In [49]: !pip install plotly Requirement already satisfied: plotly in c:\users\christopher bradway\anac onda3\lib\site-packages (4.12.0) Requirement already satisfied: six in c:\users\christopher bradway\anacond a3\lib\site-packages (from plotly) (1.15.0) Requirement already satisfied: retrying>=1.3.3 in c:\users\christopher bra dway\anaconda3\lib\site-packages (from plotly) (1.3.3) In [50]: !pip install chart studio Requirement already satisfied: chart studio in c:\users\christopher bradwa y\anaconda3\lib\site-packages (1.1.0) Requirement already satisfied: plotly in c:\users\christopher bradway\anac onda3\lib\site-packages (from chart studio) (4.12.0) Requirement already satisfied: requests in c:\users\christopher bradway\an aconda3\lib\site-packages (from chart studio) (2.24.0) Requirement already satisfied: retrying>=1.3.3 in c:\users\christopher bra dway\anaconda3\lib\site-packages (from chart studio) (1.3.3) Requirement already satisfied: six in c:\users\christopher bradway\anacond a3\lib\site-packages (from chart studio) (1.15.0) Requirement already satisfied: idna<3,>=2.5 in c:\users\christopher bradwa y\anaconda3\lib\site-packages (from requests->chart studio) (2.10) Requirement already satisfied: certifi>=2017.4.17 in c:\users\christopher bradway\anaconda3\lib\site-packages (from requests->chart studio) (2020.6. 20) Requirement already satisfied: chardet<4,>=3.0.2 in c:\users\christopher b radway\anaconda3\lib\site-packages (from requests->chart studio) (3.0.4) Requirement already satisfied: urllib3!=1.25.0, !=1.25.1, <1.26, >=1.21.1 in c:\users\christopher bradway\anaconda3\lib\site-packages (from requests->c hart studio) (1.25.9) In [51]: !pip install discover feature relationships Requirement already satisfied: discover feature relationships in c:\users\ christopher bradway\anaconda3\lib\site-packages (1.0.3) In [52]: import pandas as pd import numpy as np from matplotlib import pyplot as plt from sklearn.model selection import train test split from sklearn import datasets from sklearn.linear_model import LinearRegression from sklearn.metrics import mean squared error import seaborn as sns import plotly.graph objs as pgo import plotly.express as px from plotly.offline import download plotlyjs, init notebook mode, plot,ipl ot. import chart studio.plotly as csp init notebook mode (connected=True) import statsmodels.formula.api as stats

```
from statsmodels.formula.api import ols
from discover feature relationships import discover
from sklearn.model selection import cross val score
```

```
In [53]: df2015=pd.read csv(r'C:\Users\Christopher Bradway\Downloads\2015.csv')
         df2015['Year'] = 2015
         df2015.head()
```

Out[53]:

	Country	Region	Rank	Score	GDP	Support	Health	Freedom	Corruption	Generosity	Y
0	Switzerland	Western Europe	1	7.587	1.39651	1.34951	0.94143	0.66557	0.41978	0.29678	2
1	Iceland	Western Europe	2	7.561	1.30232	1.40223	0.94784	0.62877	0.14145	0.43630	21
2	Denmark	Western Europe	3	7.527	1.32548	1.36058	0.87464	0.64938	0.48357	0.34139	2
3	Norway	Western Europe	4	7.522	1.45900	1.33095	0.88521	0.66973	0.36503	0.34699	2
4	Canada	North America	5	7.427	1.32629	1.32261	0.90563	0.63297	0.32957	0.45811	2

```
In [54]: df2016=pd.read csv(r'C:\Users\Christopher Bradway\Downloads\2016.csv')
         df2017=pd.read csv(r'C:\Users\Christopher Bradway\Downloads\2017.csv')
         df2018=pd.read csv(r'C:\Users\Christopher Bradway\Downloads\2018.csv')
         df2019=pd.read csv(r'C:\Users\Christopher Bradway\Downloads\2019.csv')
         df2020=pd.read csv(r'C:\Users\Christopher Bradway\Downloads\2020.csv')
```

```
In [55]: for col in df2016.columns:
             print(col)
```

Country

Region

Rank

Score

GDP

Support

Health

Freedom

Corruption

Generosity

```
In [56]: for col in df2019.columns:
             print(col)
```

Rank

Country

Score

GDP

Support

Health

Freedom

Generosity Corruption

In [57]: df2020.head()

Out[57]:

	Country	Regional	Score	GDP	Support	Health	Freedom	Generosity	Corruption	R
0	Finland	Western Europe	7.8087	10.639267	0.954330	71.900825	0.949172	-0.059482	0.195445	
1	Denmark	Western Europe	7.6456	10.774001	0.955991	72.402504	0.951444	0.066202	0.168489	
2	Switzerland	Western Europe	7.5599	10.979933	0.942847	74.102448	0.921337	0.105911	0.303728	
3	Iceland	Western Europe	7.5045	10.772559	0.974670	73.000000	0.948892	0.246944	0.711710	
4	Norway	Western Europe	7.4880	11.087804	0.952487	73.200783	0.955750	0.134533	0.263218	

In [58]: df2017.head()

Out[58]:

	Country	Rank	Score	GDP	Support	Health	Freedom	Generosity	Corruption
0	Norway	1	7.537	1.616463	1.533524	0.796667	0.635423	0.362012	0.315964
1	Denmark	2	7.522	1.482383	1.551122	0.792566	0.626007	0.355280	0.400770
2	Iceland	3	7.504	1.480633	1.610574	0.833552	0.627163	0.475540	0.153527
3	Switzerland	4	7.494	1.564980	1.516912	0.858131	0.620071	0.290549	0.367007
4	Finland	5	7.469	1.443572	1.540247	0.809158	0.617951	0.245483	0.382612

In [59]: df2015.describe()

Out[59]:

		Rank	Score	GDP	Support	Health	Freedom	Corruption	Generosi
С	ount	158.000000	158.000000	158.000000	158.000000	158.000000	158.000000	158.000000	158.0000
n	nean	79.493671	5.375734	0.846137	0.991046	0.630259	0.428615	0.143422	0.2372
	std	45.754363	1.145010	0.403121	0.272369	0.247078	0.150693	0.120034	0.1266
	min	1.000000	2.839000	0.000000	0.000000	0.000000	0.000000	0.000000	0.00000
	25%	40.250000	4.526000	0.545808	0.856823	0.439185	0.328330	0.061675	0.1505
	50%	79.500000	5.232500	0.910245	1.029510	0.696705	0.435515	0.107220	0.2161
	75%	118.750000	6.243750	1.158448	1.214405	0.811013	0.549092	0.180255	0.3098
	max	158.000000	7.587000	1.690420	1.402230	1.025250	0.669730	0.551910	0.7958

In [60]: df2015.info()

```
<class 'pandas.core.frame.DataFrame'>
          RangeIndex: 158 entries, 0 to 157
          Data columns (total 11 columns):
                             Non-Null Count
           #
                Column
                                                Dtype
                _____
                              -----
                                                ----
                              158 non-null
           0
                Country
                                                object
           1
                Region
                              158 non-null
                                                object
           2
                Rank
                              158 non-null
                                                int64
           3
                Score
                              158 non-null
                                                float64
           4
                GDP
                              158 non-null
                                                float64
           5
                Support
                              158 non-null
                                                float64
                Health
                              158 non-null
                                                float64
           6
           7
                Freedom
                              158 non-null
                                                float64
                Corruption 158 non-null
                                                float64
           8
           9
                Generosity 158 non-null
                                                float64
           10 Year
                             158 non-null
                                                int64
          dtypes: float64(7), int64(2), object(2)
          memory usage: 13.7+ KB
In [61]: df2016['Year']=2016
          df2017['Year']=2017
          df2018['Year']=2018
          df2019['Year']=2019
          df2020['Year']=2020
In [62]: df2018.head()
Out[62]:
              Rank
                      Country Score
                                    GDP Support Health Freedom Generosity Corruption Year
           0
                 1
                      Finland
                              7.632 1.305
                                            1.592
                                                  0.874
                                                           0.681
                                                                     0.202
                                                                               0.393 2018
           1
                 2
                      Norway
                              7.594
                                   1.456
                                            1.582
                                                  0.861
                                                           0.686
                                                                     0.286
                                                                               0.340 2018
           2
                 3
                     Denmark
                             7.555 1.351
                                            1.590
                                                  0.868
                                                                               0.408 2018
                                                           0.683
                                                                     0.284
                                                                               0.138 2018
           3
                 4
                       Iceland 7.495 1.343
                                            1.644
                                                  0.914
                                                           0.677
                                                                     0.353
                 5 Switzerland 7.487 1.420
                                            1.549
                                                  0.927
                                                           0.660
                                                                     0.256
                                                                               0.357 2018
In [63]: df2020.head()
Out[63]:
                Country Regional
                                 Score
                                           GDP
                                                 Support
                                                            Health Freedom Generosity Corruption R
                         Western
           0
                 Finland
                                7.8087 10.639267 0.954330 71.900825 0.949172
                                                                            -0.059482
                                                                                       0.195445
                         Europe
                         Western
                                7.6456 10.774001 0.955991 72.402504 0.951444
           1
               Denmark
                                                                             0.066202
                                                                                       0.168489
                         Europe
                         Western
                                7.5599
           2 Switzerland
                                      10.979933 0.942847 74.102448 0.921337
                                                                             0.105911
                                                                                       0.303728
                         Europe
```

7.5045 10.772559 0.974670 73.000000 0.948892

0.246944

0.711710

Western

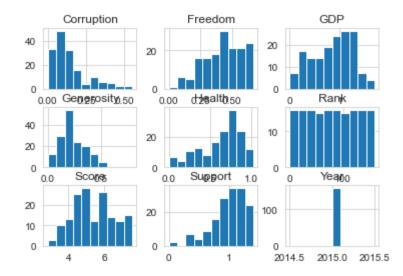
Europe

3

Iceland

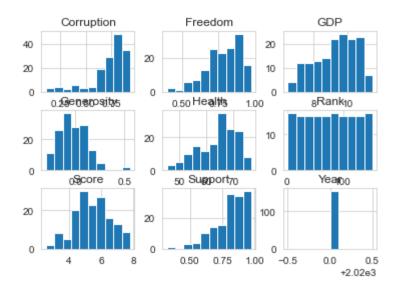
Western Furope 7.4880 11.087804 0.952487 73.200783 0.955750 0.134533 0.263218

```
In [64]: df2015.shape
Out[64]: (158, 11)
In [65]: df2016.shape
Out[65]: (157, 11)
In [66]: df2017.shape
Out[66]: (155, 10)
In [67]: df2018.shape
Out[67]: (156, 10)
In [68]: df2019.shape
Out[68]: (156, 10)
In [69]: df2020.shape
Out[69]: (153, 11)
In [70]: df2015.hist()
Out[70]: array([[<matplotlib.axes. subplots.AxesSubplot object at 0x0000023D8A5B494
         0>,
                  <matplotlib.axes. subplots.AxesSubplot object at 0x0000023D8A7B4EE</pre>
         0>,
                  <matplotlib.axes. subplots.AxesSubplot object at 0x0000023D8CBA44C</pre>
         0>],
                 [<matplotlib.axes. subplots.AxesSubplot object at 0x0000023D8A1CC91
         0>,
                  <matplotlib.axes. subplots.AxesSubplot object at 0x0000023D8A19DD6</pre>
         0>,
                  <matplotlib.axes. subplots.AxesSubplot object at 0x0000023D8CB8D13</pre>
         0>],
                 [<matplotlib.axes. subplots.AxesSubplot object at 0x0000023D8CB8D22
         0>,
                  <matplotlib.axes. subplots.AxesSubplot object at 0x0000023D8B0556D</pre>
         0>,
                  <matplotlib.axes. subplots.AxesSubplot object at 0x0000023D8AE97EE</pre>
         0>]],
                dtype=object)
```



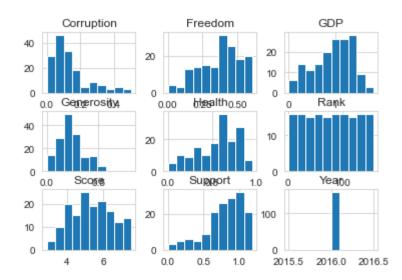
In [71]: df2020.hist()

Out[71]: array([[<matplotlib.axes. subplots.AxesSubplot object at 0x0000023D8C8A89D 0>, <matplotlib.axes. subplots.AxesSubplot object at 0x0000023D8A1F761</pre> 0>, <matplotlib.axes. subplots.AxesSubplot object at 0x0000023D8C93C85</pre> 0>], [<matplotlib.axes. subplots.AxesSubplot object at 0x0000023D8A3A7CA 0>, <matplotlib.axes. subplots.AxesSubplot object at 0x0000023D8CAE013</pre> 0>, <matplotlib.axes. subplots.AxesSubplot object at 0x0000023D8AFF84C</pre> 0>], [<matplotlib.axes. subplots.AxesSubplot object at 0x0000023D8AFF85B 0>, <matplotlib.axes. subplots.AxesSubplot object at 0x0000023D8CA90A6</pre> 0>, <matplotlib.axes. subplots.AxesSubplot object at 0x0000023D8CA8A2B</pre> 0>]], dtype=object)



In [72]: df2016.hist()

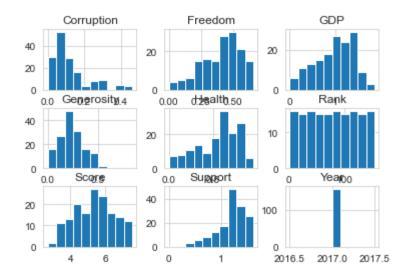
Out[72]: array([[<matplotlib.axes. subplots.AxesSubplot object at 0x0000023D8A19F82 0>, <matplotlib.axes. subplots.AxesSubplot object at 0x0000023D8D05837</pre> 0>, <matplotlib.axes. subplots.AxesSubplot object at 0x0000023D8CC877F</pre> 0>1, [<matplotlib.axes. subplots.AxesSubplot object at 0x0000023D8CFC96D 0>, <matplotlib.axes. subplots.AxesSubplot object at 0x0000023D8C98B91</pre> 0>, <matplotlib.axes. subplots.AxesSubplot object at 0x0000023D8D0C567</pre> 0>], [<matplotlib.axes. subplots.AxesSubplot object at 0x0000023D8D042A0 0>, <matplotlib.axes. subplots.AxesSubplot object at 0x0000023D8ACBD82</pre> 0>, <matplotlib.axes. subplots.AxesSubplot object at 0x0000023D8B19085</pre> 0>]], dtype=object)



In [73]: df2017.hist()

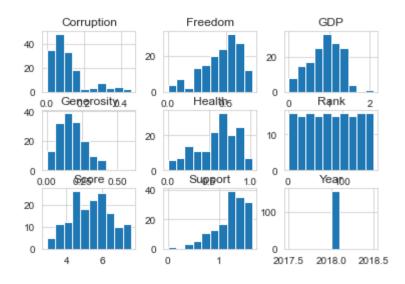
Out[73]: array([[<matplotlib.axes. subplots.AxesSubplot object at 0x0000023D8CFC06A 0>, <matplotlib.axes. subplots.AxesSubplot object at 0x0000023D8CFAABB</pre> 0>, <matplotlib.axes. subplots.AxesSubplot object at 0x0000023D8A1460A</pre> 0>1, [<matplotlib.axes. subplots.AxesSubplot object at 0x0000023D8C8AD49 0>, <matplotlib.axes. subplots.AxesSubplot object at 0x0000023D8B0A18E</pre> 0>, <matplotlib.axes. subplots.AxesSubplot object at 0x0000023D8B104C7</pre> 0>], [<matplotlib.axes. subplots.AxesSubplot object at 0x0000023D8B104D6 0>, <matplotlib.axes. subplots.AxesSubplot object at 0x0000023D8A22E0A</pre> 0>, <matplotlib.axes. subplots.AxesSubplot object at 0x0000023D8AFBEF1</pre> 0>]],

dtype=object)



In [74]: df2018.hist()

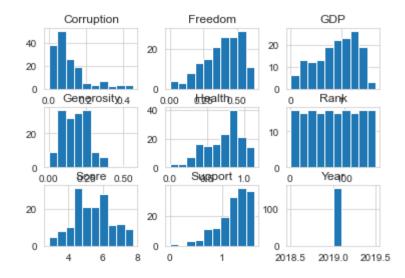
Out[74]: array([[<matplotlib.axes. subplots.AxesSubplot object at 0x0000023D8B0D0CD <matplotlib.axes. subplots.AxesSubplot object at 0x0000023D8B0DA58</pre> 0>, <matplotlib.axes. subplots.AxesSubplot object at 0x0000023D8A4CC8B</pre> 0>1, [<matplotlib.axes. subplots.AxesSubplot object at 0x0000023D8B2110D 0>, <matplotlib.axes. subplots.AxesSubplot object at 0x0000023D8CD787F</pre> 0>, <matplotlib.axes. subplots.AxesSubplot object at 0x0000023D8B1CB04</pre> 0>], [<matplotlib.axes. subplots.AxesSubplot object at 0x0000023D8B1CBFA 0>, <matplotlib.axes. subplots.AxesSubplot object at 0x0000023D8CE917C</pre> 0>, <matplotlib.axes. subplots.AxesSubplot object at 0x0000023D8CD1667</pre> 0>]],



In [75]: df2019.hist()

dtype=object)

```
Out[75]: array([[<matplotlib.axes. subplots.AxesSubplot object at 0x0000023D8CD168E
          0>,
                  <matplotlib.axes. subplots.AxesSubplot object at 0x0000023D8D3E91C</pre>
          0>,
                  <matplotlib.axes. subplots.AxesSubplot object at 0x0000023D8CFD694</pre>
          0>1,
                 [<matplotlib.axes. subplots.AxesSubplot object at 0x0000023D8CFFA16
          0>,
                  <matplotlib.axes. subplots.AxesSubplot object at 0x0000023D8CCCA88</pre>
          0>,
                  <matplotlib.axes. subplots.AxesSubplot object at 0x0000023D8A6180A</pre>
          0>],
                 [<matplotlib.axes. subplots.AxesSubplot object at 0x0000023D8A62407
          0>,
                  <matplotlib.axes. subplots.AxesSubplot object at 0x0000023D8A64B85</pre>
          0>,
                  <matplotlib.axes. subplots.AxesSubplot object at 0x0000023D8A6AB70</pre>
          0>]],
                dtype=object)
```



```
In [76]: target = ['Top','Top-Mid', 'Low-Mid', 'Low']
  target_n = [4, 3, 2, 1]
  df2015["target"] = pd.qcut(df2015['Rank'], len(target), labels=target)
  df2015["target_n"] = pd.qcut(df2015['Rank'], len(target), labels=target_n)
```

In [77]: df = df2015.append([df2016,df2017,df2018,df2019])

In [78]: df.isnull().any()

Out[78]: Country False Region True Rank False False Score GDP False Support False Health False Freedom False Corruption True Generosity False Year False

```
In [79]: df.Corruption.fillna((df.Corruption.mean()), inplace = True)
    df.head()
```

Out[79]:

	Country	Region	Rank	Score	GDP	Support	Health	Freedom	Corruption	Generosity	Y
0	Switzerland	Western Europe	1	7.587	1.39651	1.34951	0.94143	0.66557	0.41978	0.29678	2
1	Iceland	Western Europe	2	7.561	1.30232	1.40223	0.94784	0.62877	0.14145	0.43630	21
2	Denmark	Western Europe	3	7.527	1.32548	1.36058	0.87464	0.64938	0.48357	0.34139	2
3	Norway	Western Europe	4	7.522	1.45900	1.33095	0.88521	0.66973	0.36503	0.34699	2
4	Canada	North America	5	7.427	1.32629	1.32261	0.90563	0.63297	0.32957	0.45811	2

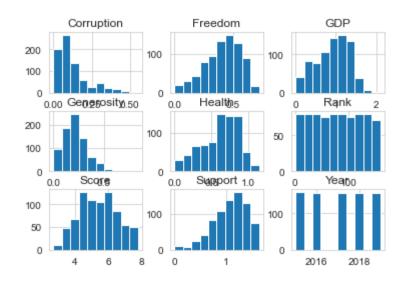
In [81]: df.describe()

Out[81]:

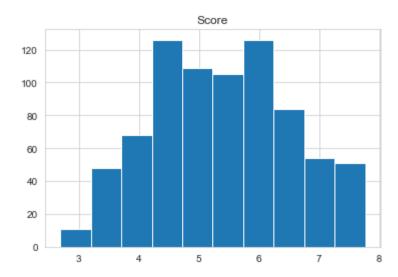
	Rank	Score	GDP	Support	Health	Freedom	Corruption	Generosi
count	782.000000	782.000000	782.000000	782.000000	782.000000	782.000000	782.000000	782.0000
mean	78.698210	5.379018	0.916047	1.078392	0.612416	0.411091	0.125436	0.2185
std	45.182384	1.127456	0.407340	0.329548	0.248309	0.152880	0.105749	0.1223
min	1.000000	2.693000	0.000000	0.000000	0.000000	0.000000	0.000000	0.00000
25%	40.000000	4.509750	0.606500	0.869363	0.440183	0.309768	0.054250	0.1300
50%	79.000000	5.322000	0.982205	1.124735	0.647310	0.431000	0.091033	0.2019
75%	118.000000	6.189500	1.236187	1.327250	0.808000	0.531000	0.155861	0.2788
max	158.000000	7.769000	2.096000	1.644000	1.141000	0.724000	0.551910	0.8380

```
In [82]: df.hist()
```

Out[82]: array([[<matplotlib.axes. subplots.AxesSubplot object at 0x0000023D8CB07F7 0>, <matplotlib.axes. subplots.AxesSubplot object at 0x0000023D8A875C1</pre> 0>, <matplotlib.axes. subplots.AxesSubplot object at 0x0000023D8A89F13</pre> 0>], [<matplotlib.axes. subplots.AxesSubplot object at 0x0000023D8A8D652 0>, <matplotlib.axes. subplots.AxesSubplot object at 0x0000023D8CCD397</pre> 0>, <matplotlib.axes. subplots.AxesSubplot object at 0x0000023D8CCFED0</pre> 0>], [<matplotlib.axes. subplots.AxesSubplot object at 0x0000023D8CCFEDF 0>, <matplotlib.axes. subplots.AxesSubplot object at 0x0000023D8CE762E</pre> 0>, <matplotlib.axes. subplots.AxesSubplot object at 0x0000023D8D10EAF</pre> 0>]], dtype=object)



In [83]: df.hist('Score')



In [84]: df

Out[84]:

	Country	Region	Rank	Score	GDP	Support	Health	Freedom	Corruption	Generosity
0	Switzerland	Western Europe	1	7.587	1.39651	1.34951	0.94143	0.66557	0.41978	0.29678
1	Iceland	Western Europe	2	7.561	1.30232	1.40223	0.94784	0.62877	0.14145	0.43630
2	Denmark	Western Europe	3	7.527	1.32548	1.36058	0.87464	0.64938	0.48357	0.34139
3	Norway	Western Europe	4	7.522	1.45900	1.33095	0.88521	0.66973	0.36503	0.34699
4	Canada	North America	5	7.427	1.32629	1.32261	0.90563	0.63297	0.32957	0.45811
151	Rwanda	NaN	152	3.334	0.35900	0.71100	0.61400	0.55500	0.41100	0.21700
152	Tanzania	NaN	153	3.231	0.47600	0.88500	0.49900	0.41700	0.14700	0.27600
153	Afghanistan	NaN	154	3.203	0.35000	0.51700	0.36100	0.00000	0.02500	0.15800
154	Central African Republic	NaN	155	3.083	0.02600	0.00000	0.10500	0.22500	0.03500	0.23500
155	South Sudan	NaN	156	2.853	0.30600	0.57500	0.29500	0.01000	0.09100	0.20200

782 rows × 13 columns

```
In [85]: f,ax=plt.subplots(figsize=(16,10))
    sns.heatmap(df2015.corr(),annot=True, linewidth=.5,fmt='.2f',ax=ax)
    plt.show()
```



In [86]: f,ax=plt.subplots(figsize=(16,10))
 sns.heatmap(df.corr(),annot=True, linewidth=.5,fmt='.2f',ax=ax)
 plt.show()

										- 1.00
Rank	1.00	-0.99	-0.79	-0.64	-0.74	-0.54	-0.37	-0.12	-0.01	
Score	-0.99	1.00	0.79	0.65	0.74	0.55	0.40	0.14	0.01	- 0.75
GDP	-0.79	0.79	1.00	0.59	0.78	0.34	0.30	-0.01	0.02	- 0.50
Support	-0.64	0.65	0.59	1.00	0.57	0.42	0.13	-0.04	0.37	- 0.25
Health S	-0.74	0.74	0.78	0.57	1.00	0.34	0.25	0.01	0.13	- 0.00
Freedom	-0.54	0.55	0.34	0.42	0.34	1.00	0.46	0.29	0.01	0.25
Corruption Fr	-0.37	0.40	0.30	0.13	0.25	0.46	1.00	0.32	-0.12	0.50
Generosity Corr	-0.12	0.14	-0.01	-0.04	0.01	0.29	0.32	1.00	-0.19	0.75
Year Gene	-0.01	0.01	0.02	0.37	0.13	0.01	-0.12	-0.19	1.00	
'	Rank	Score	GDP	Support	Health	Freedom	Corruption	Generosity	Year	

```
In [87]: px.scatter(df, x="GDP", y="Score", animation frame="Year",
                    animation group="Country",
                    size="Rank", color="Country", hover name="Country",
                    trendline= "ols")
         train data, test data = train test split(df, train size = 0.8, random stat
         e = 3)
         lr = LinearRegression()
         X train = np.array(train data['GDP'],
                            dtype = pd.Series).reshape(-1,1)
         y train = np.array(train data['Score'], dtype = pd.Series)
         lr.fit(X train, y train)
         X test = np.array(test data['GDP'],
                             dtype = pd.Series).reshape(-1,1)
         y test = np.array(test data['Score'], dtype = pd.Series)
         pred = lr.predict(X test)
         rmsesm = float(format(np.sqrt(mean squared error(y test,pred)),'.3f'))
         rtrsm = float(format(lr.score(X train, y train),'.3f'))
         rtesm = float(format(lr.score(X test, y test), '.3f'))
         cv = float(format(cross val score(lr,df[['GDP']],df['Score'],cv=5).mean(),
         print ("Average Score for Test Data: {:.3f}".format(y test.mean()))
         print('Intercept: {}'.format(lr.intercept_))
         print('Coefficient: {}'.format(lr.coef ))
         r = evaluation.shape[0]
         evaluation.loc[r] = ['Simple Linear Regression','-',rmsesm,rtrsm,'-',rtesm
         ,'-',cv]
         evaluation
```

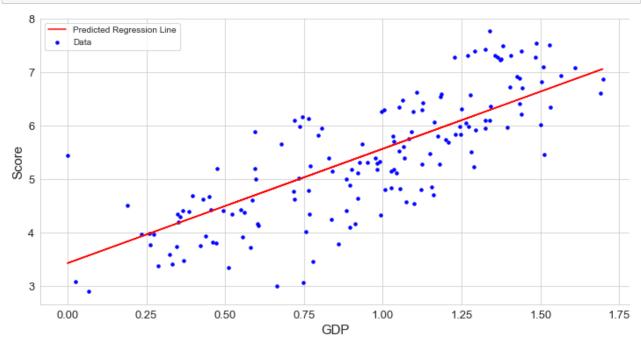
Average Score for Test Data: 5.388
Intercept: 3.430174973757425
Coefficient: [2.14115845]

Out[87]:

	Model	Details	Root Mean Squared Error (RMSE)	R- squared (training)	Adjusted R- squared (training)	R- squared (test)	Adjusted R- squared(test)	5-Fold Cross Validation
0	Simple Linear Regression	-	0.708	0.62	-	0.632	-	0.617

```
In [88]: sns.set_style(style='whitegrid')
   plt.figure(figsize=(12,6))
   plt.scatter(X_test,y_test,color='blue',label="Data", s = 12)
   plt.plot(X_test,lr.predict(X_test),color="red",label="Predicted Regression Line")
   plt.xlabel("GDP", fontsize=15)
   plt.ylabel("Score", fontsize=15)
   plt.xticks(fontsize=13)
   plt.yticks(fontsize=13)
   plt.legend()
```

```
plt.gca().spines['right'].set_visible(False)
plt.gca().spines['top'].set_visible(False)
```



```
In [89]: px.scatter(df, x="Support", y="Score", animation frame="Year",
                    animation group="Country",
                    size="Rank", color="Country", hover name="Country",
                    trendline= "ols")
         train data, test data = train test split(df, train size = 0.8, random stat
         e = 3)
         lr = LinearRegression()
         X train = np.array(train data['Support'],
                            dtype = pd.Series).reshape(-1,1)
         y_train = np.array(train_data['Score'], dtype = pd.Series)
         lr.fit(X train, y train)
         X test = np.array(test data['Support'],
                             dtype = pd.Series).reshape(-1,1)
         y test = np.array(test data['Score'], dtype = pd.Series)
         pred = lr.predict(X test)
         rmsesm = float(format(np.sqrt(mean squared error(y test,pred)),'.3f'))
         rtrsm = float(format(lr.score(X train, y train), '.3f'))
         rtesm = float(format(lr.score(X test, y test), '.3f'))
         cv = float(format(cross val score(lr,df[['Support']],df['Score'],cv=5).mea
         n(),'.3f'))
         print ("Average Score for Test Data: {:.3f}".format(y test.mean()))
         print('Intercept: {}'.format(lr.intercept ))
         print('Coefficient: {}'.format(lr.coef_))
         r = evaluation.shape[0]
         evaluation.loc[r] = ['Simple Linear Regression','-',rmsesm,rtrsm,'-',rtesm
         ,'-',cv]
         evaluation
```

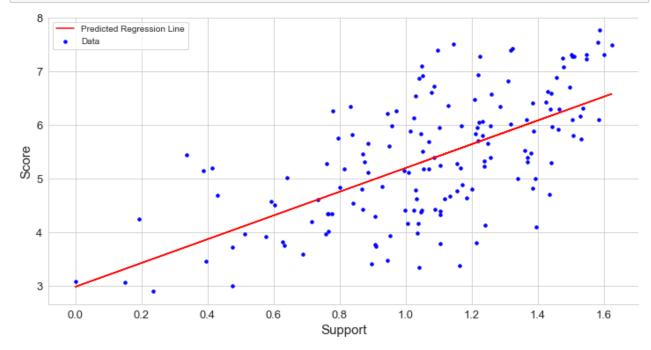
Average Score for Test Data: 5.388

Intercept: 2.987178261826246
Coefficient: [2.21482048]

Out[89]:

	Model	Details	Root Mean Squared Error (RMSE)	R- squared (training)	Adjusted R- squared (training)	R- squared (test)	Adjusted R- squared(test)	5-Fold Cross Validation
0	Simple Linear Regression	-	0.708	0.620	-	0.632	-	0.617
1	Simple Linear Regression	-	0.903	0.426	-	0.402	-	0.326

```
In [90]: sns.set_style(style='whitegrid')
    plt.figure(figsize=(12,6))
    plt.scatter(X_test,y_test,color='blue',label="Data", s = 12)
    plt.plot(X_test,lr.predict(X_test),color="red",label="Predicted Regression Line")
    plt.xlabel("Support", fontsize=15)
    plt.ylabel("Score", fontsize=15)
    plt.xticks(fontsize=13)
    plt.yticks(fontsize=13)
    plt.legend()
    plt.gca().spines['right'].set_visible(False)
    plt.gca().spines['top'].set_visible(False)
```



```
lr = LinearRegression()
X train = np.array(train data['Health'],
                   dtype = pd.Series).reshape(-1,1)
y train = np.array(train data['Score'], dtype = pd.Series)
lr.fit(X train, y train)
X test = np.array(test data['Health'],
                    dtype = pd.Series).reshape(-1,1)
y test = np.array(test data['Score'], dtype = pd.Series)
pred = lr.predict(X test)
rmsesm = float(format(np.sqrt(mean squared error(y test,pred)),'.3f'))
rtrsm = float(format(lr.score(X train, y train),'.3f'))
rtesm = float(format(lr.score(X test, y test), '.3f'))
cv = float(format(cross val score(lr,df[['Health']],df['Score'],cv=5).mean
(),'.3f'))
print ("Average Score for Test Data: {:.3f}".format(y test.mean()))
print('Intercept: {}'.format(lr.intercept_))
print('Coefficient: {}'.format(lr.coef ))
r = evaluation.shape[0]
evaluation.loc[r] = ['Simple Linear Regression','-',rmsesm,rtrsm,'-',rtesm
,'-',cv]
evaluation
```

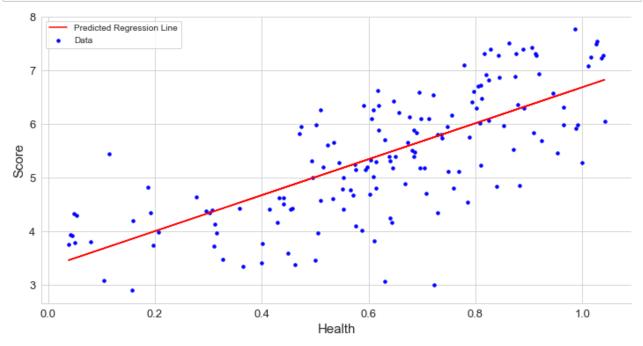
Average Score for Test Data: 5.388 Intercept: 3.3341420139695326 Coefficient: [3.35469024]

Out[91]:

	Model	Details	Root Mean Squared Error (RMSE)	R- squared (training)	Adjusted R- squared (training)	R- squared (test)	Adjusted R- squared(test)	5-Fold Cross Validation
0	Simple Linear Regression	-	0.708	0.620	-	0.632	-	0.617
1	Simple Linear Regression	-	0.903	0.426	-	0.402	-	0.326
2	Simple Linear Regression	-	0.795	0.555	-	0.536	-	0.530

```
In [92]: sns.set_style(style='whitegrid')
   plt.figure(figsize=(12,6))
   plt.scatter(X_test,y_test,color='blue',label="Data", s = 12)
   plt.plot(X_test,lr.predict(X_test),color="red",label="Predicted Regression
        Line")
   plt.xlabel("Health", fontsize=15)
   plt.ylabel("Score", fontsize=15)
   plt.xticks(fontsize=13)
   plt.yticks(fontsize=13)
```

```
plt.legend()
plt.gca().spines['right'].set_visible(False)
plt.gca().spines['top'].set_visible(False)
```



```
In [93]: px.scatter(df, x="Freedom", y="Score", animation frame="Year",
                    animation group="Country",
                    size="Rank", color="Country", hover name="Country",
                    trendline= "ols")
         train data, test data = train test split(df, train size = 0.8, random stat
         e = 3)
         lr = LinearRegression()
         X train = np.array(train data['Freedom'],
                            dtype = pd.Series).reshape(-1,1)
         y train = np.array(train data['Score'], dtype = pd.Series)
         lr.fit(X train, y train)
         X test = np.array(test data['Freedom'],
                             dtype = pd.Series).reshape(-1,1)
         y test = np.array(test data['Score'], dtype = pd.Series)
         pred = lr.predict(X test)
         rmsesm = float(format(np.sqrt(mean squared error(y test,pred)),'.3f'))
         rtrsm = float(format(lr.score(X train, y train),'.3f'))
         rtesm = float(format(lr.score(X test, y test), '.3f'))
         cv = float(format(cross val score(lr,df[['Freedom']],df['Score'],cv=5).mea
         n(),'.3f'))
         print ("Average Score for Test Data: {:.3f}".format(y test.mean()))
         print('Intercept: {}'.format(lr.intercept ))
         print('Coefficient: {}'.format(lr.coef ))
         r = evaluation.shape[0]
         evaluation.loc[r] = ['Simple Linear Regression','-',rmsesm,rtrsm,'-',rtesm
         ,'-',cv]
         evaluation
```

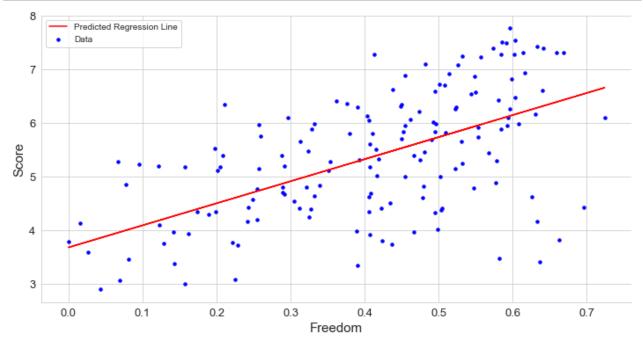
Average Score for Test Data: 5.388

Intercept: 3.683005610778345
Coefficient: [4.11588367]

Out[93]:

	Model	Details	Root Mean Squared Error (RMSE)	R- squared (training)	Adjusted R- squared (training)	R- squared (test)	Adjusted R- squared(test)	5-Fold Cross Validation
0	Simple Linear Regression	-	0.708	0.620	-	0.632	-	0.617
1	Simple Linear Regression	-	0.903	0.426	-	0.402	-	0.326
2	Simple Linear Regression	-	0.795	0.555	-	0.536	-	0.530
3	Simple Linear Regression	-	0.967	0.301	-	0.313	-	0.295

```
In [94]: sns.set_style(style='whitegrid')
    plt.figure(figsize=(12,6))
    plt.scatter(X_test,y_test,color='blue',label="Data", s = 12)
    plt.plot(X_test,lr.predict(X_test),color="red",label="Predicted Regression
        Line")
    plt.xlabel("Freedom", fontsize=15)
    plt.ylabel("Score", fontsize=15)
    plt.yticks(fontsize=13)
    plt.yticks(fontsize=13)
    plt.legend()
    plt.gca().spines['right'].set_visible(False)
    plt.gca().spines['top'].set_visible(False)
```



```
In [95]: px.scatter(df, x="Generosity", y="Score", animation frame="Year",
                    animation group="Country",
                    size="Rank", color="Country", hover name="Country",
                    trendline= "ols")
         train data, test data = train test split(df, train size = 0.8, random stat
         e = 3)
         lr = LinearRegression()
         X train = np.array(train data['Generosity'],
                            dtype = pd.Series).reshape(-1,1)
         y train = np.array(train data['Score'], dtype = pd.Series)
         lr.fit(X train, y train)
         X test = np.array(test data['Generosity'],
                             dtype = pd.Series).reshape(-1,1)
         y test = np.array(test data['Score'], dtype = pd.Series)
         pred = lr.predict(X test)
         rmsesm = float(format(np.sqrt(mean squared error(y test,pred)),'.3f'))
         rtrsm = float(format(lr.score(X train, y train), '.3f'))
         rtesm = float(format(lr.score(X test, y test),'.3f'))
         cv = float(format(cross val score(lr,df[['Generosity']],df['Score'],cv=5).
         mean(),'.3f'))
         print ("Average Score for Test Data: {:.3f}".format(y test.mean()))
         print('Intercept: {}'.format(lr.intercept ))
         print('Coefficient: {}'.format(lr.coef_))
         r = evaluation.shape[0]
         evaluation.loc[r] = ['Simple Linear Regression','-',rmsesm,rtrsm,'-',rtesm
         ,'-',cv]
         evaluation
```

Average Score for Test Data: 5.388 Intercept: 5.089485375227209 Coefficient: [1.32184138]

Out[95]:

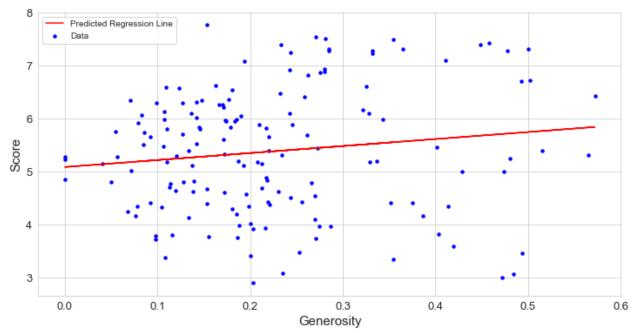
	Model	Details	Root Mean Squared Error (RMSE)	R- squared (training)	Adjusted R- squared (training)	R- squared (test)	Adjusted R- squared(test)	5-Fold Cross Validation
0	Simple Linear Regression	-	0.708	0.620	-	0.632	-	0.617
1	Simple Linear Regression	-	0.903	0.426	-	0.402	-	0.326
2	Simple Linear Regression	-	0.795	0.555	-	0.536	-	0.530
3	Simple Linear Regression	-	0.967	0.301	-	0.313	-	0.295

```
Simple

4 Linear - 1.160 0.021 - 0.012 - 0.017

Regression
```

```
In [96]: sns.set_style(style='whitegrid')
    plt.figure(figsize=(12,6))
    plt.scatter(X_test,y_test,color='blue',label="Data", s = 12)
    plt.plot(X_test,lr.predict(X_test),color="red",label="Predicted Regression Line")
    plt.xlabel("Generosity", fontsize=15)
    plt.ylabel("Score", fontsize=15)
    plt.yticks(fontsize=13)
    plt.yticks(fontsize=13)
    plt.legend()
    plt.gca().spines['right'].set_visible(False)
    plt.gca().spines['top'].set_visible(False)
```



```
In [97]: px.scatter(df, x="Corruption", y="Score", animation frame="Year",
                    animation group="Country",
                    size="Rank", color="Country", hover name="Country",
                    trendline= "ols")
         train data, test data = train test split(df, train size = 0.8, random stat
         e = 3)
         lr = LinearRegression()
         X train = np.array(train data['Corruption'],
                            dtype = pd.Series).reshape(-1,1)
         y train = np.array(train data['Score'], dtype = pd.Series)
         lr.fit(X train, y train)
         X test = np.array(test data['Corruption'],
                             dtype = pd.Series).reshape(-1,1)
         y test = np.array(test data['Score'], dtype = pd.Series)
         pred = lr.predict(X test)
         rmsesm = float(format(np.sqrt(mean squared error(y test,pred)),'.3f'))
```

```
rtrsm = float(format(lr.score(X_train, y_train),'.3f'))
rtesm = float(format(lr.score(X_test, y_test),'.3f'))
cv = float(format(cross_val_score(lr,df[['Corruption']],df['Score'],cv=5).
mean(),'.3f'))
print ("Average Score for Test Data: {:.3f}".format(y_test.mean()))
print('Intercept: {}'.format(lr.intercept_))
print('Coefficient: {}'.format(lr.coef_))
r = evaluation.shape[0]
evaluation.loc[r] = ['Simple Linear Regression','-',rmsesm,rtrsm,'-',rtesm,'-',cv]
evaluation
```

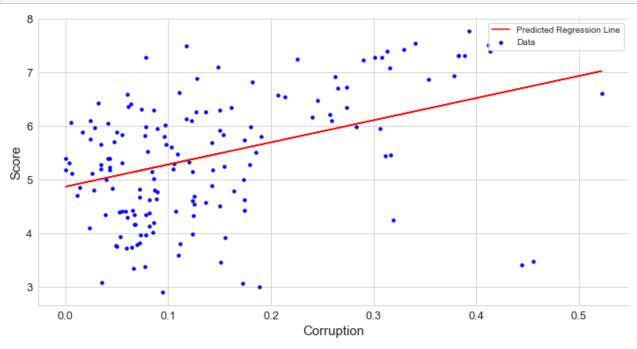
Average Score for Test Data: 5.388 Intercept: 4.872968925819756 Coefficient: [4.1304732]

Out[97]:

	Model	Details	Root Mean Squared Error (RMSE)	R- squared (training)	Adjusted R- squared (training)	R- squared (test)	Adjusted R- squared(test)	5-Fold Cross Validation
0	Simple Linear Regression	-	0.708	0.620	-	0.632	-	0.617
1	Simple Linear Regression	-	0.903	0.426	-	0.402	-	0.326
2	Simple Linear Regression	-	0.795	0.555	-	0.536	-	0.530
3	Simple Linear Regression	-	0.967	0.301	-	0.313	-	0.295
4	Simple Linear Regression	-	1.160	0.021	-	0.012	-	0.017
5	Simple Linear Regression	-	1.046	0.148	-	0.196	-	0.155

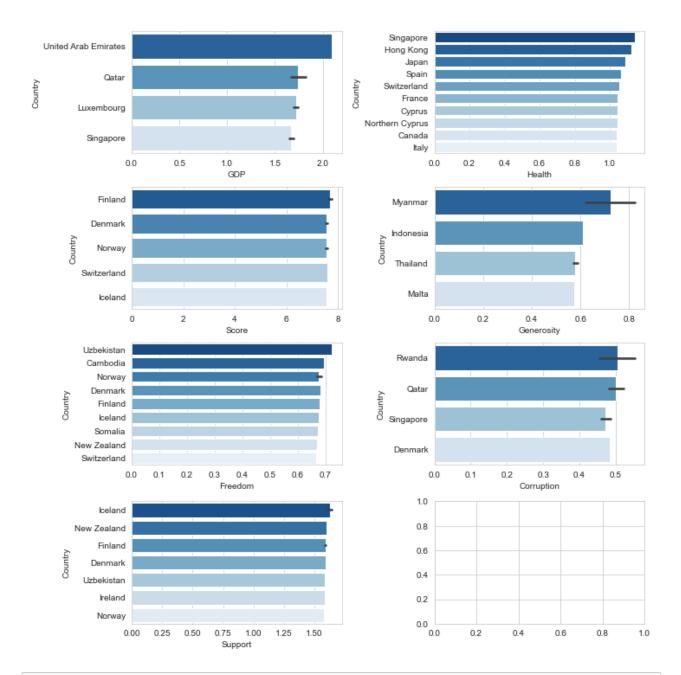
```
In [98]: sns.set_style(style='whitegrid')
   plt.figure(figsize=(12,6))
   plt.scatter(X_test,y_test,color='blue',label="Data", s = 12)
   plt.plot(X_test,lr.predict(X_test),color="red",label="Predicted Regression Line")
   plt.xlabel("Corruption", fontsize=15)
   plt.ylabel("Score", fontsize=15)
   plt.xticks(fontsize=13)
   plt.yticks(fontsize=13)
   plt.legend()
```

```
plt.gca().spines['right'].set_visible(False)
plt.gca().spines['top'].set_visible(False)
```



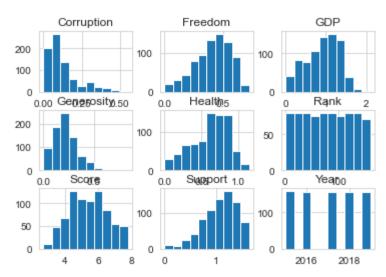
```
In [103]: fig, axes = plt.subplots(nrows=4, ncols=2,constrained layout=True,figsize=
          (10, 10))
          sns.barplot(x='GDP', y='Country',
                                   data=df.nlargest(10, 'GDP'),
                                   ax=axes[0,0],palette="Blues r")
          sns.barplot(x='Health', y='Country',
                                   data=df.nlargest(10, 'Health'),
                                   ax=axes[0,1],palette='Blues r')
          sns.barplot(x='Score', y='Country',
                                   data=df.nlargest(10, 'Score'),
                                   ax=axes[1,0],palette='Blues r')
          sns.barplot(x='Generosity' , y='Country',
                                   data=df.nlargest(10, 'Generosity'),
                                   ax=axes[1,1],palette='Blues r')
          sns.barplot(x='Freedom' , y='Country',
                                   data=df.nlargest(10, 'Freedom'),
                                   ax=axes[2,0],palette='Blues r')
          sns.barplot(x='Corruption', y='Country',
                                   data=df.nlargest(10, 'Corruption'),
                                   ax=axes[2,1],palette='Blues r')
          sns.barplot(x='Support', y='Country',
                                   data=df.nlargest(10, 'Support'),
                                   ax=axes[3,0],palette='Blues r')
```

Out[103]: <matplotlib.axes. subplots.AxesSubplot at 0x23d8f1daa90>



In [100]: df.hist()

<matplotlib.axes. subplots.AxesSubplot object at 0x0000023D8E68E13</pre>



New Happiness

In []: