

World Happiness Report 2024

June 22, 2025

```
[44]: import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
```

```
[46]: df = pd.read_csv('../data/WHR2024.csv')
```

```
[82]: # Remove all rows where 'Country name' is 'Israel'
df = df[df['Country name'] != 'Israel']

print(df.head())
print(df.info())
```

	Country name	Ladder score	upperwhisker	lowerwhisker	\
0	Finland	7.741	7.815	7.667	
1	Denmark	7.583	7.665	7.500	
2	Iceland	7.525	7.618	7.433	
3	Sweden	7.344	7.422	7.267	
5	Netherlands	7.319	7.383	7.256	

	Explained by: Log GDP per capita	Explained by: Social support	\
0	1.844	1.572	
1	1.908	1.520	
2	1.881	1.617	
3	1.878	1.501	
5	1.901	1.462	

	Explained by: Healthy life expectancy	\
0	0.695	
1	0.699	
2	0.718	
3	0.724	
5	0.706	

	Explained by: Freedom to make life choices	Explained by: Generosity	\
0	0.859	0.142	
1	0.823	0.204	
2	0.819	0.258	

3	0.838	0.221
5	0.725	0.247

	Explained by: Perceptions of corruption	Dystopia + residual	Region
0	0.546	2.082	Europe
1	0.548	1.881	Europe
2	0.182	2.050	Europe
3	0.524	1.658	Europe
5	0.372	1.906	Europe

<class 'pandas.core.frame.DataFrame'>

Index: 142 entries, 0 to 142

Data columns (total 12 columns):

#	Column	Non-Null Count	Dtype
0	Country name	142 non-null	object
1	Ladder score	142 non-null	float64
2	upperwhisker	142 non-null	float64
3	lowerwhisker	142 non-null	float64
4	Explained by: Log GDP per capita	139 non-null	float64
5	Explained by: Social support	139 non-null	float64
6	Explained by: Healthy life expectancy	139 non-null	float64
7	Explained by: Freedom to make life choices	139 non-null	float64
8	Explained by: Generosity	139 non-null	float64
9	Explained by: Perceptions of corruption	139 non-null	float64
10	Dystopia + residual	139 non-null	float64
11	Region	142 non-null	object

dtypes: float64(10), object(2)

memory usage: 14.4+ KB

None

```
[86]: # Create a mapping dictionary for countries and their regions
region_mapping = {
    # Europe
    'Finland': 'Europe', 'Denmark': 'Europe', 'Iceland': 'Europe', 'Sweden': 'Europe',
    'Netherlands': 'Europe', 'Norway': 'Europe', 'Luxembourg': 'Europe',
    'Switzerland': 'Europe',
    'Austria': 'Europe', 'Belgium': 'Europe', 'Ireland': 'Europe', 'Czechia': 'Europe',
    'Lithuania': 'Europe', 'United Kingdom': 'Europe', 'Slovenia': 'Europe',
    'France': 'Europe',
    'Kosovo': 'Europe', 'Romania': 'Europe', 'Estonia': 'Europe', 'Poland': 'Europe',
    'Spain': 'Europe', 'Serbia': 'Europe', 'Malta': 'Europe', 'Italy': 'Europe',
    'Slovakia': 'Europe', 'Latvia': 'Europe', 'Cyprus': 'Europe', 'Portugal': 'Europe',
}
```

```

    'Hungary': 'Europe', 'Croatia': 'Europe', 'Greece': 'Europe', 'Bosnia and
↪Herzegovina': 'Europe',
    'Moldova': 'Europe', 'Montenegro': 'Europe', 'Bulgaria': 'Europe', 'North
↪Macedonia': 'Europe',
    'Albania': 'Europe', 'Ukraine': 'Europe',

# Middle East
    'Kuwait': 'Middle East', 'Saudi Arabia': 'Middle East',
    'United Arab Emirates': 'Middle East', 'Bahrain': 'Middle East', 'Iraq':
↪'Middle East',
    'Iran': 'Middle East', 'State of Palestine': 'Middle East', 'Jordan':
↪'Middle East',
    'Yemen': 'Middle East', 'Lebanon': 'Middle East',

# Asia
    'Singapore': 'Asia', 'Taiwan Province of China': 'Asia', 'Uzbekistan':
↪'Asia',
    'Kazakhstan': 'Asia', 'Japan': 'Asia', 'South Korea': 'Asia', 'Philippines':
↪'Asia',
    'Vietnam': 'Asia', 'Thailand': 'Asia', 'Malaysia': 'Asia', 'China': 'Asia',
    'Kyrgyzstan': 'Asia', 'Mongolia': 'Asia', 'Armenia': 'Asia', 'Georgia':
↪'Asia',
    'Nepal': 'Asia', 'Laos': 'Asia', 'Azerbaijan': 'Asia', 'Pakistan': 'Asia',
    'Myanmar': 'Asia', 'Cambodia': 'Asia', 'India': 'Asia', 'Sri Lanka': 'Asia',
    'Bangladesh': 'Asia', 'Hong Kong S.A.R. of China': 'Asia', 'Tajikistan':
↪'Asia',
    'Indonesia': 'Asia',

# Latin America
    'Costa Rica': 'Latin America', 'Mexico': 'Latin America', 'Uruguay': 'Latin
↪America',
    'El Salvador': 'Latin America', 'Chile': 'Latin America', 'Panama': 'Latin
↪America',
    'Guatemala': 'Latin America', 'Nicaragua': 'Latin America', 'Brazil': 'Latin
↪America',
    'Argentina': 'Latin America', 'Paraguay': 'Latin America', 'Honduras':
↪'Latin America',
    'Jamaica': 'Latin America', 'Peru': 'Latin America', 'Dominican Republic':
↪'Latin America',
    'Bolivia': 'Latin America', 'Ecuador': 'Latin America', 'Colombia': 'Latin
↪America',
    'Venezuela': 'Latin America',

# Africa
    'Libya': 'Africa', 'Mauritius': 'Africa', 'South Africa': 'Africa',
    'Algeria': 'Africa', 'Congo (Brazzaville)': 'Africa', 'Mozambique': 'Africa',

```

```

    'Gabon': 'Africa', 'Ivory Coast': 'Africa', 'Guinea': 'Africa', 'Nigeria':␣
↪ 'Africa',
    'Cameroon': 'Africa', 'Namibia': 'Africa', 'Morocco': 'Africa', 'Niger':␣
↪ 'Africa',
    'Burkina Faso': 'Africa', 'Mauritania': 'Africa', 'Gambia': 'Africa', 'Chad':
↪ 'Africa',
    'Kenya': 'Africa', 'Tunisia': 'Africa', 'Benin': 'Africa', 'Uganda':␣
↪ 'Africa',
    'Ghana': 'Africa', 'Liberia': 'Africa', 'Mali': 'Africa', 'Madagascar':␣
↪ 'Africa',
    'Togo': 'Africa', 'Ethiopia': 'Africa', 'Tanzania': 'Africa', 'Comoros':␣
↪ 'Africa',
    'Zambia': 'Africa', 'Eswatini': 'Africa', 'Malawi': 'Africa', 'Botswana':␣
↪ 'Africa',
    'Zimbabwe': 'Africa', 'Congo (Kinshasa)': 'Africa', 'Sierra Leone': 'Africa',
    'Lesotho': 'Africa', 'Senegal': 'Africa', 'Egypt': 'Africa',

    # Oceania
    'Australia': 'Oceania', 'New Zealand': 'Oceania',

    # Mixed region
    'Russia': 'Europe/Asia', 'Turkiye': 'Europe/Asia'
}

# Add a new 'Region' column based on the mapping
df['Region'] = df['Country name'].map(region_mapping)

# Replace missing region values with 'Other'
df['Region'] = df['Region'].fillna('Other')

# Print the first rows of the dataframe to verify the new Region column
print(df[['Country name', 'Region']].head())

```

	Country name	Region
0	Finland	Europe
1	Denmark	Europe
2	Iceland	Europe
3	Sweden	Europe
5	Netherlands	Europe

```
[98]: # Compute mean happiness score by region
mean_scores = df.groupby('Region')['Ladder score'].mean().
    ↪sort_values(ascending=False)
print(mean_scores)
```

```
Region
Oceania      7.043000
Europe       6.466026
Latin America 6.143368
Other        5.516250
Asia         5.412481
Europe/Asia  5.380000
Middle East  5.165900
Africa       4.399075
Name: Ladder score, dtype: float64
```

```
[104]: # Average Happiness Score by Region
mean_scores.plot(kind='barh', figsize=(8,5), color='skyblue')
plt.title('Average Happiness Score by Region')
plt.xlabel('Average Ladder Score')
plt.grid(axis='x', linestyle='--', alpha=0.7)
plt.show()
```



```
[112]: # Distribution of countries within each region
region_counts = df['Region'].value_counts()
print(region_counts)
```

```
Region
Africa      40
Europe      38
Asia        27
Latin America 19
Middle East  10
Other         4
Oceania       2
Europe/Asia   2
Name: count, dtype: int64
```

```
[108]: # Top country in each region
top_countries = df.loc[df.groupby('Region')['Ladder score'].idxmax()][['Region', 'Country name', 'Ladder score']]
print("Top countries by region:\n", top_countries)
```

```
Top countries by region:
```

	Region	Country name	Ladder score
65	Africa	Libya	5.866
29	Asia	Singapore	6.523
0	Europe	Finland	7.741
71	Europe/Asia	Russia	5.785
11	Latin America	Costa Rica	6.955
12	Middle East	Kuwait	6.951
9	Oceania	Australia	7.057
14	Other	Canada	6.900

```
[110]: # Bottom country in each region
bottom_countries = df.loc[df.groupby('Region')['Ladder score'].idxmin()][['Region', 'Country name', 'Ladder score']]
print("\nBottom countries by region:\n", bottom_countries)
```

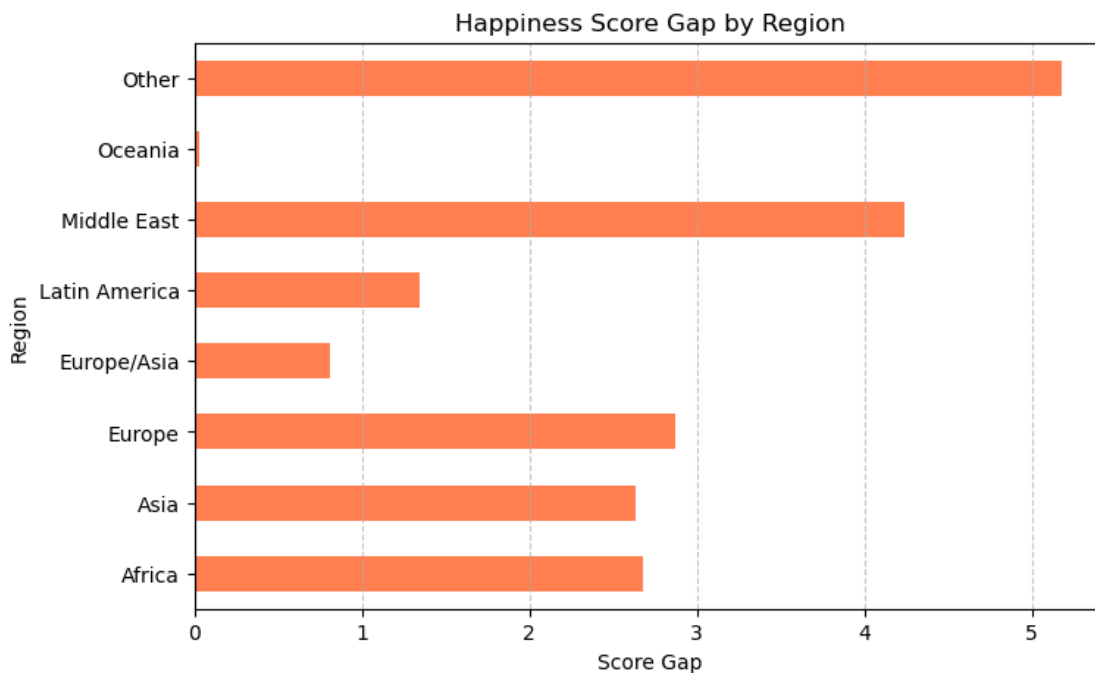
```
Bottom countries by region:
```

	Region	Country name	Ladder score
140	Africa	Lesotho	3.186
128	Asia	Bangladesh	3.886
104	Europe	Ukraine	4.873
97	Europe/Asia	Turkiye	4.975
78	Latin America	Venezuela	5.607
141	Middle East	Lebanon	2.707
10	Oceania	New Zealand	7.029
142	Other	Afghanistan	1.721

```
[120]: # The gap between the highest and lowest happiness scores in each region is
        ↪calculated
happiness_gap = df.groupby('Region')['Ladder score'].agg(['max', 'min'])
happiness_gap['Gap'] = happiness_gap['max'] - happiness_gap['min']

# The happiness score gap by region is plotted
happiness_gap['Gap'].plot(kind='barh', figsize=(8, 5), color='coral')
plt.title('Happiness Score Gap by Region')
plt.xlabel('Score Gap')
plt.grid(axis='x', linestyle='--', alpha=0.7)
plt.show()

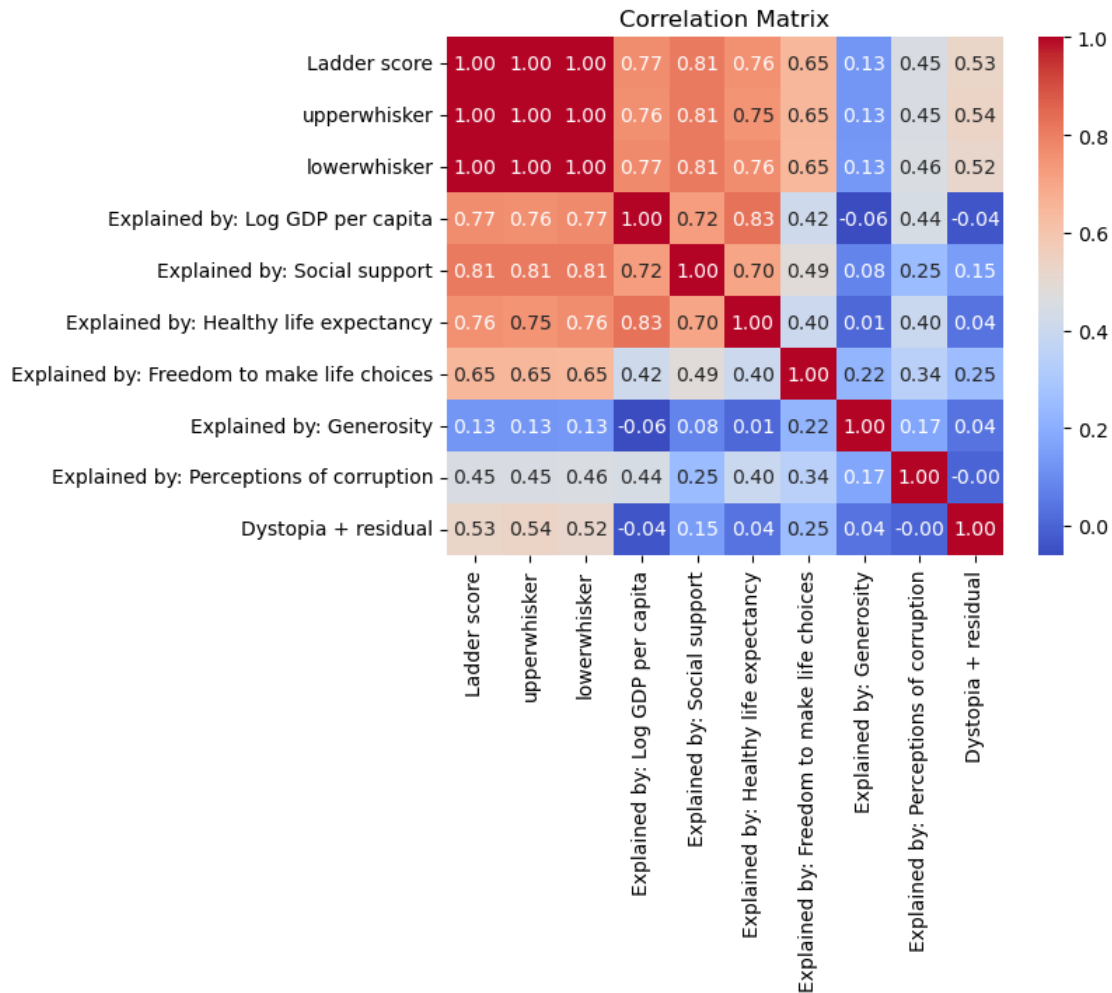
print("Happiness Score Gaps by Region:\n", happiness_gap)
```



Happiness Score Gaps by Region:

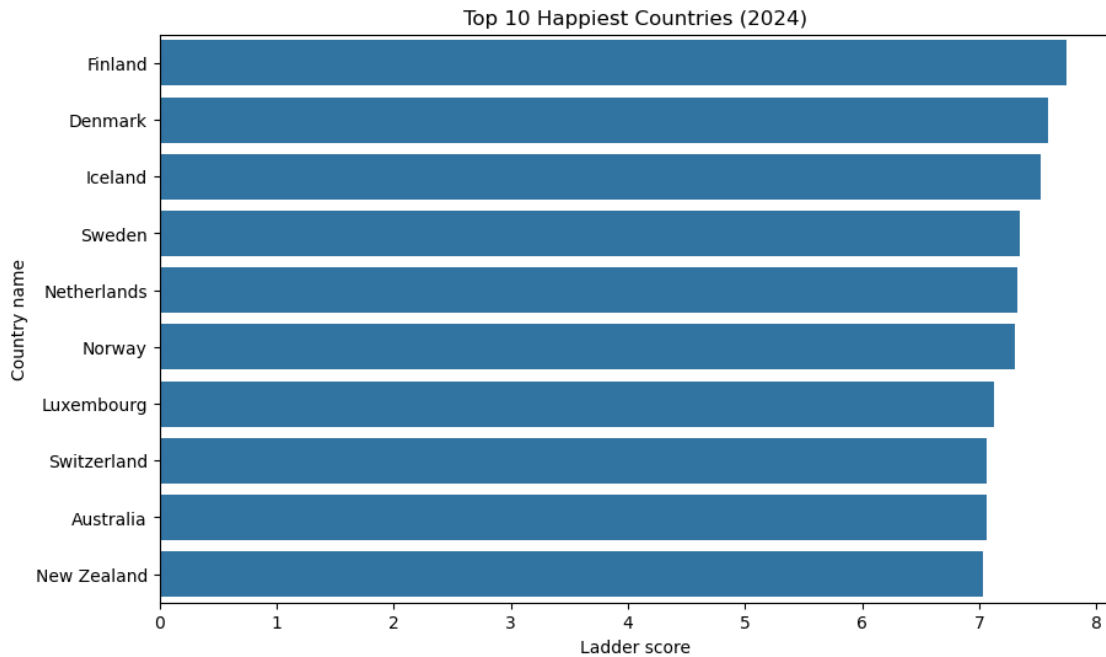
	max	min	Gap
Region			
Africa	5.866	3.186	2.680
Asia	6.523	3.886	2.637
Europe	7.741	4.873	2.868
Europe/Asia	5.785	4.975	0.810
Latin America	6.955	5.607	1.348
Middle East	6.951	2.707	4.244
Oceania	7.057	7.029	0.028
Other	6.900	1.721	5.179

```
[118]: # Heatmap of the correlation between variables
corr = df.select_dtypes(include='float64').corr()
sns.heatmap(corr, annot=True, fmt='.2f', cmap='coolwarm')
plt.title('Correlation Matrix')
plt.show()
```



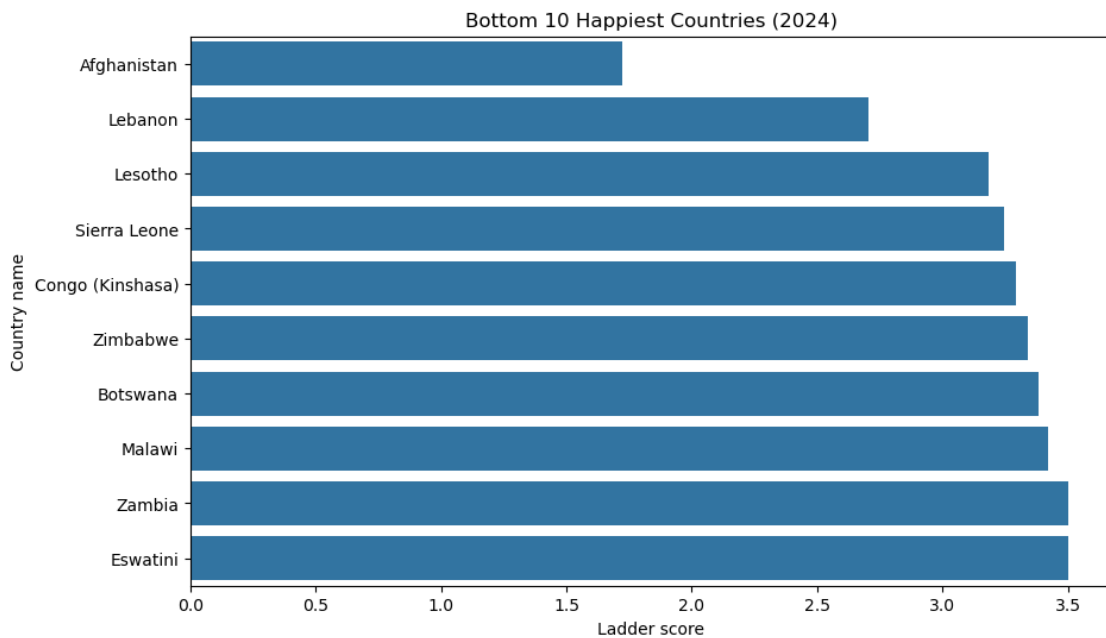
```
[88]: # Top 10 Happiest Countries (2024)
top10 = df.nlargest(10, 'Ladder score')
plt.figure(figsize=(10,6))
sns.barplot(x='Ladder score', y='Country name', data=top10)

plt.title('Top 10 Happiest Countries (2024)')
plt.show()
```

```
[90]: # Bottom 10 Happiest Countries (2024)
bottom10 = df.nsmallest(10, 'Ladder score')
plt.figure(figsize=(10,6))
sns.barplot(x='Ladder score', y='Country name', data=bottom10)

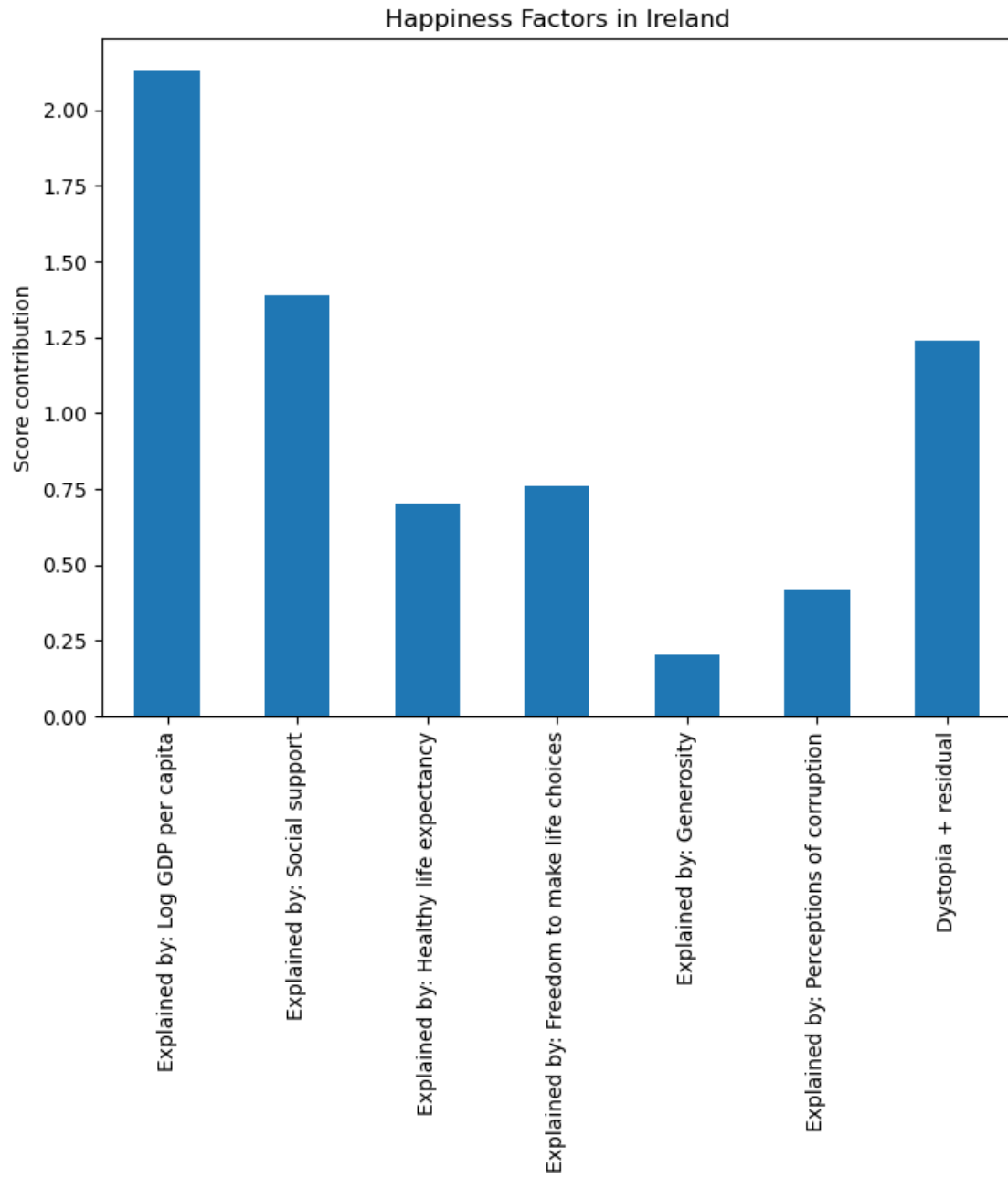
plt.title('Bottom 10 Happiest Countries (2024)')
plt.show()
```



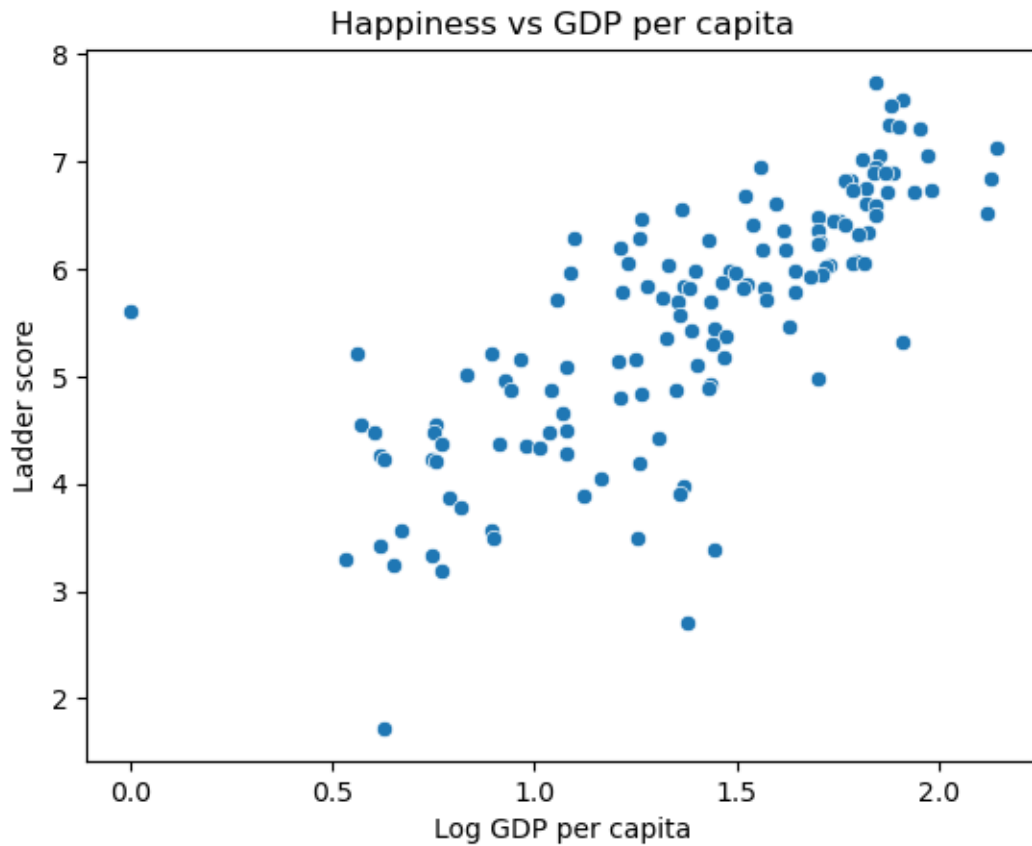
```
[92]: # Factor Contribution Analysis
country = 'Ireland'
df_country = df[df['Country name'] == country]
df_country_factors = df_country[['Explained by: Log GDP per capita',
                                   'Explained by: Social support',
                                   'Explained by: Healthy life expectancy',
                                   'Explained by: Freedom to make life choices',
                                   'Explained by: Generosity',
                                   'Explained by: Perceptions of corruption',
                                   'Dystopia + residual']].T

df_country_factors.columns = ['Score']

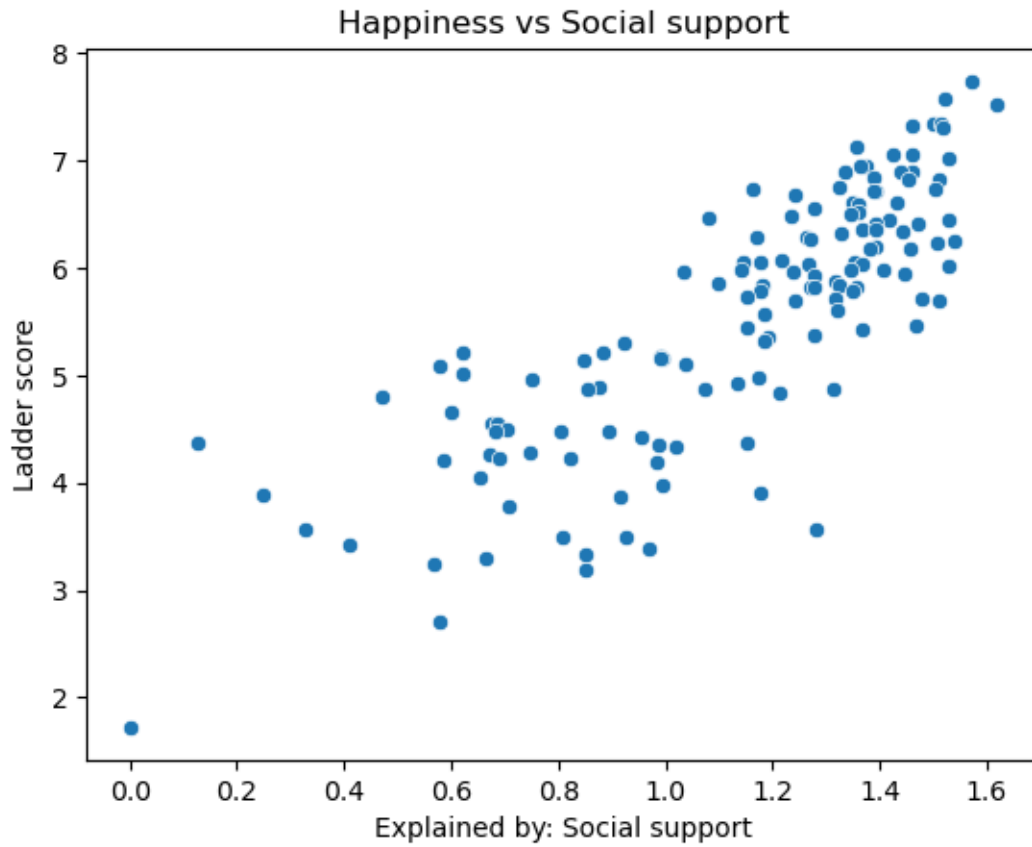
df_country_factors.plot(kind='bar', legend=False, figsize=(8,6))
plt.title(f'Happiness Factors in {country}')
plt.ylabel('Score contribution')
plt.show()
```



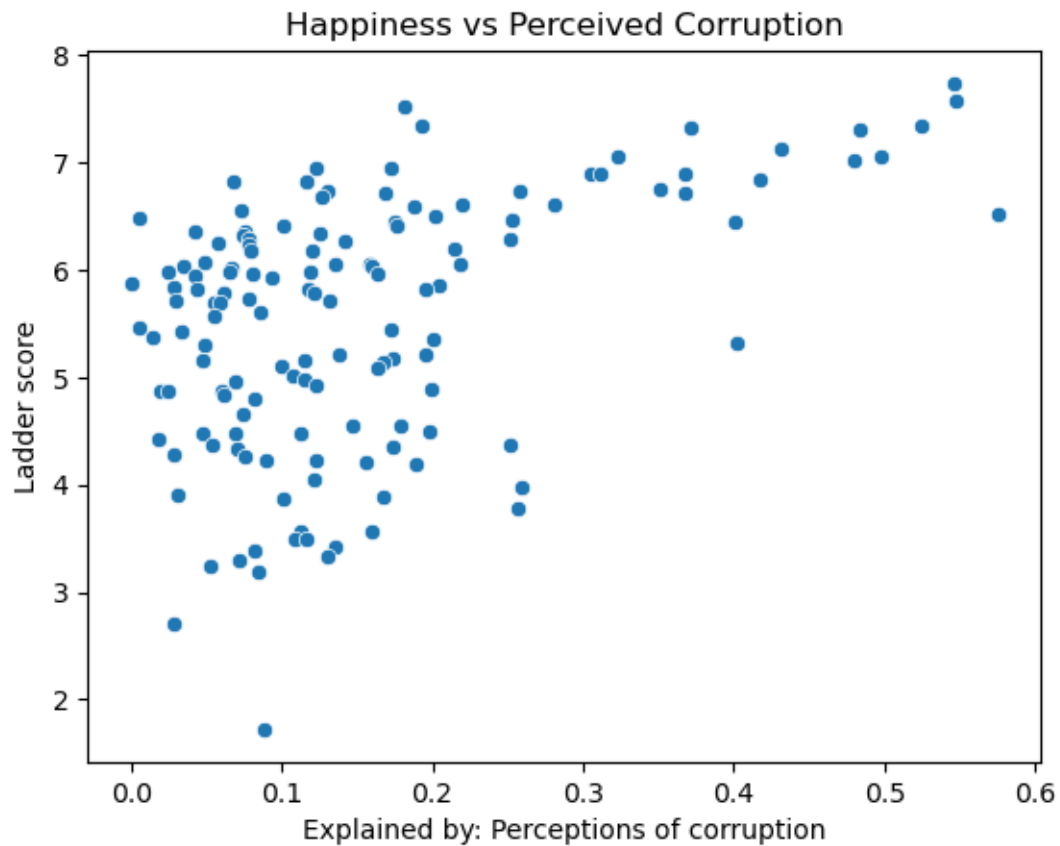
```
[94]: # Relationship between happiness and Income
sns.scatterplot(x='Explained by: Log GDP per capita', y='Ladder score', data=df)
plt.title('Happiness vs GDP per capita')
plt.xlabel('Log GDP per capita')
plt.ylabel('Ladder score')
plt.show()
```



```
[74]: # The effect of social support on happiness (Happiness vs Social Support)
sns.scatterplot(x='Explained by: Social support', y='Ladder score', data=df)
plt.title('Happiness vs Social support')
plt.show()
```



```
[76]: # The effect of perceived corruption on happiness (Happiness vs Perceived
      ↪Corruption)
      sns.scatterplot(x='Explained by: Perceptions of corruption', y='Ladder score',
      ↪data=df)
      plt.title('Happiness vs Perceived Corruption')
      plt.show()
```



```
[130]: # The categories and their labels are defined
bins = [0, 4, 5.5, 7.5, 10]
labels = ['Low happiness', 'Medium happiness', 'High happiness', 'Very high
↪happiness']

# A 'Happiness level' column is created based on the happiness score
df['Happiness level'] = pd.cut(df['Ladder score'], bins=bins, labels=labels)

# All countries along with their happiness scores and levels are printed
print(df[['Country name', 'Ladder score', 'Happiness level']])
```

	Country name	Ladder score	Happiness level
0	Finland	7.741	Very high happiness
1	Denmark	7.583	Very high happiness
2	Iceland	7.525	Very high happiness
3	Sweden	7.344	High happiness
5	Netherlands	7.319	High happiness
..
138	Congo (Kinshasa)	3.295	Low happiness
139	Sierra Leone	3.245	Low happiness
140	Lesotho	3.186	Low happiness
141	Lebanon	2.707	Low happiness
142	Afghanistan	1.721	Low happiness

[142 rows x 3 columns]