GDP and life satisfaction

May 14, 2021

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[]: # Here are imported the libraries/modules that are used below for the analysis.
     %matplotlib inline
     import numpy as np
     import matplotlib.pyplot as plt
     import scipy as sp
     import pandas as pd
     import sklearn.neighbors
[]: # Here I use my local file path on reading the .csv files.
     # The reader must use the appropriate path where files are located in their
     \rightarrow computers.
     # I select as header the third row of the file qdp per capita (n=2). The
      \hookrightarrow delimiter is ","
     GDP=pd.read_csv("/Users/damianejlli/Downloads/gdp per capita.csv", __

delimiter=",", header=2)
     # There is no need to specify the header for the "better life index.csv" file.
     LS=pd.read_csv("/Users/damianejlli/Downloads/better life index.csv")
[]: #I display some of the content of the GDP dataframe.
     GDP.head()
[]: # I select the columns "Country Name" and year "2015" for the analysis in the
      \hookrightarrow GDP dataframe.
     GPD1=gdp_per_capita.loc[:,["Country Name", '2015']]
[]: # I display the GDP1 dataframe content.
     GPD1
[]: # I set as index the "Country Name" column and rename the column "2015" to "GPD_{\sqcup}
      →per capita 2015 (USD)"
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→2015 (USD)"})
[]: # I show the first ten rows of the GDP2 dataframe as a matter of example
     GDP2.head(10)
[]: # I show the first ten rows of the LS dataframe as a matter of example.
     LS.head()
[]: # I show the shape of the LS dataframe. It has 2369 rows and 17 columns.
     LS.shape
[]: # I use a conditional to choose all those rows with values "Life satisfaction"
     → in the column "Indicator"
     # and all those equal to "TOT" in the "INEQUALITY" column in the LS dataframe.
     # "TOT" is the total value of life satisfaction for men and women in a given_
     \hookrightarrow country.
    LS1=LS[(LS["Indicator"]=="Life satisfaction") & (LS["INEQUALITY"]=="TOT")]
[]: # I show the first 10 entries of the LS1 dataframe as a matter of exmple.
     LS1.head(10)
[]: # First, in the LS1 dataframe, I rename the columns "Country" and "Value"
     →respectively to "Country Name" and "Life Satisfaction Value".
     # Second, I set as index of the new dataframe the "Country Name" and after I_{\sqcup}
     ⇒select all rows in the "Indicator" column
     # with entries equal to "Life Satisfaction Value".
     LS2=LS1.rename(columns={"Country" : "Country Name", "Value": "Life Satisfaction_
     →Value"}).set_index("Country Name").loc[:, ["Life Satisfaction Value"]]
[]: # I show the first 10 entries of the LS2 dataframe as a matter of exmple.
     LS2.head(10)
[]: # I remove the entry "OECD-Total" country index from the LS2 dataframe because
     →it is unneccessary for the analysis.
    LS3=LS2[LS2.index != "OECD - Total"]
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GDP2=GPD1.set_index("Country Name").rename(columns={"2015": "GPD per capita_

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[]: # I show the first 10 entries of the LS3 dataframe as a matter of exmple.
     LS3.head(10)
[]: # I join the LS3 dataframe with the GPD2 dataframe in order to form the final \Box
      \rightarrow dataframe, df.
     df=LS3.join(GPD2)
[]: # I display the entries in the joint dataframe, df.
     df
[]: \# I remove the NaN values from the "df" dataframe to form the final dataframe.
      \rightarrow for the analysis, "df1".
     df1=df.dropna()
[]: # I display the df1 dataframe.
     df1
[]: # I calculate the shape of the df1 dataframe. The dataframe has 38 rows and 2_{\sqcup}
      →columns.
     df1.shape
[]: # I create a scatter plot for the data in the df1 dataframe.
     df1.plot(kind="scatter", x="GPD per capita 2015 (USD)", y="Life Satisfaction⊔
      →Value", color="b", figsize=(10,6))
[]: \# I calculate the Pearson correlation coeffeicient r for the data in the df1_{\sqcup}
     \rightarrow dataframe
     # and display the correlation dataframe.
     df1.corr()
[]: # I extract all values of the "GPD per capita 2015 (USD)" and "Life,
      \rightarrow Satisfaction Value" columns and
     # form new (38x1) column arrays "a" and "b".
     a=df1.loc[:, ["GPD per capita 2015 (USD)"]].values
     b=df1.loc[:, ["Life Satisfaction Value"]].values
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[]: # I reshape the original (38x1) column arrary "a" to a (1x38) row array "X".
     X=a.reshape(38)
[]: # I display the "X" array.
     Х
[]: # I reshape the original (38x1) column "b" array to a (1x38) row array "y".
     y=b.reshape(38)
[]: # I display the "y" array.
     У
[]: \# First, I assume a simple linear regression model for the data in "X" and "y"
     \hookrightarrow arrays
     # and calculate the slope, intercept etc., of the linear regression method.
     # Here I use the "stats" module of "Scipy" library and its linear regression \Box
     \rightarrow built in method.
     result = sp.stats.linregress(X, y)
[]: # I print the results of the simple linear regression method.
     print(result)
[]: \# I create a figure with a single subplot where the original data of the df1_{\sqcup}
      \rightarrow dataframe
     # and the linear regression line Y(X) are shown.
     fig, ax=plt.subplots(figsize=(10, 6))
     ax.scatter(X, y, color='b', label="Original data")
     ax.plot(X, result.intercept + (result.slope)*X, color="m", label="Linear"
     \rightarrowregression line: Y(X)=5.74+2.39 \cdot 10^{-5} X")
     ax.set_xlabel("GPD per capita 2015 (USD)")
     ax.set_ylabel("Life Satifaction Value")
     plt.legend()
[]: | # I calculate the t-score in order to estimate the Confidence Intervals (CIs)
     # of the linear regression coefficients "beta_0" and "beta_1" at significance
     # level of alpha=0.05 and Confidence Level (CL) of 95%. The number of degrees \Box
     \rightarrow of freedom for the data is n=38.
     n=38
     alpha=0.05
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[]: # Second, I use the KNN regression method to find a relationship between the
     \rightarrow data for K=5 (default value).
     model=sklearn.neighbors.KNeighborsRegressor()
[]: # I use the fit() function to fit the data of the KNN method
     # and use the original column vectors "a" and "b" and redefine
     # them as "X1" and "y1" to use in the KNN regression method.
     X1=a
     y1=b
    model.fit(X1,y1)
[]: # I calculate the predicted values of the KNN method for
     # the GDP data "X" not present in the df1 dataframe for the countries of
     → Albania, United Arab Emirates and Armenia.
     X_new=[[3607.296697],[38663.383807],[3607.296697]]
[]: # I Print the predicted values of "Life Satisfaction Value" respectively
     # for Albania, United Arab Emirates and Armenia.
     print(model.predict(X_new))
[]: # I print the value of the generalized correlation coefficient R^2.
     print(model.score(X1, y1, sample_weight=None))
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t_score = sp.stats.t.ppf(1-alpha/2, n-2)

print(t)