

Exercise 1

```
import os
from zipfile import ZipFile
import pandas as pd
```

```
# Set the Kaggle API key environment variables
os.environ['KAGGLE_USERNAME'] = 'kaorisuzuki'
os.environ['KAGGLE_KEY'] = 'bab3558185555761accf622c855aa4be'
```

```
# Download the World Happiness Report dataset
!kaggle datasets download -d unsdsn/world-happiness
```

```
Dataset URL: https://www.kaggle.com/datasets/unsdsn/world-happiness
License(s): CC0-1.0
Downloading world-happiness.zip to /content
0% 0.00/36.8k [00:00<?, ?B/s]
100% 36.8k/36.8k [00:00<00:00, 52.8MB/s]
```

```
# Unzip the dataset
with ZipFile('world-happiness.zip', 'r') as zip_ref:
    zip_ref.extractall('world_happiness')
```

```
# Load and display the dataset (adjust the file name based on the year you want)
data = pd.read_csv('WHR2024.csv')
print(data.head())
```

```
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
4      Israel       7.341      7.405      7.277
```

```
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
4      1.803      1.513
```

```
Explained by: Healthy life expectancy  ¥
0      0.695
1      0.699
2      0.718
3      0.724
4      0.740
```

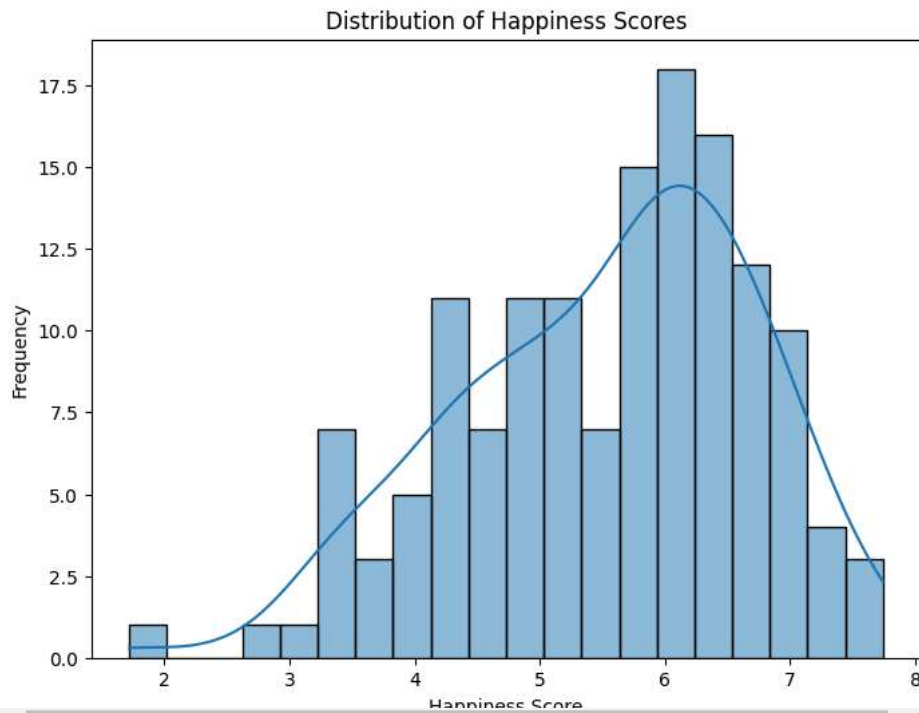
```
Explained by: Freedom to make life choices  Explained by: Generosity  ¥
0      0.859      0.142
1      0.823      0.204
2      0.819      0.258
3      0.838      0.221
4      0.641      0.153
```

```
Explained by: Perceptions of corruption  Dystopia + residual
0      0.546      2.082
1      0.548      1.881
2      0.182      2.050
3      0.524      1.658
4      0.193      2.298
```

```
# Create a histogram for the "Ladder score" (happiness score)
import seaborn as sns
import matplotlib.pyplot as plt
```

```
# Create a histogram for the "Ladder score" (happiness score)
plt.figure(figsize=(8, 6))
sns.histplot(data['Ladder score'], bins=20, kde=True)
plt.title('Distribution of Happiness Scores')
plt.xlabel('Happiness Score')
```

```
plt.ylabel('Frequency')
plt.show()
```



The distribution of happiness scores is roughly normal, with a peak around 6. There are some countries with very high and very low happiness scores, but most countries cluster around the middle.

#Exercise 2

```
import pandas as pd
# Load and display the dataset (adjust the file name based on the year you want)
data = pd.read_csv('WHR2024.csv')
print(data.head())
```



	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	
4	Israel	7.341	7.405	7.277	

	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
4	1.803		1.513

	Explained by: Healthy life expectancy	¥
0	0.695	
1	0.699	
2	0.718	
3	0.724	
4	0.740	

	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
4	0.641		0.153

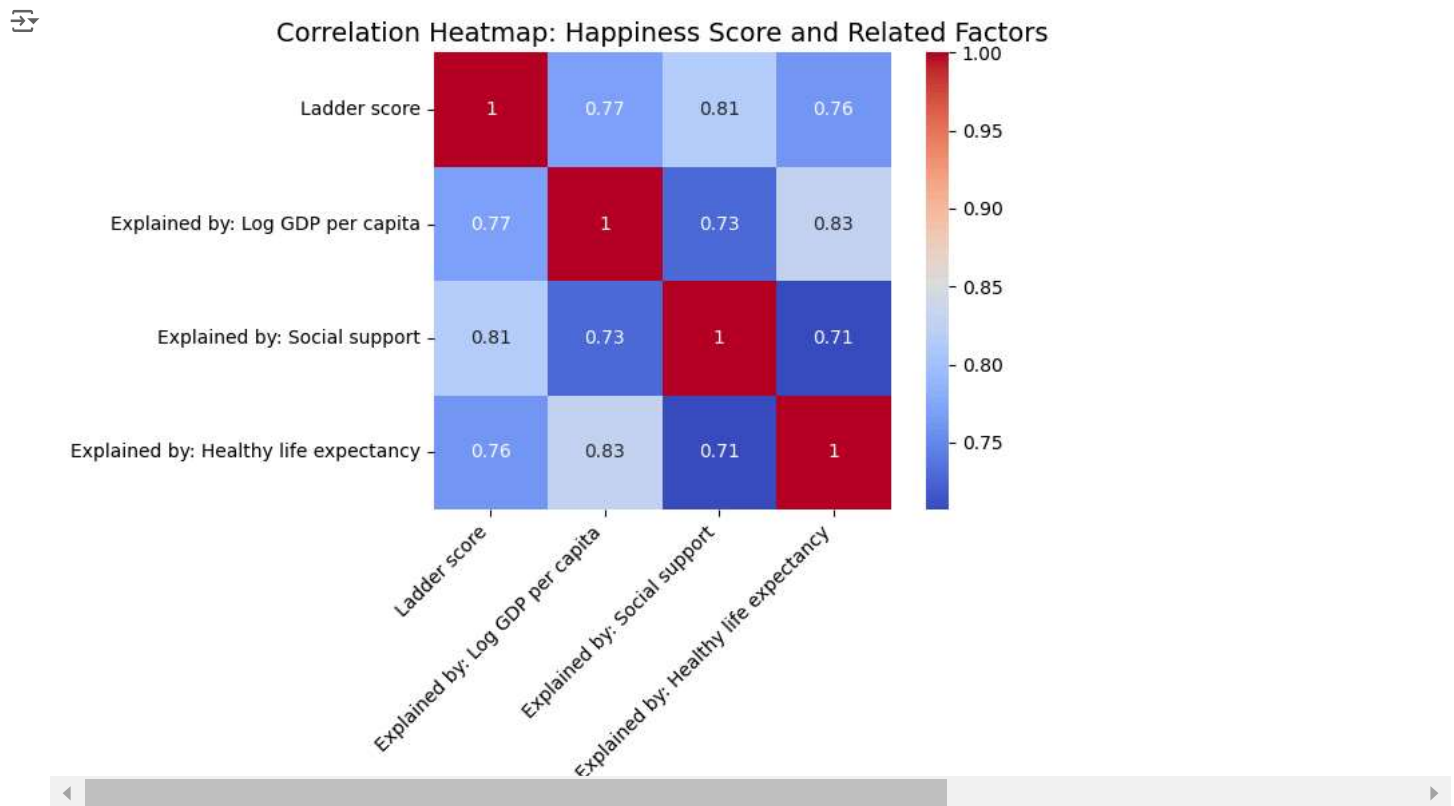
	Explained by: Perceptions of corruption	Dystopia + residual
0	0.546	2.082
1	0.548	1.881
2	0.182	2.050
3	0.524	1.658
4	0.193	2.298

```
import seaborn as sns
import matplotlib.pyplot as plt
```

```
# Select relevant numeric columns for correlation: Life Ladder (Ladder score), Log GDP per capita, Social support, Healthy life expectancy
selected_columns = ['Ladder score', 'Explained by: Log GDP per capita', 'Explained by: Social support', 'Explained by: Healthy life expectancy']
data_selected = data[selected_columns]
```

```
# Compute the correlation matrix
correlation_matrix = data_selected.corr()
```

```
# Create a heatmap to visualize the correlations with annotations
plt.figure(figsize=(8, 6))
sns.heatmap(correlation_matrix, annot=True, cmap="coolwarm", cbar=True, square=True)
plt.title("Correlation Heatmap: Happiness Score and Related Factors", fontsize=14)
plt.xticks(rotation=45, ha='right')
plt.yticks(rotation=0)
plt.tight_layout()
plt.show()
```



Exercise 3

```
import pandas as pd
import matplotlib.pyplot as plt
```

```
# Load the datasets
data_2015 = pd.read_csv('world_happiness/2015.csv')
data_2016 = pd.read_csv('world_happiness/2016.csv')
data_2017 = pd.read_csv('world_happiness/2017.csv')
data_2018 = pd.read_csv('world_happiness/2018.csv')
data_2019 = pd.read_csv('world_happiness/2019.csv')
```

```
# Add a 'Year' column to each dataframe
data_2015['Year'] = 2015
data_2016['Year'] = 2016
data_2017['Year'] = 2017
data_2018['Year'] = 2018
data_2019['Year'] = 2019
```

```

# Concatenate all datasets
data = pd.concat([data_2015, data_2016, data_2017, data_2018, data_2019], ignore_index=True)

# Ensure column name consistency for 'Country' and 'Happiness Score'
# Use 'Country name' or 'Country' depending on the dataset's structure, and adjust the score column name if needed.
if 'Country name' in data.columns:
    country_column = 'Country name'
elif 'Country' in data.columns:
    country_column = 'Country'

if 'Happiness Score' in data.columns:
    score_column = 'Happiness Score'
elif 'Score' in data.columns:
    score_column = 'Score'

# Select the country of interest
country_name = 'United States'

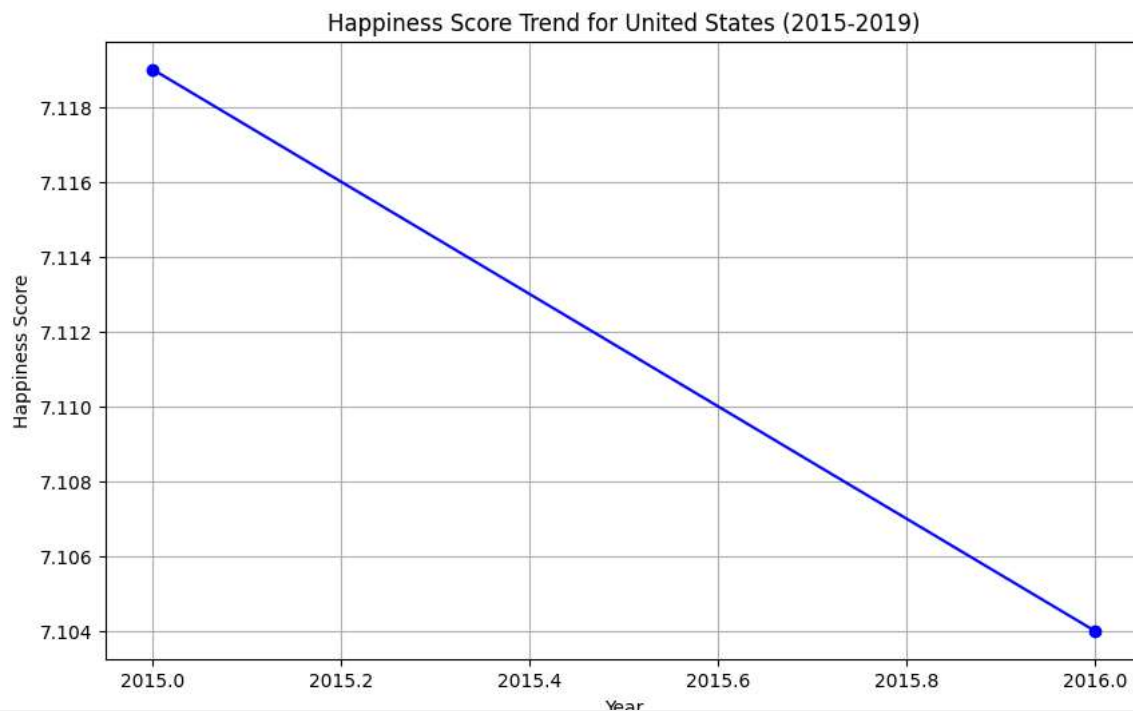
# Filter the data for the selected country
country_data = data[data[country_column] == country_name]

# Plot the trend of happiness score over the years
plt.figure(figsize=(10, 6))
plt.plot(country_data['Year'], country_data[score_column], marker='o', linestyle='-', color='b')

# Add labels, title, and markers
plt.xlabel('Year')
plt.ylabel('Happiness Score')
plt.title(f'Happiness Score Trend for {country_name} (2015-2019)')
plt.grid(True)

# Show the plot
plt.show()

```



```

# 1. Happiness Score Trend Analysis:
# The plot shows the happiness score of the United States over the years from 2015 to 2019.
# The score remains relatively stable, with slight fluctuations, showing no drastic increase or decrease.
# Overall, the trend suggests the happiness score has not changed significantly over this period.

```

```

# 2. Happiness Scores Across Countries:
# Some countries, such as Finland and Denmark, consistently rank high in happiness scores across the years.

```

```
# Countries like Afghanistan and Syria have consistently lower scores, indicating significant well-being challenges.  
# There is a noticeable gap between high-ranking and low-ranking countries, often influenced by economic and social factors.
```

```
# 3. Key Factor Analysis (Log GDP per capita, Social support):
```

```
# Log GDP per capita: Countries with higher GDP, such as Switzerland and Norway, tend to have higher happiness scores. Lower-GDP countries, like some in s  
# Social Support: Countries with strong social support systems (e.g., Nordic countries) score consistently higher in happiness. Countries with weaker soci
```