# **Problem Statement**

### In [1]:

```
# import libraies
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
```

# In [40]:

```
d=pd.read_csv(r"C:\Users\user\Downloads\2015.csv")
d
```

# Out[40]:

	Country	Region	Happiness Rank	Happiness Score	Standard Error	Economy (GDP per Capita)	Family	Health (Life Expectancy)
0	Switzerland	Western Europe	1	7.587	0.03411	1.39651	1.34951	0.94143
1	Iceland	Western Europe	2	7.561	0.04884	1.30232	1.40223	0.94784
2	Denmark	Western Europe	3	7.527	0.03328	1.32548	1.36058	0.87464
3	Norway	Western Europe	4	7.522	0.03880	1.45900	1.33095	0.88521
4	Canada	North America	5	7.427	0.03553	1.32629	1.32261	0.90563
153	Rwanda	Sub- Saharan Africa	154	3.465	0.03464	0.22208	0.77370	0.42864
154	Benin	Sub- Saharan Africa	155	3.340	0.03656	0.28665	0.35386	0.31910
155	Syria	Middle East and Northern Africa	156	3.006	0.05015	0.66320	0.47489	0.72193
156	Burundi	Sub- Saharan Africa	157	2.905	0.08658	0.01530	0.41587	0.22396
157	Togo	Sub- Saharan Africa	158	2.839	0.06727	0.20868	0.13995	0.28443
158 rows × 12 columns								

# In [41]:

d.head(10)

# Out[41]:

	Country	Region	Happiness Rank	Happiness Score	Standard Error	Economy (GDP per Capita)	Family	Health (Life Expectancy)	F
0	Switzerland	Western Europe	1	7.587	0.03411	1.39651	1.34951	0.94143	
1	Iceland	Western Europe	2	7.561	0.04884	1.30232	1.40223	0.94784	
2	Denmark	Western Europe	3	7.527	0.03328	1.32548	1.36058	0.87464	
3	Norway	Western Europe	4	7.522	0.03880	1.45900	1.33095	0.88521	
4	Canada	North America	5	7.427	0.03553	1.32629	1.32261	0.90563	
5	Finland	Western Europe	6	7.406	0.03140	1.29025	1.31826	0.88911	
6	Netherlands	Western Europe	7	7.378	0.02799	1.32944	1.28017	0.89284	
7	Sweden	Western Europe	8	7.364	0.03157	1.33171	1.28907	0.91087	
8	New Zealand	Australia and New Zealand	9	7.286	0.03371	1.25018	1.31967	0.90837	
9	Australia	Australia and New Zealand	10	7.284	0.04083	1.33358	1.30923	0.93156	
4									•

### In [42]:

#### d.describe()

#### Out[42]:

	Happiness Rank	Happiness Score	Standard Error	Economy (GDP per Capita)	Family	Health (Life Expectancy)	Freedom
coun	158.000000	158.000000	158.000000	158.000000	158.000000	158.000000	158.000000
mear	79.493671	5.375734	0.047885	0.846137	0.991046	0.630259	0.428615
sto	45.754363	1.145010	0.017146	0.403121	0.272369	0.247078	0.150693
mir	1.000000	2.839000	0.018480	0.000000	0.000000	0.000000	0.000000
25%	40.250000	4.526000	0.037268	0.545808	0.856823	0.439185	0.328330
50%	79.500000	5.232500	0.043940	0.910245	1.029510	0.696705	0.435515
75%	118.750000	6.243750	0.052300	1.158448	1.214405	0.811013	0.549092
max	158.000000	7.587000	0.136930	1.690420	1.402230	1.025250	0.669730
4							<b>)</b>

### In [43]:

#### d.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 158 entries, 0 to 157
Data columns (total 12 columns):

#	Column	Non-Null Count	Dtype
0	Country	158 non-null	object
1	Region	158 non-null	object
2	Happiness Rank	158 non-null	int64
3	Happiness Score	158 non-null	float64
4	Standard Error	158 non-null	float64
5	Economy (GDP per Capita)	158 non-null	float64
6	Family	158 non-null	float64
7	Health (Life Expectancy)	158 non-null	float64
8	Freedom	158 non-null	float64
9	Trust (Government Corruption)	158 non-null	float64
10	Generosity	158 non-null	float64
11	Dystopia Residual	158 non-null	float64
	(1 164/0) 1164/4) 11	1 (2)	

dtypes: float64(9), int64(1), object(2)

memory usage: 14.9+ KB

#### In [44]:

#### d.columns

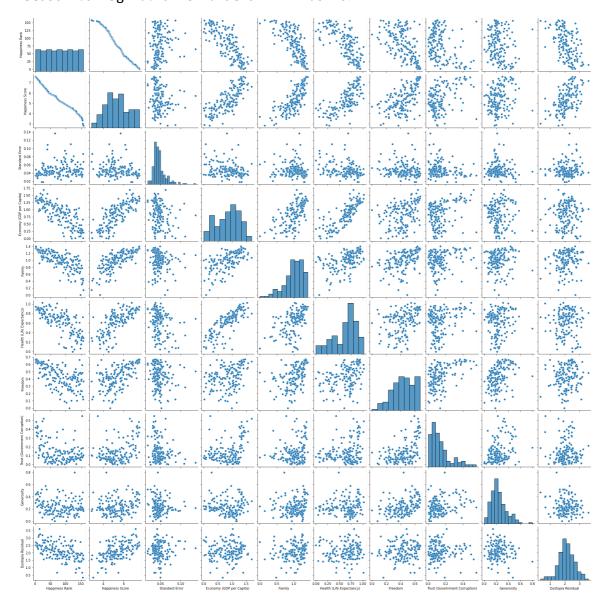
#### Out[44]:

# In [45]:

sns.pairplot(d)

# Out[45]:

<seaborn.axisgrid.PairGrid at 0x171924db4f0>



#### In [46]:

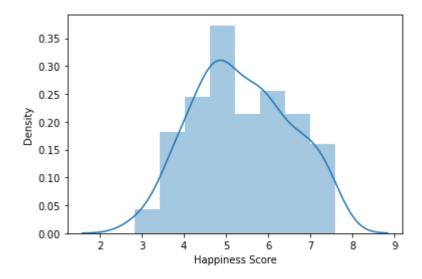
```
sns.distplot(d['Happiness Score'])
```

C:\ProgramData\Anaconda3\lib\site-packages\seaborn\distributions.py:2557: FutureWarning: `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 flexibility) or `histplot` (an axes-level function for histograms).

warnings.warn(msg, FutureWarning)

#### Out[46]:

<AxesSubplot:xlabel='Happiness Score', ylabel='Density'>



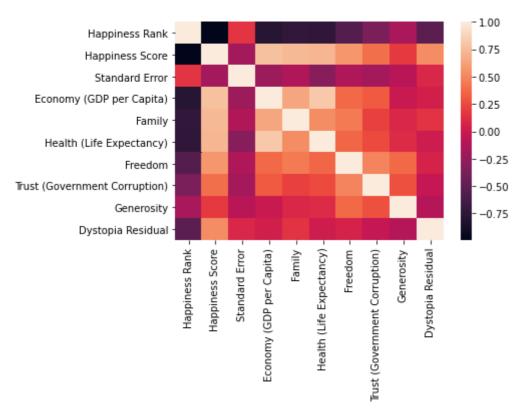
#### In [47]:

#### In [48]:

```
# relation
sns.heatmap(da.corr())
```

#### Out[48]:

#### <AxesSubplot:>



# to train the model

we are going to train linear regresssion model; we need to split out data into two values varible x and y where x is independent(input) and y is dependent on x (output) we could ignore adrees column as it not requird foer model

#### In [55]:

```
x=da[['Happiness Rank', 'Happiness Score']]
y=da['Family']
```

# In [56]:

```
# to split my dataset into test and train data
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 [57]:

```
from sklearn.linear_model import LinearRegression
lr=LinearRegression()
lr.fit(x_train,y_train)
```

#### Out[57]:

LinearRegression()

### In [58]:

```
print(lr.intercept_)
```

#### 1.641269439798164

### In [59]:

```
coeff=pd.DataFrame(lr.coef_,x.columns,columns=['Co-effecient'])
coeff
```

#### Out[59]:

#### Co-effecient

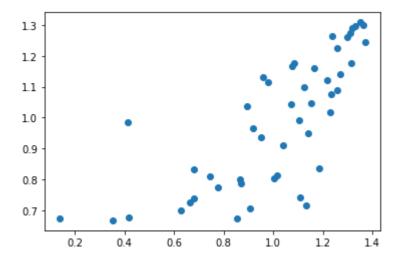
Happiness Rank -0.005355
Happiness Score -0.043140

#### In [60]:

```
prediction=lr.predict(x_test)
plt.scatter(y_test,prediction)
```

#### Out[60]:

<matplotlib.collections.PathCollection at 0x171970c0d00>



#### In [61]:

```
print(lr.score(x_test,y_test))
```

#### 0.5625425945166421

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