

```
!pip install matplotlib seaborn pandas
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns

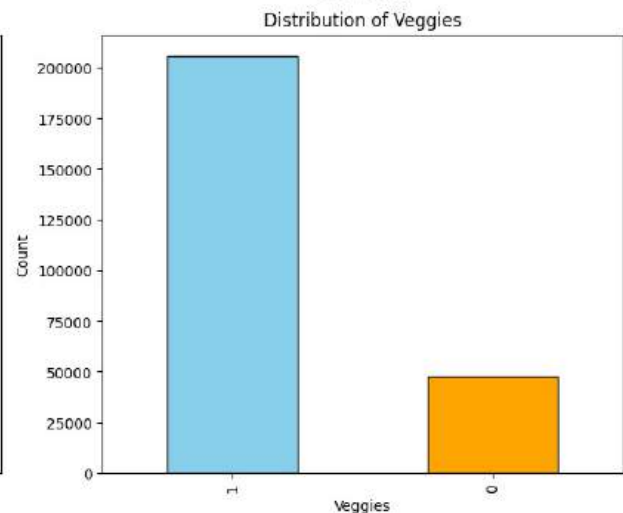
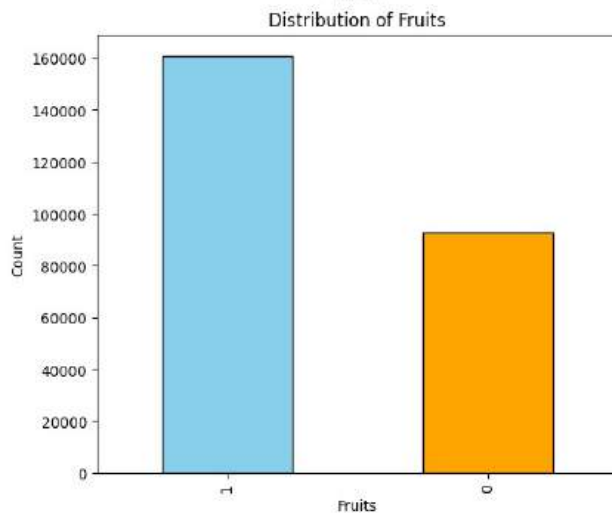
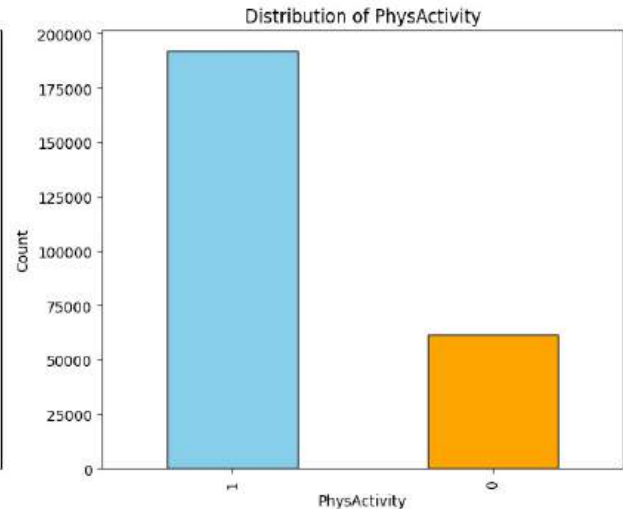
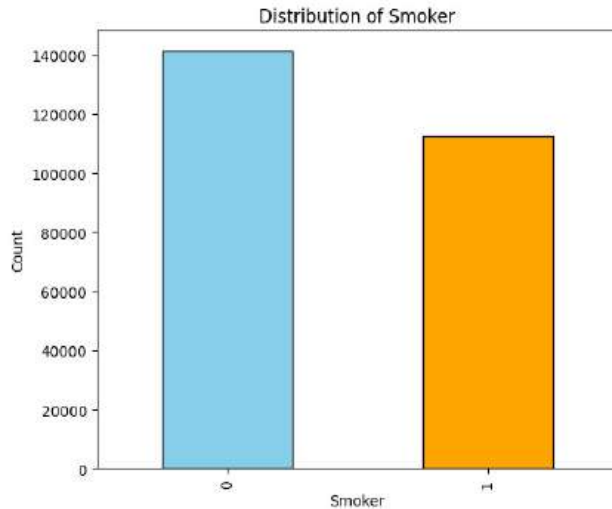
data = pd.read_csv("healthcare_data_set.csv")
data.head()

# Categorical variables
categorical_vars = ['Smoker', 'PhysActivity', 'Fruits', 'Veggies']

# --- Bar Charts (2x2 grid) ---
fig, axes = plt.subplots(2, 2, figsize=(12, 10))

for ax, var in zip(axes.flatten(), categorical_vars):
    counts = data[var].value_counts()
    counts.plot(kind='bar', ax=ax, edgecolor='black',
color=['skyblue', 'orange'])
    ax.set_title(f"Distribution of {var}")
    ax.set_xlabel(var)
    ax.set_ylabel("Count")

plt.tight_layout()
plt.show()
```



```
# categorical variables
```

```
target = 'HeartDiseaseorAttack'
```

```
categorical_vars = ['Smoker', 'PhysActivity', 'Fruits', 'Veggies']
```

```
# Plot grouped bar charts
```

```
fig, axes = plt.subplots(2, 2, figsize=(12, 10))
```

```
for ax, var in zip(axes.flatten(), categorical_vars):
```

```
    # Create crosstab: counts of HeartDisease by category
```

```
    ct = pd.crosstab(data[var], data[target], normalize='index') * 100
```

```
# percentage
```

```
    # Plot grouped bar chart
```

```
    ct.plot(
```

```
        kind='bar',
```

```
        stacked=False,
```

```
        ax=ax,
```

```
        edgecolor='black',
```

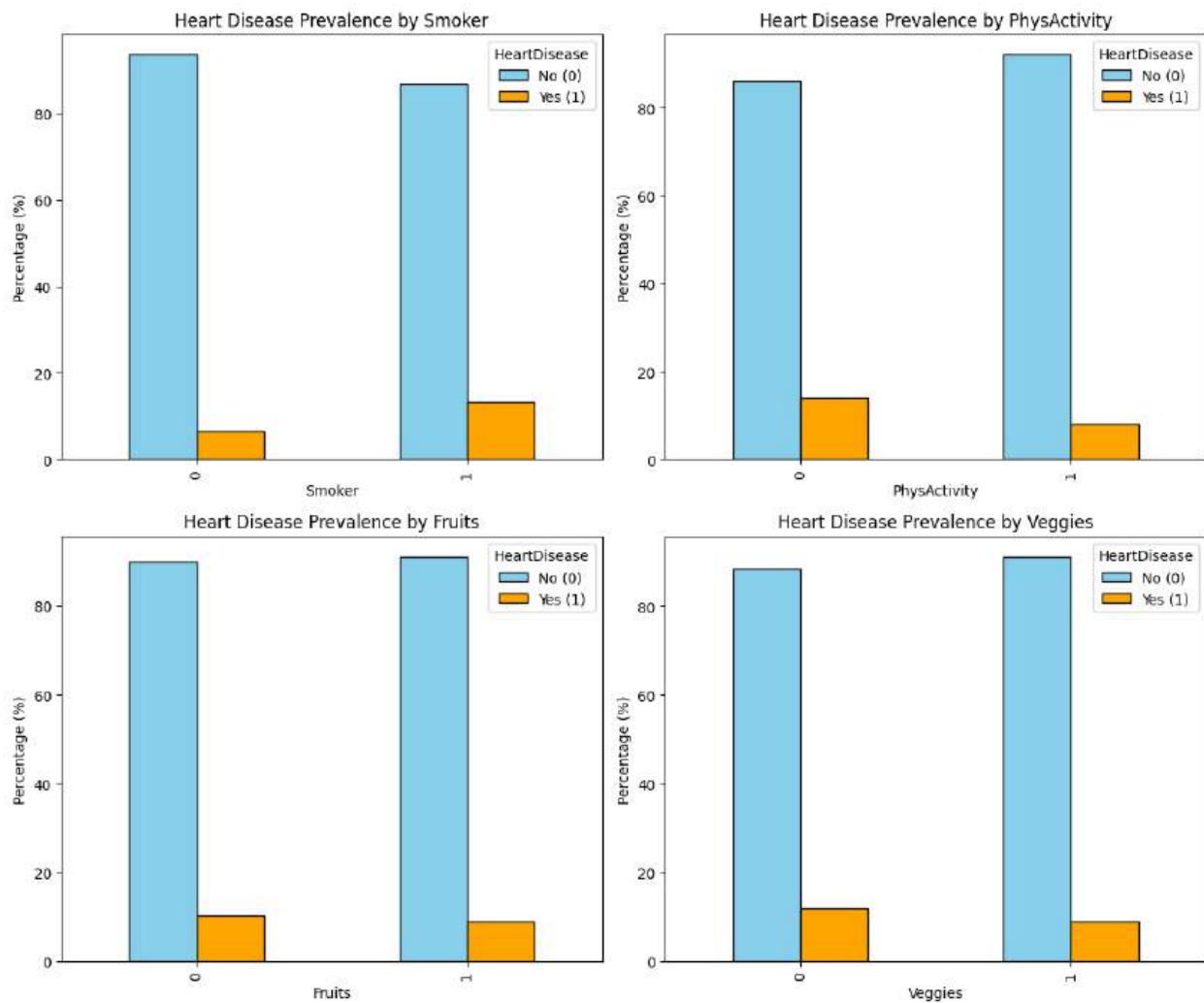
```

        color=['skyblue', 'orange']
    )

    ax.set_title(f"Heart Disease Prevalence by {var}")
    ax.set_xlabel(var)
    ax.set_ylabel("Percentage (%)")
    ax.legend(title='HeartDisease', labels=['No (0)', 'Yes (1)'])

plt.tight_layout()
plt.show()

