Assignment 1 Group 85

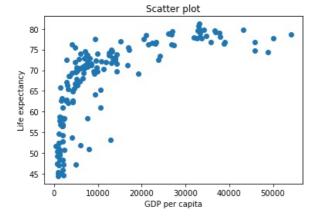
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```
In [1]:
          import pandas as pd
          import matplotlib.pyplot as plt
         import numpy as np
In [2]:
         GDP per capita = pd.read csv("./life-expectancy.csv")
         life_expect = pd.read_csv("./maddison-data-gdp-per-capita-in-2011us.csv")
         GDP = pd.read_csv("./gdp-in-british-pounds.csv")
In [3]:
         d = GDP_per_capita.merge(life_expect, on=('Year', 'Entity'))
         d = d.merge(GDP, on=('Year',
                                         'Entity'))
          reduced_d = d[d.Year == 2000]
          reduced d.rename(columns={'GDP (Fouquin and Hugot (CEPII 2016))': 'GDP'}, inplace = True)
          reduced d
         /Users/nick/opt/anaconda3/lib/python3.9/site-packages/pandas/core/frame.py:5039: SettingWithCopyWarning:
         A value is trying to be set on a copy of a slice from a DataFrame
         See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user guide/indexing.html#retur
         ning-a-view-versus-a-copy
         return super().rename(
Out[3]:
                  Entity Code_x Year Life expectancy at birth (historical) Code_y GDP per capita 417485-annotations Code
                                                                                                                       GDP
           98
                 Albania
                            ALB 2000
                                                              75.4
                                                                      ALB
                                                                               4808.4795
                                                                                                     NaN
                                                                                                          ALB 2.400530e+09
           163
                  Algeria
                           DZA 2000
                                                              70.5
                                                                      DZA
                                                                               6834.5537
                                                                                                    NaN
                                                                                                          DZA 3.621245e+10
           203
                           AGO 2000
                                                              46.0
                                                                     AGO
                                                                               2013.6362
                                                                                                          AGO
                                                                                                               6.034059e+09
                  Angola
                                                                                                    NaN
           281
                Argentina
                           ARG 2000
                                                              73.9
                                                                      ARG
                                                                              14368.9420
                                                                                                     NaN
                                                                                                          ARG
                                                                                                               1.878391e+11
           306
                 Armenia
                           ARM 2000
                                                              70.6
                                                                     ARM
                                                                               5139.8257
                                                                                                     NaN
                                                                                                          ARM
                                                                                                               1.263412e+09
          9882
               Venezuela
                           VEN 2000
                                                              72.5
                                                                      VEN
                                                                              13992.6080
                                                                                                     NaN
                                                                                                          VEN 7.742649e+10
          9942
                 Vietnam
                           VNM 2000
                                                              72.5
                                                                     VNM
                                                                               2773.1016
                                                                                                    NaN
                                                                                                          VNM 2.223377e+10
                           YEM 2000
                                                                      YEM
                                                                                                          YEM 6.368957e+09
          9967
                  Yemen
                                                              62.6
                                                                               4212.1055
                                                                                                     NaN
         10031
                 Zambia
                           ZMB
                               2000
                                                              45.2
                                                                      ZMB
                                                                               1428.5024
                                                                                                     NaN
                                                                                                          ZMB 2.379803e+09
         10095 Zimbabwe
                           ZWE 2000
                                                              44.7
                                                                     ZWE
                                                                               2211.1963
                                                                                                          ZWE 4.421600e+09
                                                                                                    NaN
```

158 rows × 9 columns

```
In [4]:
    x,y = reduced_d['GDP per capita'], reduced_d['Life expectancy at birth (historical)']
    plt.scatter(x,y)
    plt.xlabel('GDP per capita')
    plt.ylabel('Life expectancy')
    plt.title('Scatter plot')
```

Out[4]: Text(0.5, 1.0, 'Scatter plot')



```
In [5]:
         std_exp = np.std(a=reduced_d['Life expectancy at birth (historical)'])
         mean exp = np.mean(a=reduced_d['Life expectancy at birth (historical)'])
         mean_std_exp = mean_exp + std_exp
In [6]:
         return_list = []
         for i, r in reduced d.iterrows():
             country = r["Entity"]
             life_exp = r['Life expectancy at birth (historical)']
             if life_exp > mean_std_exp:
                 return_list.append(country)
         return list
Out[6]: ['Australia',
          'Austria',
         'Belgium',
         'Canada',
         'Costa Rica',
          'Finland',
         'France',
          'Germany',
          'Greece'
         'Hong Kong',
          'Iceland',
         'Israel',
          'Italy',
          'Japan',
          'Luxembourg',
          'Malta'
          'Netherlands',
         'New Zealand',
          'Norway',
          'Singapore',
         'Spain',
          'Sweden'
          'Switzerland',
          'United Kingdom']
```

c) Which countries have high life expectancy but have low GDP? Motivate how you have chosen to define "high" and "low".

```
In [7]:
         std_gdp = np.std(a=reduced_d['GDP'])
         mean_gdp = np.mean(a=reduced_d['GDP'])
         std exp = np.std(a=reduced d['Life expectancy at birth (historical)'])
         mean_exp = np.mean(a=reduced_d['Life expectancy at birth (historical)'])
         mean std exp = mean exp + std exp
         mean_std_gdp = mean_gdp - std_gdp
In [8]:
         return_list = []
         for i, r in reduced_d.iterrows():
             country = r["Entity"]
             life_exp = r['Life expectancy at birth (historical)']
             gdp = r['GDP']
             if life exp > mean std exp and gdp < mean std gdp:</pre>
                 return_list.append(country)
         return_list
Out[8]: []
```

Answer: No countries fit our criteria. For life expectancy we have decided to define "high" as 1 standard deviation above the mean. For GDP we have decided to define "low" as 1 standard deviation below the mean.

d) Does every strong economy (normally indicated by GDP) have high life expectancy?

```
In [9]:
    std_gdp = np.std(a=reduced_d['GDP'])
    mean_gdp = np.mean(a=reduced_d['GDP'])

    std_exp = np.std(a=reduced_d['Life expectancy at birth (historical)'])
    mean_exp = np.mean(a=reduced_d['Life expectancy at birth (historical)'])

    mean_std_gdp = mean_gdp + std_gdp
    mean_std_exp = mean_exp + std_exp
In [10]:

return list high = []
```

```
return_list_low = []
for i, r in reduced_d.iterrows():
    country = r["Entity"]
    life_exp = r['Life expectancy at birth (historical)']
    gdp = r['GDP']
    if gdp > mean_std_gdp:
        if life_exp > mean_std_exp:
            return_list_high.append(country)
    else:
        return_list_low.append(country)
```

```
In [11]:
    tot_len = len(return_list_low + return_list_high)
    percentage = len(return_list_high)/tot_len
    print(f'Percentage of countries with high life expectancy and high GDP is {round(percentage*100, 2)}%')
```

Percentage of countries with high life expectancy and high GDP is 71.43%

life_exp = r['Life expectancy at birth (historical)']

return_list_high.append(country)

return list low.append(country)

gdp = r['GDP per capita']
if gdp > mean std gdp:

else:

if life exp > mean std exp:

Answer: Yes, mostly. 71,43% of the countries with high GDP have high life expectancy, indicating a weak correlation.

e) Related to question d, what would happen if you use GDP per capita as an indicator of strong economy? Explain the results you obtained, and discuss any insights you get from comparing the results of d and e.

```
In [12]:
    std_gdp = np.std(a=reduced_d['GDP per capita'])
    mean_gdp = np.mean(a=reduced_d['GDP per capita'])
    std_exp = np.std(a=reduced_d['Life expectancy at birth (historical)'])
    mean_exp = np.mean(a=reduced_d['Life expectancy at birth (historical)'])
    mean_std_gdp = mean_gdp + std_gdp
    mean_std_exp = mean_exp + std_exp

In [13]:
    return_list_high = []
    return_list_low = []
    for i, r in reduced_d.iterrows():
        country = r["Entity"]
```

```
In [14]:
    tot_len = len(return_list_low + return_list_high)
    percentage = len(return_list_high)/tot_len
    print(f'Percentage of countries with high life expectancy and high GDP per capita is {round(percentage*100, 2)}%'
```

Percentage of countries with high life expectancy and high GDP per capita is 72.41%

Answer: From our results GDP per capita and GDP are decent indicators of strong economies. A factor that GDP per capita/GDP doesn't concider, is where the money is coming from, if it's one industry (eg oil) or multiple intustries. Having multiple strong industries are much preferred as it reduces the risk of an economic collapse.

We tested with diffrent years and found that the percentage varies a lot. From the 1900s the percentage has stedily decreased, indicating weaker coorelation between strong economies and and life expectancy. This might be because either poor countries economies grow faster than their life expectancy, or that their life expectancy grow faster than the economy.

We used the same deviation every time (1 standart deviation), so that we could compare results between questions.

We choose to reduce our dataset to one year to simplify the calculations. The year 2000 was chosen because it is recent and we had data for that year.