# **ASSIGNMENT - 2**

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```
In []: import pandas as pd
    import numpy as np
    import wbgapi as wb
    import matplotlib.pyplot as plt
    import scipy.stats as stats
    %matplotlib inline
    import seaborn as sns
```

## **Defining Functions**

```
In [ ]: """Defines a function which takes a filename as argument, reads a dataframe in World bank format and returns
two dataframes: one with years as columns and one with countries as columns"""

def wb_dataframe(Filename,country_list):
    df1 = wb.data.DataFrame(Filename,country_codes,time=range(2010,2019),columns='economy')
    df2= wb.data.DataFrame(Filename,country_codes,time=range(2010,2019))
    return df1,df2
```

```
In [ ]: """Defines a function to provide Xlabel ,Y label and Titler for the Plots"""

def plot_labels(x_label,y_label,title):
    plt.xticks( rotation='vertical')
    plt.xlabel(x_label)
    plt.ylabel(y_label)
    plt.title(title,fontsize= 15, fontweight='bold')
```

```
"""Function to print the Different summary statitics of dataframe Attributes"""
In [ ]:
        def Display stats(Attributes):
                                   ", np.average(Attributes))
            print("average:
            print("std. deviations:", np.std(Attributes))
                               ", np.max(Attributes))
            print("maximum :
                                   ", np.min(Attributes))
            print("minimum :
                               ", stats.skew(Attributes))
            print("skewness:
            print("kurtosis:
                                   ", stats.kurtosis(Attributes))
            return
In [ ]: """Defining Function for the Subplots"""
        def subplot(data 1,data 2,data 3,data 4):
            plt.style.use('fivethirtyeight')
            plt.figure(figsize=(15,8), constrained layout=True)
            plt.subplot(4,1,1)
            plt.plot(data 1)
            plot labels('Years','values','Agricultural land (% of land area)')
            plt.subplot(4,1,2)
            plt.plot(data 2)
            plot labels('Years','values','Forest Area')
            plt.subplot(4,1,3)
            plt.plot(data 3)
            plot labels('Years','values','% of total territorial area')
            plt.subplot(4,1,4)
            plt.plot(data 4)
            plot labels('Years','values','value added (% of GDP)')
```

```
"""Function defining the Pie Plots Dimensions"""
        def Region wise(df1,df2,subtitle):
            fig, axes = plt.subplots(1, 2, figsize=(3,2),dpi=144)
            plt.suptitle(subtitle, size=10)
            for ax, df in zip(axes, (df1,df2)):
                ax.pie(df, labels=df.index,shadow=False,startangle=90, autopct="%1.1f%"
                            ,textprops={'fontsize': 5})
       """Prints Pearon's and Spearmans correction coefficients between the different attributes of the dataframe"""
In [ ]:
        def print corr(df):
            print("Pearson's correlation coefficients")
            r, p = stats.spearmanr(df["Agricultural land (% of land area)"], df["Forest Area"])
            print("\n Agricultural land (% of land area) vs. Forest Area coefficient ", r, "probability", p)
            r, p = stats.spearmanr(df["Forest Area"], df["% of total territorial area"])
            print("\n Forest Area vs. % of total territorial area coefficient", r, "probability", p)
            r, p = stats.spearmanr(df["% of total territorial area"], df["value added (% of GDP)"])
            print("\n % of total territorial area vs. value added (% of GDP) coefficient", r, "probability", p)
            r, p = stats.spearmanr(df["Agricultural land (% of land area)"], df["value added (% of GDP)"])
            print("\n Agricultural land (% of land area) vs. value added (% of GDP) coefficient", r, "probability", p)
```

**Defining and Creating DataFrame** 

```
In [ ]: idx=pd.IndexSlice
        #Defining Multiindex for Dataframe 2
        df1,df2 = wb dataframe(indicator list,country codes)
        df2
        index1 = pd.MultiIndex.from product([['Agricultural land (% of land area)', 'Forest Area', 'value added (% of GDP)'
                                               ,'% of total territorial area'],['GBR','NLD','ESP','JPN','KOR','IND','LKA','USA',
                                               ,names=['Attributes','Country'])
        df2.index = index1
In [ ]: #Defining Multiindex for Dataframe 1
        index = pd.MultiIndex.from product([['Agricultural land (% of land area)','Forest Area',
               'value added (% of GDP)','% of total territorial area'],['YR2010','YR2011','YR2012','YR2013','YR2014','YR2015',
                                                                        'YR2016', 'YR2017', 'YR2018']], names=['Attributes', 'year']
        df1.index=index
        Data Cleaning
In [ ]: # Checks for the null values in both the dataframes
        print(df1.isna().sum())
        print(df1.info())
        print(df2.isna().sum())
        print(df2.info())
In [ ]: # Filling the Nan Values with backward fill method
        df2.fillna(method='bfill',inplace=True)
        df1.fillna(method='bfill',inplace=True)
```

#### **Data Preperation**

### **Plotting and Analysis**

df 1

Analysing and Plotting the Averge Variations in the Agricultural, Tesrrestial, Forest And Value added GDP across the world

```
In [ ]: print()
        print("Agricultural land (% of land area)")
        Display stats(df1.loc[idx['Agricultural land (% of land area)', ("YR2014","YR2015","YR2016","YR2017","YR2018")], :])
        print()
        print("Forest Area")
        Display stats(df1.loc[idx['Forest Area', ("YR2014","YR2015","YR2016","YR2017","YR2018")], :])
        print()
        print("value added (% of GDP)")
        Display stats(df1.loc[idx['value added (% of GDP)', ("YR2014", "YR2015", "YR2016", "YR2017", "YR2018")], :])
        print()
        print("% of total territorial area")
        Display stats(df1.loc[idx['% of total territorial area', ("YR2014","YR2015","YR2016","YR2017","YR2018")], :])
In [ ]: """THe average values of different constrains during the year of consideration assigning as dataframe dd"""
        dd=pd.DataFrame((df2.groupby('Attributes')).mean())
        ddd=dd.T
        ddd.index
In [ ]: import seaborn as sns
        # Declaring the cm variable by the color palette from seaborn
        cm = sns.light_palette("orange", as_cmap=True)
        # Visualizing the DataFrame with set precision
        ddd.style.background gradient(cmap=cm)
```

#### Analysing and Plotting the Attributes for selected Countries(GBR,USA,IND,JPN) in Different regions of world

```
In [ ]: #Defing the datas for the individual plots of Country USA
        idx=pd.IndexSlice
        data=df 1.loc[idx['Agricultural land (% of land area)','USA']]
        data1=df 1.loc[idx['Forest Area','USA']]*.4
        data2=df 1.loc[idx['% of total territorial area','USA']]
        data3=df 1.loc[idx['value added (% of GDP)','USA']]
        #Function Call for Subplotting
        subplot(data,data1,data2,data3)
        #Display its Statistics as well
        print()
        print("USA")
        pd.DataFrame(Display stats(df1.loc[idx['Agricultural land (% of land area)', 'USA']]))
In [ ]: idx=pd.IndexSlice
        #Defing the datas for the individual plots of Country INDIA
        data=df 1.loc[idx['Agricultural land (% of land area)','IND']]
        data1=df 1.loc[idx['Forest Area','IND']]*.4
        data2=df 1.loc[idx['% of total territorial area','IND']]
        data3=df 1.loc[idx['value added (% of GDP)','IND']]
        #Function Call for Subplotting
        subplot(data,data1,data2,data3)
        #Display its Statistics as well
        print()
        print("IND")
        pd.DataFrame(Display stats(df1.loc[idx['Agricultural land (% of land area)', 'IND']]))
```

```
In [ ]: #Defing the datas for the individual plots of Country JAPAN
        idx=pd.IndexSlice
        data=df 1.loc[idx['Agricultural land (% of land area)','JPN']]
        data1=df 1.loc[idx['Forest Area','JPN']]*.4
        data2=df 1.loc[idx['% of total territorial area','JPN']]
        data3=df 1.loc[idx['value added (% of GDP)','JPN']]
        #Function Call for Subplotting
        subplot(data,data1,data2,data3)
        #Display its Statistics as well
        print()
        print("JPN")
        pd.DataFrame(Display stats(df1.loc[idx['Agricultural land (% of land area)', 'JPN']]))
In [ ]: #Defining an dAnalysing the Variation in GDP for the above defined Countries
        d1=df 1.loc[idx['value added (% of GDP)',['USA','GBR','KOR','JPN']]]
        print(d1)
        d1.index=['2010','2011','2012','2013','2014','2015','2016','2017','2018']
        d1.index = pd.to datetime(d1.index)
        print(d1)
        d1.plot(subplots=True, figsize=(12, 15))
```

Analysing and plotting variations in attribute Across different Regions of World(Europe,North America,South and East Asia)

```
In [ ]: # Mapping the Existing Datframe with the Country groups Dictionary to have a detailed Analysis.
        df2['Region'] = df2.index.get level values(1).map(country groups)
        df2.groupby('Region')
        # Defining the Slices For Plotting
        agri=df2.loc[idx[['Agricultural land (% of land area)'],:]].groupby('Region').mean()
        forest=df2.loc[idx[['Forest Area'],:]].groupby('Region').mean()
        terrestial=df2.loc[idx[['% of total territorial area'],:]].groupby('Region').mean()
        # Locating the Slices For Desired years
        agri a = agri.loc[:, "YR2015":"YR2018"].mean(axis=1)
        agri b= agri.loc[:, "YR2010":"YR2014"].mean(axis=1)
        forest a = forest.loc[:, "YR2015":"YR2018"].mean(axis=1)
        forest b= forest.loc[:, "YR2010":"YR2014"].mean(axis=1)
        Terrestial a = terrestial.loc[:, "YR2015":"YR2018"].mean(axis=1)
        Terrestial b= terrestial.loc[:, "YR2010":"YR2014"].mean(axis=1)
        # Plotting with the Function Region wise
        Region wise(agri b,agri a,'Agricultural land before and after 2015')
        Region wise(forest b, forest a, 'Forest Area before and after 2015')
        Region wise(Terrestial b, Terrestial_a, 'Total territorial area before and after 2015')
In [ ]: # Printing the Correlation between Attributes
        print corr(ddd)
In [ ]: # plotting the Correlation using Heatmap
        sns.heatmap(ddd.corr(), annot = True,cmap="YlOrRd r")
```