```



```

##### a.Display Distributions of Continuous Variables

# Continuous variables
continuous_vars = ['BMI', 'MentHlth', 'PhysHlth']

# --- Histograms (1 row) ---
fig, axes = plt.subplots(1, 3, figsize=(15, 5))

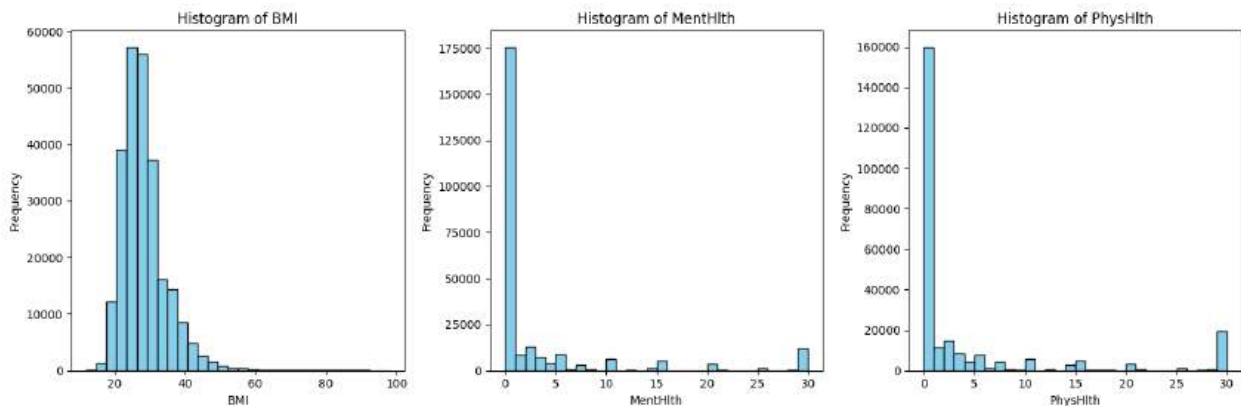
```

```

for ax, var in zip(axes, continuous_vars):
    ax.hist(data[var].dropna(), bins=30, color='skyblue',
edgecolor='black')
    ax.set_title(f"Histogram of {var}")
    ax.set_xlabel(var)
    ax.set_ylabel("Frequency")

plt.tight_layout()
plt.show()

```



#### ##### b. Heatmaps and Correlation Plots: Explore Variable Relationships

```

numeric_vars = ['HeartDiseaseorAttack', 'HighBP', 'HighChol',
'CholCheck',
                'BMI', 'Smoker', 'Stroke', 'Diabetes', 'PhysActivity',
                'Fruits', 'Veggies', 'HvyAlcoholConsump',
'AnyHealthcare',
                'NoDocbcCost', 'GenHlth', 'MentHlth', 'PhysHlth']

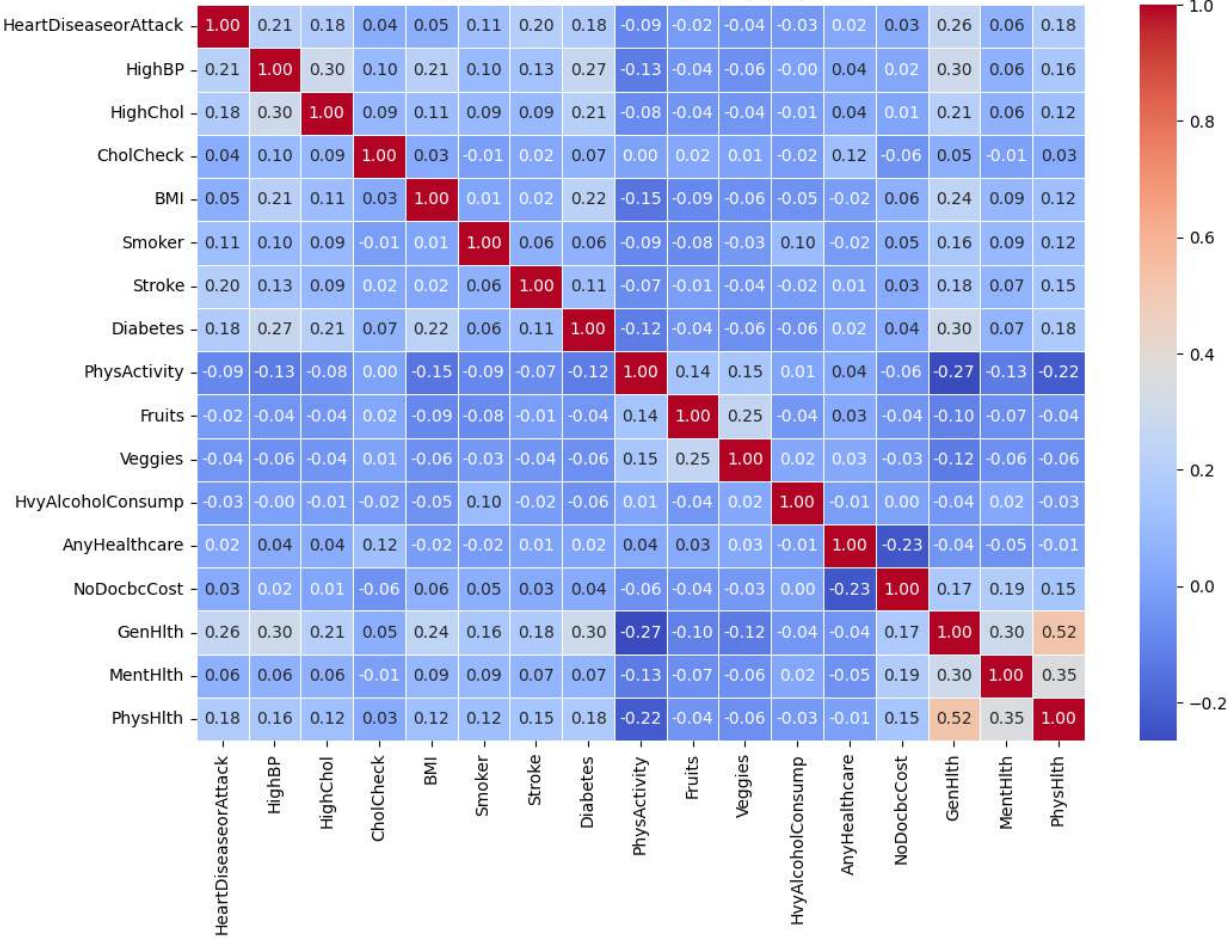
data_numeric = data[numeric_vars]

corr_matrix = data_numeric.corr()

plt.figure(figsize=(12, 8))
sns.heatmap(corr_matrix, annot=True, fmt=".2f", cmap="coolwarm",
linewidths=0.5)
plt.title("Correlation Heatmap of Variables", fontsize=16)
plt.show()

```

Correlation Heatmap of Variables



```
!pip install matplotlib seaborn pandas
```

```
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
```

```
data = pd.read_csv("healthcare_data_set.csv")
```

```
data.head()
```

	HeartDiseaseorAttack	HighBP	HighChol	CholCheck	BMI	Smoker
Stroke \						
0	0	1	1	1	40	1
0						
1	0	0	0	0	25	1
0						
2	0	1	1	1	28	0
0						
3	0	1	0	1	27	0
0						
4	0	1	1	1	24	0
0						

	Diabetes	PhysActivity	Fruits	...	AnyHealthcare	NoDocbcCost
GenHlth \						
0	0	0	0	...	1	0
5						
1	0	1	0	...	0	1
3						
2	0	0	1	...	1	1
5						
3	0	1	1	...	1	0
2						
4	0	1	1	...	1	0
2						

	MentHlth	PhysHlth	DiffWalk	Sex	Age	Education	Income
0	18	15	1	0	9	4	3
1	0	0	0	0	7	6	1
2	30	30	1	0	9	4	8
3	0	0	0	0	11	3	6
4	3	0	0	0	11	5	4

```
[5 rows x 22 columns]
```

```
##### c. Scatter Plots: Investigate Relationships Between Variables
```

```
risk_factors = ['BMI', 'PhysActivity', 'MentHlth', 'PhysHlth'] # independent variables
target = 'HeartDiseaseorAttack' # dependent variable
```

```

missing = [col for col in risk_factors + [target] if col not in
data.columns]
if missing:
    raise ValueError(f"Missing columns in dataset: {missing}")

#####Scatter plots for each risk factor vs. heart disease

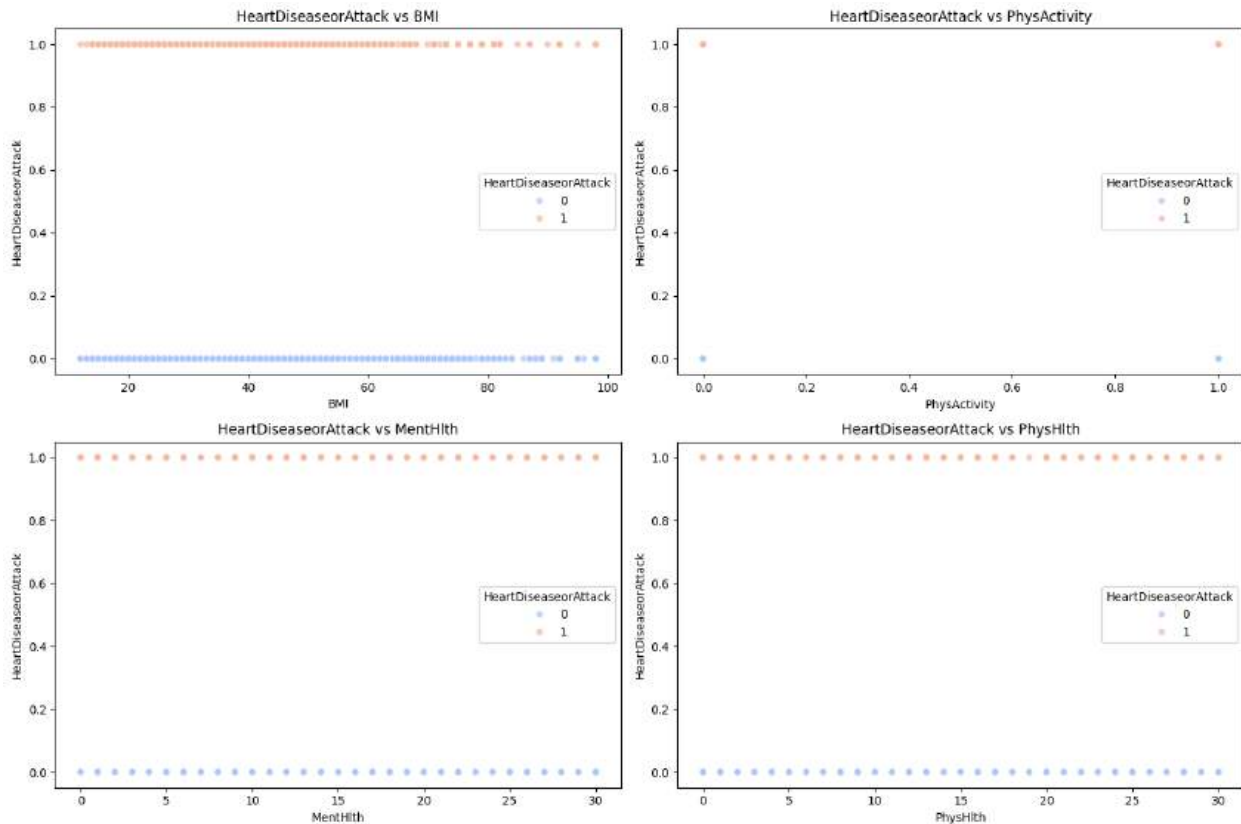
plt.figure(figsize=(15, 10))

for i, var in enumerate(risk_factors):
    plt.subplot(2, 2, i+1)
    sns.scatterplot(
        data=data, x=var, y=target,
        hue=target, palette="coolwarm", alpha=0.7
    )
    plt.title(f"{target} vs {var}")
    plt.xlabel(var)
    plt.ylabel(target)

plt.tight_layout()
plt.show()

sns.pairplot(data, vars=risk_factors + [target], hue=target,
palette="coolwarm")
plt.show()

```



#####d. Line Plots: Display trends over time or across ordered categories.

```
target = 'HeartDiseaseorAttack'
```

```
# Create BMI categories
```

```
bins = [0, 18.5, 25, 30, 35, 40, 100]
```

```
labels = ['Underweight', 'Normal', 'Overweight', 'Obese I', 'Obese II', 'Obese III']
```

```
data['BMI_Category'] = pd.cut(data['BMI'], bins=bins, labels=labels, right=False)
```

```
# Calculate prevalence of heart disease by BMI category
```

```
prevalence = data.groupby('BMI_Category')[target].mean() * 100
```

```
# Line plot with seaborn style
```

```
plt.figure(figsize=(8, 5))
```

```
sns.set_style("whitegrid")
```

```
plt.plot(prevalence.index, prevalence.values, marker='o', linestyle='-', color='blue', label='Heart Disease Prevalence')
```

```
# Add data labels on top of each point
```

```
for i, value in enumerate(prevalence.values):
    plt.text(i, value + 0.5, f"{value:.1f}%", ha='center',
```



```

va='bottom', fontsize=9)

# Optional: Add a shaded region for visual emphasis
plt.fill_between(prevalence.index, 0, prevalence.values, color='blue',
alpha=0.1)

# Titles and labels
plt.xticks(rotation=45)
plt.title("Heart Disease Prevalence Across BMI Categories",
fontsize=14)
plt.xlabel("BMI Category", fontsize=12)
plt.ylabel("Prevalence (%)", fontsize=12)
plt.ylim(0, prevalence.max() + 5)
plt.legend()
plt.tight_layout()
plt.show()

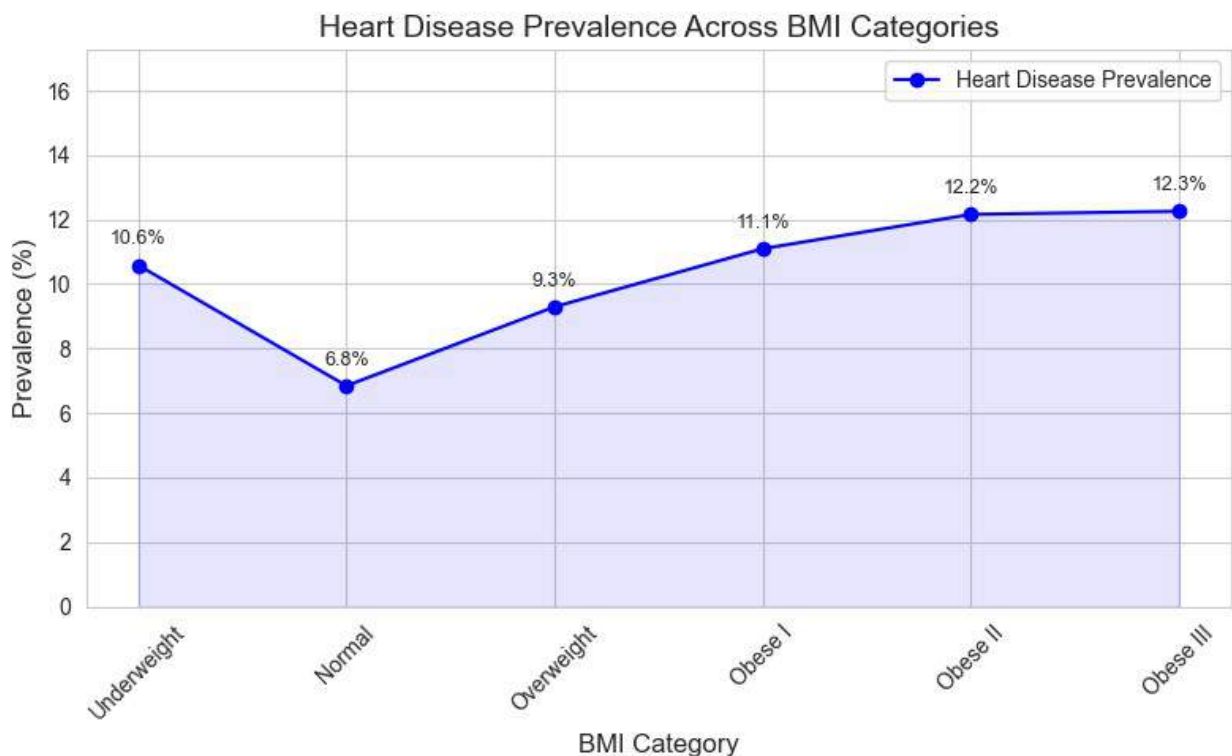
```

C:\Users\seeth\AppData\Local\Temp\ipykernel\_24032\2932911419.py:9:  
FutureWarning: The default of observed=False is deprecated and will be  
changed to True in a future version of pandas. Pass observed=False to  
retain current behavior or observed=True to adopt the future default  
and silence this warning.

```

prevalence = data.groupby('BMI_Category')[target].mean() * 100

```



#####e. Violin Plots: Combine aspects of box plots and density plots to show the distribution of data.

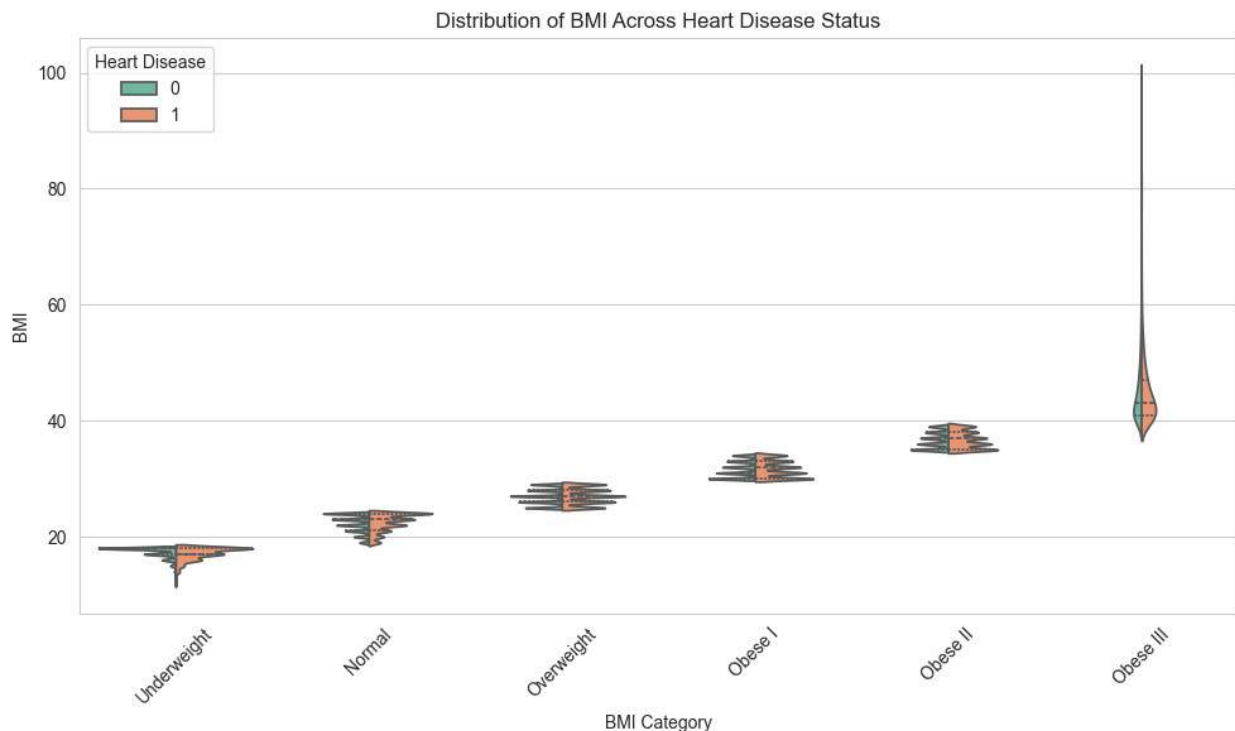
```

# Create BMI categories (ordered)
bins = [0, 18.5, 25, 30, 35, 40, 100]
labels = ['Underweight', 'Normal', 'Overweight', 'Obese I', 'Obese II', 'Obese III']
data['BMI_Category'] = pd.cut(data['BMI'], bins=bins, labels=labels,
right=False)

# Violin plot: Distribution of BMI by Heart Disease status
plt.figure(figsize=(10, 6))
sns.violinplot(x='BMI_Category', y='BMI', hue='HeartDiseaseorAttack',
data=data,
               split=True, inner='quartile', palette='Set2')

plt.title("Distribution of BMI Across Heart Disease Status")
plt.xlabel("BMI Category")
plt.ylabel("BMI")
plt.xticks(rotation=45)
plt.legend(title='Heart Disease', loc='upper left')
plt.tight_layout()
plt.show()

```



```

##### f. Pair Plots: Plot pairwise relationships across
multiple variables
%matplotlib inline

```

```

continuous_vars = ['BMI', 'MentHlth', 'PhysHlth', 'Age']
target = 'HeartDiseaseorAttack'

# Display first few rows to check
print(data[continuous_vars + [target]].head())

# Pair plot
sns.pairplot(data[continuous_vars + [target]], hue=target,
diag_kind='kde', palette='Set1')
plt.suptitle("Pairwise Relationships Between Continuous Variables",
y=1.02)
plt.show()

```

	BMI	MentHlth	PhysHlth	Age	HeartDiseaseorAttack
0	40	18	15	9	0
1	25	0	0	7	0
2	28	30	30	9	0
3	27	0	0	11	0
4	24	3	0	11	0

Pairwise Relationships Between Continuous Variables

