

# 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	(GDP per Capita)
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()
```

	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
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 Africa	157	2.905	0.08658	0.01530	0.41587	0.22396	0.11850
157	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

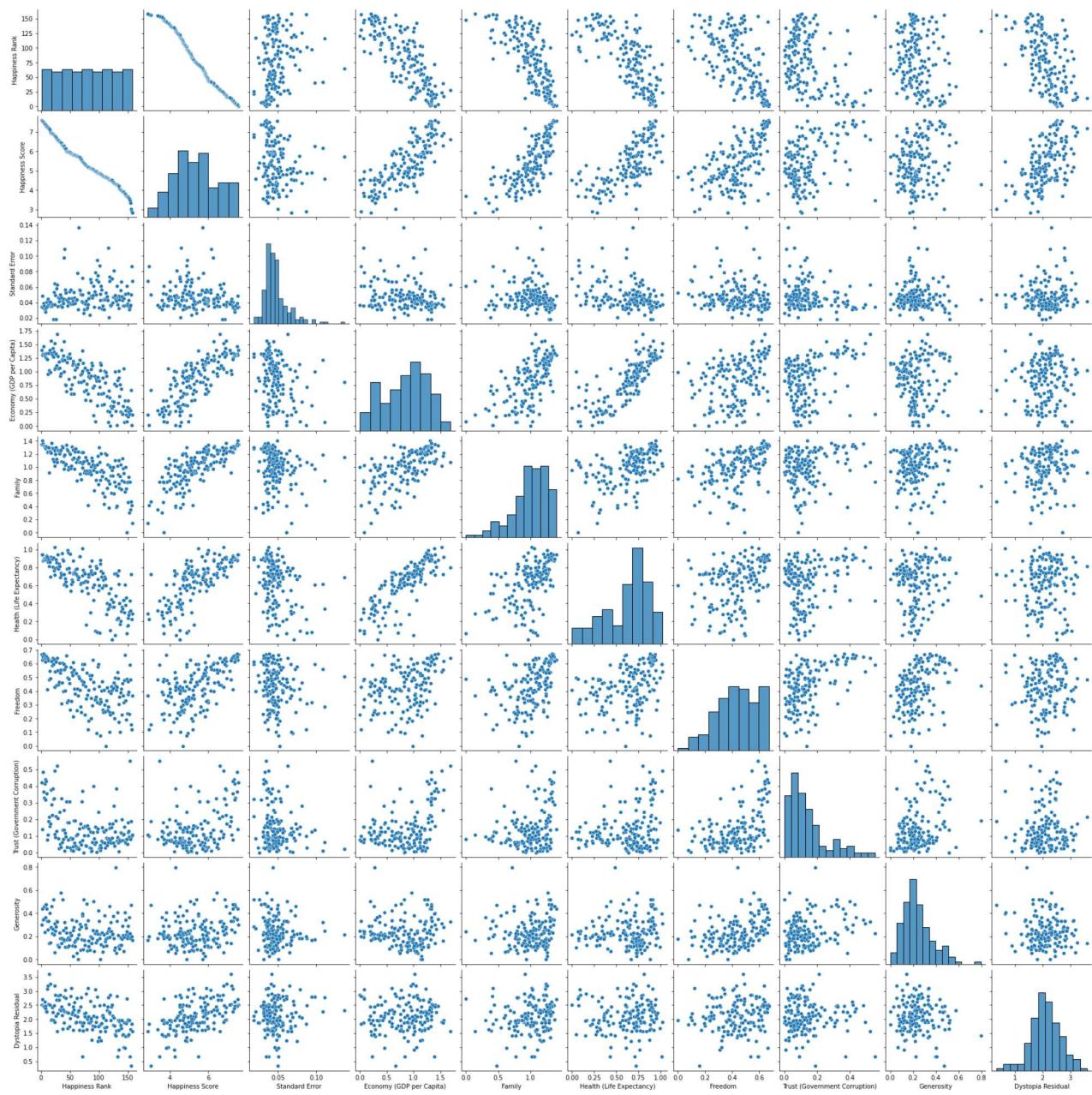
```
Out[8]: Index(['Country', 'Region', 'Happiness Rank', 'Happiness Score',
       'Standard Error', 'Economy (GDP per Capita)', 'Family',
       'Health (Life Expectancy)', 'Freedom', 'Trust (Government Corruption)',
       'Generosity', 'Dystopia Residual'],
      dtype='object')
```

## EDA and Visualization

In [9]:

sns.pairplot(a)

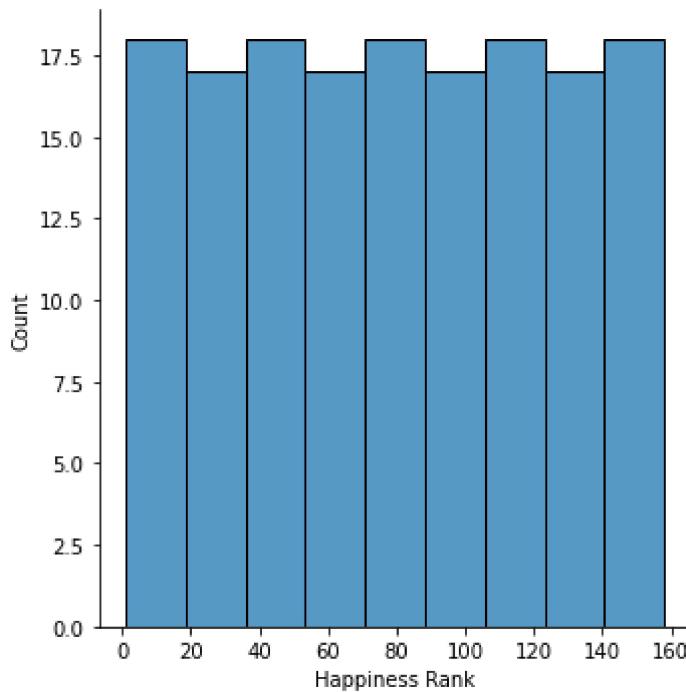
```
Out[9]: <seaborn.axisgrid.PairGrid at 0x25e54418b20>
```



## distribution plot

```
In [10]: sns.displot(a["Happiness Rank"])
```

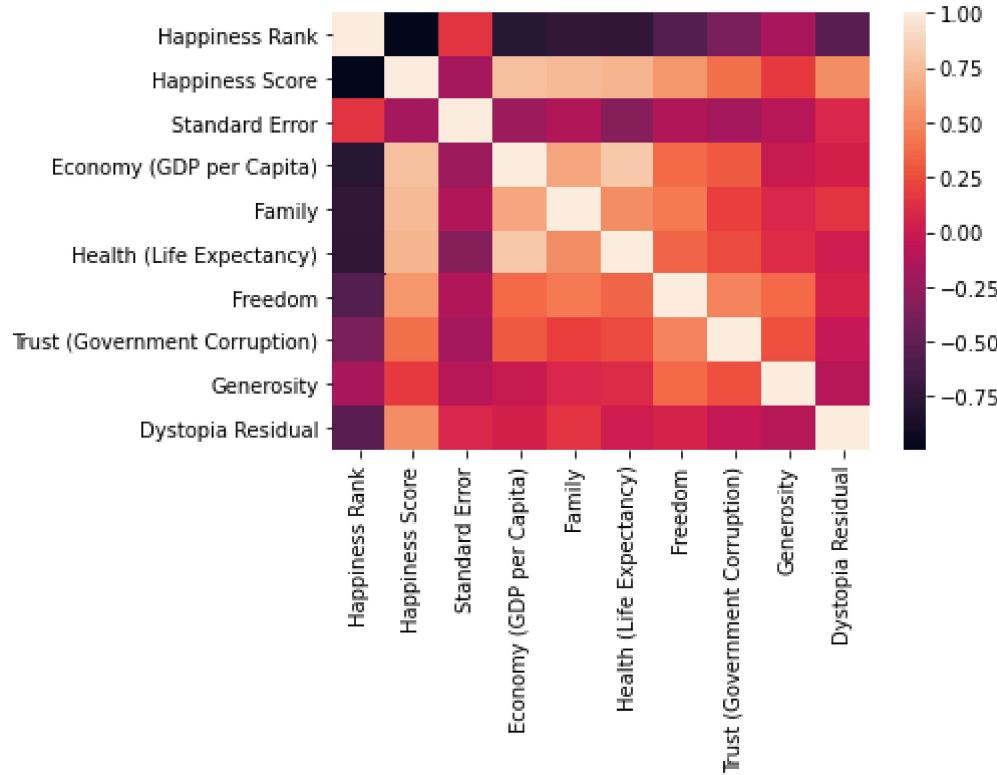
```
Out[10]: <seaborn.axisgrid.FacetGrid at 0x25e5992fa00>
```



## correlation

```
In [11]: 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[11]: <AxesSubplot:>



# To train the model-Model Building

```
In [12]: x=a[['Happiness Rank']]
y=a['Happiness Rank']
```

```
In [13]: # 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 [14]: from sklearn.linear_model import LinearRegression
lr=LinearRegression()
lr.fit(x_train,y_train)
```

```
Out[14]: LinearRegression()
```

```
In [15]: print(lr.intercept_)
```

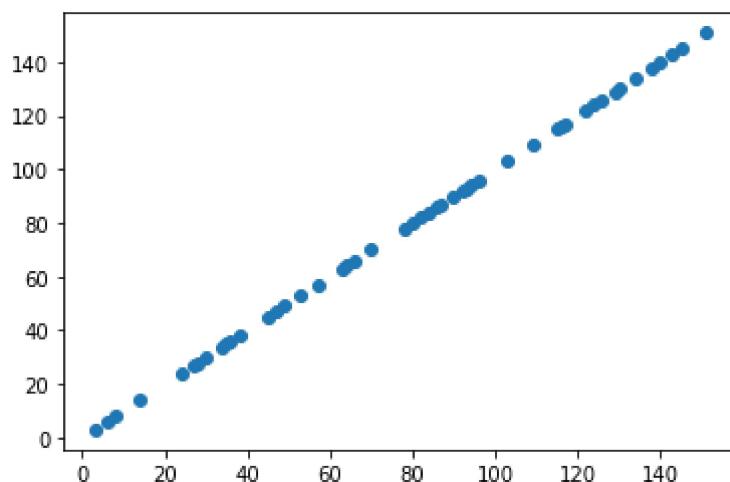
```
1.4210854715202004e-14
```

```
In [16]: coeff=pd.DataFrame(lr.coef_,x.columns,columns=['Co-efficient'])
coeff
```

	Co-efficient
Happiness Rank	1.0

```
In [17]: prediction=lr.predict(x_test)
plt.scatter(y_test,prediction)
```

```
Out[17]: <matplotlib.collections.PathCollection at 0x25e5af58d00>
```



```
In [18]: print(lr.score(x_test,y_test))
```

1.0

```
In [19]: lr.score(x_train,y_train)
```

```
Out[19]: 1.0
```

## Ridge regression

```
In [20]: from sklearn.linear_model import Ridge,Lasso
```

```
In [21]: rr=Ridge(alpha=10)
rr.fit(x_train,y_train)
rr.score(x_test,y_test)
```

```
Out[21]: 0.999999983109494
```

```
In [22]: rr.score(x_train,y_train)
```

```
Out[22]: 0.999999983111332
```

## Lasso regression

```
In [23]: la=Lasso(alpha=10)
la.fit(x_train,y_train)
la.score(x_train,y_train)
```

```
Out[23]: 0.9999795630312011
```

```
In [24]: la.score(x_test,y_test)
```

```
Out[24]: 0.9999795608081737
```

```
In [ ]:
```