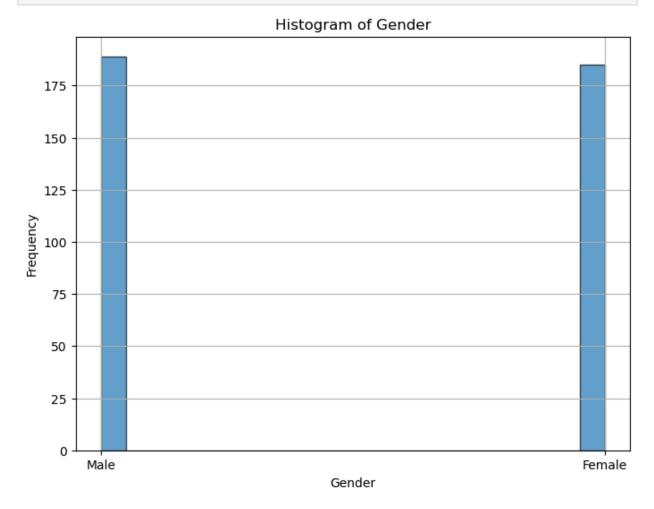
```
In [ ]: # DSC 530
         # Gracie Inman
         # Final Project
         # 08/12/23
In [49]: import pandas as pd
         import matplotlib.pyplot as plt
         import numpy as np
         import scipy.stats as stats
         import statsmodels.api as sm
         NUM BINS = 20
         # Step 1: Load the dataset
         def load dataset(filename):
             return pd.read_csv(filename)
         # Step 2: Describe the 5 variables
         # Heart rate is the patient's heart rate.
         ## Sleep duration is how long the patient slept
         ## Sleep quality quantifies the quality of the sleep
         ## Physical activity level quantifies how active the patient is.
         ## Age is how old the subject is.
         ## Stress Level is how stressed the subject feels quantified.
         # Step 3: Create histograms and identify outliers (Chapter 2)
         def plot histogram with outliers(data, column):
             plt.figure(figsize=(8, 6))
             plt.hist(data[column], bins=NUM BINS, edgecolor='k', alpha=0.7)
             plt.xlabel(column)
             plt.ylabel('Frequency')
             plt.title(f'Histogram of {column}')
             plt.grid(True)
             plt.show()
         # Plot histograms for each variable and identify outliers
         variables = ['Gender', 'Age', 'Sleep Duration', 'Quality of Sleep', 'Physical A
         for var in variables:
             plot histogram with outliers(data, var)
         # Step 4: Calculate descriptive characteristics (Chapter 2)
         def calculate descriptive stats(data, column):
             mean = data[column].mean()
             mode = data[column].mode()[0]
             spread = data[column].std()
             tails = data[column].skew()
             return mean, mode, spread, tails
         # Step 5: Compare scenarios using PMF (Chapter 3)
         def plot pmf(data, column, scenario column, scenario value):
             filtered data = data[data[scenario column] == scenario value]
             pmf = filtered data[column].value counts(normalize=True).sort index()
             plt.bar(pmf.index, pmf.values)
             plt.xlabel(column)
             plt.ylabel('PMF')
             plt.title(f'PMF of {column} for {scenario column} = {scenario value}')
             plt.grid(True)
             plt.show()
```

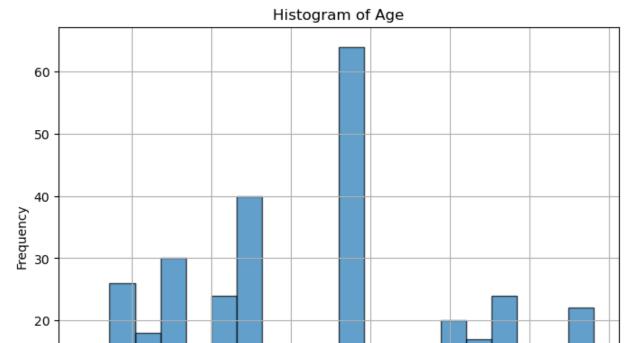
```
# Step 6: Create a CDF (Chapter 4)
def plot cdf(data, column):
    plt.figure(figsize=(8, 6))
    sorted_data = np.sort(data[column])
    cdf = np.arange(1, len(sorted data) + 1) / len(sorted data)
    plt.plot(sorted_data, cdf)
    plt.xlabel(column)
    plt.ylabel('CDF')
    plt.title(f'CDF of {column}')
    plt.grid(True)
    plt.show()
# Step 7: Plot an analytical distribution and provide analysis (Chapter 5)
def plot_analytical_distribution(data, column, distribution=stats.norm):
    plt.figure(figsize=(8, 6))
    # Plot the histogram or KDE of the data
    plt.hist(data[column], bins=20, density=True, alpha=0.7, label='Data')
    # Plot the analytical distribution
    xmin, xmax = plt.xlim()
    x = np.linspace(xmin, xmax, 100)
    params = distribution.fit(data[column])
    pdf = distribution.pdf(x, *params)
    plt.plot(x, pdf, 'r', label='Analytical Distribution')
    plt.xlabel(column)
    plt.ylabel('Density')
    plt.title(f'Analytical Distribution vs. Data Distribution of {column}')
    plt.legend()
    plt.grid(True)
    # Adjust y-axis limits for better visibility
    plt.ylim([0, max(pdf) * 1.0])
    plt.show()
# Step 8: Create scatter plots and analyze correlation and causation (Chapter
def plot scatter(data, x column, y column):
    plt.figure(figsize=(8, 6))
    plt.scatter(data[x column], data[y column], alpha=0.5)
    plt.xlabel(x column)
    plt.ylabel(y column)
    plt.title(f'Scatter Plot of {x column} vs {y column}')
    plt.grid(True)
    plt.show()
# Step 9: Conduct a hypothesis test (Chapter 9)
def conduct hypothesis test(data, variable1, variable2):
    group1 = data[data[variable2] == 'Male'][variable1]
    group2 = data[data[variable2] == 'Female'][variable1]
    # Perform a hypothesis test (t-test)
    t stat, p value = stats.ttest ind(group1, group2)
    # Provide interpretation of results
```

```
if p value < 0.05:
        result = "Reject null hypothesis: There is a significant difference bet
        result = "Fail to reject null hypothesis: There is no significant diffe
    print(f"T-Stat: {t stat}")
    print(f"P-Value: {p value}")
    print(result)
# Step 10: Conduct a regression analysis (Chapter 10 & 11)
def conduct regression analysis(data, dependent variable, explanatory variables
    X = sm.add_constant(data[explanatory_variables])
    y = data[dependent_variable]
    model = sm.OLS(y, X).fit()
    # Print regression summary and interpret results
    print(model.summary())
def analyze variable(column data):
    if column_data.dtype.kind in "biufc":
        # If the data is numerical, perform statistical analysis
        stats = {
            "Mean": column_data.mean(),
            "Median": column data.median(),
            "Standard Deviation": column_data.std(),
            "Minimum": column_data.min(),
            "Maximum": column_data.max(),
    else:
        # If the data is non-numerical, count occurrences of each value
        counts = column_data.value_counts()
        stats = {value: count for value, count in counts.items()}
    return stats
def correlation matrix(data frame):
    correlation_matrix = data_frame.corr()
    return correlation matrix
def main():
    # Call the functions to perform the analyses
    data = load dataset('Sleep health and lifestyle dataset.csv')
    # Calculate and display descriptive statistics
    variables_to_analyze = ['Age', 'Sleep Duration', 'Quality of Sleep', 'Heart'
    for variable in variables to analyze:
        stats = calculate descriptive stats(data, variable)
        print(f"Descriptive Statistics for {variable}:")
        print(f" Mean: {stats[0]:.2f}")
        print(f" Mode: {stats[1]:.2f}")
        print(f" Standard Deviation: {stats[2]:.2f}")
        print(f" Skewness: {stats[3]:.2f}")
        print()
    plot_analytical_distribution(data, 'Quality of Sleep')
    plot_histogram_with_outliers(data, 'Physical Activity Level')
    plot_pmf(data, 'Sleep Duration', 'Gender', 'Male')
    plot cdf(data, 'Quality of Sleep')
    plot_analytical_distribution(data, 'Physical Activity Level')
    plot scatter(data, 'Heart Rate', 'Age')
    plot_scatter(data, 'Quality of Sleep', 'Heart Rate')
```

```
conduct_hypothesis_test(data, 'Quality of Sleep', 'Gender')
  conduct_regression_analysis(data, 'Heart Rate', ['Sleep Duration', 'Age'])

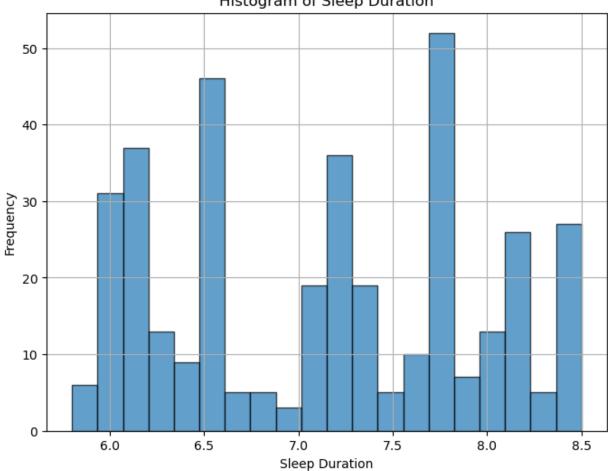
if __name__ == "__main__":
  main()
```

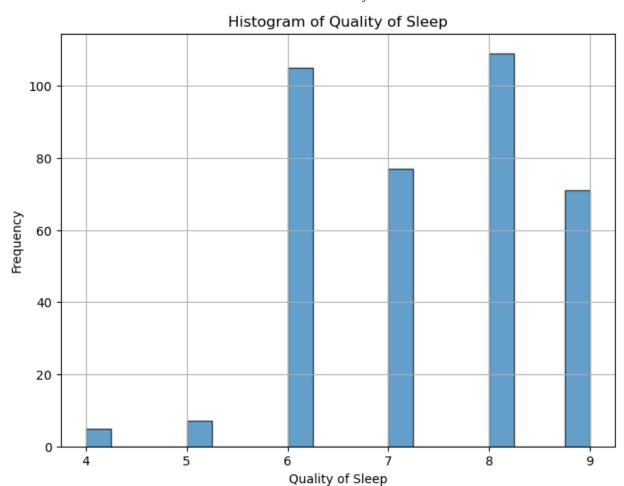


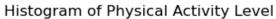


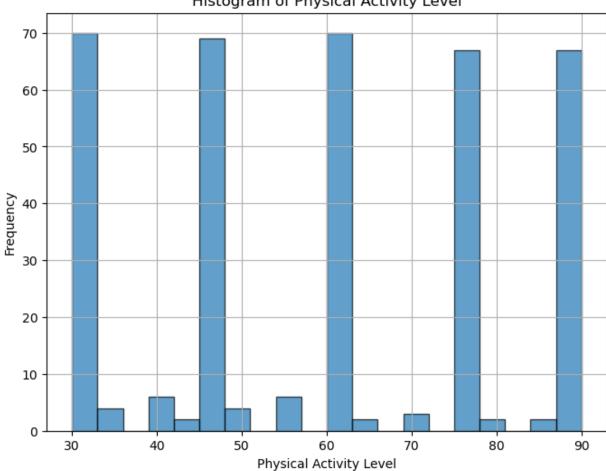
Age



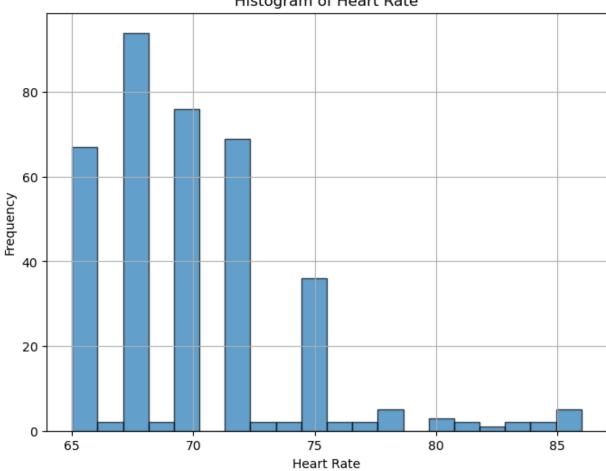








Histogram of Heart Rate



Descriptive Statistics for Age:

Mean: 42.18 Mode: 43.00

Standard Deviation: 8.67

Skewness: 0.26

Descriptive Statistics for Sleep Duration:

Mean: 7.13 Mode: 7.20

Standard Deviation: 0.80

Skewness: 0.04

Descriptive Statistics for Quality of Sleep:

Mean: 7.31 Mode: 8.00

Standard Deviation: 1.20

Skewness: -0.21

Descriptive Statistics for Heart Rate:

Mean: 70.17 Mode: 68.00

Standard Deviation: 4.14

Skewness: 1.22

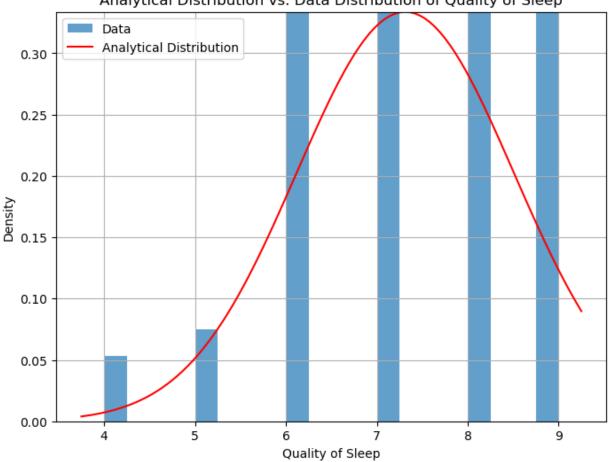
Descriptive Statistics for Physical Activity Level:

Mean: 59.17 Mode: 60.00

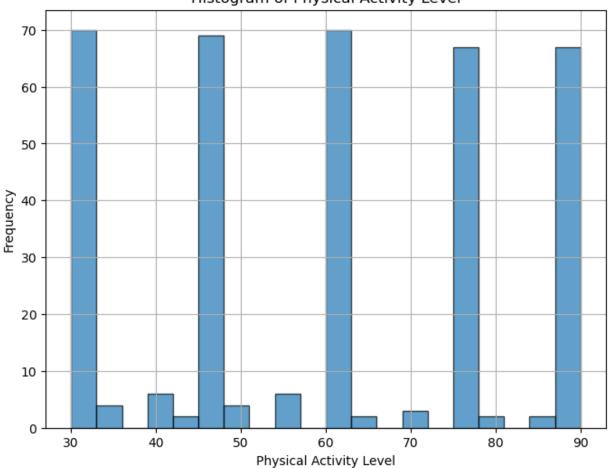
Standard Deviation: 20.83

Skewness: 0.07

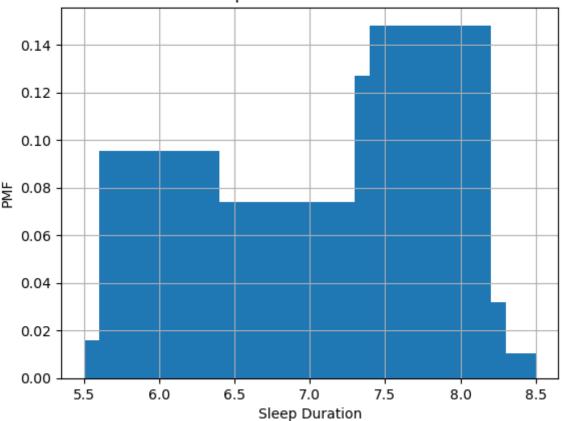
Analytical Distribution vs. Data Distribution of Quality of Sleep



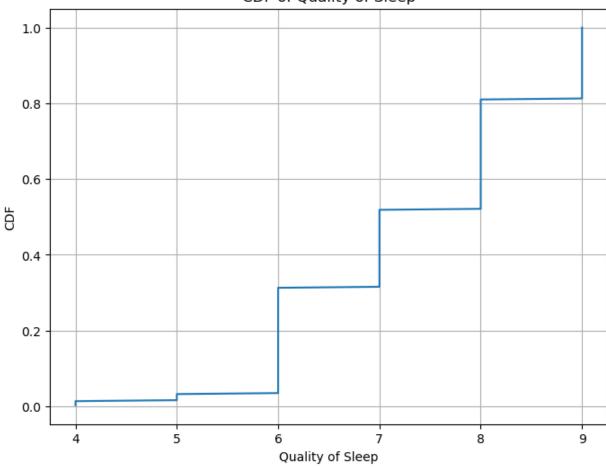


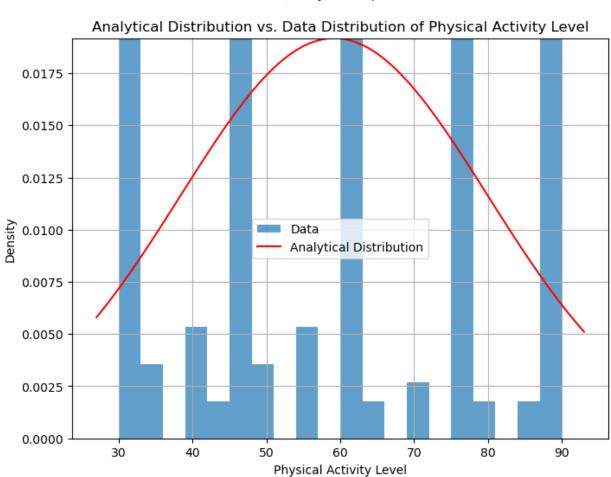




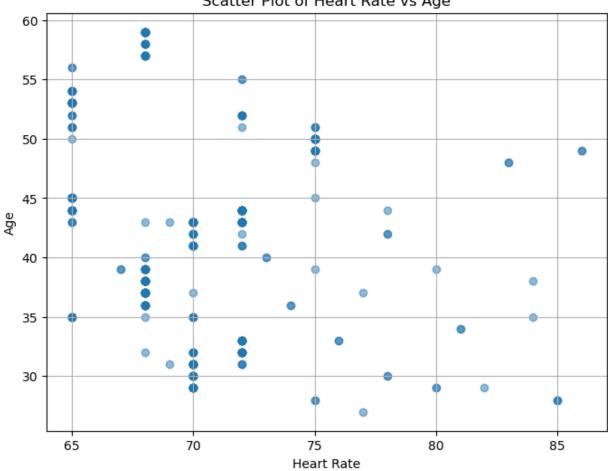




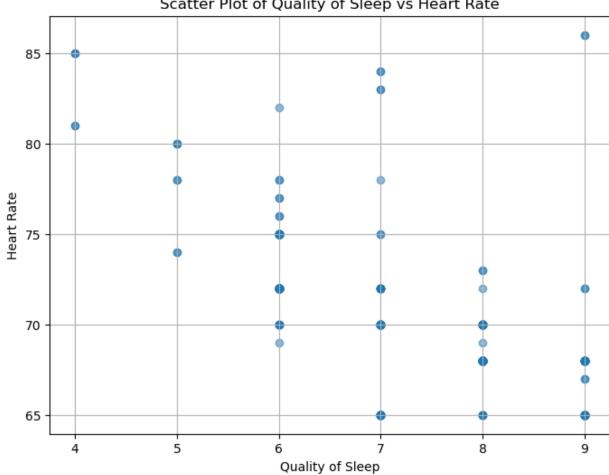




Scatter Plot of Heart Rate vs Age



Scatter Plot of Quality of Sleep vs Heart Rate



T-Stat: -5.874547760454642 P-Value: 9.416446532689304e-09

Reject null hypothesis: There is a significant difference between the groups.

OLS Regression Results

		OLS Regles:	sion Results =======	, .=======		
Dep. Variable:		Heart Rate	R-squared:		0.269	
Model:	OLS		Adj. R-squared:		0.265	
Method:	Least Squares		F-statistic:		68.36	
Date:	Sat, 12 Aug 2023		Prob (F-statistic):		5.30e-26	
Time:	20:13:07		Log-Likelihood:		-1002.5	
No. Observations:	374		AIC:		2011.	
Df Residuals:	371		BIC:		2023.	
Df Model:		2				
Covariance Type:		nonrobust				
====	=======	========	=======	=======	=======	=====
	coef	std err	t	P> t	[0.025	0.
975]	0001	504 011	C	1, 101	[0.023	
const	89.7075	1.691	53.039	0.000	86.382	9
3.033						
Sleep Duration 2.104	-2.5877	0.246	-10.530	0.000	-3.071	-
2.104 Age	-0.0257	0.023	-1.142	0.254	-0.070	
0.019	-0.0237	0.023	-1.142	0.234	-0.070	
=======================================	=======	========	-=======	:========	=========	=====
Omnibus:		136.960	Durbin-Watson:		1.076	
Prob(Omnibus):		0.000	Jarque-Bera (JB):		706.619	
Skew:		1.473	Prob(JB):		3.63e-154	
Kurtosis:		9.055	•			406.

Notes:

[1] Standard Errors assume that the covariance matrix of the errors is correct ly specified.

In []: