

```
1 !pip install hvplot
```

```
Collecting hvplot
  Downloading hvplot-0.9.2-py2.py3-none-any.whl (1.8 MB)
    1.8/1.8 MB 9.8 MB/s eta 0:00:00
Requirement already satisfied: bokeh>=1.0.0 in /usr/local/lib/python3.10/dist-packages (from hvplot) (3.3.4)
Requirement already satisfied: colorcet>=2 in /usr/local/lib/python3.10/dist-packages (from hvplot) (3.1.0)
Requirement already satisfied: holoviews>=1.11.0 in /usr/local/lib/python3.10/dist-packages (from hvplot) (1.17.1)
Requirement already satisfied: pandas in /usr/local/lib/python3.10/dist-packages (from hvplot) (2.0.3)
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Requirement already satisfied: panel>=0.11.0 in /usr/local/lib/python3.10/dist-packages (from hvplot) (1.3.8)
Requirement already satisfied: param<3.0,>=1.12.0 in /usr/local/lib/python3.10/dist-packages (from hvplot) (2.1.0)
Requirement already satisfied: Jinja2>=2.9 in /usr/local/lib/python3.10/dist-packages (from bokeh>=1.0.0->hvplot) (3.1.3)
Requirement already satisfied: contourpy>=1 in /usr/local/lib/python3.10/dist-packages (from bokeh>=1.0.0->hvplot) (1.2.1)
Requirement already satisfied: pillow>=7.1.0 in /usr/local/lib/python3.10/dist-packages (from bokeh>=1.0.0->hvplot) (9.4.0)
Requirement already satisfied: PyYAML>=3.10 in /usr/local/lib/python3.10/dist-packages (from bokeh>=1.0.0->hvplot) (6.0.1)
Requirement already satisfied: tornado>=5.1 in /usr/local/lib/python3.10/dist-packages (from bokeh>=1.0.0->hvplot) (6.3.3)
Requirement already satisfied: xyzservices>=2021.09.1 in /usr/local/lib/python3.10/dist-packages (from bokeh>=1.0.0->hvplot) (2024.4.0)
Requirement already satisfied: pyviz-commms>=0.7.4 in /usr/local/lib/python3.10/dist-packages (from holoviews>=1.11.0->hvplot) (3.0.2)
Requirement already satisfied: python-dateutil>=2.8.2 in /usr/local/lib/python3.10/dist-packages (from pandas->hvplot) (2.8.2)
Requirement already satisfied: pytz>=2020.1 in /usr/local/lib/python3.10/dist-packages (from pandas->hvplot) (2023.4)
Requirement already satisfied: tzdata>=2022.1 in /usr/local/lib/python3.10/dist-packages (from pandas->hvplot) (2024.1)
Requirement already satisfied: markdown in /usr/local/lib/python3.10/dist-packages (from panel>=0.11.0->hvplot) (3.6)
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Requirement already satisfied: requests in /usr/local/lib/python3.10/dist-packages (from panel>=0.11.0->hvplot) (2.31.0)
Requirement already satisfied: tqdm>=4.48.0 in /usr/local/lib/python3.10/dist-packages (from panel>=0.11.0->hvplot) (4.66.2)
Requirement already satisfied: bleach in /usr/local/lib/python3.10/dist-packages (from panel>=0.11.0->hvplot) (6.1.0)
Requirement already satisfied: typing-extensions in /usr/local/lib/python3.10/dist-packages (from panel>=0.11.0->hvplot) (4.11.0)
Requirement already satisfied: MarkupSafe>=2.0 in /usr/local/lib/python3.10/dist-packages (from Jinja2>=2.9->bokeh>=1.0.0->hvplot) (2.1.5)
Requirement already satisfied: six>=1.5 in /usr/local/lib/python3.10/dist-packages (from python-dateutil>=2.8.2->pandas->hvplot) (1.16.0)
Requirement already satisfied: webencodings in /usr/local/lib/python3.10/dist-packages (from bleach->panel>=0.11.0->hvplot) (0.5.1)
Requirement already satisfied: uc-micro-py in /usr/local/lib/python3.10/dist-packages (from linkify-it-py->panel>=0.11.0->hvplot) (1.0.3)
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Requirement already satisfied: charset-normalizer<4,>=2 in /usr/local/lib/python3.10/dist-packages (from requests->panel>=0.11.0->hvplot) (3.3.2)
Requirement already satisfied: idna<4,>=2.5 in /usr/local/lib/python3.10/dist-packages (from requests->panel>=0.11.0->hvplot) (3.7)
Requirement already satisfied: urllib3<3,>=1.21.1 in /usr/local/lib/python3.10/dist-packages (from requests->panel>=0.11.0->hvplot) (2.0.7)
Requirement already satisfied: certifi>=2017.4.17 in /usr/local/lib/python3.10/dist-packages (from requests->panel>=0.11.0->hvplot) (2024.2.2)
Installing collected packages: hvplot
Successfully installed hvplot-0.9.2
```

```
1 import pandas as pd
2 import numpy as np
3 import matplotlib.pyplot as plt
4 import seaborn as sns
5 import hvplot.pandas
6
7 from sklearn.model_selection import train_test_split
8
9 from sklearn import metrics
10
11 from sklearn.linear_model import LinearRegression
12
13 %matplotlib inline
```

✓ Check Out Data

```
1 df=pd.read_csv('/content/Life Expectancy Data.csv')
```

```
1 df.head()
```

	Country	Year	Status	Life expectancy	Adult Mortality	infant deaths	Alcohol	percentage expenditure	Hepatitis B	Measles	BMI	under-five deaths	Po
0	Afghanistan	2015	Developing	65.0	263.0	62	0.01	71.279624	65.0	1154	19.1	83	
1	Afghanistan	2014	Developing	59.9	271.0	64	0.01	73.523582	62.0	492	18.6	86	
2	Afghanistan	2013	Developing	59.9	268.0	66	0.01	73.219243	64.0	430	18.1	89	
3	Afghanistan	2012	Developing	59.5	272.0	69	0.01	78.184215	67.0	2787	17.6	93	
4	Afghanistan	2011	Developing	59.2	275.0	71	0.01	7.097109	68.0	3013	17.2	97	

```
1 df.shape
```

```
(2938, 22)
```

```
1 df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 2938 entries, 0 to 2937
Data columns (total 22 columns):
 #   Column           Non-Null Count  Dtype  
--- 
 0   Country          2938 non-null   object 
 1   Year              2938 non-null   int64  
 2   Status             2938 non-null   object 
 3   Life expectancy    2928 non-null   float64
 4   Adult Mortality    2928 non-null   float64
 5   infant deaths     2938 non-null   int64  
 6   Alcohol            2744 non-null   float64
 7   percentage expenditure  2938 non-null   float64
 8   Hepatitis B        2385 non-null   float64
 9   Measles             2938 non-null   int64  
 10  BMI                2904 non-null   float64
 11  under-five deaths  2938 non-null   int64  
 12  Polio               2919 non-null   float64
 13  Total expenditure   2712 non-null   float64
 14  Diphtheria          2919 non-null   float64
 15  HIV/AIDS            2938 non-null   float64
 16  GDP                 2490 non-null   float64
 17  Population           2286 non-null   float64
 18  thinness 1-19 years  2904 non-null   float64
 19  thinness 5-9 years   2904 non-null   float64
 20  Income composition of resources 2771 non-null   float64
 21  Schooling            2775 non-null   float64
dtypes: float64(16), int64(4), object(2)
memory usage: 505.1+ KB
```

```
1 print(df.dtypes)
Country          object
Year            int64
Status          object
Life expectancy float64
Adult Mortality float64
infant deaths   int64
Alcohol         float64
percentage expenditure float64
Hepatitis B    float64
Measles          int64
BMI             float64
under-five deaths int64
Polio            float64
Total expenditure float64
Diphtheria      float64
HIV/AIDS        float64
GDP             float64
Population      float64
  thinness 1-19 years float64
  thinness 5-9 years  float64
Income composition of resources float64
Schooling        float64
dtype: object
```

```
1 numeric_df = df.select_dtypes(include=[np.number])
2
3 print(numeric_df.isnull().sum())
4
5 numeric_df = numeric_df.dropna()
6
7 correlation_matrix = numeric_df.corr()
8
9 print(correlation_matrix)

Year          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
Year      ... Schooling
Year     1.000000 ... 0.088732
Life expectancy 0.050771 ... 0.727630
```

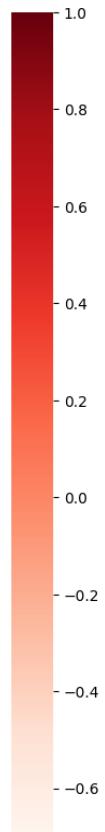
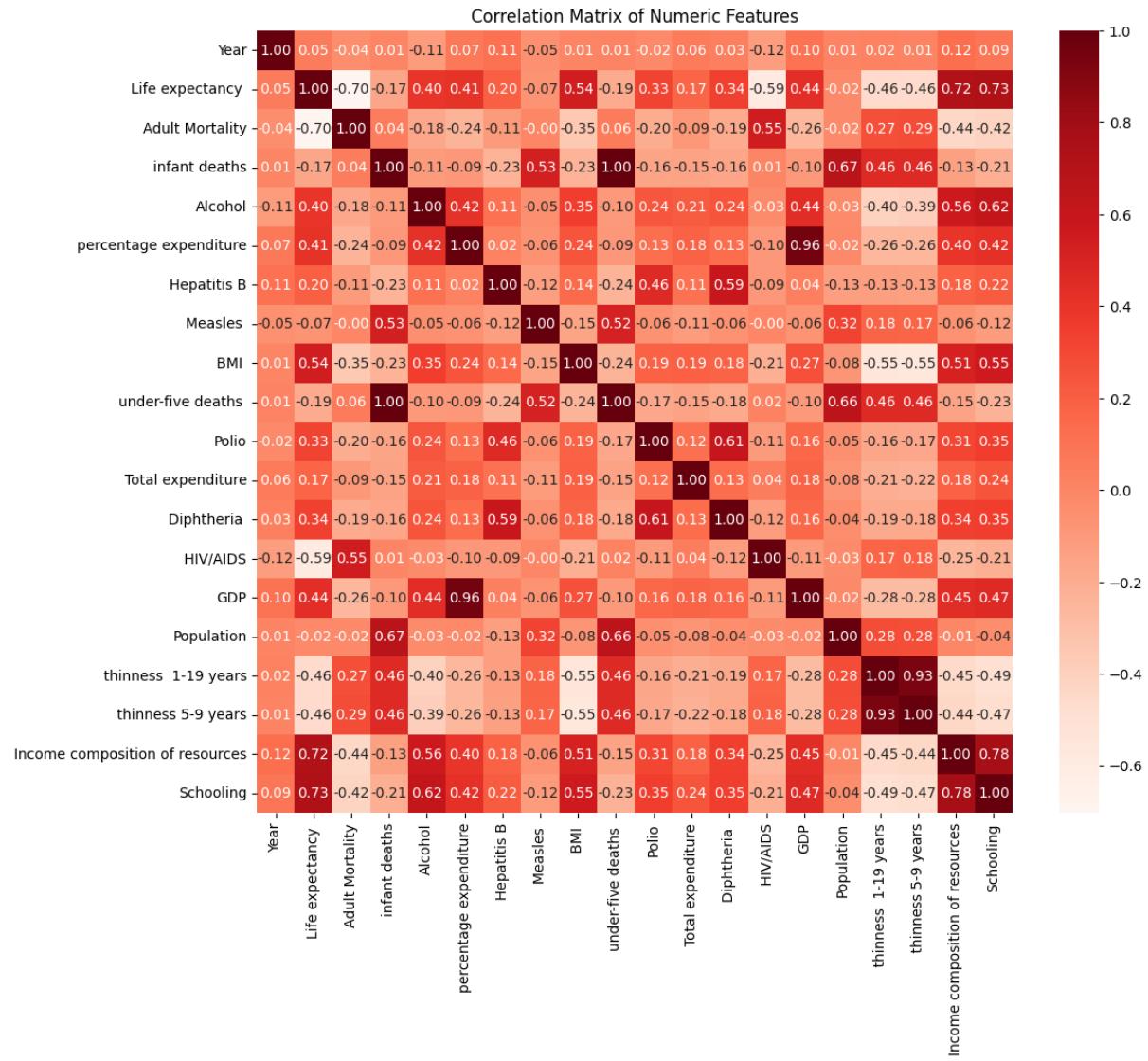
Adult Mortality	-0.037092	...	-0.421171
infant deaths	0.008029	...	-0.214372
Alcohol	-0.113365	...	0.616975
percentage expenditure	0.069553	...	0.422088
Hepatitis B	0.114897	...	0.215182
Measles	-0.053822	...	-0.115660
BMI	0.005739	...	0.554844
under-five deaths	0.010479	...	-0.226013
Polio	-0.016699	...	0.350147
Total expenditure	0.059493	...	0.243783
Diphtheria	0.029641	...	0.350398
HIV/AIDS	-0.123405	...	-0.211840
GDP	0.096421	...	0.467947
Population	0.012567	...	-0.040312
thinness 1-19 years	0.019757	...	-0.491199
thinness 5-9 years	0.014122	...	-0.472482
Income composition of resources	0.122892	...	0.784741
Schooling	0.088732	...	1.000000

[20 rows x 20 columns]

```

1 numeric_df = df.select_dtypes(include=[np.number])
2
3 numeric_df = numeric_df.dropna()
4
5 correlation_matrix = numeric_df.corr()
6
7 plt.figure(figsize=(12, 10))
8 sns.heatmap(correlation_matrix, annot=True, cmap='Reds', fmt=".2f")
9 plt.title('Correlation Matrix of Numeric Features')
10 plt.show()

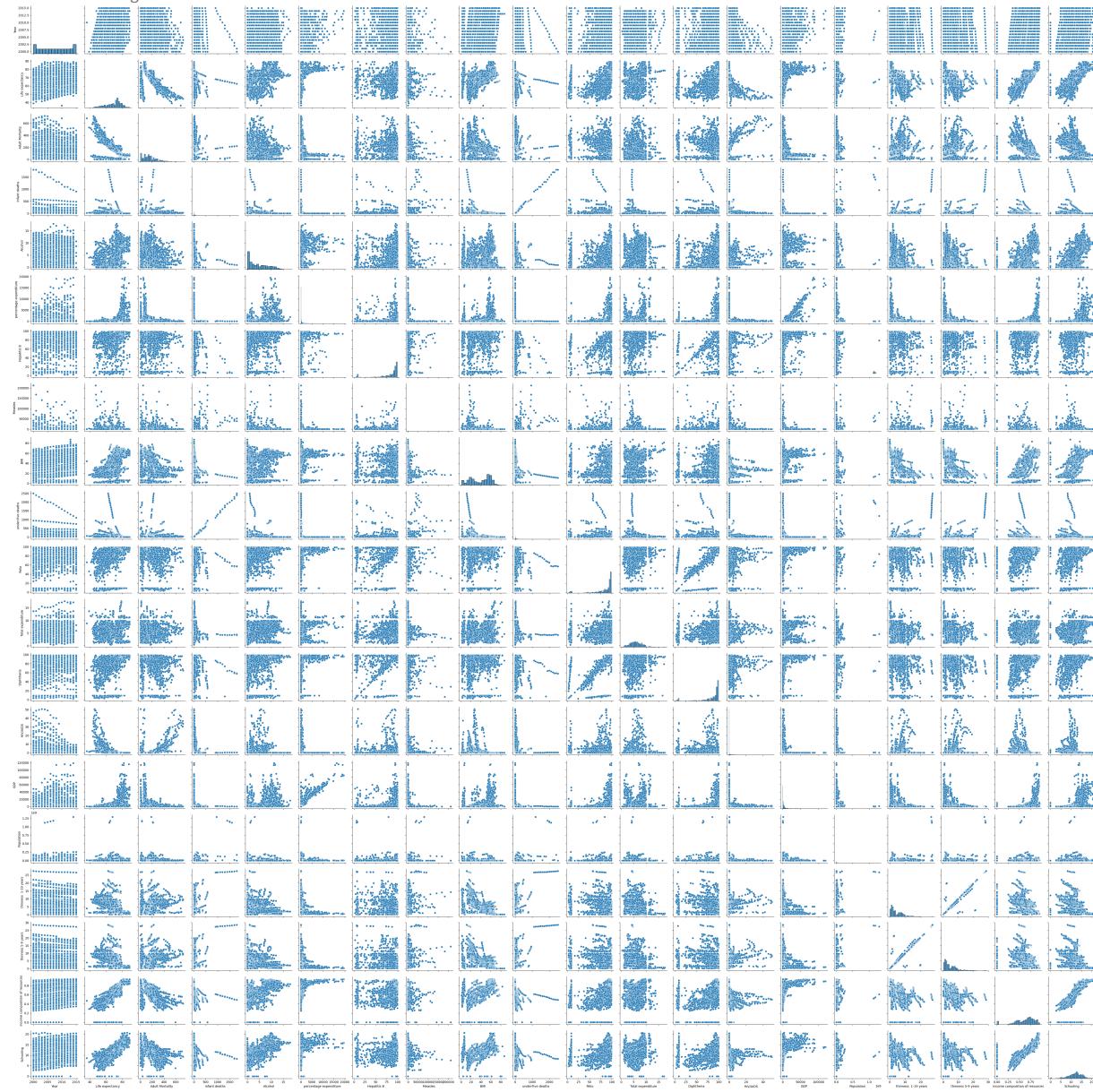
```



✓ Exploratory Data Analysis (EDA)

```
1 sns.pairplot(df)
```

```
<seaborn.axisgrid.PairGrid at 0x7ad6ddc71450>
```



⌄ Training a Linear Regression Model

X and y Arrays

```
1 df.columns = df.columns.str.strip()  
  
1 X = df.drop('Life expectancy', axis=1)  
2 y = df['Life expectancy']
```

Train Test Split

```
1 X_train, X_test, y_train, y_test = train_test_split(features, target, test_size=0.3, random_state=101)  
  
1 X_train.shape  
(2056, 214)  
  
1 X_test.shape  
(882, 214)
```

⌄ Linear Regression

```
1 model = LinearRegression()

1 model.fit(X_train, y_train)

▼ LinearRegression
LinearRegression()

1 model.coef_

array([[ 2.63644186e-01, -2.40133942e-03,   1.08407317e-01, -7.83409090e-02,
       1.30738853e-04, -2.95212685e-03, -8.22343021e-06, -2.12118090e-03,
      -8.20382163e-02,   3.74677532e-03, -8.96052195e-03,   7.06028674e-03,
     -3.11921426e-01, -1.09325914e-05,   1.61644920e-10, -1.20523567e-02,
      5.69775053e-02, -6.17774814e-01,   1.54759935e-01, -9.19236963e+00,
      6.84769665e+00,   4.69103310e+00, -1.56854428e+01,   7.30302397e+00,
     6.53967147e+00,   5.16622440e+00,   2.18934862e+00,   3.22964390e+00,
     2.14940857e+00, -6.76677706e+00,   6.95652980e+00,   7.26011021e-01,
     6.36158412e+00,   2.52950580e+00,   2.34057845e+00,   1.63617795e+00,
    -9.15499070e+00, -2.90262876e+00, -6.39344610e-01,   7.84883228e+00,
    -6.96377887e+00,   4.19736970e+00,   7.60374996e+00, -5.65427026e+00,
    -9.20182925e+00, -1.11026005e+01,   4.24190169e+00, -3.50378887e+00,
   -1.05605207e+01,   1.29552172e+01, -1.59255036e+01, -1.41953143e+01,
   1.12294437e+01,   3.33281363e+00,   5.05269237e+00, -6.40929171e+00,
   -6.69514427e+00, -3.81781754e-01,   1.01604014e+01, -2.36603017e+00,
   9.32509045e+00,   1.16040157e+00, -1.48942418e+00, -1.51676393e+01,
   9.76401254e-01, -9.89303407e+00,   2.34167936e-01, -5.16308489e+00,
   -6.38134354e-01,   5.48409809e+00,   6.56050194e+00,   3.39306828e+00,
   3.39915717e+00, -9.71743039e+00, -6.20375775e+00,   6.58188318e+00,
   -4.53384527e+00,  2.09661601e-02,   1.26087171e+01,   1.35596993e+01,
   -2.73694706e+00, -7.08322193e+00,   5.38412343e+00,   2.44290770e+00,
   -6.08236038e+00,   1.30556636e+01,   5.24386382e+00,   4.09747147e+00,
   -1.01341676e+01, -1.05853184e+01, -1.41015567e+00, -6.77998127e+00,
   5.14785781e+00, -4.80169690e+00,   1.41398402e+00, -3.57898247e+00,
   -1.57804065e+00,   4.78632246e+00,   2.03167027e+00,   9.96404504e-01,
   1.26263005e+01,   3.12129376e+00,   6.56443485e+00,   3.17462693e+00,
   4.10005159e+00, -1.13831824e+00, -6.64117443e+00, -2.95050759e+00,
   5.40916443e+00,   1.01870392e+00, -5.12234561e+00, -4.75792212e+00,
   5.51563580e+00, -1.11787330e+01, -1.04491170e+01,   3.23765381e+00,
   -5.87787987e+00,   1.48602681e+00, -4.78306951e+00, -1.22212749e+01,
   5.00356828e+00,   6.93450872e+00, -1.04394447e+01,   9.98709909e-01,
   1.37746955e+00, -4.24688940e+00,   4.18629021e+00,   7.05505037e+00,
   7.29305385e-01, -6.92209177e-01, -2.28652413e+00,   6.25329573e+00,
   3.03963553e+00, -9.51724006e+00, -3.83351510e+00, -3.32208676e+00,
   1.17019408e-02, -1.95405865e+00,   8.66653552e-01,   1.99515495e+00,
   5.69891311e+00, -6.24915423e+00, -2.42975528e+00, -3.68866852e-01,
   2.39165160e+00,   6.02045198e+00, -5.64869851e+00, -3.29499283e-01,
   8.58386805e+00, -5.12031167e+00,   4.52674127e+00,   5.51513999e+00,
   -8.78902025e-01, -3.14959549e+00,   1.05871413e+00,   8.67882482e+00,
   1.24566229e+01,   3.22265557e+00, -5.33286845e+00, -1.50496598e-01,
   -7.07353054e+00,   8.24229573e-13,   6.11434659e+00,   5.82285591e+00,
   6.05944355e+00, -9.17719519e-01, -2.63321128e+00,   4.48468835e+00,
   -4.55361061e+00,   5.99894531e+00,   4.00075634e+00, -2.01447082e+01,
   2.32442180e+00, -3.77286218e+00,   8.51494966e-01, -1.10664859e-01,
   -1.28728570e+01, -3.27556478e+00, -1.01984139e+01,   2.69385000e+00,
   4.35477362e+00, -4.01099114e+00,   1.82097273e+00, -5.63850273e+00,
   3.31601326e+00,   2.70136815e+00,   3.75321132e+00, -1.55046256e+00,
   4.61516142e+00,   5.53381643e+00, -3.46372550e+00, -9.96225176e+00,
   4.04835946e+00,   3.61459631e+00,   5.36095257e+00,   5.29857155e+00,
   -3.67613801e+00, -1.89537214e-07, -7.22960068e+00,   3.01355678e+00,
```

```
7.11759563e+00, 2.29954940e+00, -7.74457607e+00, -7.30631570e-01,
7.76850750e+00, -5.65020584e-01, 3.38159937e+00, 6.07821567e+00,
6.04426543e+00, -4.10937147e+00, -8.99853937e+00, -8.87538376e+00,
5.35379703e+00, -5.35377051e+00])
```

```
1 pd.DataFrame(model.coef_, X.columns, columns=['Coedicients'])
```

	Coedicients
Year	0.263644
Adult Mortality	-0.002401
infant deaths	0.108407
Alcohol	-0.078341
percentage expenditure	0.000131
...	...
Country_Yemen	-4.109371
Country_Zambia	-8.998539
Country_Zimbabwe	-8.875384
Status_Developed	5.353797
Status_Developing	-5.353771

214 rows × 1 columns

▼ Prediction from our Model

```
1 y_pred = model.predict(X_test)
```

▼ Registration Evaluation Metrics

- **Mean Absolute Error** (MAE) is the mean of the absolute value of the errors:

$$\frac{1}{n} \sum_{i=1}^n |y_i - \hat{y}_i|$$

- **Mean Squared Error** (MSE) is the mean of the squared errors:

$$\frac{1}{n} \sum_{i=1}^n (y_i - \hat{y}_i)^2$$

```
1 MAE = metrics.mean_absolute_error(y_test, y_pred)
2 MSE = metrics.mean_squared_error(y_test, y_pred)
3 RMSE = np.sqrt(MSE)
```

1 MAE

1.2139823762077666

1 MSE

3.68331185459279

1 RMSE

1.9191954526465715

```
1 df['Life expectancy'].mean()
```

69.22493169398906

⌄ Residual Histogram

```
1 test_residual = y_test - y_pred
```