

# Problem statement

## Data collection

## Importing libraries

```
In [1]: import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
```

## Importing dataset

```
In [2]: data=pd.read_csv(r"C:\Users\user\Downloads\VE.CSV.csv")
data
```

Out[2]:

	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
...	...	...	...	...	...	...	...	...	...
153	Rwanda	Sub-Saharan Africa	154	3.465	0.03464	0.22208	0.77370	0.42864	0.59201
154	Benin	Sub-Saharan Africa	155	3.340	0.03656	0.28665	0.35386	0.31910	0.48450
155	Syria	Middle East and Northern Africa	156	3.006	0.05015	0.66320	0.47489	0.72193	0.15684
156	Burundi	Sub-Saharan	157	2.905	0.08658	0.01530	0.41587	0.22396	0.11850

	Country	Region	Happiness Rank	Happiness Score	Standard Error	Economy (GDP per Capita)	Family	Health (Life Expectancy)	Freedom
		Africa							
157	Togo	Sub-Saharan Africa	158	2.839	0.06727	0.20868	0.13995	0.28443	0.36453

158 rows × 12 columns

## head

```
In [3]: # to display first 8 dataset values
da=data.head(8)
da
```

```
Out[3]:
```

	Country	Region	Happiness Rank	Happiness Score	Standard Error	Economy (GDP per Capita)	Family	Health (Life Expectancy)	Freedom	(G	C
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		

## info

```
In [4]: # to identify missing values
data.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
dtypes: float64(9), int64(1), object(2)
memory usage: 14.9+ KB

```

## describe

```
In [5]: # to display summary of the dataset
data.describe()
```

```
Out[5]:
```

	Happiness Rank	Happiness Score	Standard Error	Economy (GDP per Capita)	Family	Health (Life Expectancy)	Freedom	Trust (Government Corruption)
<b>count</b>	158.000000	158.000000	158.000000	158.000000	158.000000	158.000000	158.000000	158.000000
<b>mean</b>	79.493671	5.375734	0.047885	0.846137	0.991046	0.630259	0.428615	0.14342
<b>std</b>	45.754363	1.145010	0.017146	0.403121	0.272369	0.247078	0.150693	0.12003
<b>min</b>	1.000000	2.839000	0.018480	0.000000	0.000000	0.000000	0.000000	0.00000
<b>25%</b>	40.250000	4.526000	0.037268	0.545808	0.856823	0.439185	0.328330	0.06167
<b>50%</b>	79.500000	5.232500	0.043940	0.910245	1.029510	0.696705	0.435515	0.10722
<b>75%</b>	118.750000	6.243750	0.052300	1.158448	1.214405	0.811013	0.549092	0.18025
<b>max</b>	158.000000	7.587000	0.136930	1.690420	1.402230	1.025250	0.669730	0.55191

## columns

```
In [6]: # to display headings of the dataset
data.columns
```

```
Out[6]: Index(['Country', 'Region', 'Happiness Rank', 'Happiness Score',
              'Standard Error', 'Economy (GDP per Capita)', 'Family',
              'Health (Life Expectancy)', 'Freedom', 'Trust (Government Corruption)',
              'Generosity', 'Dystopia Residual'],
              dtype='object')
```

```
In [7]: a=data.dropna(axis=1)
a
```

Out[7]:

	Country	Region	Happiness Rank	Happiness Score	Standard Error	Economy (GDP per Capita)	Family	Health (Life Expectancy)	Freedom
<b>0</b>	Switzerland	Western Europe	1	7.587	0.03411	1.39651	1.34951	0.94143	0.66557
<b>1</b>	Iceland	Western Europe	2	7.561	0.04884	1.30232	1.40223	0.94784	0.62877
<b>2</b>	Denmark	Western Europe	3	7.527	0.03328	1.32548	1.36058	0.87464	0.64938
<b>3</b>	Norway	Western Europe	4	7.522	0.03880	1.45900	1.33095	0.88521	0.66973
<b>4</b>	Canada	North America	5	7.427	0.03553	1.32629	1.32261	0.90563	0.63297
...	...	...	...	...	...	...	...	...	...
<b>153</b>	Rwanda	Sub-Saharan Africa	154	3.465	0.03464	0.22208	0.77370	0.42864	0.59201
<b>154</b>	Benin	Sub-Saharan Africa	155	3.340	0.03656	0.28665	0.35386	0.31910	0.48450
<b>155</b>	Syria	Middle East and Northern Africa	156	3.006	0.05015	0.66320	0.47489	0.72193	0.15684
<b>156</b>	Burundi	Sub-Saharan Africa	157	2.905	0.08658	0.01530	0.41587	0.22396	0.11850
<b>157</b>	Togo	Sub-Saharan Africa	158	2.839	0.06727	0.20868	0.13995	0.28443	0.36453

158 rows × 12 columns



In [8]:

a.columns

Out[8]:

```
Index(['Country', 'Region', 'Happiness Rank', 'Happiness Score',
      'Standard Error', 'Economy (GDP per Capita)', 'Family',
      'Health (Life Expectancy)', 'Freedom', 'Trust (Government Corruption)',
      'Generosity', 'Dystopia Residual'],
      dtype='object')
```

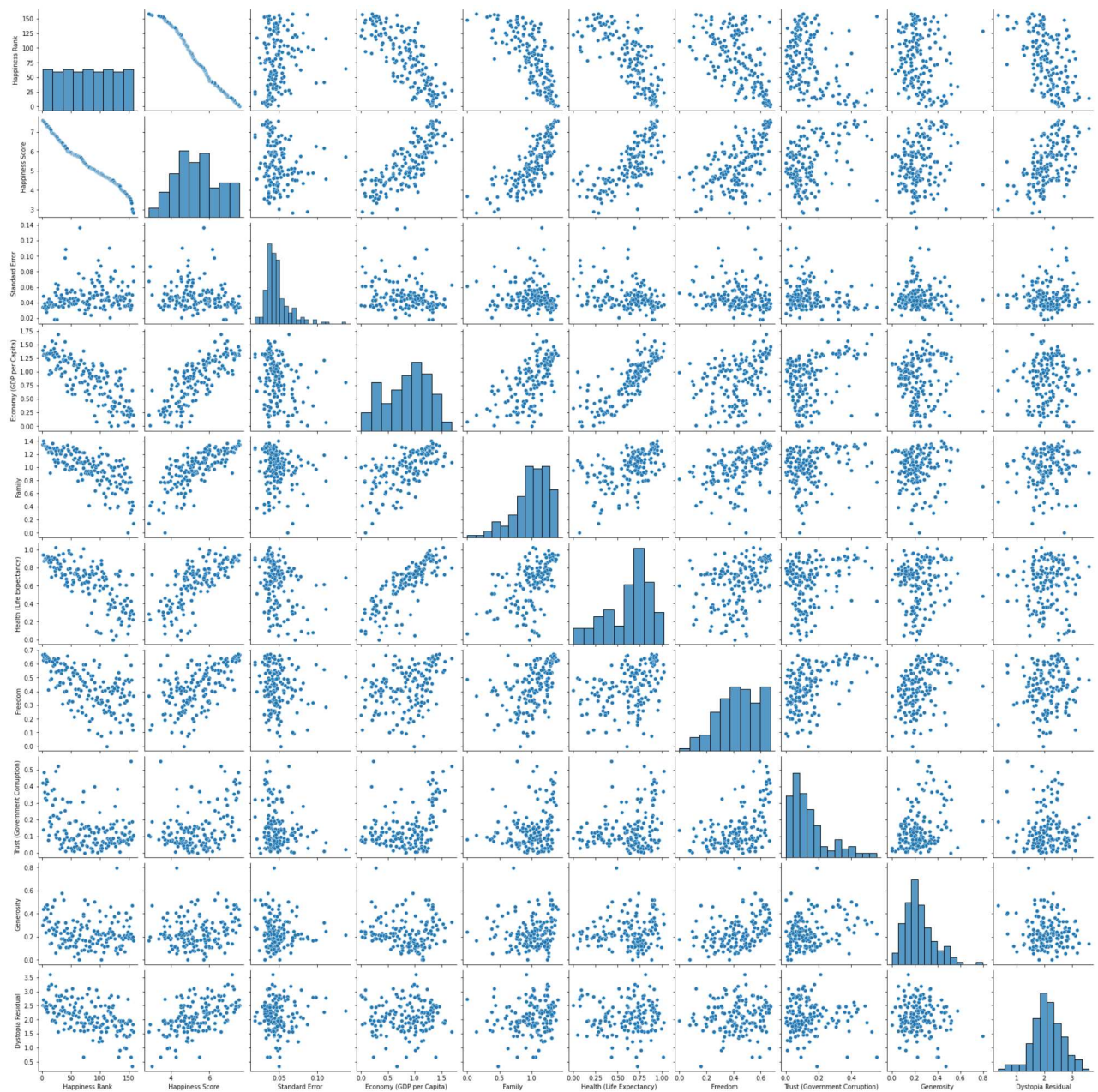
## EDA and Visualization

In [9]:

sns.pairplot(a)

Out[9]:

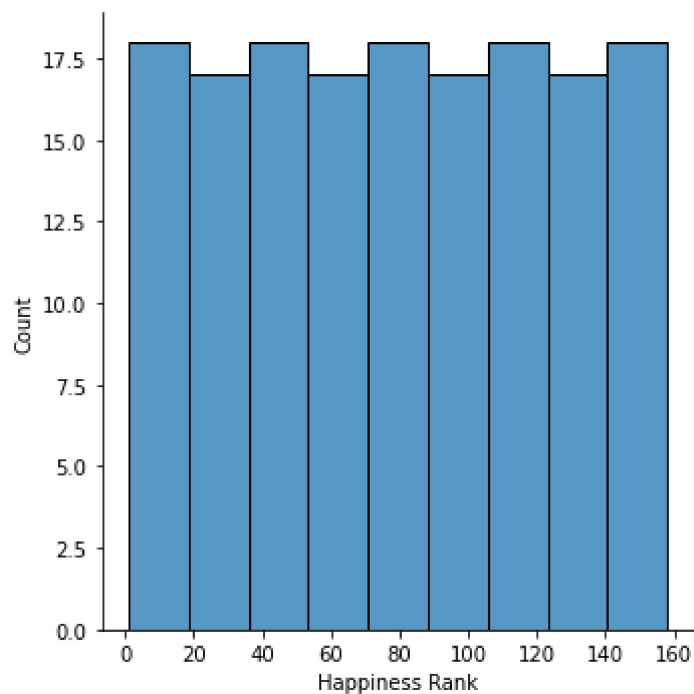
&lt;seaborn.axisgrid.PairGrid at 0x1dab334d6a0&gt;



## distribution plot

```
In [11]: sns.displot(a["Happiness Rank"])
```

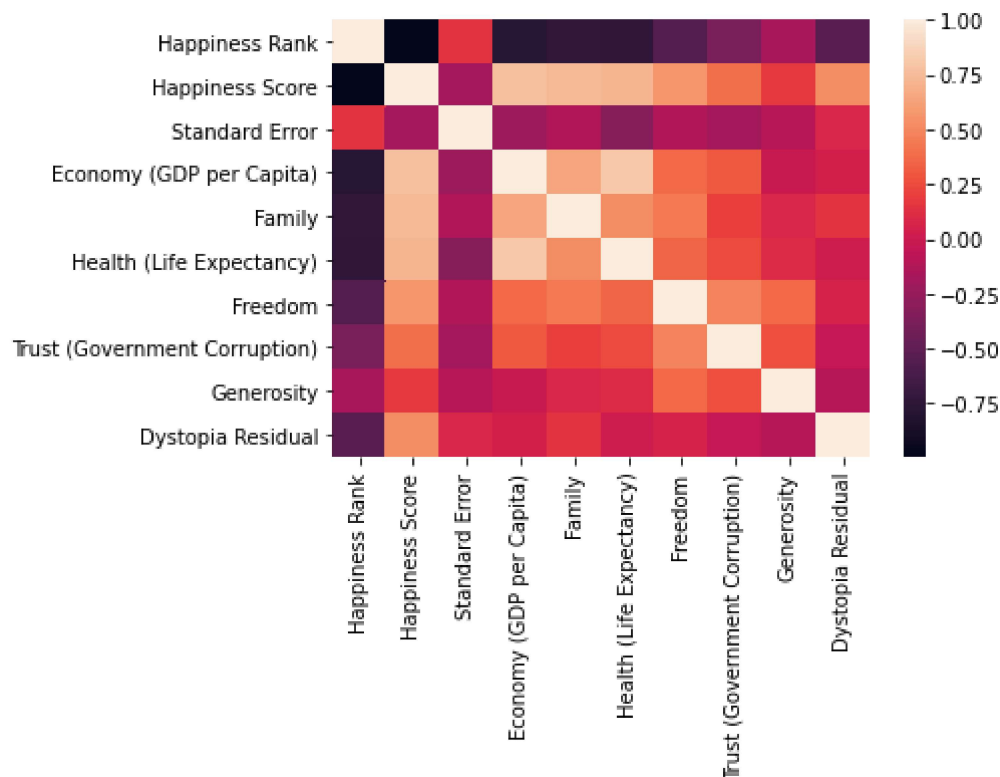
```
Out[11]: <seaborn.axisgrid.FacetGrid at 0x1dabc146a60>
```



## correlation

```
In [12]: dat=data[['Country', 'Region', 'Happiness Rank', 'Happiness Score',
                  'Standard Error', 'Economy (GDP per Capita)', 'Family',
                  'Health (Life Expectancy)', 'Freedom', 'Trust (Government Corruption)',
                  'Generosity', 'Dystopia Residual']]
sns.heatmap(dat.corr())
```

Out[12]: <AxesSubplot:>





# To train the model-Model Building

```
In [17]: x=a[['Happiness Rank']]
         y=a['Happiness Rank']
```

```
In [18]: # to split my dataset into training and test 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 [21]: from sklearn.linear_model import LinearRegression
         lr= LinearRegression()
         lr.fit(x_train,y_train)
```

Out[21]: LinearRegression()

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

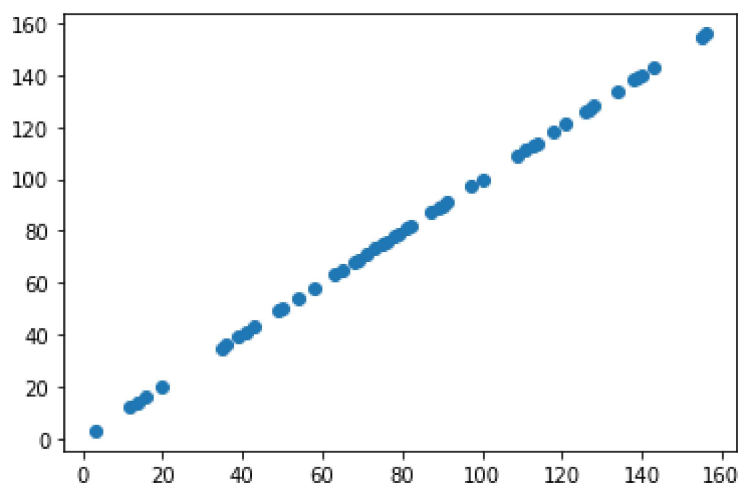
```
In [22]: coeff=pd.DataFrame(lr.coef_,x.columns,columns=['Co-efficient'])
         coeff
```

Out[22]:

	Co-efficient
Happiness Rank	1.0

```
In [24]: prediction=lr.predict(x_test)
         plt.scatter(y_test,prediction)
```

Out[24]: <matplotlib.collections.PathCollection at 0x1dabdec2430>



```
In [23]: print(lr.score(x_test,y_test))
```

1.0