

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

```
In [3]: df=pd.read_csv(r"C:\Users\user\Downloads\2_2015.csv")
df.fillna(0,inplace=True)
df
```

Out[3]:

	Country	Region	Happiness Rank	Happiness Score	Standard Error	Economy (GDP per Capita)	Family	Health (Life Expectancy)	Frei
0	Switzerland	Western Europe	1	7.587	0.03411	1.39651	1.34951	0.94143	0.6
1	Iceland	Western Europe	2	7.561	0.04884	1.30232	1.40223	0.94784	0.6
2	Denmark	Western Europe	3	7.527	0.03328	1.32548	1.36058	0.87464	0.6
3	Norway	Western Europe	4	7.522	0.03880	1.45900	1.33095	0.88521	0.6
4	Canada	North America	5	7.427	0.03553	1.32629	1.32261	0.90563	0.6
...
153	Rwanda	Sub-Saharan Africa	154	3.465	0.03464	0.22208	0.77370	0.42864	0.5
154	Benin	Sub-Saharan Africa	155	3.340	0.03656	0.28665	0.35386	0.31910	0.4
155	Syria	Middle East and Northern Africa	156	3.006	0.05015	0.66320	0.47489	0.72193	0.1
156	Burundi	Sub-Saharan Africa	157	2.905	0.08658	0.01530	0.41587	0.22396	0.1
157	Togo	Sub-Saharan Africa	158	2.839	0.06727	0.20868	0.13995	0.28443	0.3

158 rows × 12 columns

In [4]: `df.head()`

Out[4]:

	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	0.665
1	Iceland	Western Europe	2	7.561	0.04884	1.30232	1.40223	0.94784	0.628
2	Denmark	Western Europe	3	7.527	0.03328	1.32548	1.36058	0.87464	0.649
3	Norway	Western Europe	4	7.522	0.03880	1.45900	1.33095	0.88521	0.669
4	Canada	North America	5	7.427	0.03553	1.32629	1.32261	0.90563	0.632

In [5]: `df.info()`

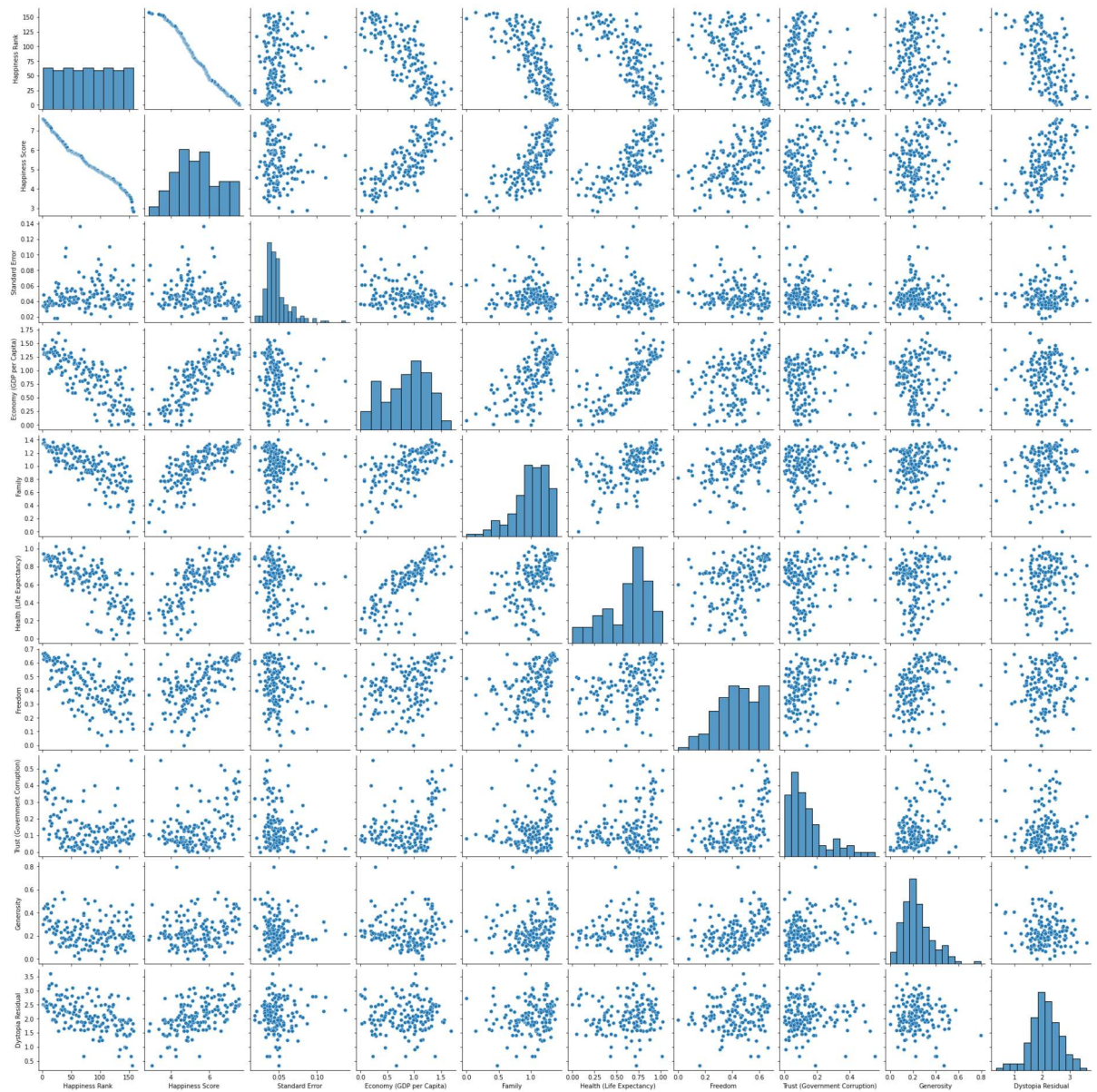
```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 158 entries, 0 to 157
Data columns (total 12 columns):
#   Column                                Non-Null Count  Dtype
---  -
0   Country                               158 non-null    object
1   Region                                158 non-null    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
```

In [6]: `df=pd.read_csv("2_2015.csv")`

In [7]: `import seaborn as sns`

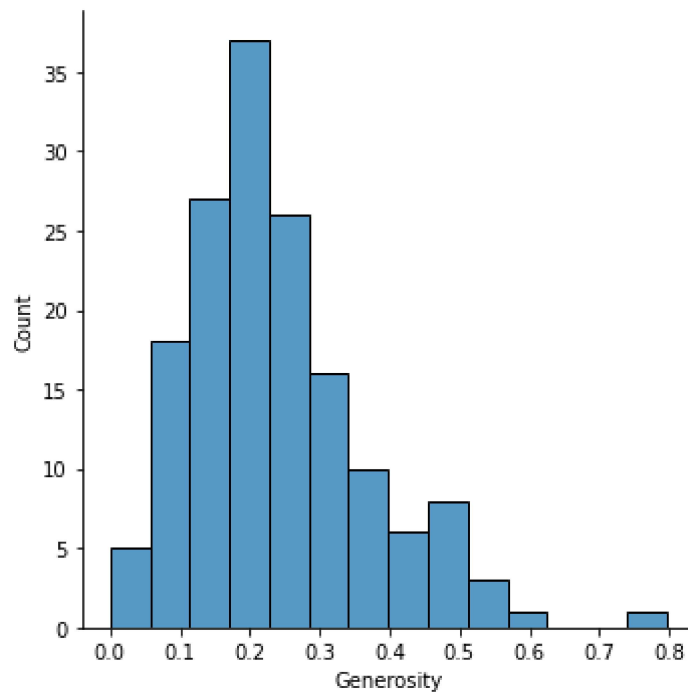
```
In [8]: sns.pairplot(df)
```

```
Out[8]: <seaborn.axisgrid.PairGrid at 0x1e0a1080c10>
```



```
In [9]: sns.displot(df['Generosity'])
```

```
Out[9]: <seaborn.axisgrid.FacetGrid at 0x1e0a6d47dc0>
```

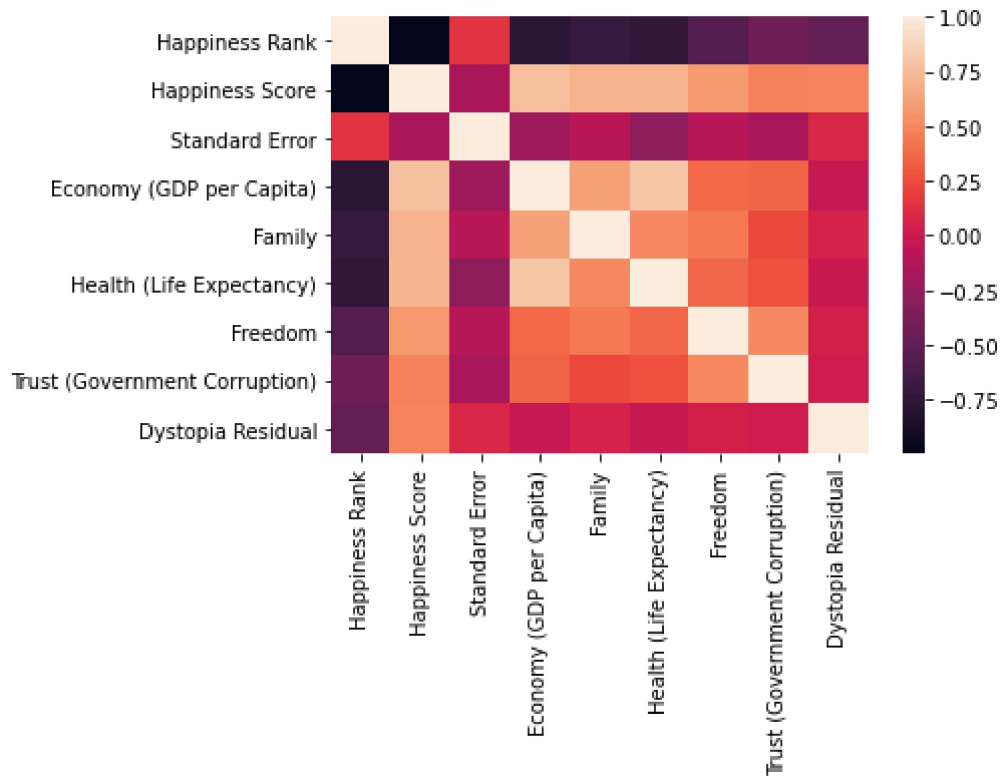


```
In [10]: df1=df.drop(['Generosity'],axis=1)
df1
df1=df1.drop(df1.index[153:])
df1.isna().sum()
```

```
Out[10]: Country      0
Region      0
Happiness Rank      0
Happiness Score      0
Standard Error      0
Economy (GDP per Capita)      0
Family      0
Health (Life Expectancy)      0
Freedom      0
Trust (Government Corruption)      0
Dystopia Residual      0
dtype: int64
```

```
In [11]: sns.heatmap(df1.corr())
```

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



```
In [12]: from sklearn.model_selection import train_test_split
from sklearn.linear_model import LinearRegression
```

```
In [13]: df1.isna().sum()
```

```
Out[13]: Country          0
Region          0
Happiness Rank    0
Happiness Score   0
Standard Error    0
Economy (GDP per Capita)  0
Family           0
Health (Life Expectancy)  0
Freedom          0
Trust (Government Corruption)  0
Dystopia Residual  0
dtype: int64
```

```
In [23]: y=df1['Happiness Rank']
x=df1.drop(['Country','Happiness Rank','Region'],axis=1)
x_train,x_test,y_train,y_test=train_test_split(x,y,test_size=0.3)
print(x_train)
```

	Happiness Score	Standard Error	Economy (GDP per Capita)	Family \
138	3.989	0.06682	0.67866	0.66290
22	6.810	0.06476	1.04424	1.25596
70	5.477	0.07197	1.00761	0.98521
101	4.857	0.05062	1.15406	0.92933
6	7.378	0.02799	1.32944	1.28017
..
48	5.960	0.05412	1.32376	1.21624
49	5.948	0.03914	1.25114	1.19777
69	5.548	0.04175	0.95847	1.22668
37	6.298	0.03868	1.29098	1.07617
87	5.102	0.04802	1.15991	1.13935

	Health (Life Expectancy)	Freedom	Trust (Government Corruption)	\
138	0.31051	0.41466	0.11686	
22	0.72052	0.42908	0.11069	
70	0.70950	0.56066	0.07521	
101	0.88213	0.07699	0.01397	
6	0.89284	0.61576	0.31814	
..	
48	0.74716	0.45492	0.30600	
49	0.95446	0.26236	0.02901	
69	0.53886	0.47610	0.30844	
37	0.87530	0.39740	0.08129	
87	0.87519	0.51469	0.01078	

	Dystopia Residual
138	1.68135
22	3.19131
70	1.76145
101	1.80101
6	2.46570
..	...
48	1.73797
49	2.02518
69	1.86984
37	2.32323
87	1.26462

[107 rows x 8 columns]

```
In [24]: model=LinearRegression()
model.fit(x_train,y_train)
model.intercept_
```

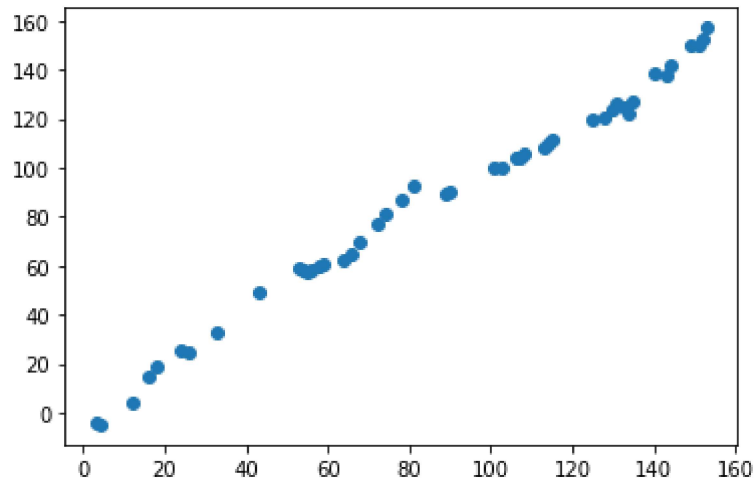
Out[24]: 297.1524499401098

```
In [25]: model.coef_
```

```
Out[25]: array([-26.92401634, -29.70457572, -12.69173816, -15.35442221,  
              -17.13611956, -18.78902373, -2.93847116, -12.41537647])
```

```
In [26]: prediction=model.predict(x_test)  
plt.scatter(y_test,prediction)
```

```
Out[26]: <matplotlib.collections.PathCollection at 0x1e0a8dd45b0>
```



```
In [27]: model.score(x_test,y_test)
```

```
Out[27]: 0.9876499329152798
```

```
In [ ]:
```