

In [7]: # display the missing values count

df.isnull().sum()

Out[7]:

```
Country      0
Year         0
Status       0
Life expectancy  10
Adult Mortality  10
infant deaths   0
Alcohol      194
percentage expenditure  0
Hepatitis B    553
Measles       0
BMI          34
under-five deaths  0
Polio        19
Total expenditure  226
Diphtheria    19
HIV/AIDS      0
GDP          448
Population    652
thinness 1-19 years  34
thinness 5-9 years  34
Income composition of resources  167
Schooling     163
dtype: int64
```

In [8]: # display the missing values percentage
round(df.isnull().sum() / df.shape[0]*100,2)

Out[8]:

```
Country      0.00
Year         0.00
Status       0.00
Life expectancy  0.34
Adult Mortality  0.34
infant deaths   0.00
Alcohol      6.60
percentage expenditure  0.00
Hepatitis B   18.82
Measles       0.00
BMI          1.16
under-five deaths  0.00
Polio        0.65
Total expenditure  7.69
Diphtheria    0.65
HIV/AIDS      0.00
GDP          15.25
Population    22.19
thinness 1-19 years  1.16
thinness 5-9 years  1.16
Income composition of resources  5.68
Schooling     5.55
dtype: float64
```

In [9]: # check the duplicate value

df.duplicated().sum()

Out[9]: 0

In [10]:

```
# identify the garbage value
for i in df.select_dtypes(include='object').columns:
    print(df[i].value_counts())
    print(' '*30)
```

```
Afghanistan      16
Peru              16
Nicaragua         16
Niger            16
Nigeria          16
..              ..
Niue              1
San Marino        1
Nauru             1
Saint Kitts and Nevis  1
Dominica          1
Name: Country, Length: 193, dtype: int64
*****
Developing      2426
Developed       512
Name: Status, dtype: int64
*****
```

In [11]: *# describe numerical features*
df.describe().T

Out[11]:

	count	mean	std	min	25%	50%
Year	2938.0	2.007519e+03	4.613841e+00	2000.000000	2004.000000	2.008000e+03
Life expectancy	2928.0	6.922493e+01	9.523867e+00	36.300000	63.100000	7.210000e+01
Adult Mortality	2928.0	1.647964e+02	1.242921e+02	1.000000	74.000000	1.440000e+02
infant deaths	2938.0	3.030395e+01	1.179265e+02	0.000000	0.000000	3.000000e+00
Alcohol	2744.0	4.602861e+00	4.052413e+00	0.010000	0.877500	3.755000e+00
percentage expenditure	2938.0	7.382513e+02	1.987915e+03	0.000000	4.885343	6.491291e+01
Hepatitis B	2385.0	8.094046e+01	2.507002e+01	1.000000	77.000000	9.200000e+01
Measles	2938.0	2.419592e+03	1.146727e+04	0.000000	0.000000	1.700000e+01
BMI	2904.0	3.832125e+01	2.004403e+01	1.000000	19.300000	4.350000e+01
under-five deaths	2938.0	4.203574e+01	1.604455e+02	0.000000	0.000000	4.000000e+00
Polio	2919.0	8.255019e+01	2.342805e+01	3.000000	78.000000	9.300000e+01
Total expenditure	2712.0	5.938190e+00	2.498320e+00	0.370000	4.260000	5.755000e+00
Diphtheria	2919.0	8.232408e+01	2.371691e+01	2.000000	78.000000	9.300000e+01
HIV/AIDS	2938.0	1.742103e+00	5.077785e+00	0.100000	0.100000	1.000000e-01
GDP	2490.0	7.483158e+03	1.427017e+04	1.681350	463.935626	1.766948e+03
Population	2286.0	1.2753338e+07	6.101210e+07	34.000000	195793.250000	1.386542e+06
thinness 1-19 years	2904.0	4.839704e+00	4.420195e+00	0.100000	1.600000	3.300000e+00
thinness 5-9 years	2904.0	4.870317e+00	4.508882e+00	0.100000	1.500000	3.300000e+00
Income composition of resources	2771.0	6.275511e-01	2.109036e-01	0.000000	0.493000	6.770000e-01
Schooling	2775.0	1.199279e+01	3.358920e+00	0.000000	10.100000	1.230000e+01

In [12]: *# describing categorical features*
df.describe(include="object").T

Out[12]:

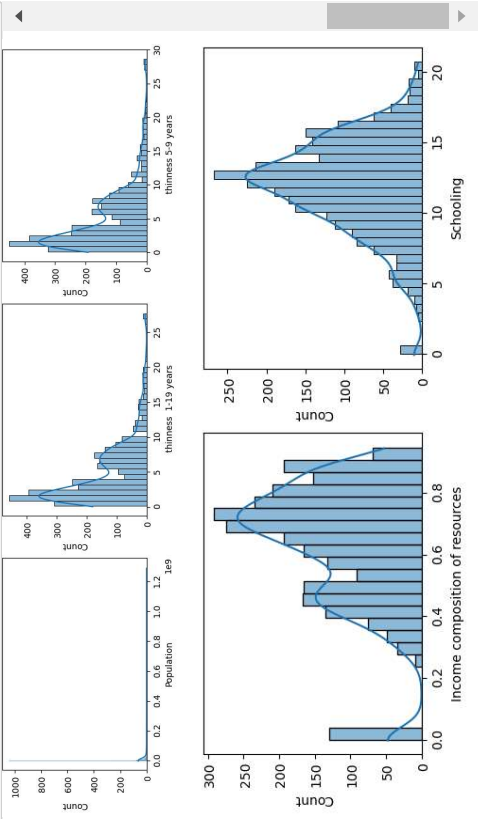
	count	unique	top	freq
Country	2938	193	Afghanistan	16
Status	2938	2	Developing	2426

Exploratory Data Analysis

- check data distribution

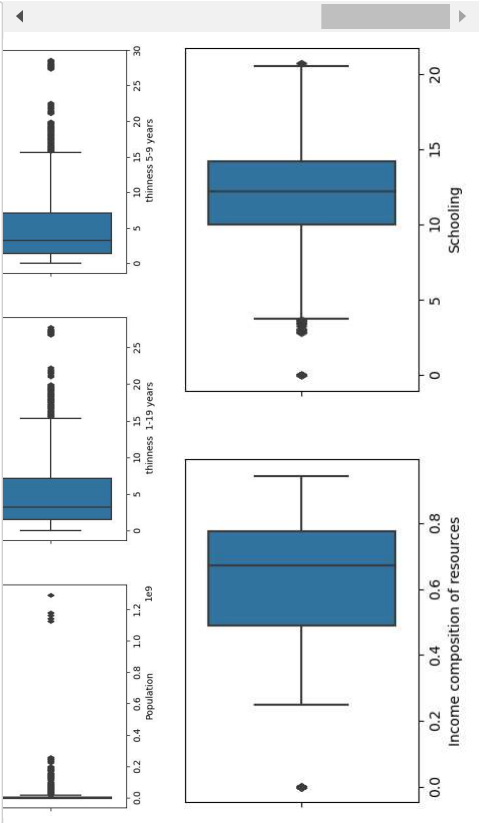
In [13]:

```
import warnings
warnings.filterwarnings("ignore")
f = 1
plt.figure(figsize=(10,4))
for i in df.select_dtypes(include='number').columns:
    plt.subplot(1,3,f)
    sns.histplot(data=df, x=i, kde=True)
    if f<3:
        f += 1
    else:
        f = 1
    plt.show()
plt.figure(figsize=(15,3))
```



Identify Outlier

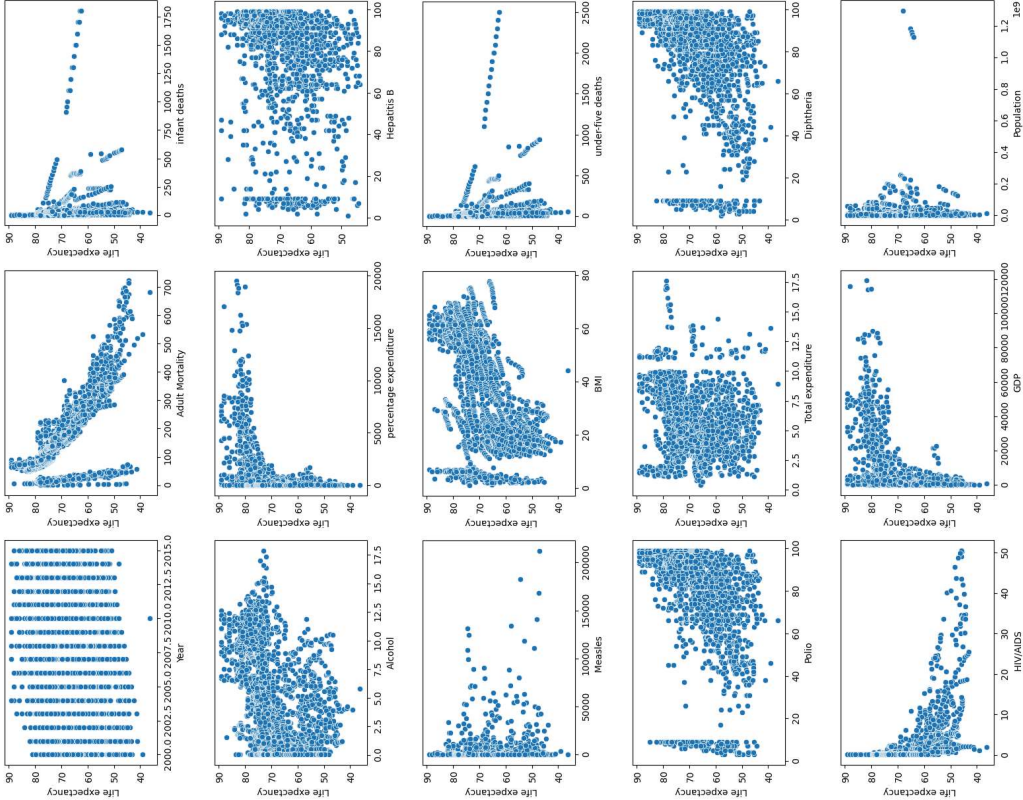
```
In [14]: f = 1
plt.figure(figsize=(15,3))
for i in df.select_dtypes(include='number').columns:
    plt.subplot(1,3,f)
    sns.boxplot(data=df, x=i)
    if f<3:
        f += 1
    else:
        f = 1
plt.show()
plt.figure(figsize=(15,3))
```

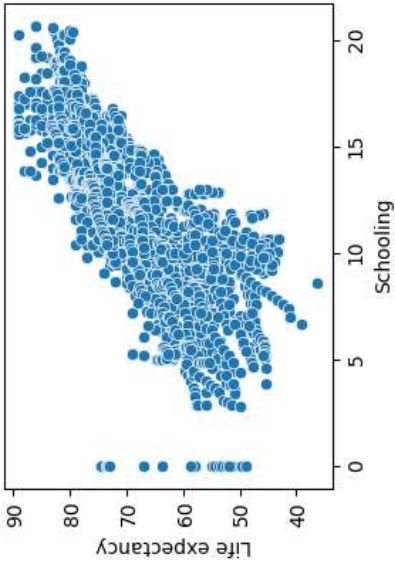
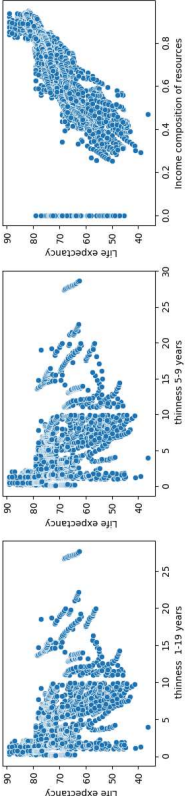


Relation between Feature Matrix and Target Vector

```
In [15]: x = df.select_dtypes(include="number").columns
x = list(x)
target = 'Life expectancy'
x.remove(target) # removing target vector
```

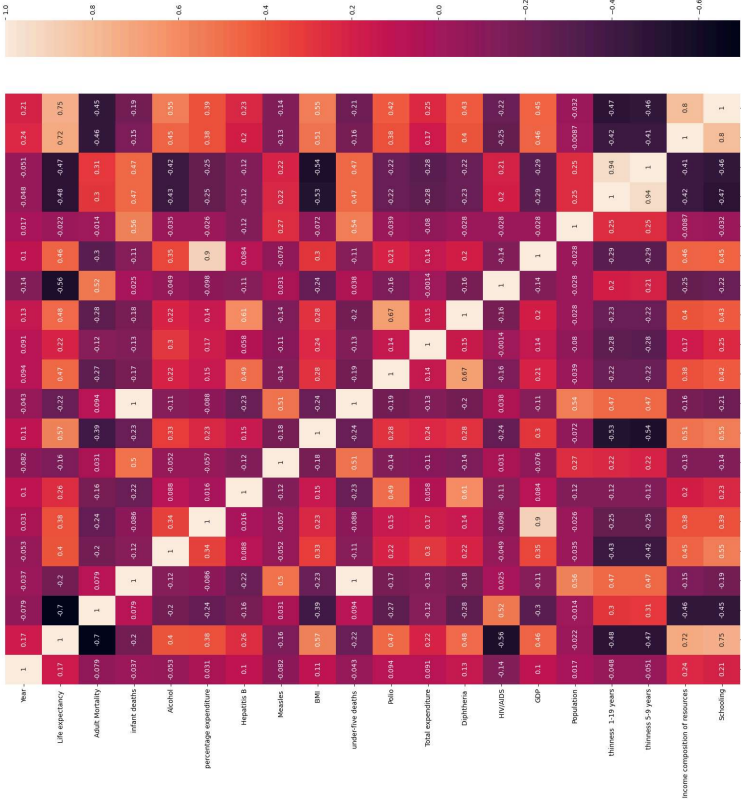
```
In [16]: f = 1
plt.figure(figsize=(15,3))
for i in x:
    plt.subplot(1,3,f)
    sns.scatterplot(data=df,x=i,y=target)
    if f<3:
        f += 1
    else:
        f = 1
plt.show()
plt.figure(figsize=(15,3))
```





In [17]:

```
corr_matrix = df.select_dtypes(include='number').corr()  
plt.figure(figsize=(20,20))  
sns.heatmap(corr_matrix, annot=True)  
plt.show()
```



Missing Value Treatment

- Traditional Method - (Mean, Mode, Median)
- New Method - KNNImputer

In [18]:

```
for i in ['BMI', 'Polio', 'Income composition of resources']:  
    df[i].fillna(df[i].median(), inplace=True)
```

In [19]: df.isna().sum()

Out[19]: Country 0
Year 0
Status 0
Life expectancy 10
Adult Mortality 10
Infant deaths 0
Alcohol 194
percentage expenditure 0
Hepatitis B 553
Measles 0
BMI 0
under-five deaths 0
Polio 0
Total expenditure 226
Diphtheria 19
HIV/AIDS 0
GDP 448
Population 652
thinness 1-19 years 34
thinness 5-9 years 34
Income composition of resources 0
Schooling 163
dtype: int64

In [20]: # using KNNImputer
from sklearn.impute import KNNImputer
imputer = KNNImputer()

In [21]: for i in df.select_dtypes(include='number').columns:
df[i] = imputer.fit_transform(df[[i]])

In [22]: df.isna().sum()

Out[22]: Country 0
Year 0
Status 0
Life expectancy 0
Adult Mortality 0
Infant deaths 0
Alcohol 0
percentage expenditure 0
Hepatitis B 0
Measles 0
BMI 0
under-five deaths 0
Polio 0
Total expenditure 0
Diphtheria 0
HIV/AIDS 0
GDP 0
Population 0
thinness 1-19 years 0
thinness 5-9 years 0
Income composition of resources 0
Schooling 0
dtype: int64

In [23]: imputer.n_neighbors
Out[23]: 5

• Outlier Treatment

In [24]: def wisker(col):
q1,q3 = np.percentile(col,[25,75])
iqr = q3 - q1
hf = q3 + 1.5 * iqr
lf = q1 - 1.5 * iqr
return lf,hf

In [25]: wisker(df['GDP'])

Out[25]: (-9773.52021495771, 17837.165679596183)

In [26]: df_outlier_cols = list(df.select_dtypes(include='number').columns)
df_outlier_cols.remove('Year')
df_outlier_cols.remove(' BMI ')
df_outlier_cols.remove('Alcohol')
df_outlier_cols[:2]

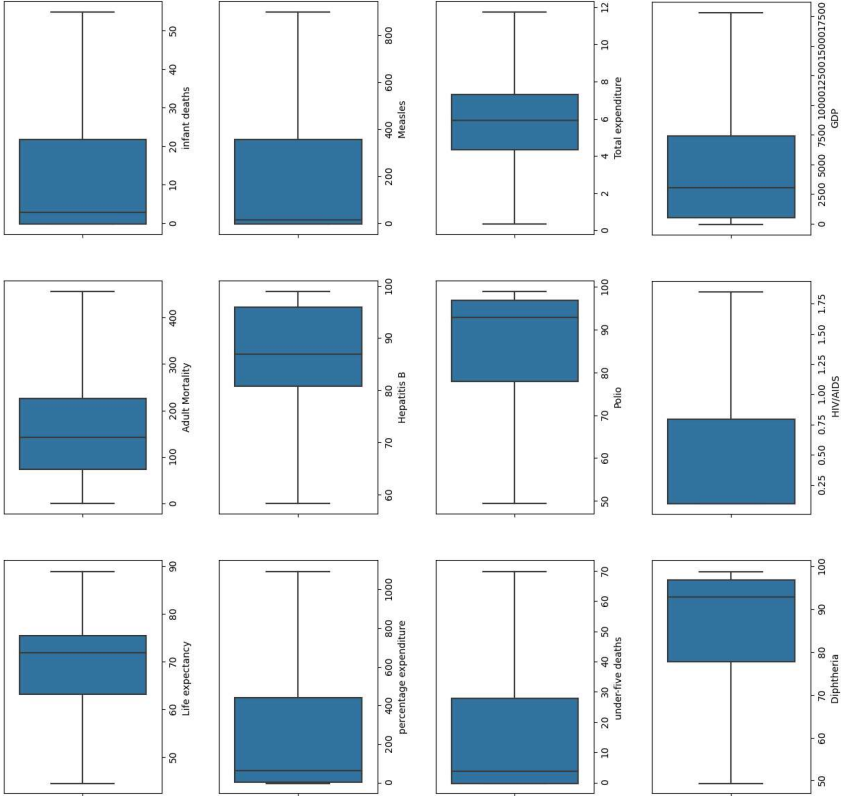
Out[26]: ['Life expectancy ', 'Adult Mortality']

In [28]:

```
# removing outlier
for i in df_outlier_cols:
    lw, uw = whisker(df[i])
    df[i] = np.where(df[i]<lw, lw, np.where(df[i]>uw, uw, df[i]))

# boxplot after removing outlier

f = 1
plt.figure(figsize=(15,3))
for i in df_outlier_cols:
    plt.subplot(1,3,f)
    sns.boxplot(data=df, x=i)
    if f<3:
        f += 1
    else:
        f = 1
plt.show()
plt.figure(figsize=(15,3))
```



In [29]:

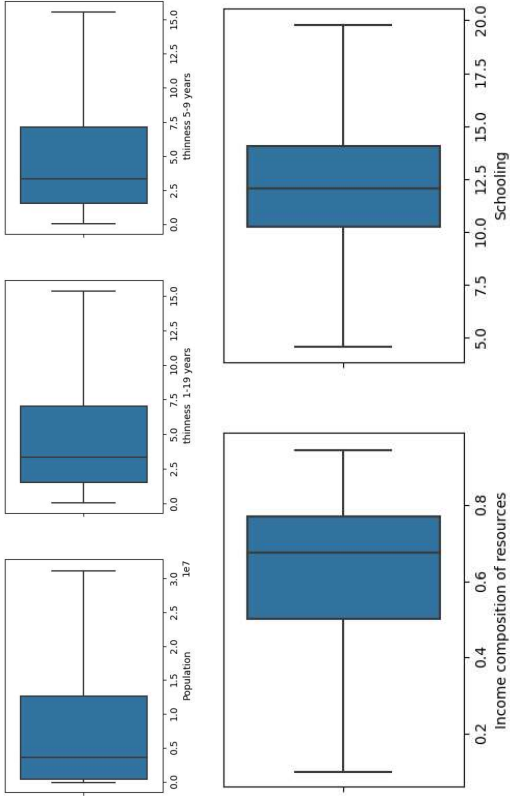
```
print(df.shape)
df.drop_duplicates(inplace=True)
df.shape
```

(2938, 22)

Out[29]:

(2938, 22)

- Duplicates & garbage value treatment



- Encoding of data

```
In [30]: mydata = pd.get_dummies(data=df,columns=['Country', 'Status'], drop_first=True)
print(mydata.shape)
mydata.head()
```

(2938, 213)

Out[30]:

	Year	Life expectancy	Adult Mortality	infant deaths	Alcohol	percentage expenditure	Hepatitis B	Measles	BMI	under-five deaths
0	2015.0	65.0	263.0	55.0	0.01	71.279624	65.0	900.625	19.1	7
1	2014.0	59.9	271.0	55.0	0.01	73.523582	62.0	492.000	18.6	7
2	2013.0	59.9	268.0	55.0	0.01	73.219243	64.0	430.000	18.1	7
3	2012.0	59.5	272.0	55.0	0.01	78.184215	67.0	900.625	17.6	7
4	2011.0	59.2	275.0	55.0	0.01	7.097109	68.0	900.625	17.2	7

5 rows × 213 columns

- Normalization

```
In [31]: X = mydata.drop(['Life expectancy'],axis=1)
y = mydata['Life expectancy']
```

```
In [32]: # perform Standardization using StandardScaler
from sklearn.preprocessing import StandardScaler

scaler = StandardScaler()

X_scaled = pd.DataFrame(scaler.fit_transform(X), columns=X.columns)
```

In [33]: X_scaled.head(2)

Out[33]:

	Year	Adult Mortality	infant deaths	Alcohol	percentage expenditure	Hepatitis B	Measles	BMI	un	de
0	1.621762	0.874521	2.165057	-1.172958	-0.546410	-1.534064	1.886225	-0.967349	2.061	
1	1.404986	0.943807	2.165057	-1.172958	-0.540647	-1.768413	0.730456	-0.992434	2.061	

2 rows × 212 columns

In [34]:

```
# or perform Normalization using MinMaxScaler
from sklearn.preprocessing import MinMaxScaler

scaler = MinMaxScaler(feature_range=(-1,1)) # default range is 0 to 1
X_scaled_normal = pd.DataFrame(scaler.fit_transform(X), columns=X.columns)
```

In [35]: X_scaled_normal.head(4)

Out[35]:

	Year	Adult Mortality	infant deaths	Alcohol	percentage expenditure	Hepatitis B	Measles	BMI	under-five deaths
0	1.000000	0.150384	1.0	-1.0	-0.870023	-0.672864	1.000000	-0.580533	1.0
1	0.866667	0.185510	1.0	-1.0	-0.865932	-0.820470	0.092575	-0.592121	1.0
2	0.733333	0.172338	1.0	-1.0	-0.866487	-0.722066	-0.045108	-0.603708	1.0
3	0.600000	0.189901	1.0	-1.0	-0.857433	-0.574460	1.000000	-0.615295	1.0

4 rows × 212 columns

- Split dataset into Train and Test

```
In [36]: from sklearn.model_selection import train_test_split
X_train, X_test, y_train, y_test = train_test_split(X_scaled_normal, y, ran
```

In [37]: X_test.shape, X_train.shape

Out[37]: ((882, 212), (2056, 212))

In []: