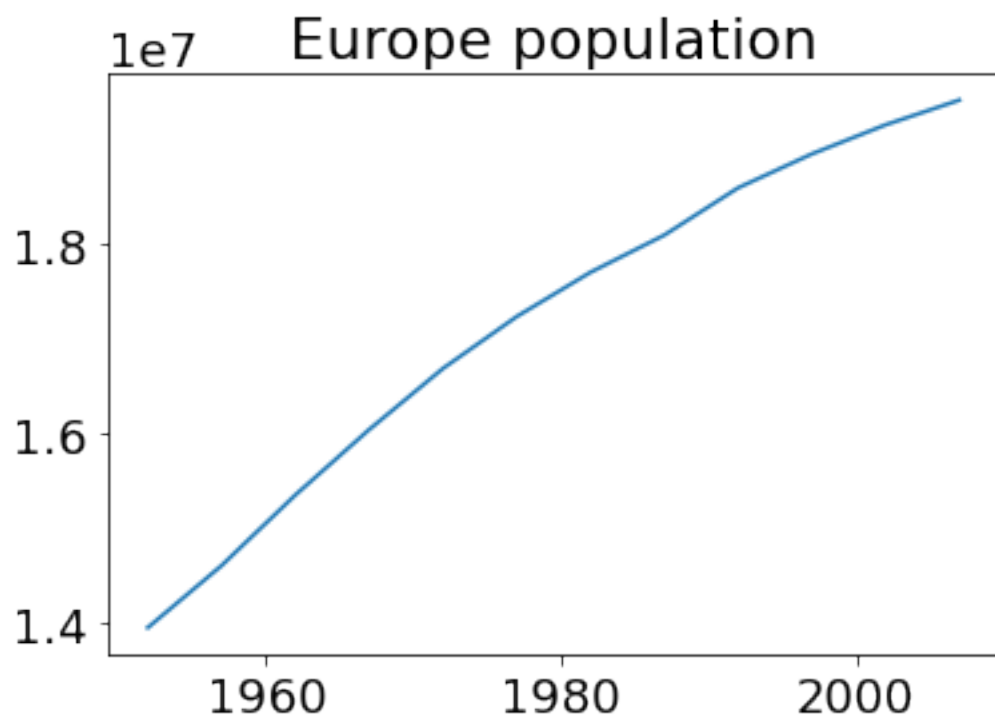
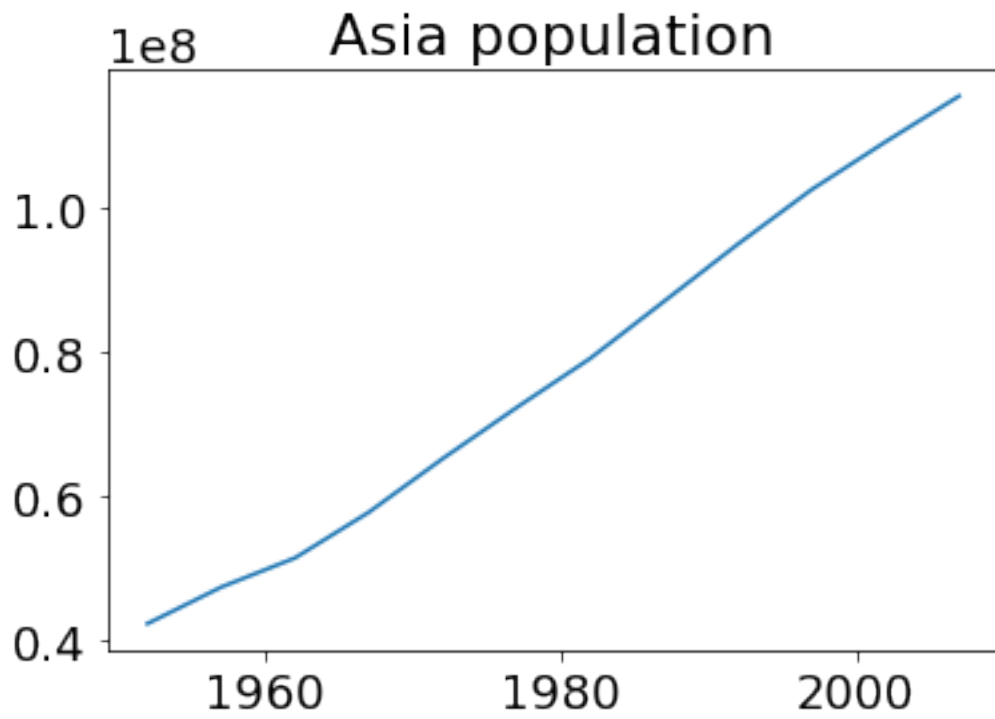
Observations “Q2 : Which country have the highest increase ?” : - It’s Asia. - Plus, we can see that no matter the continent, the life expectancy has considerably risen over the year.

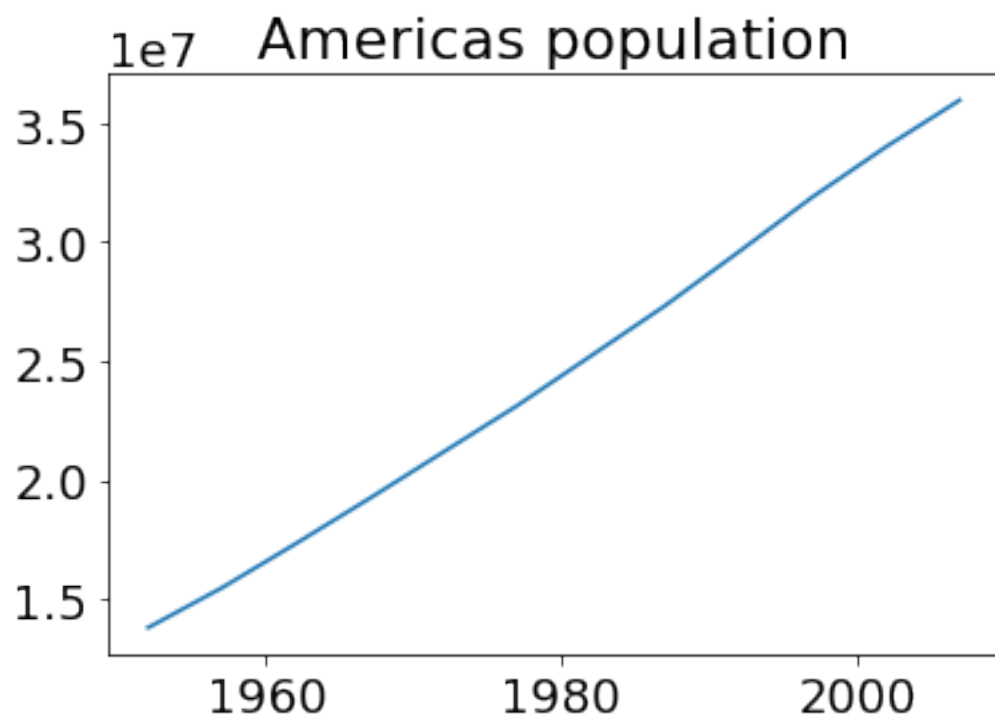
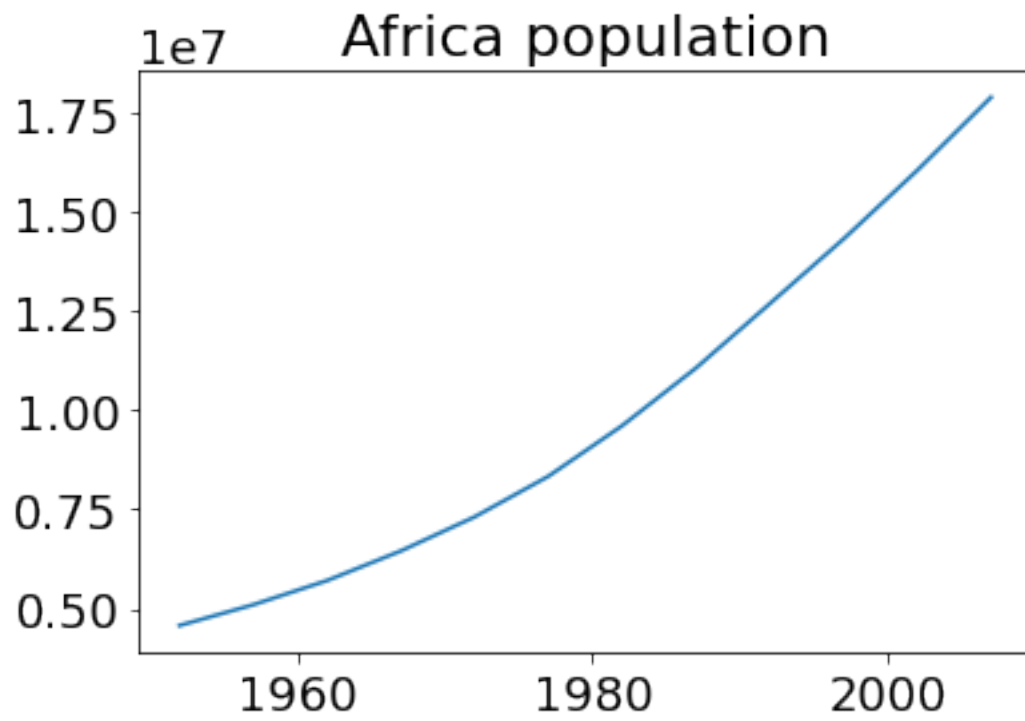
1.1.7 I.7) Generate a plot showing the evolution between 1952 & 2007 of the population for each CONTINENT.

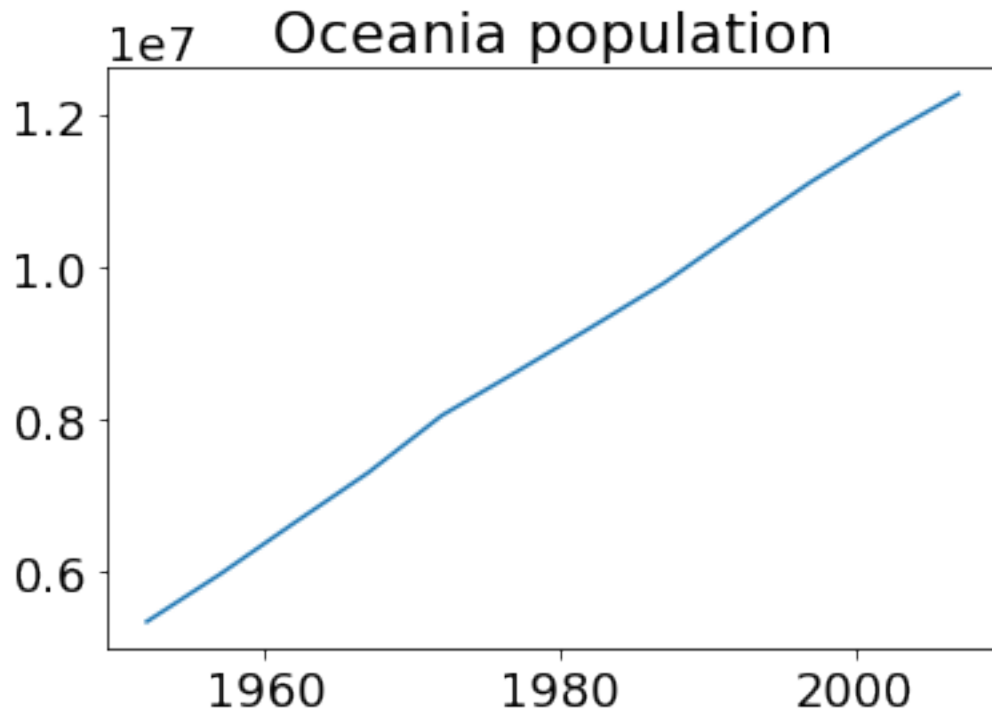
```
[49]: ## List of the YEAR
ls_years = pd.unique(gap["year"])

## For each CONTINENT ..
for cont in pd.unique(gap["continent"]):
    ## Recover the mean value of population for each YEAR listed
    df_means = [pd.DataFrame.mean(d) for d in [gap[(gap["year"] == y) &
    →(gap["continent"] == cont)]["pop"] \
                                for y in ls_years]]

    ## Set plot's infos
    plt.plot(ls_years, df_means)
    ## Set plot title
    plt.title(f"{cont} population")
    ## Show plot
    plt.show()
```







histograms ? Those are not histograms

Observations “Q3 : What can you conclude about those histograms ?” : - With the previous graphes, they show us that the Average population has drastically increased through the ages. - Plus, the Africa is the continent that has gain the most population over the decades.

Really, Africa ? Check the units.

1.1.8 I.8) Generate 2 sub-plots that will contain scatters with :

the gdpPerCap as the ‘X’ axis &

the life Expectancy as the ‘Y’ axis

for 1952 & 2007 (make 2 plots and make a color for each CONTINENT)

```
[50]: ## Secondary dataframes depending of the YEAR wanted :
df_1952 = gap[(gap["year"] == 1952)]
df_2007 = gap[(gap["year"] == 2007)]

## Set colors for each CONTINENT
colors = {"Asia" : "gold", "Europe" : "lime", "Africa" : "saddlebrown", "Americas" : "cyan", "Oceania" : "magenta"}

## Set sub-plots and plot size
fig, axs = plt.subplots(2, figsize=(13,13))
## For each CONTINENT ..
for cont in pd.unique(gap["continent"]):
    ## 1952 ----->
```

```

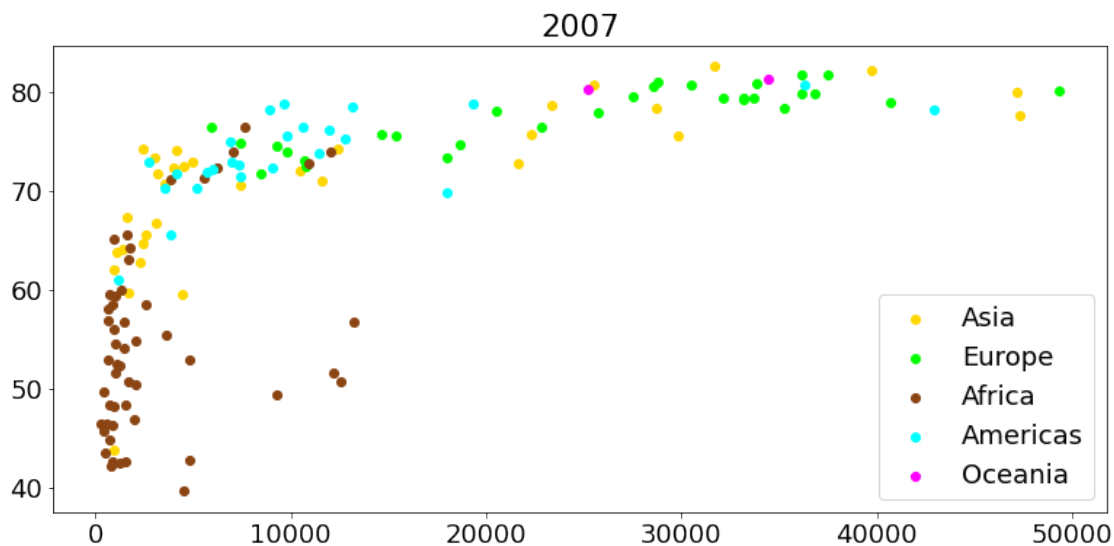
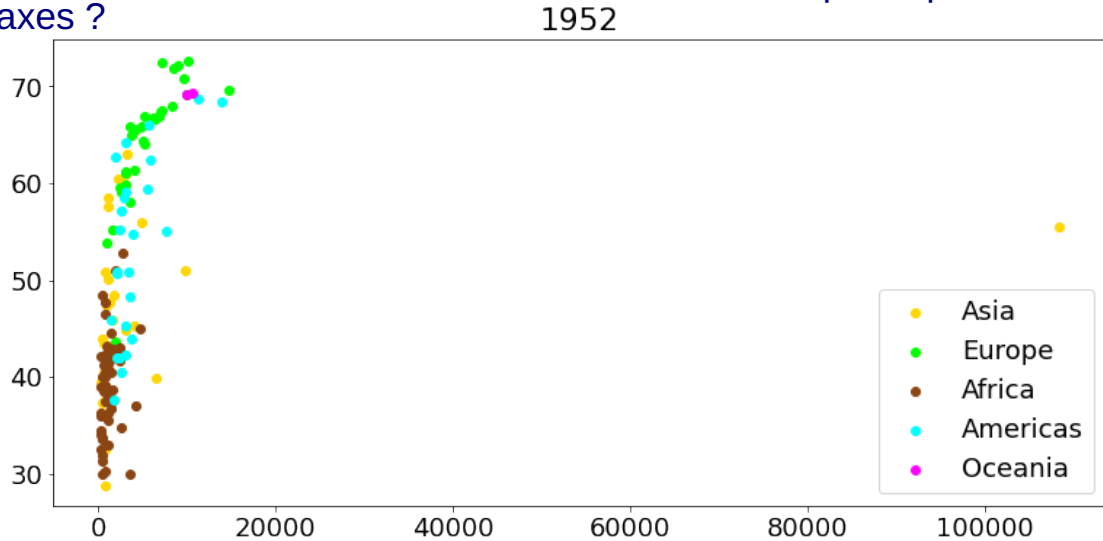
## Make a list of the X & Y values
x_1952 = df_1952[df_1952["continent"] == cont]["gdpPercap"]
y_1952 = df_1952[df_1952["continent"] == cont]["lifeExp"]
## Set first plot
axs[0].scatter(x_1952, y_1952, c=colors[cont])
axs[0].title.set_text("1952")
axs[0].legend(colors)

## 2007 ----->
## Make a list of the X & Y values
x_2007 = df_2007[df_2007["continent"] == cont]["gdpPercap"]
y_2007 = df_2007[df_2007["continent"] == cont]["lifeExp"]
## Set second plot
axs[1].scatter(x_2007, y_2007, c=colors[cont])
axs[1].title.set_text("2007")
axs[1].legend(colors)
## Show plots
plt.show()

```

Title? What does this plot represent?

What are the axes ?



```
[51]: ## Find the value over 100k for the gdpPerCap
print(max(df_1952["gdpPerCap"]))
```

108382.3529

```
[52]: ## Find which country as the previous value
#print(df_1952[df_1952["gdpPerCap"] == max(df_1952["gdpPerCap"])])
print(df_1952[df_1952["gdpPerCap"] == max(df_1952["gdpPerCap"])][["country"]])
```

852 Kuwait

Name: country, dtype: object

On the scatter plot of 1952, there is a gdpPerCap's value that is bigger than the other (the over 100k one), that we think of it is absurd.

So we wanted to know which one was it. And we found that it's the **Kuwait's gdpPerCap**.

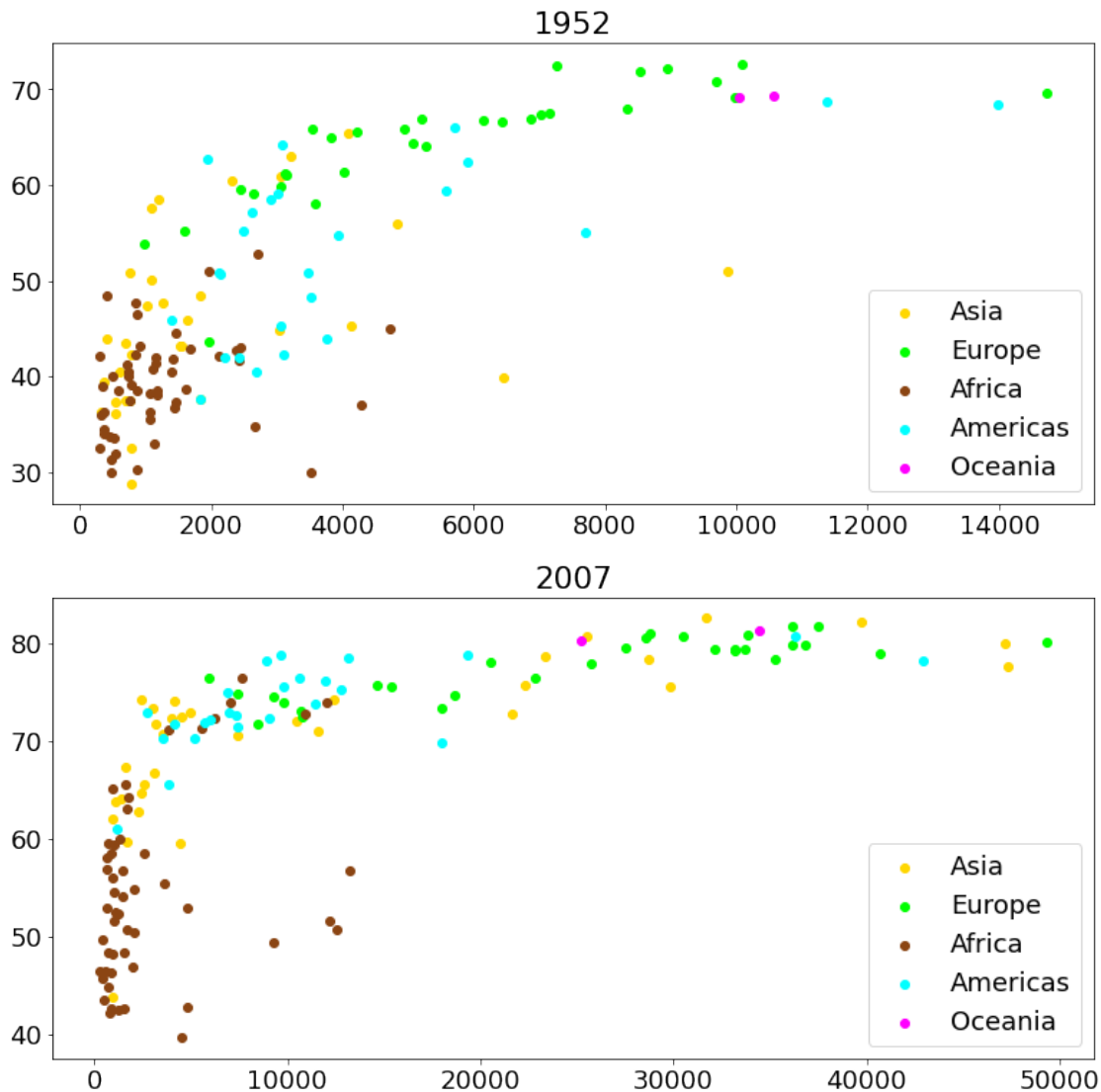
We assume that it is a wrong value, so we decided to make another plot without it (only for 1952).

From now, on every 1952 plot, Kuwait will be omitted.

```
[53]: ## Replacing dataframe of 1952 and don't take Kuwait in consideration :
df_1952 = gap[(gap["year"] == 1952) & (gap["country"] != "Kuwait")]

## Set sub-plots and plot size
fig, axs = plt.subplots(2, figsize=(13,13))
## For each CONTINENT ..
for cont in pd.unique(gap["continent"]):
    ## 1952 ----->
    ## Make a list of the X & Y values
    x_1952 = df_1952[df_1952["continent"] == cont]["gdpPercap"]
    y_1952 = df_1952[df_1952["continent"] == cont]["lifeExp"]
    ## Set first plot
    axs[0].scatter(x_1952, y_1952, c=colors[cont])
    axs[0].title.set_text("1952")
    axs[0].legend(colors)

    ## 2007 ----->
    ## Make a list of the X & Y values
    x_2007 = df_2007[df_2007["continent"] == cont]["gdpPercap"]
    y_2007 = df_2007[df_2007["continent"] == cont]["lifeExp"]
    ## Set second plot
    axs[1].scatter(x_2007, y_2007, c=colors[cont])
    axs[1].title.set_text("2007")
    axs[1].legend(colors)
## Show plots
plt.show()
```



Observations “Q4 : What are your conclusion about those histograms ?” : - We can see that the continent with **the lowest life Expectancy & gdpPerCap** is Africa. - Oceania doesn't have a lot of country, so there's not a lot of point. - The continent with **the highest life Expectancy** is Europe. But its gdpPerCap is very diversified. - Between the two plots, we can say that **life Expectancy** has really increased. Except for Africa and Asia, some countries still have a low life Expectancy.

1.1.9 I.9) Generate scatter plots of the life Expectancy over the gdpPerCap

Separate each CONTINENT by a color.

```
[54]: # Remove Kuwait for 2007, because both list must have the same size.
df_2007 = gap[(gap["year"] == 2007) & (gap["country"] != "Kuwait")]
```



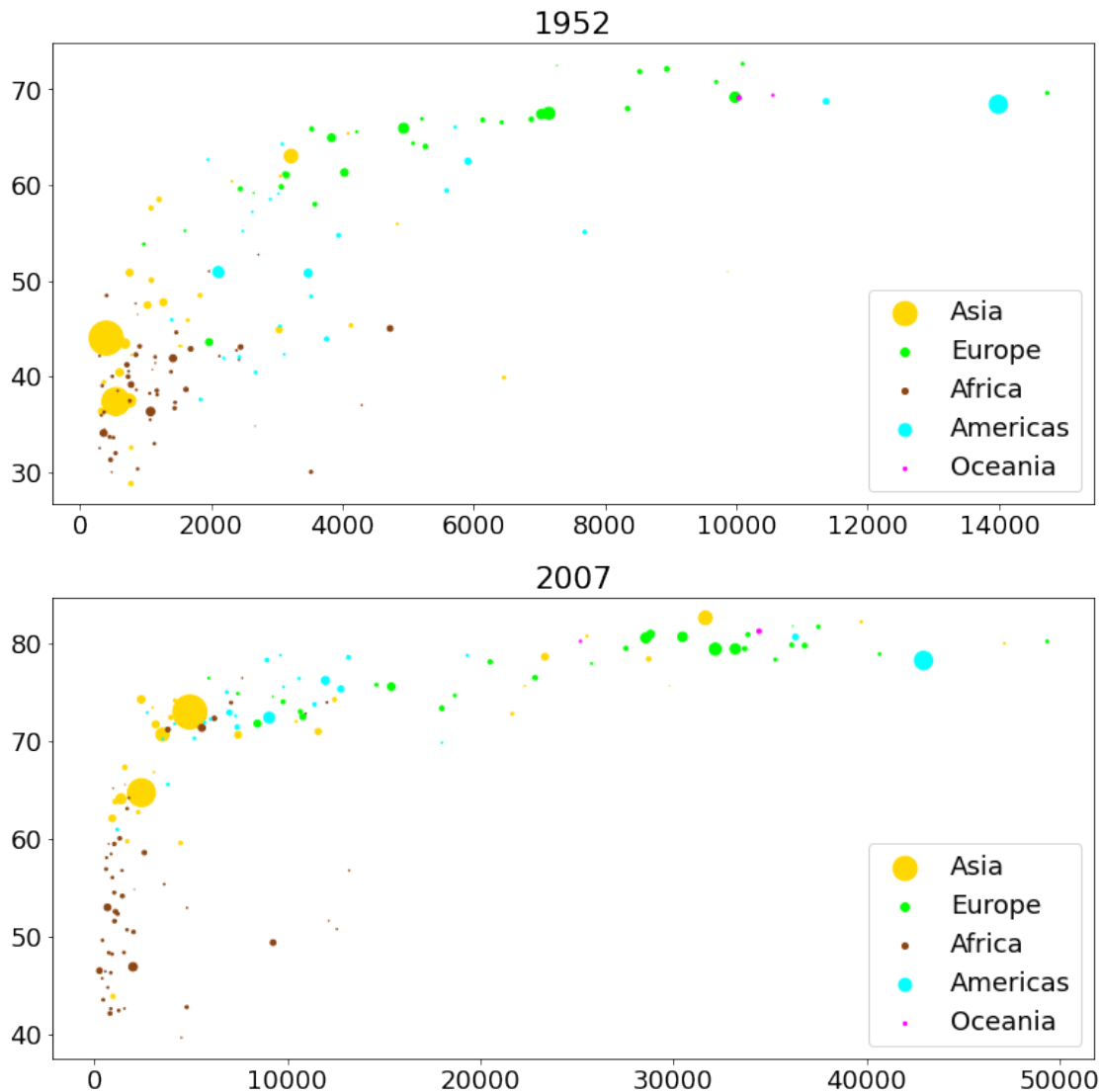
```

## Set a certain reference for the dot size of the scatter plots
dotRef = 1_000_000

## Set sub-plots and plot size
fig, axs = plt.subplots(2, figsize=(13,13))
## For each CONTINENT ..
for cont in pd.unique(gap["continent"]):
    ## 1952 ----->
    ## Make a list of the X & Y values
    x_1952 = df_1952[df_1952["continent"] == cont]["gdpPercap"]
    y_1952 = df_1952[df_1952["continent"] == cont]["lifeExp"]
    ## Manage the scatter' size
    dotSize_1952 = df_1952[df_1952["continent"] == cont]["pop"] / dotRef
    ## Set first plot
    axs[0].scatter(x_1952, y_1952, s=dotSize_1952, c=colors[cont])
    axs[0].title.set_text("1952")
    axs[0].legend(colors)

    ## 2007 ----->
    ## Make a list of the X & Y values
    x_2007 = df_2007[df_2007["continent"] == cont]["gdpPercap"]
    y_2007 = df_2007[df_2007["continent"] == cont]["lifeExp"]
    ## Manage the scatter' size
    dotSize_2007 = df_1952[df_1952["continent"] == cont]["pop"] / dotRef
    ## Set second plot
    axs[1].scatter(x_2007, y_2007, s=dotSize_2007, c=colors[cont])
    axs[1].title.set_text("2007")
    axs[1].legend(colors)
## Show plots
plt.show()

```



1.1.10 I.10) Generate a scatter plot with the increase between 1952 & 2007 with :
the gdpPerCap as the 'X' axis &

the population as the 'Y' axis

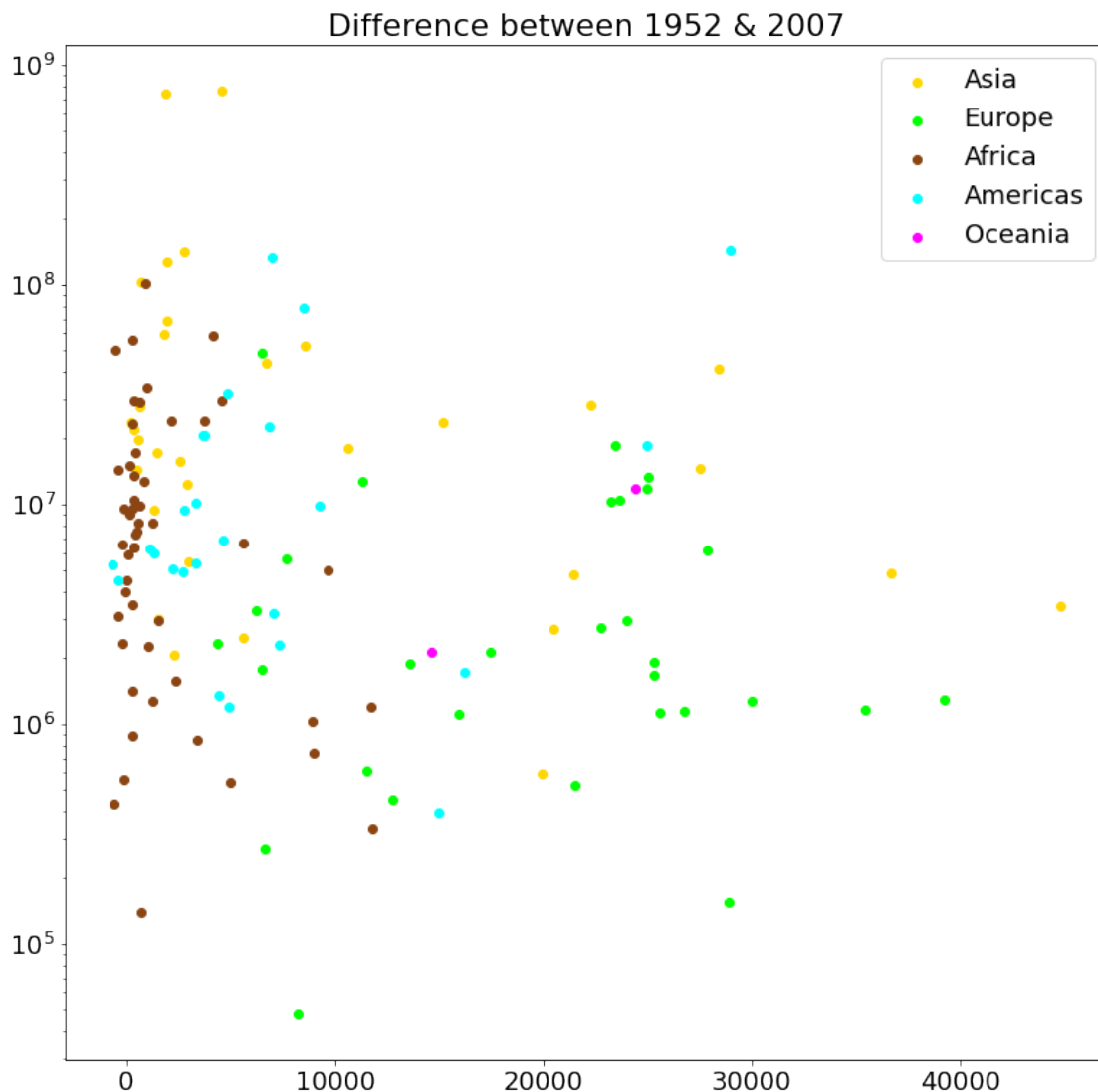
```
[55]: ## Set sub-plots and plot size
plt.figure(figsize=(13,13))
## For each CONTINENT ..
for cont in pd.unique(gap["continent"]):
    ## Make list out of the gdpPerCap for each YEAR concerned
    x_1952 = df_1952[df_1952["continent"] == cont]["gdpPerCap"]
    x_2007 = df_2007[df_2007["continent"] == cont]["gdpPerCap"]
```

```

## Make list out of the population for each YEAR concerned
y_1952 = df_1952[df_1952["continent"] == cont]["pop"]
y_2007 = df_2007[df_2007["continent"] == cont]["pop"]

## Make array out of the difference between both YEAR concerned
x = np.array([g for g in x_2007]) - np.array([g for g in x_1952])
y = np.array([p for p in y_2007]) - np.array([p for p in y_1952])
plt.scatter(x, y, c=colors[cont])
## Show plot with a specific title, a log. scale on the 'Y' axis and add the
→ legend to the plot.
plt.title("Difference between 1952 & 2007")
plt.yscale("log")
plt.legend(colors)
plt.show()

```



Which variables are on the axes?

Observation “Q5 : What can be seen in the previous scatter plot ?” :

- The trend is as follow : the lower the gdpPerCap, the higher population increase.

That’s what we understand, but it is not clear enough for us with this plot.

It is a **trend** and not an equation or else.

[]:

[]:

[]: