WorldHappinessRecord

March 5, 2024

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

[2]:
```
# 2015 Analysis
```

[3]:
```
y15 = pd.read_csv("2015.csv")
```

[4]:
```
y15.head(5)
```

[4]:

	Country	Region	Happiness Rank	Happiness Score	Standard Error	Economy (GDP per Capita)	Family	Health (Life Expectancy)	Freedom	Trust (Government Corruption)	Generosity	Dystopia Residual
0	Switzerland	Western Europe	1	7.587	0.03411	1.39651	1.34951	0.94143	0.66557	0.41978	0.29678	2.51738
1	Iceland	Western Europe	2	7.561	0.04884	1.30232	1.40223	0.94784	0.62877	0.14145	0.43630	2.70201
2	Denmark	Western Europe	3	7.527	0.03328	1.32548	1.36058	0.87464	0.64938	0.48357	0.34139	2.49204
3	Norway	Western Europe	4	7.522	0.03880	1.45900	1.33095	0.88521	0.66973	0.36503	0.34699	2.46531
4	Canada	North America	5	7.427	0.03553	1.32629	1.32261	0.90563	0.63297	0.32957	0.45811	2.45176

[5]:
```
y15.shape
```

[5]:
```
(158, 12)
```

[6]:
```
y15.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
```

[7]:
```
y15.isnull().sum()
```

[7]:
```
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
Generosity                       0
Dystopia Residual                0
dtype: int64
```

[8]:
```
y15 = y15.drop_duplicates()
```

[9]:
```
y15 = y15.drop( columns= ['Happiness Rank','Standard Error','Dystopia␣
↪Residual'])
```

[10]:
```
# 2016 Analysis
```

[11]:
```
y16 = pd.read_csv("2016.csv")
```

[12]:
```
y16.head(5)
```

[12]:

	Country	Region	Happiness Rank	Happiness Score	\
0	Denmark	Western Europe	1	7.526	
1	Switzerland	Western Europe	2	7.509	
2	Iceland	Western Europe	3	7.501	
3	Norway	Western Europe	4	7.498	
4	Finland	Western Europe	5	7.413	

	Lower Confidence Interval	Upper Confidence Interval	\
0	7.460	7.592	
1	7.428	7.590	
2	7.333	7.669	
3	7.421	7.575	
4	7.351	7.475	

	Economy (GDP per Capita)	Family	Health (Life Expectancy)	Freedom	\
0	1.44178	1.16374	0.79504	0.57941	
1	1.52733	1.14524	0.86303	0.58557	
2	1.42666	1.18326	0.86733	0.56624	
3	1.57744	1.12690	0.79579	0.59609	
4	1.40598	1.13464	0.81091	0.57104	

	Trust (Government Corruption)	Generosity	Dystopia Residual
0	0.44453	0.36171	2.73939
1	0.41203	0.28083	2.69463
2	0.14975	0.47678	2.83137
3	0.35776	0.37895	2.66465
4	0.41004	0.25492	2.82596

[13]:
```
y16.shape
```

[13]:
```
(157, 13)
```

[14]:
```
y16.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 157 entries, 0 to 156
Data columns (total 13 columns):
 #   Column                         Non-Null Count  Dtype
---  ------                         --------------  -----
 0   Country                        157 non-null    object
 1   Region                         157 non-null    object
 2   Happiness Rank                 157 non-null    int64
 3   Happiness Score                157 non-null    float64
 4   Lower Confidence Interval      157 non-null    float64
```

```
 5   Upper Confidence Interval      157 non-null    float64
 6   Economy (GDP per Capita)       157 non-null    float64
 7   Family                         157 non-null    float64
 8   Health (Life Expectancy)       157 non-null    float64
 9   Freedom                        157 non-null    float64
 10  Trust (Government Corruption)  157 non-null    float64
 11  Generosity                     157 non-null    float64
 12  Dystopia Residual              157 non-null    float64
dtypes: float64(10), int64(1), object(2)
memory usage: 16.1+ KB
```

[15]:
```
y16.isnull().sum()
```

[15]:
```
Country                          0
Region                           0
Happiness Rank                   0
Happiness Score                  0
Lower Confidence Interval        0
Upper Confidence Interval        0
Economy (GDP per Capita)         0
Family                           0
Health (Life Expectancy)         0
Freedom                          0
Trust (Government Corruption)    0
Generosity                       0
Dystopia Residual                0
dtype: int64
```

[16]:
```
y16 = y16.drop_duplicates()
```

[17]:
```
y16 = y16.drop( columns= ['Happiness Rank','Dystopia Residual','Lower␣
↪Confidence Interval', 'Upper Confidence Interval'])
```

[18]:
```
# 2017 Analysis
```

[19]:
```
y17 = pd.read_csv("2017.csv")
```

[20]:
```
y17.head(5)
```

[20]:

	Country	Happiness.Rank	Happiness.Score	Whisker.high	Whisker.low	\
0	Norway	1	7.537	7.594445	7.479556	
1	Denmark	2	7.522	7.581728	7.462272	
2	Iceland	3	7.504	7.622030	7.385970	
3	Switzerland	4	7.494	7.561772	7.426227	
4	Finland	5	7.469	7.527542	7.410458	

