In [2]: import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns

In [75]: a=pd.read_csv(r"C:\Users\user\Downloads\2_2015.csv")
a

Out[75]:

	Country	Region	Happiness Rank	Happiness Score	Standard Error	(GDP per Capita)	Family	Health (Life Expectancy)	Freedom	((
0	Switzerland	Western Europe	1	7.587	0.03411	1.39651	1.34951	0.94143	0.66557	
1	Iceland	Western Europe	2	7.561	0.04884	1.30232	1.40223	0.94784	0.62877	
2	Denmark	Western Europe	3	7.527	0.03328	1.32548	1.36058	0.87464	0.64938	
3	Norway	Western Europe	4	7.522	0.03880	1.45900	1.33095	0.88521	0.66973	
4	Canada	North America	5	7.427	0.03553	1.32629	1.32261	0.90563	0.63297	
		Sub-								_

In [76]: a=a.head(50)

Out[76]:

	Country	Region	Happiness Rank	Happiness Score	Standard Error	Economy (GDP per Capita)	Family	Health (Life Expectancy)	Freedom (
0	Switzerland	Western Europe	1	7.587	0.03411	1.39651	1.34951	0.94143	0.66557
1	Iceland	Western Europe	2	7.561	0.04884	1.30232	1.40223	0.94784	0.62877
2	Denmark	Western Europe	3	7.527	0.03328	1.32548	1.36058	0.87464	0.64938
3	Norway	Western Europe	4	7.522	0.03880	1.45900	1.33095	0.88521	0.66973
4	Canada	North America	5	7.427	0.03553	1.32629	1.32261	0.90563	0.63297
5	Finland	Western Europe	6	7.406	0.03140	1.29025	1.31826	0.88911	0.64169
6	Netherlands	Western Europe	7	7.378	0.02799	1.32944	1.28017	0.89284	0.61576
7	Sweden	Western Europe	8	7.364	0.03157	1.33171	1.28907	0.91087	0.65980
8	New Zea l and	Australia and New Zealand	9	7.286	0.03371	1.25018	1.31967	0.90837	0.63938
9	Australia	Australia and New Zealand	10	7.284	0.04083	1.33358	1.30923	0.93156	0.65124
10	Israel	Middle East and Northern Africa	11	7.278	0.03470	1.22857	1.22393	0.91387	0.41319
11	Costa Rica	Latin America and Caribbean	12	7.226	0.04454	0.95578	1.23788	0.86027	0.63376
12	Austria	Western Europe	13	7.200	0.03751	1.33723	1.29704	0.89042	0.62433
13	Mexico	Latin America and Caribbean	14	7.187	0.04176	1.02054	0.91451	0.81444	0.48181
14	United States	North America	15	7.119	0.03839	1.39451	1.24711	0.86179	0.54604
15	Brazil	Latin America and Caribbean	16	6.983	0.04076	0.98124	1.23287	0.69702	0.49049
16	Luxembourg	Western Europe	17	6.946	0.03499	1.56391	1.21963	0.91894	0.61583
17	Ireland	Western Europe	18	6.940	0.03676	1.33596	1.36948	0.89533	0.61777
18	Belgium	Western Europe	19	6.937	0.03595	1.30782	1.28566	0.89667	0.58450
19	United Arab Emirates	Middle East and Northern Africa	20	6.901	0.03729	1.42727	1.12575	0.80925	0.64157
20	United Kingdom	Western Europe	21	6.867	0.01866	1.26637	1.28548	0.90943	0.59625
21	Oman	Middle East and Northern Africa	22	6.853	0.05335	1.36011	1.08182	0.76276	0.63274

	Country	Region	Happiness Rank	Happiness Score	Standard Error	Economy (GDP per Capita)	Family	Health (Life Expectancy)	Freedom
22	Venezuela	Latin America and Caribbean	23	6.810	0.06476	1.04424	1.25596	0.72052	0.42908
23	Singapore	Southeastern Asia	24	6.798	0.03780	1.52186	1.02000	1.02525	0.54252
24	Panama	Latin America and Caribbean	25	6.786	0.04910	1.06353	1.19850	0.79661	0.54210
25	Germany	Western Europe	26	6.750	0.01848	1.32792	1.29937	0.89186	0.61477
26	Chile	Latin America and Caribbean	27	6.670	0.05800	1.10715	1.12447	0.85857	0.44132
27	Qatar	Middle East and Northern Africa	28	6.611	0.06257	1.69042	1.07860	0.79733	0.64040
28	France	Western Europe	29	6.575	0.03512	1.27778	1.26038	0.94579	0.55011
29	Argentina	Latin America and Caribbean	30	6.574	0.04612	1.05351	1.24823	0.78723	0.44974
30	Czech Repub l ic	Central and Eastern Europe	31	6.505	0.04168	1.17898	1.20643	0.84483	0.46364
31	Uruguay	Latin America and Caribbean	32	6.485	0.04539	1.06166	1.20890	0.81160	0.60362
32	Colombia	Latin America and Caribbean	33	6.477	0.05051	0.91861	1.24018	0.69077	0.53466
33	Thai l and	Southeastern Asia	34	6.455	0.03557	0.96690	1.26504	0.73850	0.55664
34	Saudi Arabia	Middle East and Northern Africa	35	6.411	0.04633	1.39541	1.08393	0.72025	0.31048
35	Spain	Western Europe	36	6.329	0.03468	1.23011	1.31379	0.95562	0.45951
36	Malta	Western Europe	37	6.302	0.04206	1.20740	1.30203	0.88721	0.60365
37	Taiwan	Eastern Asia	38	6.298	0.03868	1.29098	1.07617	0.87530	0.39740
38	Kuwait	Middle East and Northern Africa	39	6.295	0.04456	1.55422	1.16594	0.72492	0.55499
39	Suriname	Latin America and Caribbean	40	6.269	0.09811	0.99534	0.97200	0.60820	0.59657
40	Trinidad and Tobago	Latin America and Caribbean	41	6.168	0.10895	1.21183	1.18354	0.61483	0.55884
41	El Salvador	Latin America and Caribbean	42	6.130	0.05618	0.76454	1.02507	0.67737	0.40350

	Country	Region	Happiness Rank	Happiness Score	Standard Error	Economy (GDP per Capita)	Family	Health (Life Expectancy)	Freedom	(
42	Guatemala	Latin America and Caribbean	43	6.123	0.05224	0.74553	1.04356	0.64425	0.57733	
43	Uzbekistan	Central and Eastern Europe	44	6.003	0.04361	0.63244	1.34043	0.59772	0.65821	
44	Slovakia	Central and Eastern Europe	45	5.995	0.04267	1.16891	1.26999	0.78902	0.31751	
45	Japan	Eastern Asia	46	5.987	0.03581	1.27074	1.25712	0.99111	0.49615	
46	South Korea	Eastern Asia	47	5.984	0.04098	1.24461	0.95774	0.96538	0.33208	
47	Ecuador	Latin America and Caribbean	48	5.975	0.04528	0.86402	0.99903	0.79075	0.48574	
48	Bahrain	Middle East and Northern Africa	49	5.960	0.05412	1.32376	1.21624	0.74716	0.45492	
49	Italy	Western Europe	50	5.948	0.03914	1.25114	1.19777	0.95446	0.26236	

In [77]: a.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 50 entries, 0 to 49
Data columns (total 12 columns):

#	Column	Non-Null Count	Dtype			
0	Country	50 non-null	object			
1	Region	50 non-null	object			
2	Happiness Rank	50 non-null	int64			
3	Happiness Score	50 non-null	float64			
4	Standard Error	50 non-null	float64			
5	Economy (GDP per Capita)	50 non-null	float64			
6	Family	50 non-null	float64			
7	Health (Life Expectancy)	50 non-null	float64			
8	Freedom	50 non-null	float64			
9	Trust (Government Corruption)	50 non-null	float64			
10	Generosity	50 non-null	float64			
11	Dystopia Residual	50 non-null	float64			
<pre>dtypes: float64(9), int64(1), object(2)</pre>						

In [78]: a.columns

memory usage: 4.8+ KB

In [79]: a.head()

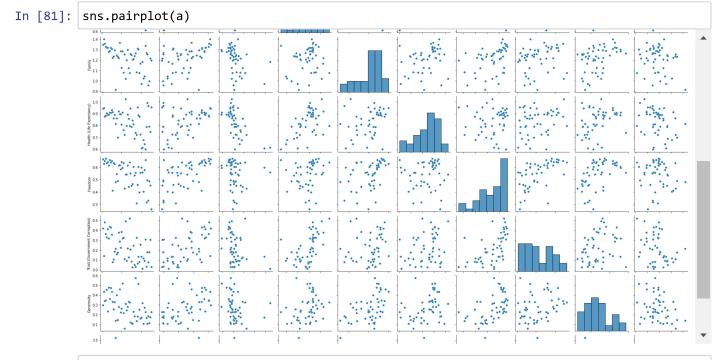
Out[79]:

	Country	Region	Happiness Rank	Happiness Score	Standard Error	Economy (GDP per Capita)	Family	Health (Life Expectancy)	Freedom	(Goverı Corru
0	Switzerland	Western Europe	1	7.587	0.03411	1.39651	1.34951	0.94143	0.66557	0.
1	Iceland	Western Europe	2	7.561	0.04884	1.30232	1.40223	0.94784	0.62877	0.
2	Denmark	Western Europe	3	7.527	0.03328	1.32548	1.36058	0.87464	0.64938	0.
3	Norway	Western Europe	4	7.522	0.03880	1.45900	1.33095	0.88521	0.66973	0.
4	Canada	North America	5	7.427	0.03553	1.32629	1.32261	0.90563	0.63297	0.
4										•

In [80]: a.describe()

Out[80]:

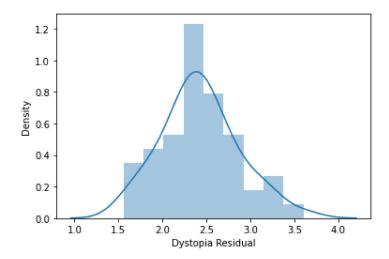
	Happiness Rank	Happiness Score	Standard Error	Economy (GDP per Capita)	Family	Health (Life Expectancy)	Freedom	Trust (Government Corruption)	Gene
count	50.00000	50.000000	50.000000	50.000000	50.000000	50.000000	50.000000	50.000000	50.00
mean	25.50000	6.729040	0.043584	1.217752	1.212277	0.835402	0.543408	0.221732	0.2
std	14.57738	0.511449	0.015482	0.216687	0.118959	0.105584	0.105589	0.142510	0.12
min	1.00000	5.948000	0.018480	0.632440	0.914510	0.597720	0.262360	0.011400	0.0
25%	13.25000	6.299000	0.035540	1.062128	1.135798	0.768877	0.468183	0.097498	0.17
50%	25.50000	6.768000	0.040795	1.268555	1.243645	0.868215	0.568085	0.196610	0.2
75%	37.75000	7.196750	0.046277	1.333112	1.298787	0.909165	0.631748	0.325240	0.30
max	50.00000	7.587000	0.108950	1.690420	1.402230	1.025250	0.669730	0.522080	0.5
4									•



In [82]: sns.distplot(a['Dystopia Residual'])

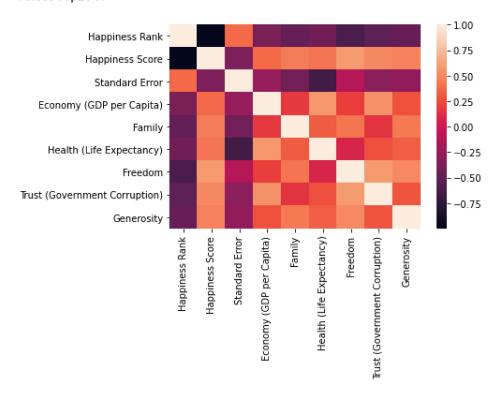
C:\ProgramData\Anaconda3\lib\site-packages\seaborn\distributions.py:2557: FutureWarnin
g: `distplot` is a deprecated function and will be removed in a future version. Please
adapt your code to use either `displot` (a figure-level function with similar flexibil
ity) or `histplot` (an axes-level function for histograms).
 warnings.warn(msg, FutureWarning)

Out[82]: <AxesSubplot:xlabel='Dystopia Residual', ylabel='Density'>



```
In [84]: sns.heatmap(x1.corr())
```

Out[84]: <AxesSubplot:>



```
In [87]: from sklearn.model_selection import train_test_split
    x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.3)
```

```
In [88]: from sklearn.linear_model import LinearRegression
lr=LinearRegression()
lr.fit(x_train,y_train)
```

Out[88]: LinearRegression()

```
In [89]: print(lr.intercept_)
```

0.004859264853705358

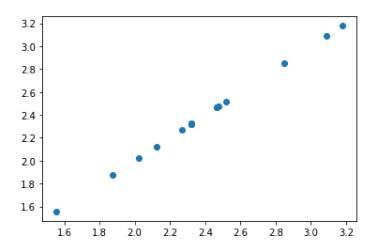
```
In [90]: coeff=pd.DataFrame(lr.coef_,x.columns,columns=['Co-efficient'])
coeff
```

```
Out[90]:
```

	Co-efficient
Happiness Rank	- 0.000021
Happiness Score	0.999387
Standard Error	- 0.002848
Economy (GDP per Capita)	-0.999397
Family	- 1.000531
Health (Life Expectancy)	- 1.000615
Freedom	-0.998902
Trust (Government Corruption)	- 1.000772
Generosity	- 1.000018

```
In [91]: prediction=lr.predict(x_test)
plt.scatter(y_test,prediction)
```

Out[91]: <matplotlib.collections.PathCollection at 0x22eb35b1220>



```
In [92]: print(lr.score(x_test,y_test))
```

0.999999523602235

```
In [93]: from sklearn.linear_model import Ridge,Lasso
```

```
In [94]: rr=Ridge(alpha=10)
rr.fit(x_train,y_train)
```

Out[94]: Ridge(alpha=10)

```
In [95]: rr.score(x_test,y_test)
```

Out[95]: 0.37967082440574806

```
In [96]: la=Lasso(alpha=10)
          la.fit(x_train,y_train)
Out[96]: Lasso(alpha=10)
In [97]: la.score(x test,y test)
Out[97]: -0.015015955172610562
In [98]: from sklearn.linear_model import ElasticNet
          en=ElasticNet()
          en.fit(x_train,y_train)
Out[98]: ElasticNet()
In [99]: |print(en.coef_)
          [-0.00351865 0.
                                     0.
                                                -0.
                                                            -0.
                                                                        -0.
           -0.
                       -0.
                                    -0.
                                               1
In [100]: print(en.intercept_)
          2.530602103288059
In [101]: print(en.predict(x_test))
          [2.50597154 2.45319176 2.39689332 2.48837828 2.38633737 2.51652749
           2.52708345 2.40393063 2.41800524 2.3546695 2.41096793 2.43207984
           2.4426358 2.49189693 2.39337467]
In [102]: |en.score(x_test,y_test)
Out[102]: 0.11248425296623765
In [103]: from sklearn import metrics
In [104]: | print("Mean Absolute Error", metrics.mean_absolute_error(y_test, prediction))
          Mean Absolute Error 0.00022193162481184045
In [105]: print("Mean Squared Error", metrics.mean squared error(y test, prediction))
          Mean Squared Error 8.079143874781493e-08
In [106]: print(" Root Mean Squared Error",np.sqrt(metrics.mean_squared_error(y_test,prediction))
           Root Mean Squared Error 0.0002842383484820705
```