	Economy..GDP.per.Capita.	Family	Health..Life.Expectancy.	Freedom	\
0	1.616463	1.533524	0.796667	0.635423	

```
1                  1.482383  1.551122                  0.792566  0.626007
2                  1.480633  1.610574                  0.833552  0.627163
3                  1.564980  1.516912                  0.858131  0.620071
4                  1.443572  1.540247                  0.809158  0.617951

   Generosity  Trust..Government.Corruption.  Dystopia.Residual
0    0.362012                       0.315964           2.277027
1    0.355280                       0.400770           2.313707
2    0.475540                       0.153527           2.322715
3    0.290549                       0.367007           2.276716
4    0.245483                       0.382612           2.430182
```

[21]:
```
y17.shape
```

[21]:
```
(155, 12)
```

[22]:
```
y17.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 155 entries, 0 to 154
Data columns (total 12 columns):
 #   Column                         Non-Null Count  Dtype
---  ------                         --------------  -----
 0   Country                        155 non-null    object
 1   Happiness.Rank                 155 non-null    int64
 2   Happiness.Score                155 non-null    float64
 3   Whisker.high                   155 non-null    float64
 4   Whisker.low                    155 non-null    float64
 5   Economy..GDP.per.Capita.       155 non-null    float64
 6   Family                         155 non-null    float64
 7   Health..Life.Expectancy.       155 non-null    float64
 8   Freedom                        155 non-null    float64
 9   Generosity                     155 non-null    float64
 10  Trust..Government.Corruption.  155 non-null    float64
 11  Dystopia.Residual              155 non-null    float64
dtypes: float64(10), int64(1), object(1)
memory usage: 14.7+ KB
```

[23]:
```
y17.isnull().sum()
```

[23]:
```
Country                        0
Happiness.Rank                 0
Happiness.Score                0
Whisker.high                   0
Whisker.low                    0
Economy..GDP.per.Capita.       0
Family                         0
Health..Life.Expectancy.       0
```

```
Freedom                          0
Generosity                       0
Trust..Government.Corruption.    0
Dystopia.Residual                0
dtype: int64
```

[24]:
```
y17 = y17.drop_duplicates()
```

[25]:
```
y17 = y17.drop( columns= ['Happiness.Rank','Dystopia.Residual','Whisker.
↪high','Whisker.low'])
```

[26]:
```
# 2018 Analysis
```

[27]:
```
y18 = pd.read_csv("2018.csv")
```

[28]:
```
y18.head(5)
```

[28]:

	Overall rank	Country or region	Score	GDP per capita	Social support	Healthy life expectancy	Freedom to make life choices	Generosity	Perceptions of corruption
0	1	Finland	7.632	1.305	1.592	0.874	0.681	0.202	0.393
1	2	Norway	7.594	1.456	1.582	0.861	0.686	0.286	0.340
2	3	Denmark	7.555	1.351	1.590	0.868	0.683	0.284	0.408
3	4	Iceland	7.495	1.343	1.644	0.914	0.677	0.353	0.138
4	5	Switzerland	7.487	1.420	1.549	0.927	0.660	0.256	0.357

[29]:
```
y18.shape
```

[29]:
```
(156, 9)
```

[30]:
```
y18.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 156 entries, 0 to 155
Data columns (total 9 columns):
 #   Column                        Non-Null Count  Dtype
```

```
---  ------                          --------------  -----
 0   Overall rank                    156 non-null    int64
 1   Country or region               156 non-null    object
 2   Score                           156 non-null    float64
 3   GDP per capita                  156 non-null    float64
 4   Social support                  156 non-null    float64
 5   Healthy life expectancy         156 non-null    float64
 6   Freedom to make life choices    156 non-null    float64
 7   Generosity                      156 non-null    float64
 8   Perceptions of corruption       155 non-null    float64
dtypes: float64(7), int64(1), object(1)
memory usage: 11.1+ KB
```

[31]:
```
y18.isnull().sum()
```

[31]:
```
Overall rank                    0
Country or region               0
Score                           0
GDP per capita                  0
Social support                  0
Healthy life expectancy         0
Freedom to make life choices    0
Generosity                      0
Perceptions of corruption       1
dtype: int64
```

[32]:
```
m = y18["Perceptions of corruption"].mean()
y18["Perceptions of corruption"] = y18["Perceptions of corruption"].fillna(m)
```

[33]:
```
y18 = y18.drop_duplicates()
```

[34]:
```
y18 = y18.drop(columns= ['Overall rank'])
```

[35]:
```
# 2019 Analysis
```

[36]:
```
y19 = pd.read_csv("2019.csv")
```

[37]:
```
y19.head(5)
```

[37]:

	Overall rank	Country or region	Score	GDP per capita	Social support	\
0	1	Finland	7.769	1.340	1.587	
1	2	Denmark	7.600	1.383	1.573	
2	3	Norway	7.554	1.488	1.582	
3	4	Iceland	7.494	1.380	1.624	
4	5	Netherlands	7.488	1.396	1.522	

	Healthy life expectancy	Freedom to make life choices	Generosity	\
0	0.986	0.596	0.153	

```
1                    0.996                          0.592       0.252
2                    1.028                          0.603       0.271
3                    1.026                          0.591       0.354
4                    0.999                          0.557       0.322

   Perceptions of corruption
0                      0.393
1                      0.410
2                      0.341
3                      0.118
4                      0.298
```

[38]:
```
y19.shape
```

[38]:
```
(156, 9)
```

[39]:
```
y19.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 156 entries, 0 to 155
Data columns (total 9 columns):
 #   Column                          Non-Null Count  Dtype
---  ------                          --------------  -----
 0   Overall rank                    156 non-null    int64
 1   Country or region               156 non-null    object
 2   Score                           156 non-null    float64
 3   GDP per capita                  156 non-null    float64
 4   Social support                  156 non-null    float64
 5   Healthy life expectancy         156 non-null    float64
 6   Freedom to make life choices    156 non-null    float64
 7   Generosity                      156 non-null    float64
 8   Perceptions of corruption       156 non-null    float64
dtypes: float64(7), int64(1), object(1)
memory usage: 11.1+ KB
```

[40]:
```
y19.isnull().sum()
```

[40]:
```
Overall rank                    0
Country or region               0
Score                           0
GDP per capita                  0
Social support                  0
Healthy life expectancy         0
Freedom to make life choices    0
Generosity                      0
Perceptions of corruption       0
dtype: int64
```

[41]:
```
y19 = y19.drop(columns= ['Overall rank'])
```

[42]:
```
y17.rename(columns={"Happiness.Score" : "Happiness Score", "Economy..GDP.per.
↪Capita.":"Economy (GDP per Capita)", "Trust..Government.Corruption.":"Trust␣
↪(Government Corruption)",  "Health..Life.Expectancy." : "Health (Life␣
↪Expectancy)"}, inplace= True)
```

[43]:
```
y18.rename(columns={"Score":"Happiness Score", "Country or region":
↪"Country","Freedom to make life choices":"Freedom", "Healthy life␣
↪expectancy":"Health (Life Expectancy)", "GDP per capita":"Economy (GDP per␣
↪Capita)", "Social support":"Family", "Perceptions of corruption":"Trust␣
↪(Government Corruption)"}, inplace= True)
```

[44]:
```
y19.rename(columns={"Score":"Happiness Score", "Country or region":
↪"Country","Freedom to make life choices":"Freedom", "Healthy life␣
↪expectancy":"Health (Life Expectancy)", "GDP per capita":"Economy (GDP per␣
↪Capita)", "Social support":"Family", "Perceptions of corruption":"Trust␣
↪(Government Corruption)"}, inplace= True)
```

[45]:
```
y15["Year"]= "2015"
y16["Year"]= "2016"
y17["Year"]= "2017"
y18["Year"]= "2018"
y19["Year"]= "2019"
```

[46]:
```
y_overall = pd.concat([y15,y16,y17,y18,y19])
y_overall
```

[46]:

	Country	Region	Happiness Score	\
0	Switzerland	Western Europe	7.587	
1	Iceland	Western Europe	7.561	
2	Denmark	Western Europe	7.527	
3	Norway	Western Europe	7.522	
4	Canada	North America	7.427	
..	…	…	…	
151	Rwanda	NaN	3.334	
152	Tanzania	NaN	3.231	
153	Afghanistan	NaN	3.203	
154	Central African Republic	NaN	3.083	
155	South Sudan	NaN	2.853	

	Economy (GDP per Capita)	Family	Health (Life Expectancy)	Freedom	\
0	1.39651	1.34951	0.94143	0.66557	
1	1.30232	1.40223	0.94784	0.62877	
2	1.32548	1.36058	0.87464	0.64938	
3	1.45900	1.33095	0.88521	0.66973	
4	1.32629	1.32261	0.90563	0.63297	

```
..                        …       …                        …       …
151                   0.35900  0.71100                  0.61400  0.55500
152                   0.47600  0.88500                  0.49900  0.41700
153                   0.35000  0.51700                  0.36100  0.00000
154                   0.02600  0.00000                  0.10500  0.22500
155                   0.30600  0.57500                  0.29500  0.01000

     Trust (Government Corruption)  Generosity  Year
0                          0.41978     0.29678  2015
1                          0.14145     0.43630  2015
2                          0.48357     0.34139  2015
3                          0.36503     0.34699  2015
4                          0.32957     0.45811  2015
..                             …         …    …
151                        0.41100     0.21700  2019
152                        0.14700     0.27600  2019
153                        0.02500     0.15800  2019
154                        0.03500     0.23500  2019
155                        0.09100     0.20200  2019

[782 rows x 10 columns]
```

[47]:
```
y_overall.isna().sum()
```

[47]:
```
Country                          0
Region                         467
Happiness Score                  0
Economy (GDP per Capita)         0
Family                           0
Health (Life Expectancy)         0
Freedom                          0
Trust (Government Corruption)    0
Generosity                       0
Year                             0
dtype: int64
```

[48]:
```
grouped = y_overall.groupby('Country')
y_overall['Region'] = grouped['Region'].ffill( )
y_overall['Region'] = grouped['Region'].bfill( )
```

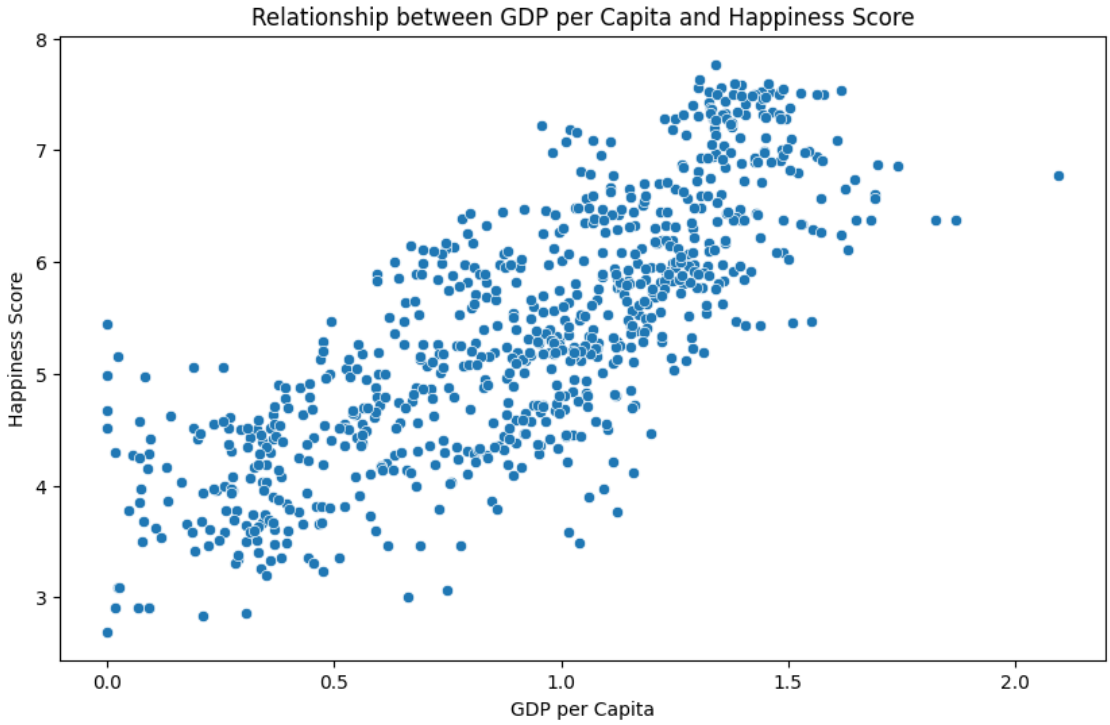
[49]:
```
missing = y_overall["Region"].isna()
indices = y_overall[missing].index
print(indices)
```

```
Index([32, 70, 37, 57, 38, 63, 83, 119], dtype='int64')
```

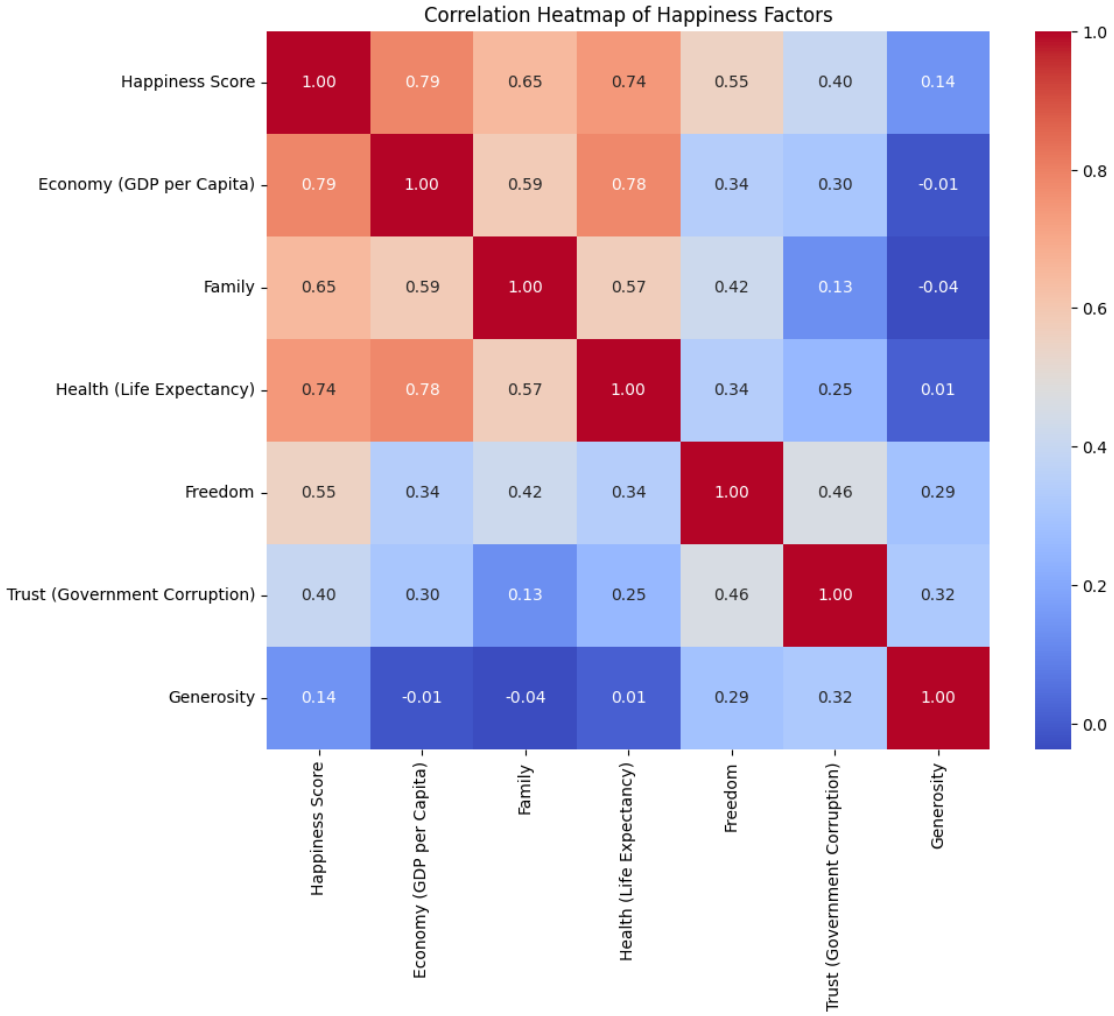
[50]:
```
y_overall.describe()
```

[50]:
```
       Happiness Score  Economy (GDP per Capita)      Family  \
count       782.000000                782.000000  782.000000
mean          5.379018                  0.916047    1.078392
std           1.127456                  0.407340    0.329548
min           2.693000                  0.000000    0.000000
25%           4.509750                  0.606500    0.869363
50%           5.322000                  0.982205    1.124735
75%           6.189500                  1.236187    1.327250
max           7.769000                  2.096000    1.644000

       Health (Life Expectancy)     Freedom  Trust (Government Corruption)  \
count                782.000000  782.000000                     782.000000
mean                   0.612416    0.411091                       0.125418
std                    0.248309    0.152880                       0.105750
min                    0.000000    0.000000                       0.000000
25%                    0.440183    0.309768                       0.054250
50%                    0.647310    0.431000                       0.091033
75%                    0.808000    0.531000                       0.155861
max                    1.141000    0.724000                       0.551910

       Generosity
count  782.000000
mean     0.218576
std      0.122321
min      0.000000
25%      0.130000
50%      0.201982
75%      0.278832
max      0.838075
```

[51]:
```
plt.figure(figsize=(10, 6))
sns.scatterplot(x='Economy (GDP per Capita)', y='Happiness Score',␣
 ↪data=y_overall)
plt.title("Relationship between GDP per Capita and Happiness Score")
plt.xlabel("GDP per Capita")
plt.ylabel("Happiness Score")
plt.show()
```

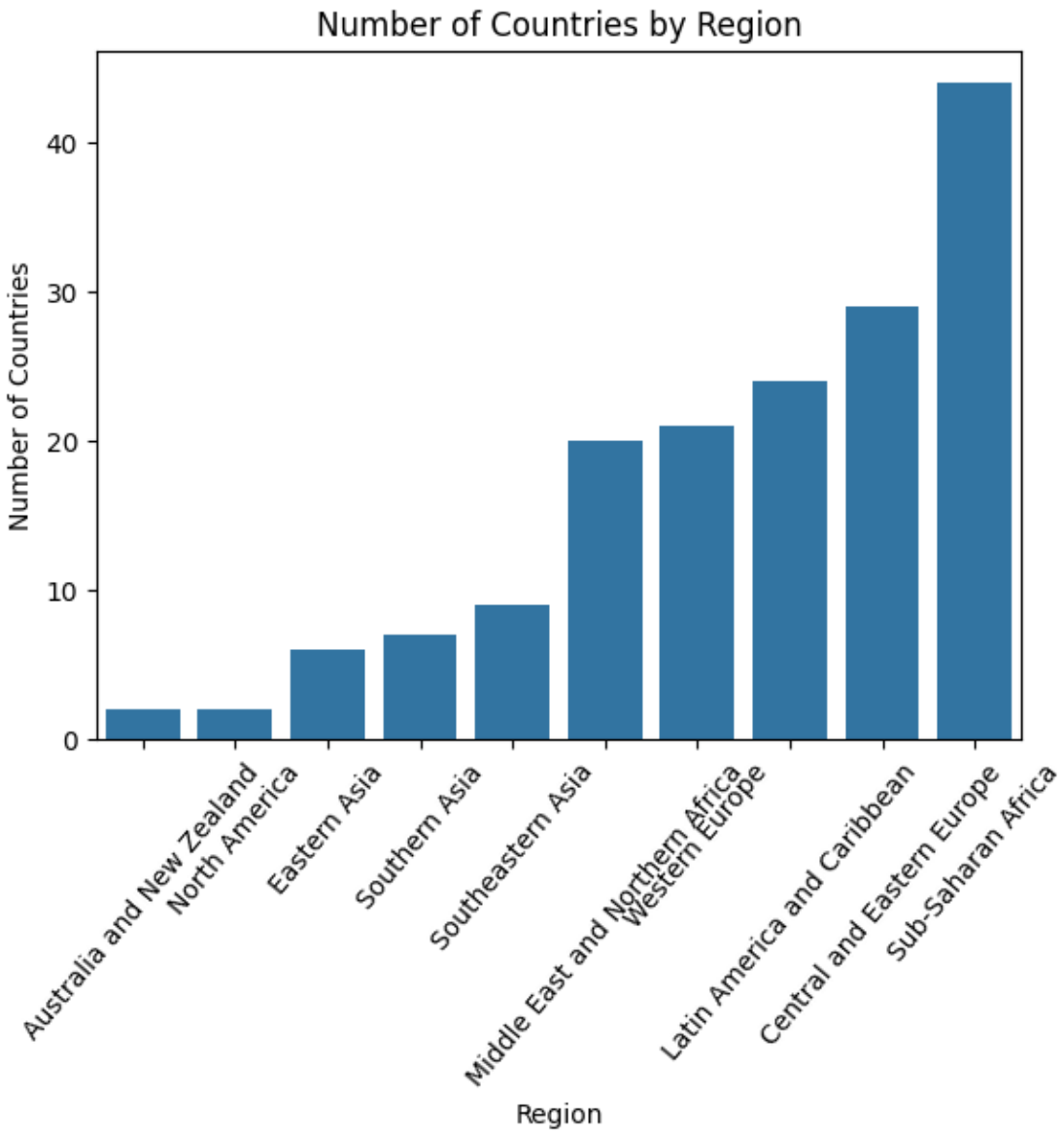


[52]:
```
number_df = y_overall.select_dtypes(include=[np.number])

# Calculating correlations
correlation = number_df.corr()

# Plotting the heatmap
plt.figure(figsize=(10, 8))
sns.heatmap(correlation, annot=True, cmap='coolwarm', fmt=".2f")
plt.title("Correlation Heatmap of Happiness Factors")
plt.show()
```



[53]:
```
df_count = y_overall.groupby('Region')['Country'].nunique().
  ↪reset_index(name='num')
df_count = df_count.sort_values( by="num")
sns.barplot(df_count, x='Region', y ='num')
plt.xticks(rotation=50)
plt.xlabel('Region')
plt.ylabel('Number of Countries')
plt.title('Number of Countries by Region')
```

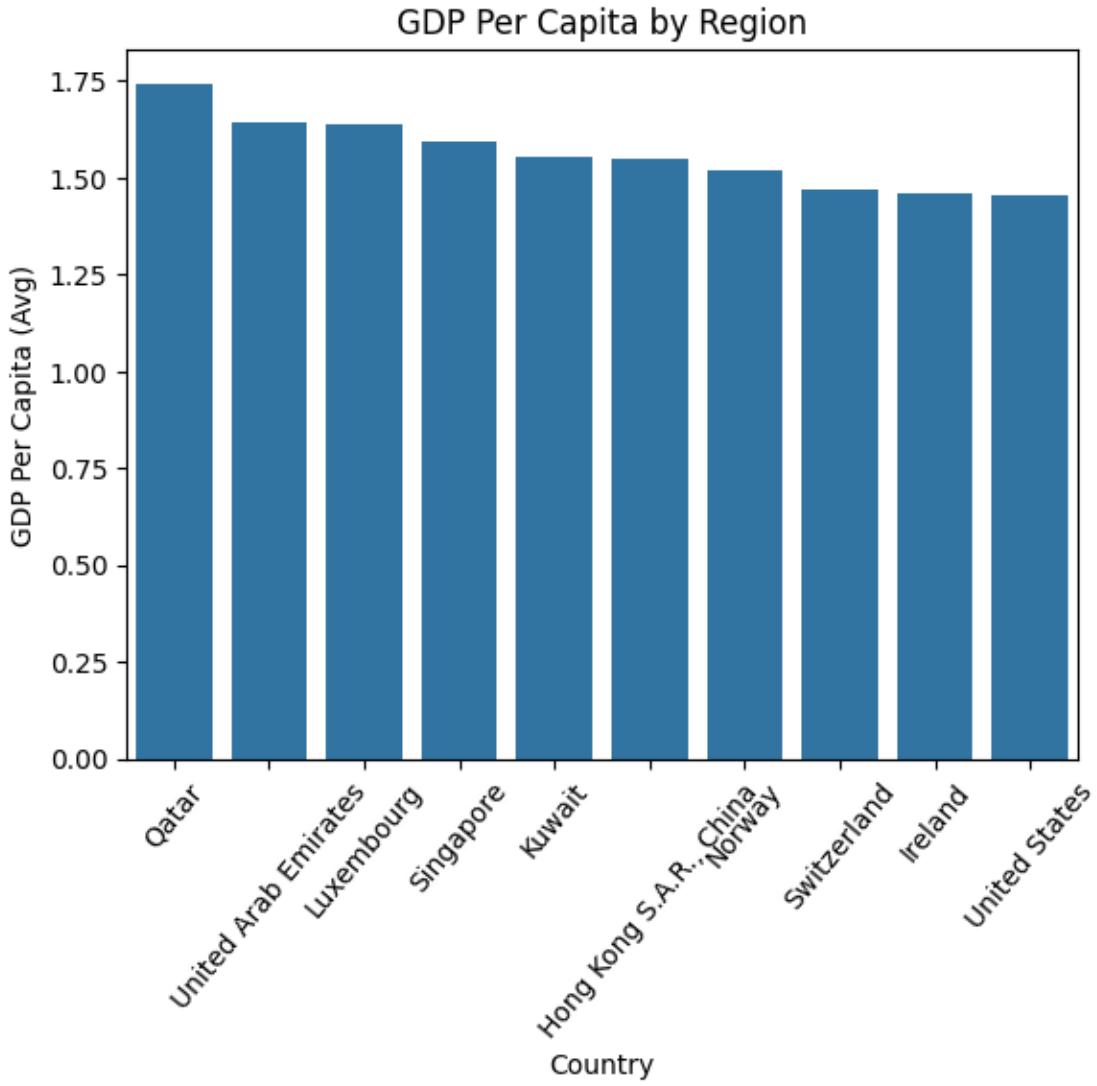
[53]: Text(0.5, 1.0, 'Number of Countries by Region')



[55]:
```
Avg_gdp = y_overall.groupby('Country')['Economy (GDP per Capita)'].mean()
Avg_gdp= Avg_gdp.sort_values(ascending=False).iloc[0:10]
sns.barplot(Avg_gdp)
plt.title('GDP Per Capita by Region')
plt.xlabel('Country')
plt.ylabel('GDP Per Capita (Avg)')
plt.xticks(rotation= 50)
```

[55]:
```
([0, 1, 2, 3, 4, 5, 6, 7, 8, 9],
 [Text(0, 0, 'Qatar'),
  Text(1, 0, 'United Arab Emirates'),
```

```
Text(2, 0, 'Luxembourg'),
Text(3, 0, 'Singapore'),
Text(4, 0, 'Kuwait'),
Text(5, 0, 'Hong Kong S.A.R., China'),
Text(6, 0, 'Norway'),
Text(7, 0, 'Switzerland'),
Text(8, 0, 'Ireland'),
Text(9, 0, 'United States')])
```



[56]:
```
reg_year = y_overall.pivot_table(index=['Region','Year'],values='Happiness␣
↪Score',aggfunc='mean')
reg_year = reg_year.reset_index()
reg_year
```

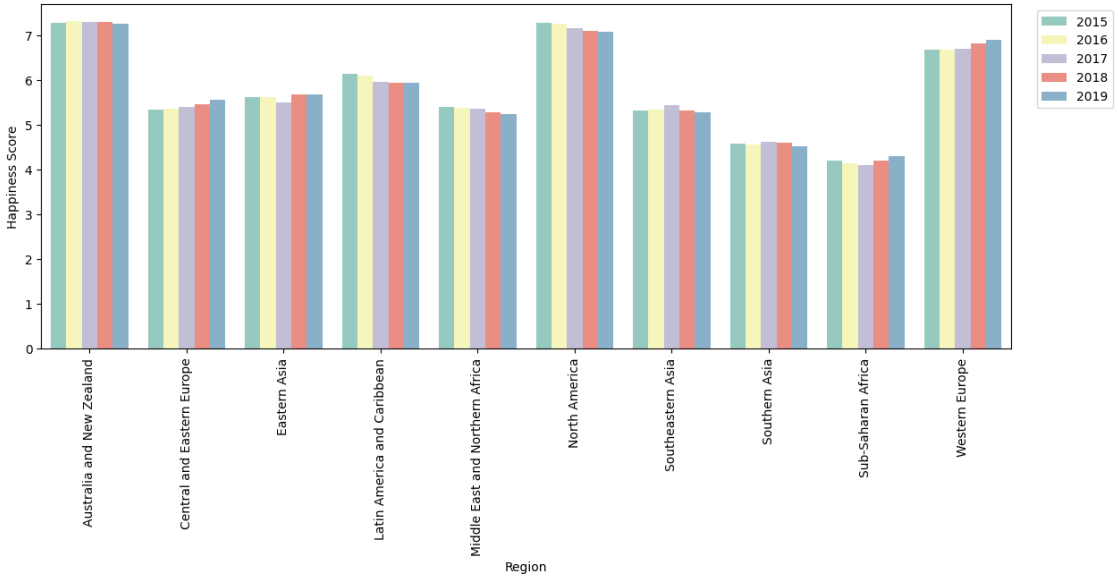
[56]:

	Region	Year	Happiness Score
0	Australia and New Zealand	2015	7.285000
1	Australia and New Zealand	2016	7.323500
2	Australia and New Zealand	2017	7.299000
3	Australia and New Zealand	2018	7.298000
4	Australia and New Zealand	2019	7.267500
5	Central and Eastern Europe	2015	5.332931
6	Central and Eastern Europe	2016	5.370690
7	Central and Eastern Europe	2017	5.409931
8	Central and Eastern Europe	2018	5.463966
9	Central and Eastern Europe	2019	5.571786
10	Eastern Asia	2015	5.626167
11	Eastern Asia	2016	5.624167
12	Eastern Asia	2017	5.496500
13	Eastern Asia	2018	5.672000
14	Eastern Asia	2019	5.688833
15	Latin America and Caribbean	2015	6.144682
16	Latin America and Caribbean	2016	6.101750
17	Latin America and Caribbean	2017	5.957818
18	Latin America and Caribbean	2018	5.938619
19	Latin America and Caribbean	2019	5.942550
20	Middle East and Northern Africa	2015	5.406900
21	Middle East and Northern Africa	2016	5.386053
22	Middle East and Northern Africa	2017	5.369684
23	Middle East and Northern Africa	2018	5.282737
24	Middle East and Northern Africa	2019	5.237000
25	North America	2015	7.273000
26	North America	2016	7.254000
27	North America	2017	7.154500
28	North America	2018	7.107000
29	North America	2019	7.085000
30	Southeastern Asia	2015	5.317444
31	Southeastern Asia	2016	5.338889
32	Southeastern Asia	2017	5.444875
33	Southeastern Asia	2018	5.313444
34	Southeastern Asia	2019	5.273667
35	Southern Asia	2015	4.580857
36	Southern Asia	2016	4.563286
37	Southern Asia	2017	4.628429
38	Southern Asia	2018	4.603857
39	Southern Asia	2019	4.526857
40	Sub-Saharan Africa	2015	4.202800
41	Sub-Saharan Africa	2016	4.136421
42	Sub-Saharan Africa	2017	4.111949
43	Sub-Saharan Africa	2018	4.195026
44	Sub-Saharan Africa	2019	4.294513
45	Western Europe	2015	6.689619

```
46                  Western Europe  2016         6.685667
47                  Western Europe  2017         6.703714
48                  Western Europe  2018         6.829100
49                  Western Europe  2019         6.898400
```

[57]:
```
plt.figure(figsize=(14,5))
g1 = sns.barplot(data=reg_year,x='Region',y='Happiness␣
 ↪Score',hue='Year',palette='Set3')
g1.set_xticklabels(g1.get_xticklabels(),rotation=90)
plt.legend(bbox_to_anchor=(1.02, 1),loc="upper left")
plt.show()
```

```
C:\Users\sapna\AppData\Local\Temp\ipykernel_8584\1993896136.py:3: UserWarning:
set_ticklabels() should only be used with a fixed number of ticks, i.e. after
set_ticks() or using a FixedLocator.
  g1.set_xticklabels(g1.get_xticklabels(),rotation=90)
```

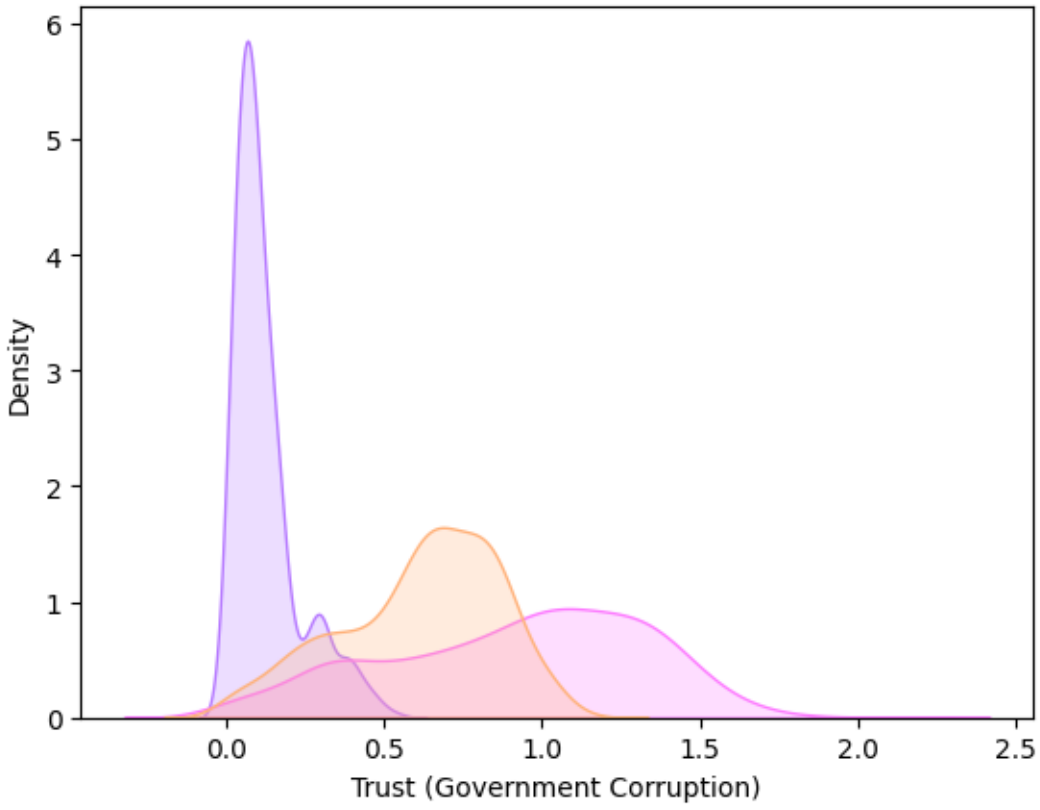


[58]:
```
color=["#B77AFF","#FD7AFF","#FFB27A","#A9FF7A","#7AFFD4","#FF7A7A"]
sns.kdeplot(y_overall['Trust (Government␣
 ↪Corruption)'],shade=True,color=color[0],label='Trust (Government␣
 ↪Corruption)')
sns.kdeplot(y_overall['Economy (GDP per␣
 ↪Capita)'],shade=True,color=color[1],label='Economy (GDP per Capita)')
sns.kdeplot(y_overall['Health (Life␣
 ↪Expectancy)'],shade=True,color=color[2],label='Health (Life Expectancy)')
sns.kdeplot(df['Freedom'],shade=True,color=color[3],label='Freedom')
sns.kdeplot(df['Generosity'],shade=True,color=color[4],label='Generosity')
```

```
plt.xlabel('factores')
plt.legend(fontsize=8)
```

```
C:\Users\sapna\AppData\Local\Temp\ipykernel_8584\1153193609.py:2: FutureWarning:

`shade` is now deprecated in favor of `fill`; setting `fill=True`.
This will become an error in seaborn v0.14.0; please update your code.

  sns.kdeplot(y_overall['Trust (Government
Corruption)'],shade=True,color=color[0],label='Trust (Government Corruption)')
C:\Users\sapna\AppData\Local\Temp\ipykernel_8584\1153193609.py:3: FutureWarning:

`shade` is now deprecated in favor of `fill`; setting `fill=True`.
This will become an error in seaborn v0.14.0; please update your code.

  sns.kdeplot(y_overall['Economy (GDP per
Capita)'],shade=True,color=color[1],label='Economy (GDP per Capita)')
C:\Users\sapna\AppData\Local\Temp\ipykernel_8584\1153193609.py:4: FutureWarning:

`shade` is now deprecated in favor of `fill`; setting `fill=True`.
This will become an error in seaborn v0.14.0; please update your code.

  sns.kdeplot(y_overall['Health (Life
Expectancy)'],shade=True,color=color[2],label='Health (Life Expectancy)')
```

```
---------------------------------------------------------------------------
NameError                                 Traceback (most recent call last)
Cell In[58], line 5
      3 sns.kdeplot(y_overall['Economy (GDP per␣
 ↪Capita)'],shade=True,color=color[1],label='Economy (GDP per Capita)')
      4 sns.kdeplot(y_overall['Health (Life␣
 ↪Expectancy)'],shade=True,color=color[2],label='Health (Life Expectancy)')
----> 5 sns.kdeplot(df['Freedom'],shade=True,color=color[3],label='Freedom')
      6 sns.
 ↪kdeplot(df['Generosity'],shade=True,color=color[4],label='Generosity')
      8 plt.xlabel('factores')

NameError: name 'df' is not defined
```


Density
Trust (Government Corruption)
0
1
2
3
4
5
6
0.0
0.5
1.0
1.5
2.0
2.5

[ ]:

[ ]:

[ ]: