

Understanding Heart Health

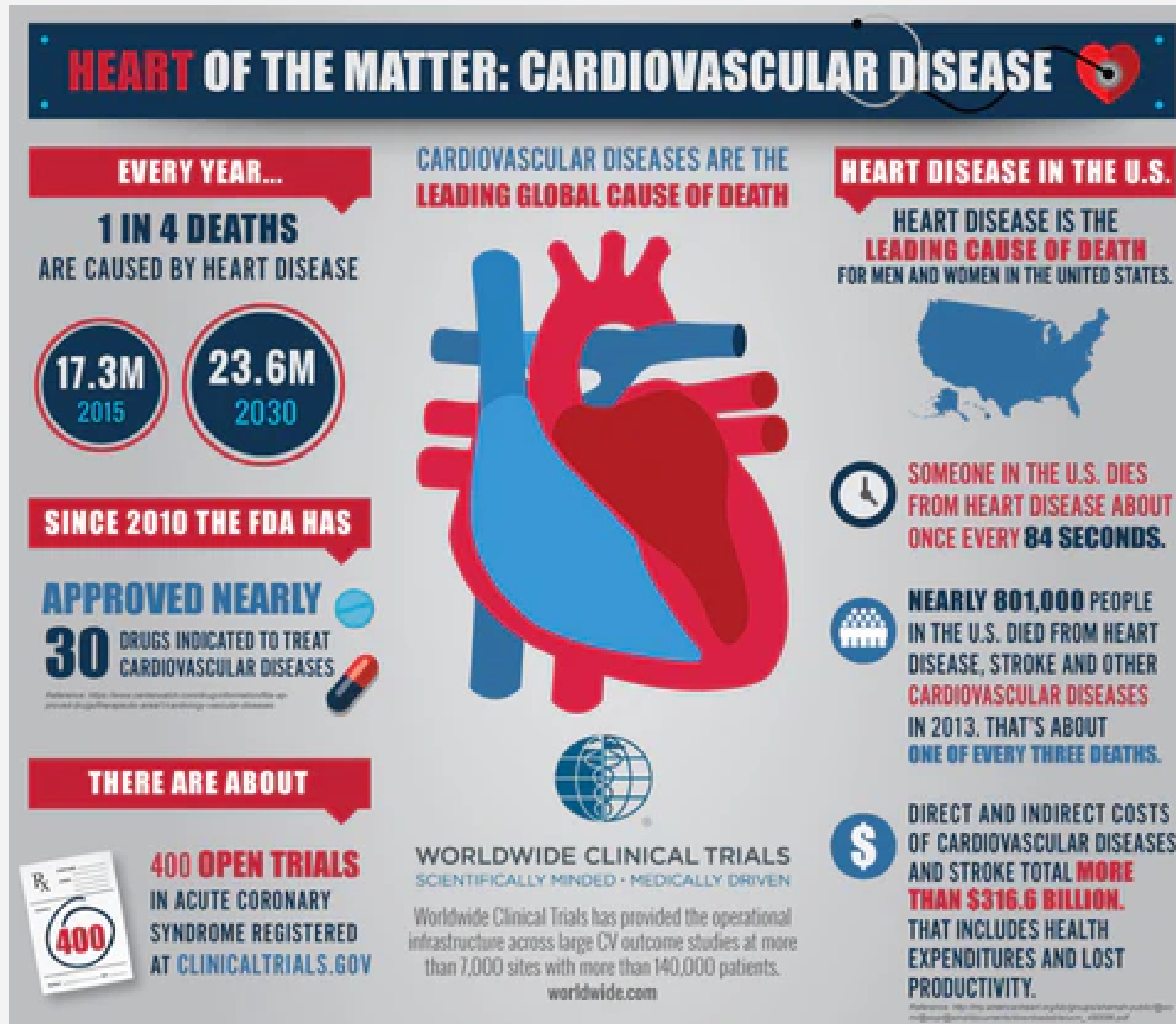
Data Analysis on Heart

INSIGHTS FOR A HEALTHIER LIFE

SAI VINAY BHIMANA, PREM MALDE, RASIKA PANDE, DEVIKA RAHEJA



INTRODUCTION



Globally, cardiovascular diseases (CVDs) are the leading cause of death, claiming an estimated 17.9 million lives annually, accounting for 32% of all global deaths. Of these, 85% are due to heart attacks and strokes.

WHO BENEFITS FROM THIS DATA?



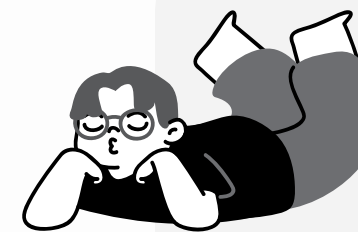
Healthcare Providers



Pharmaceutical Companies



Insurance Companies



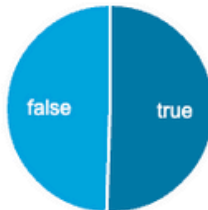






Individuals

DATASET OVERVIEW

Numerical: Age, Cholesterol, Blood Pressure, Heart Rate, Exercise Hours, Blood Sugar

Categorical: Gender, Smoking, Alcohol Intake, Family History, Diabetes, Obesity, Stress Level, Exercise-Induced Angina, Chest Pain Type, Heart Disease

<div><div>▲ Smoking</div><div></div></div>	<div><div>▲ Alcohol Intake</div><div></div></div>	<div><div># Exercise Hours</div><div></div></div>	<div><div>✓ Family History</div><div></div></div>	<div><div>✓ Diabetes</div><div></div></div>
<div><div>Never34%</div><div>Current34%</div><div>Other (326)33%</div></div>	<div><div>Heavy35%</div><div>None34%</div><div>Other (314)31%</div></div>	<div><div></div><div>09</div></div>	<div><div><div><div>true49950%</div><div>false50150%</div></div></div></div>	<div><div><div><div>true50551%</div><div>false49550%</div></div></div></div>
<div><div># Age</div><div>Age of the individual (years).</div></div>	<div><div>▲ Gender</div><div></div></div>	<div><div># Cholesterol</div><div></div></div>	<div><div># Blood Pressure</div><div></div></div>	<div><div># Heart Rate</div><div></div></div>
<div><div></div><div>2579</div></div>	<div><div>Female50%</div><div>Male50%</div></div>	<div><div></div><div>150349</div></div>	<div><div></div><div>90179</div></div>	<div><div></div><div>6099</div></div>

DATA UNDERSTANDING

Shape

The shape is 1000,16

Null Values

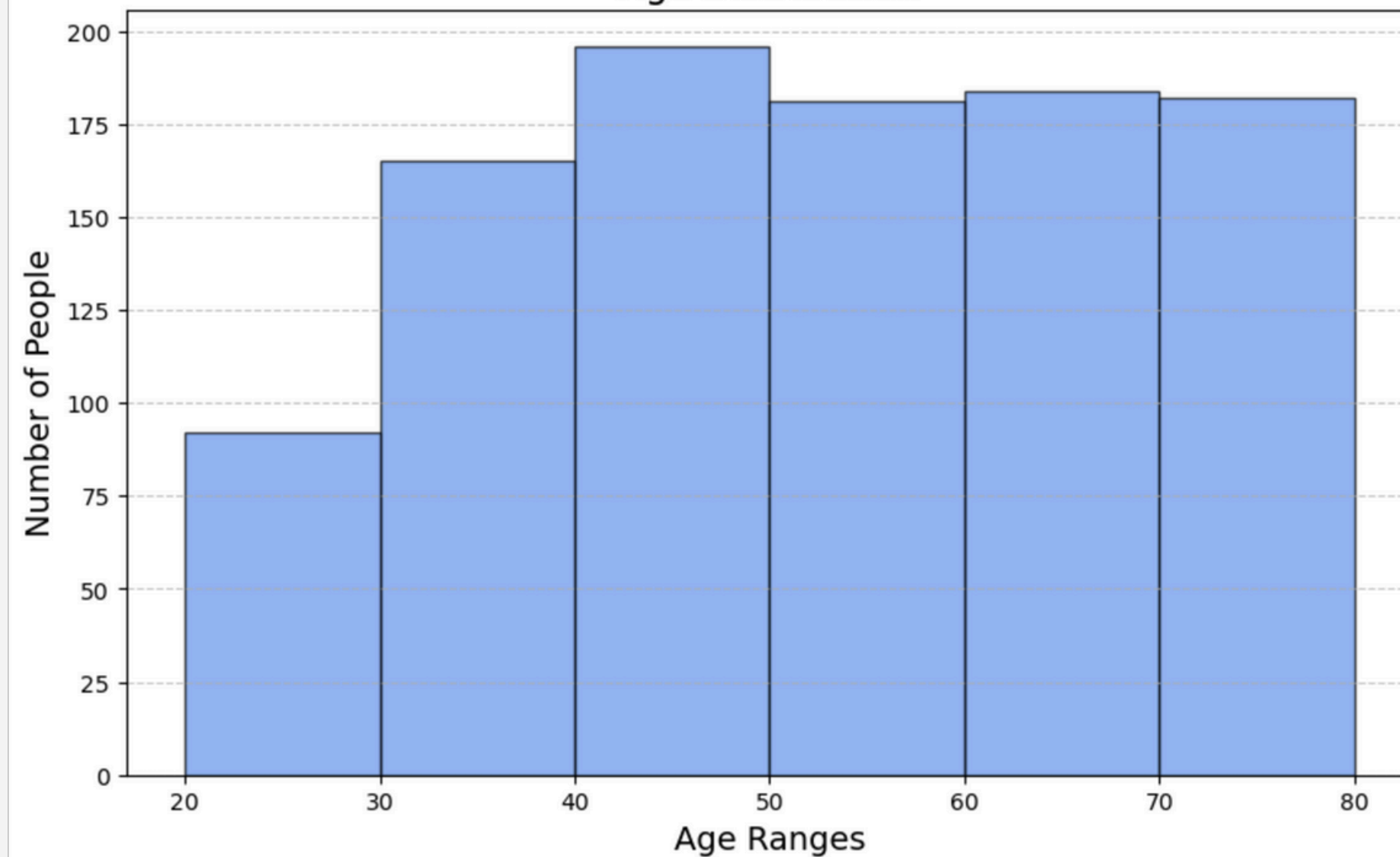
“None” being interpreted as null values in ‘Alcohol Intake’

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1000 entries, 0 to 999
Data columns (total 16 columns):
#   Column                                Non-Null Count  Dtype
---  -
0   Age                                   1000 non-null   int64
1   Gender                               1000 non-null   object
2   Cholesterol                           1000 non-null   int64
3   Blood Pressure                        1000 non-null   int64
4   Heart Rate                           1000 non-null   int64
5   Smoking                               1000 non-null   object
6   Alcohol Intake                        660 non-null    object
7   Exercise Hours                       1000 non-null   int64
8   Family History                       1000 non-null   object
9   Diabetes                             1000 non-null   object
10  Obesity                              1000 non-null   object
11  Stress Level                         1000 non-null   int64
12  Blood Sugar                          1000 non-null   int64
13  Exercise Induced Angina              1000 non-null   object
14  Chest Pain Type                      1000 non-null   object
15  Heart Disease                        1000 non-null   int64
dtypes: int64(8), object(8)
memory usage: 125.1+ KB
```

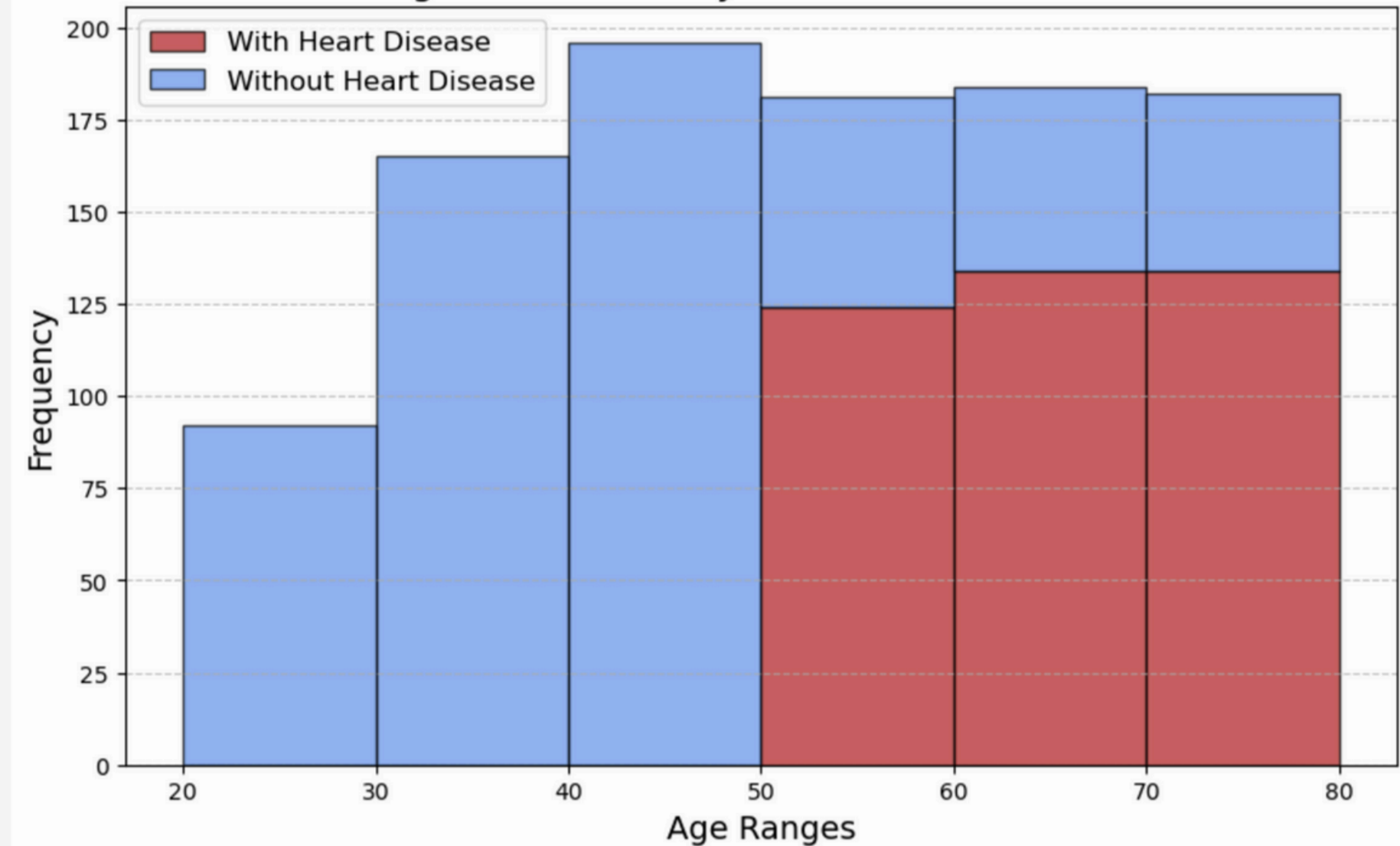
	Age	Cholesterol	Blood Pressure	Heart Rate	Exercise Hours	Stress Level	Blood Sugar	Heart Disease
count	1000.000000	1000.000000	1000.0000	1000.000000	1000.000000	1000.000000	1000.000000	1000.000000
mean	52.293000	249.939000	135.2810	79.204000	4.529000	5.646000	134.941000	0.392000
std	15.727126	57.914673	26.3883	11.486092	2.934241	2.831024	36.699624	0.488441
min	25.000000	150.000000	90.0000	60.000000	0.000000	1.000000	70.000000	0.000000
25%	39.000000	200.000000	112.7500	70.000000	2.000000	3.000000	104.000000	0.000000
50%	52.000000	248.000000	136.0000	79.000000	4.500000	6.000000	135.000000	0.000000
75%	66.000000	299.000000	159.0000	89.000000	7.000000	8.000000	167.000000	1.000000
max	79.000000	349.000000	179.0000	99.000000	9.000000	10.000000	199.000000	1.000000

EXPLORING AGE

Age Distribution

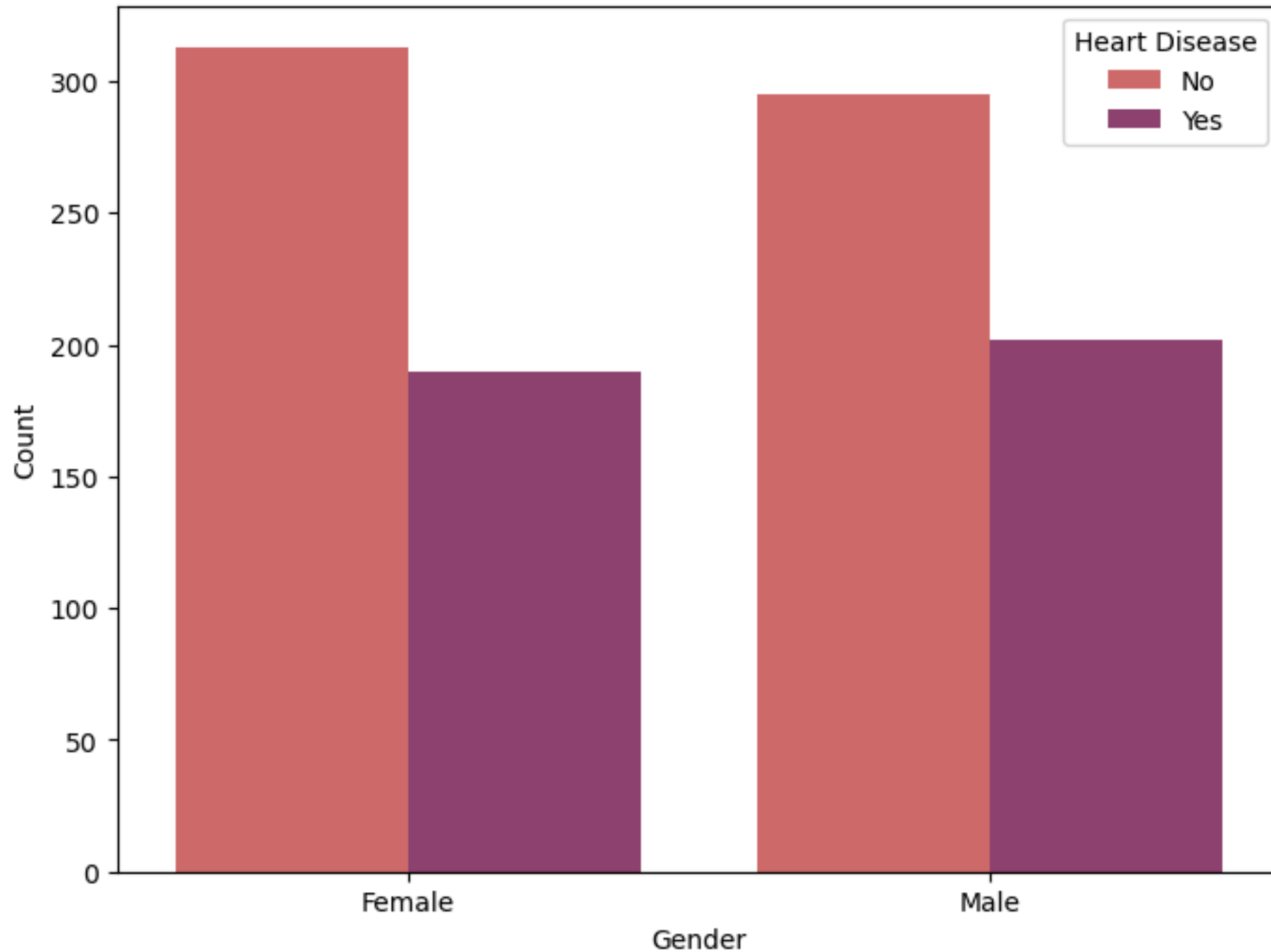


Age Distribution by Heart Disease Status

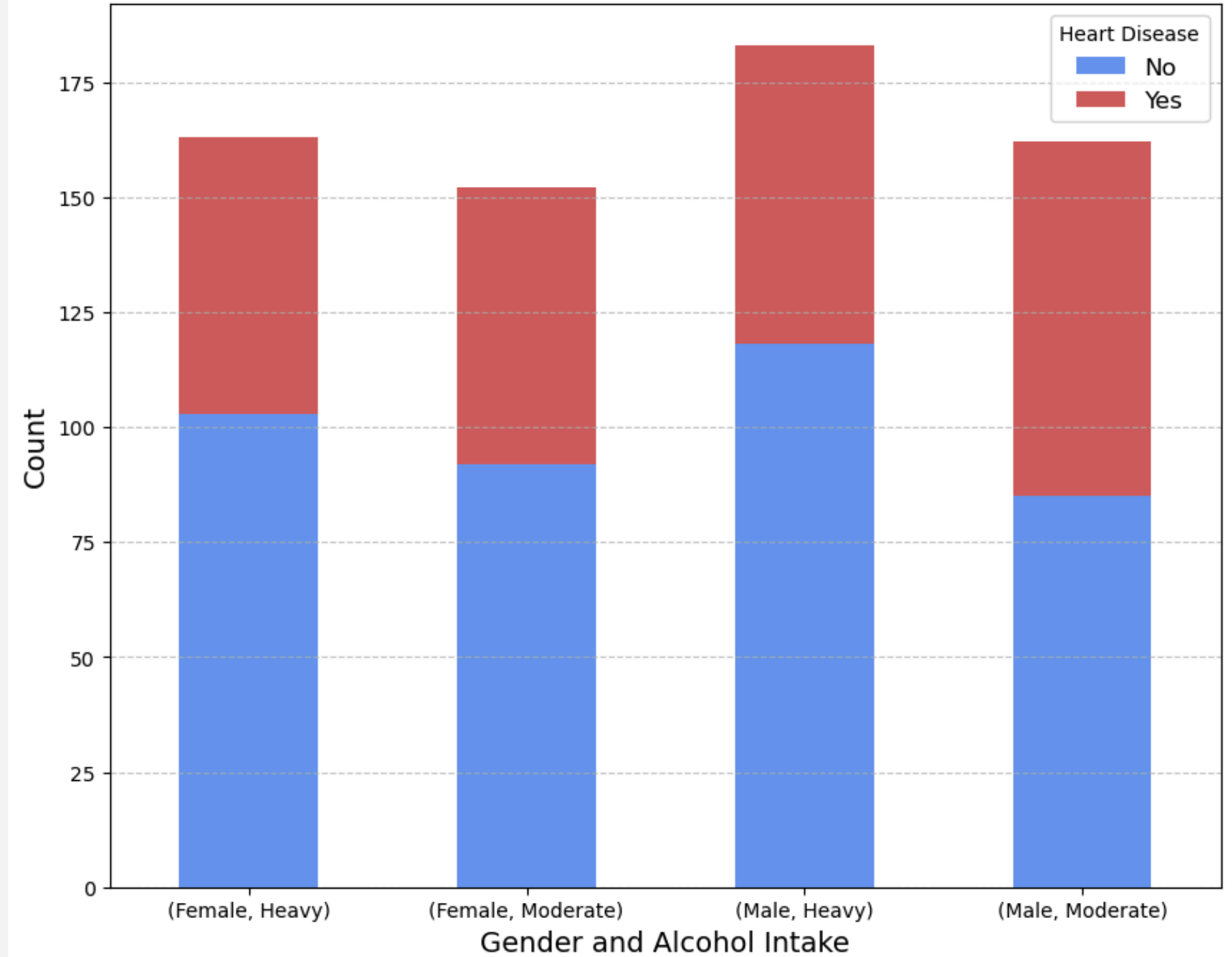


EXPLORING GENDER

Heart Disease Distribution by Gender

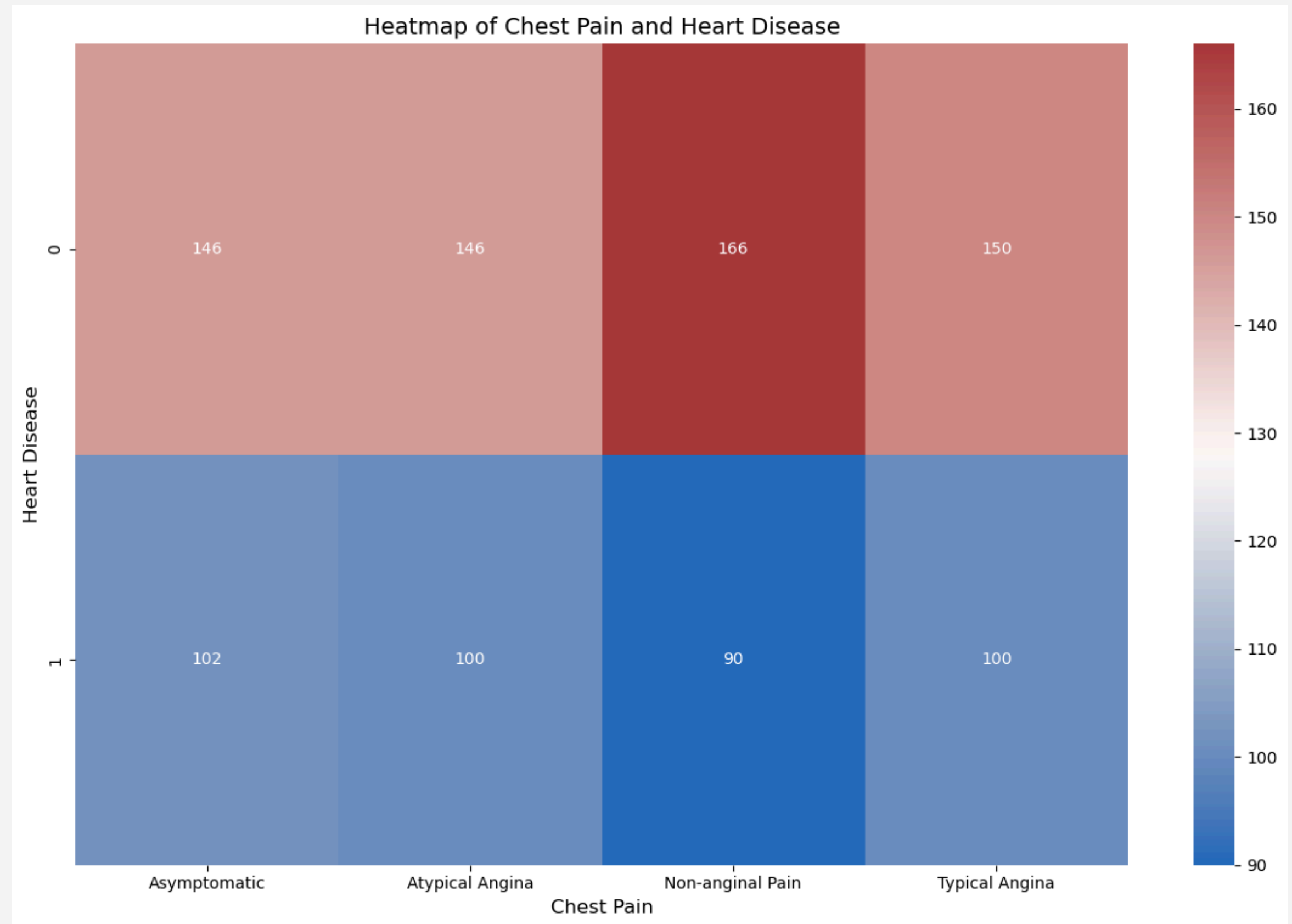


Gender, Heart Disease, and Alcohol Intake



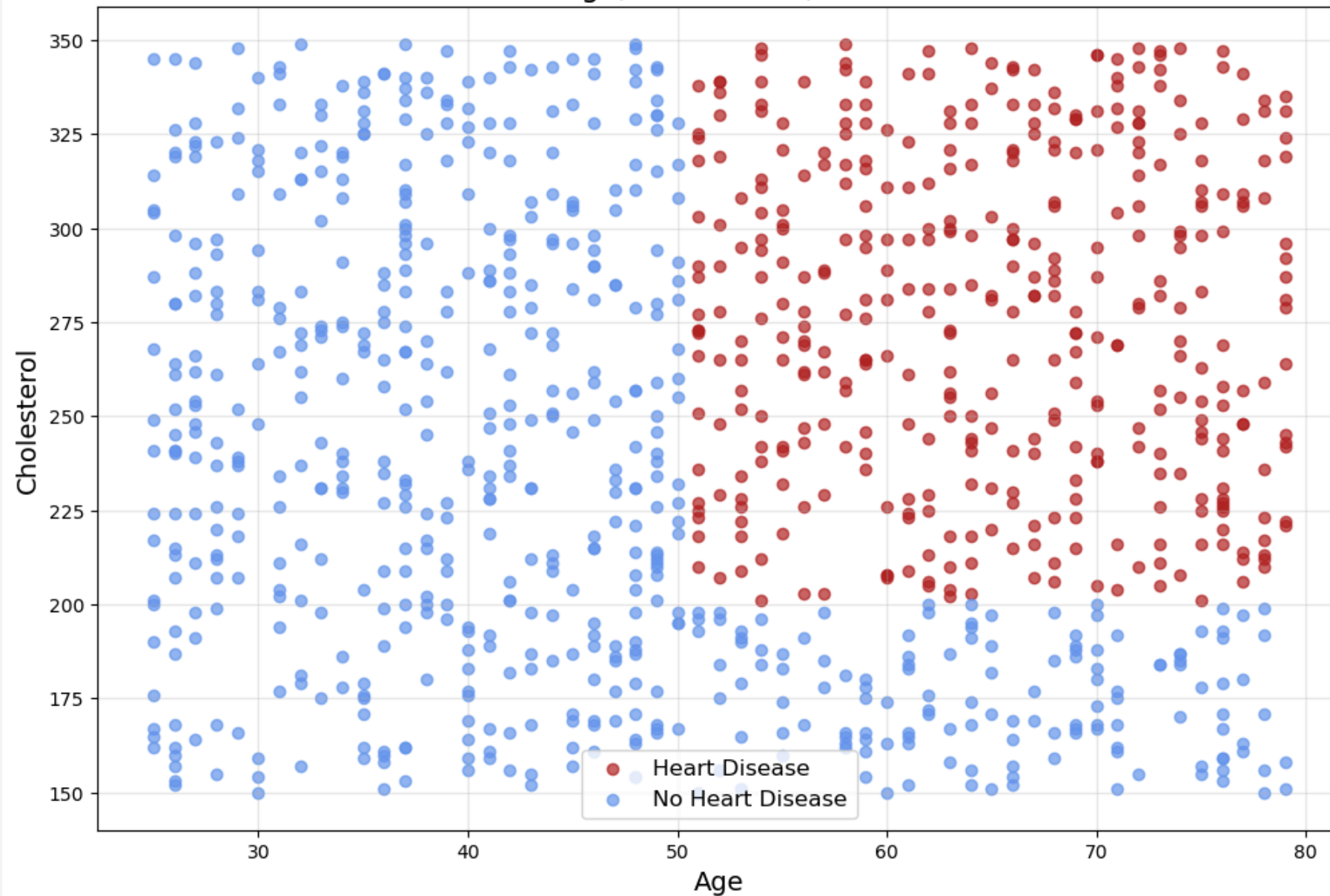
EXPLORING CHEST PAIN

- Asymptomatic and Atypical Angina have relatively balanced distributions of heart disease cases, which may warrant deeper analysis into other co-factors.
- Typical Angina, while a classic symptom, does not have the strongest association with heart disease in this dataset, suggesting potential underdiagnosis in cases with other pain types.

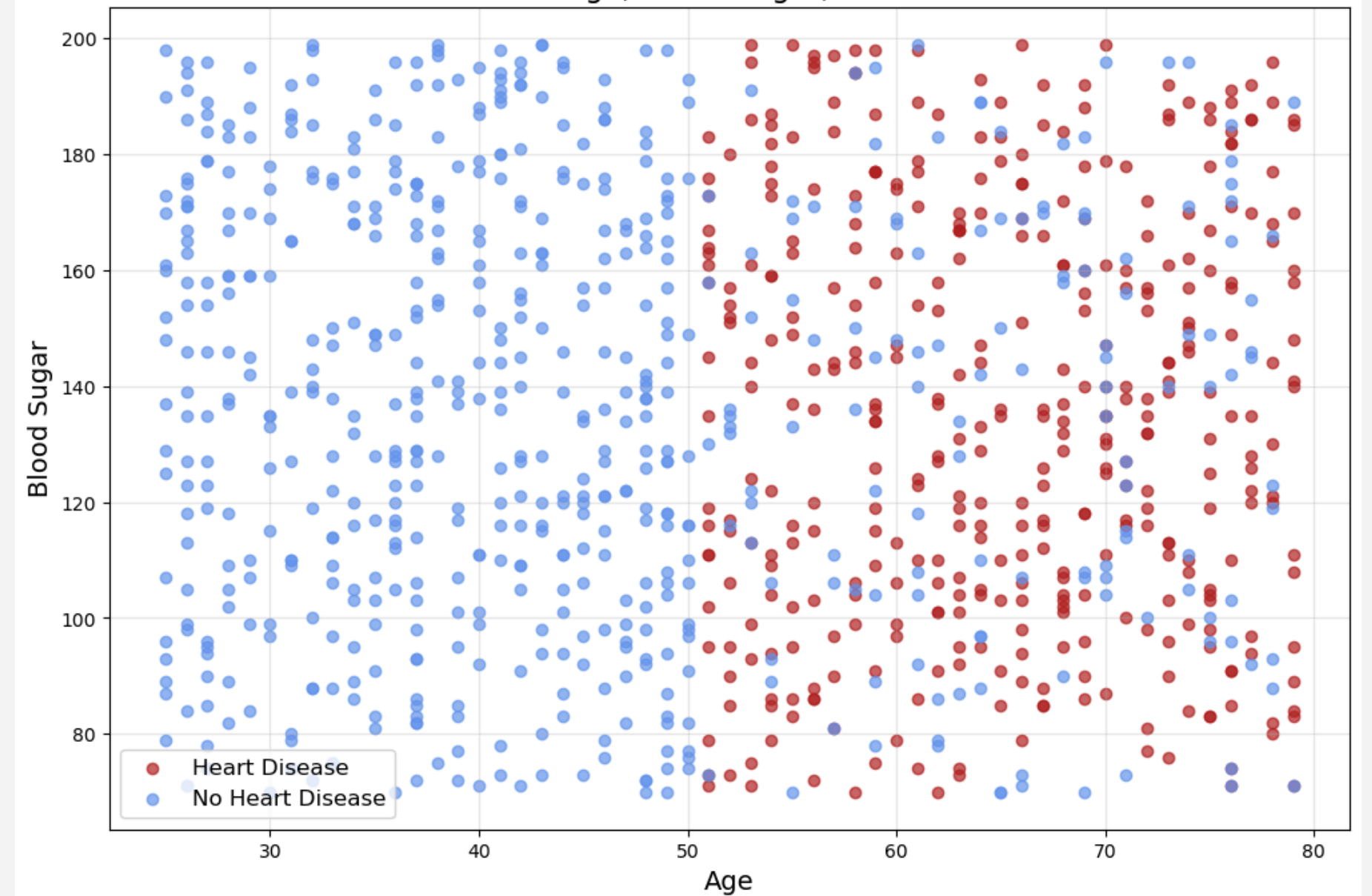


EXPLORING CHOLESTEROL & BLOOD SUGAR

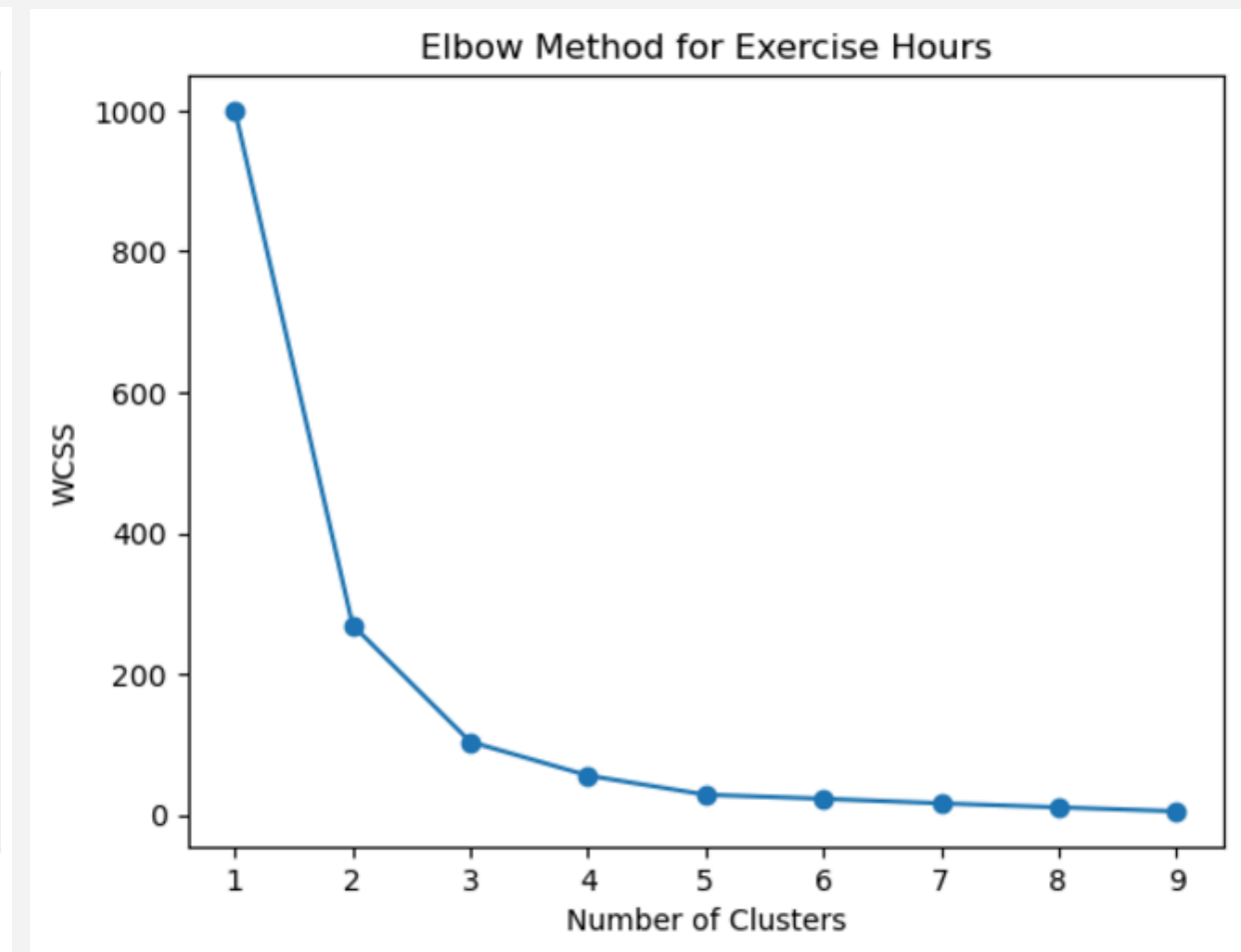
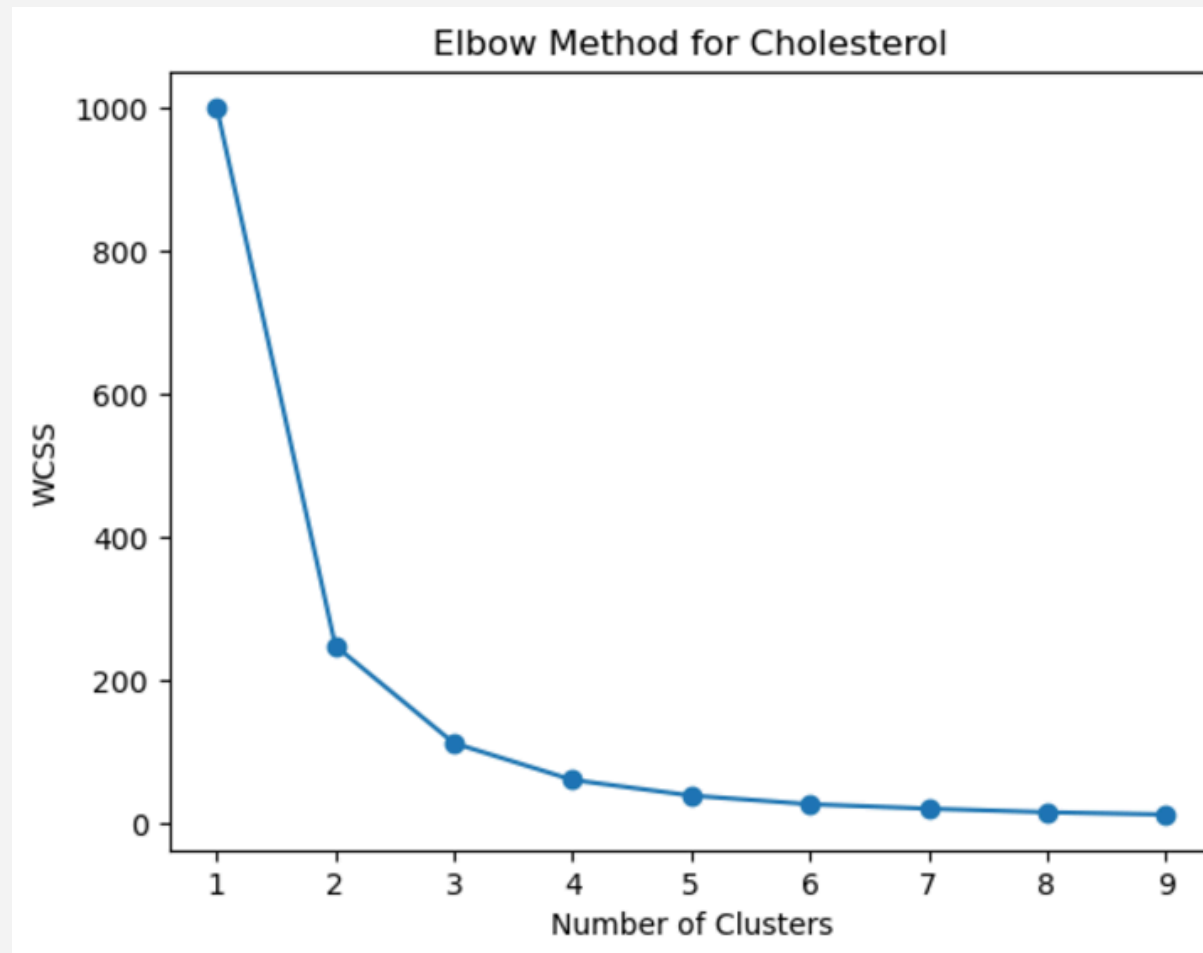
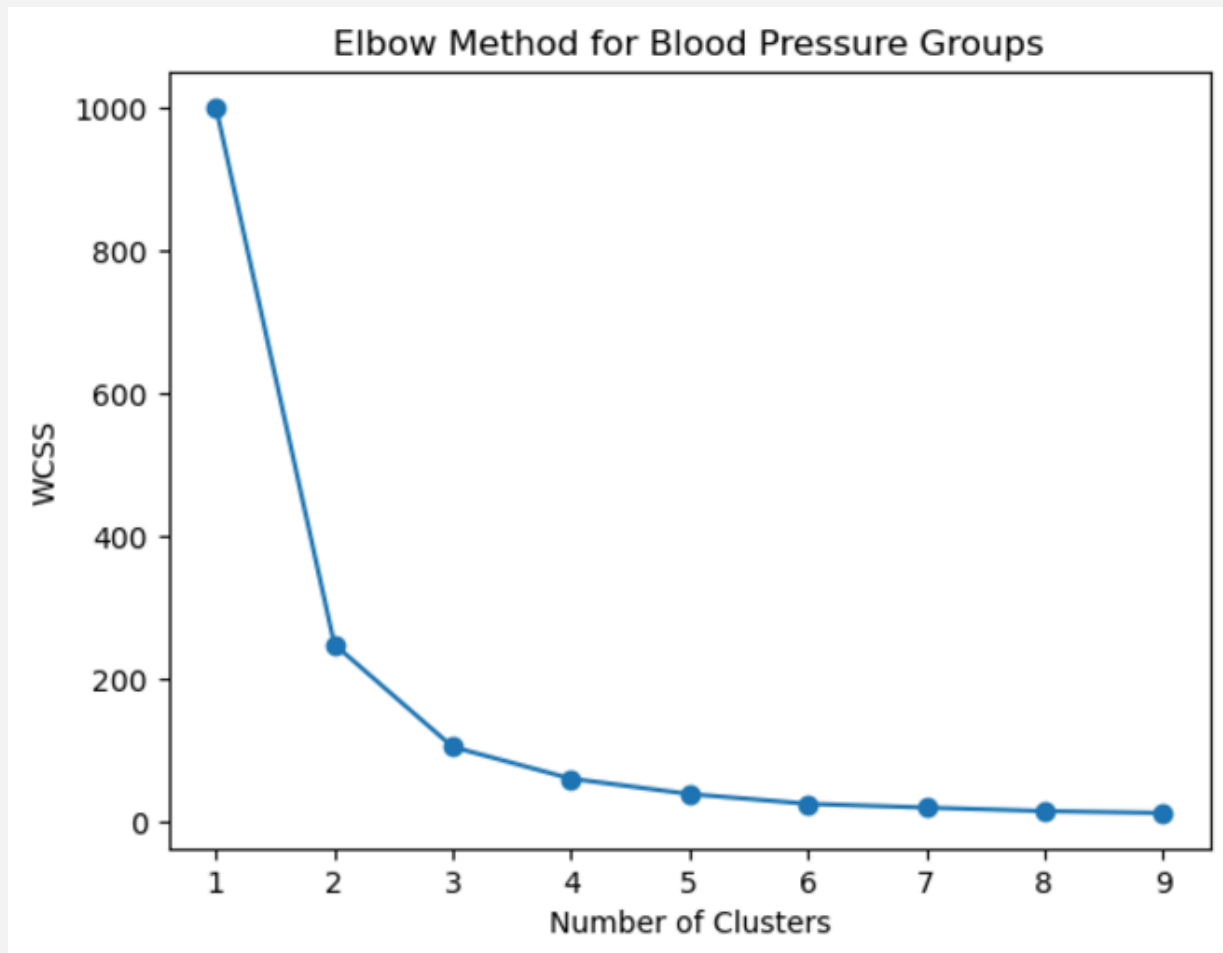
Scatter Plot of Age, Cholesterol, and Heart Disease



Scatter Plot of Age, Blood Sugar, and Heart Disease

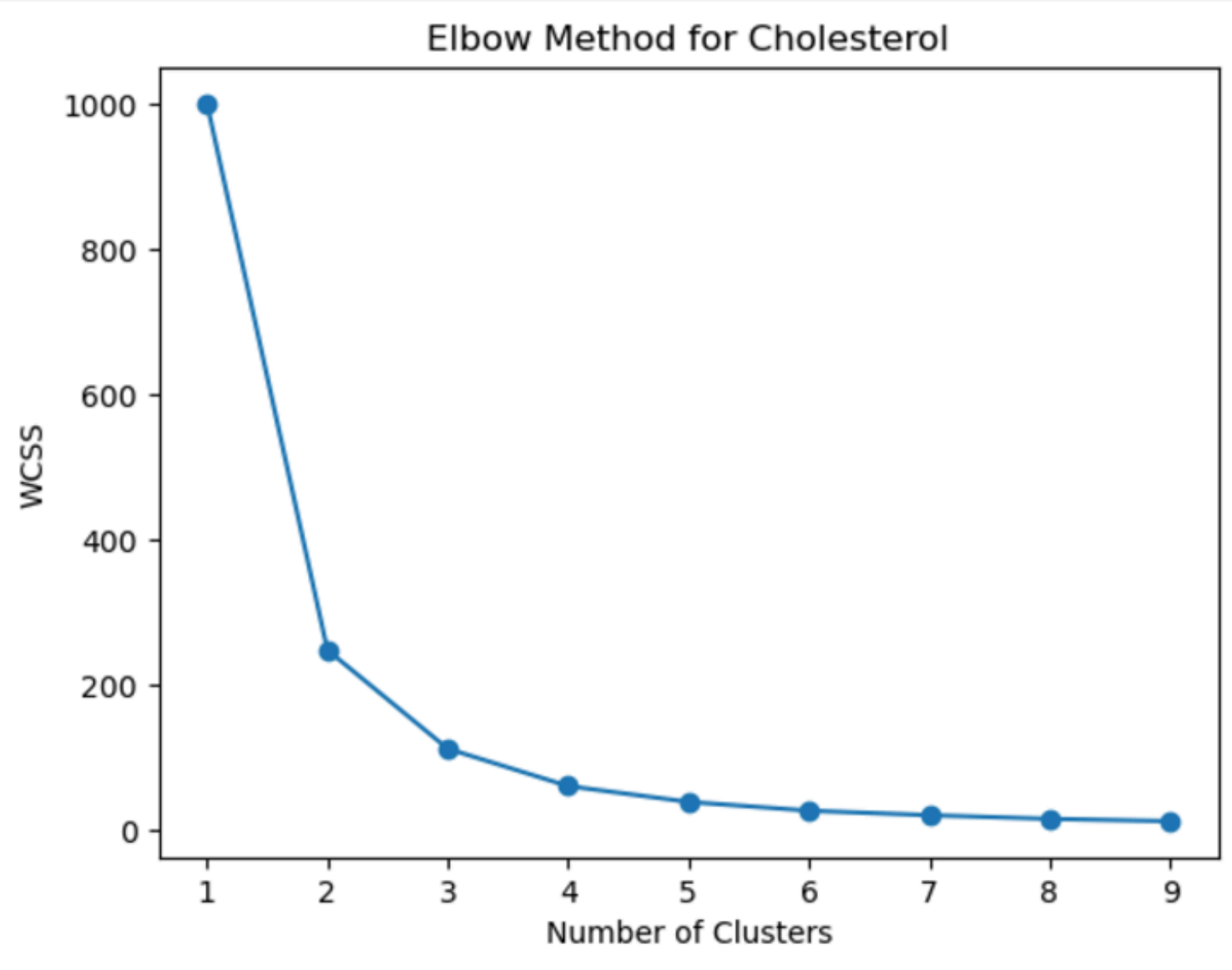


CLUSTER ANALYSIS

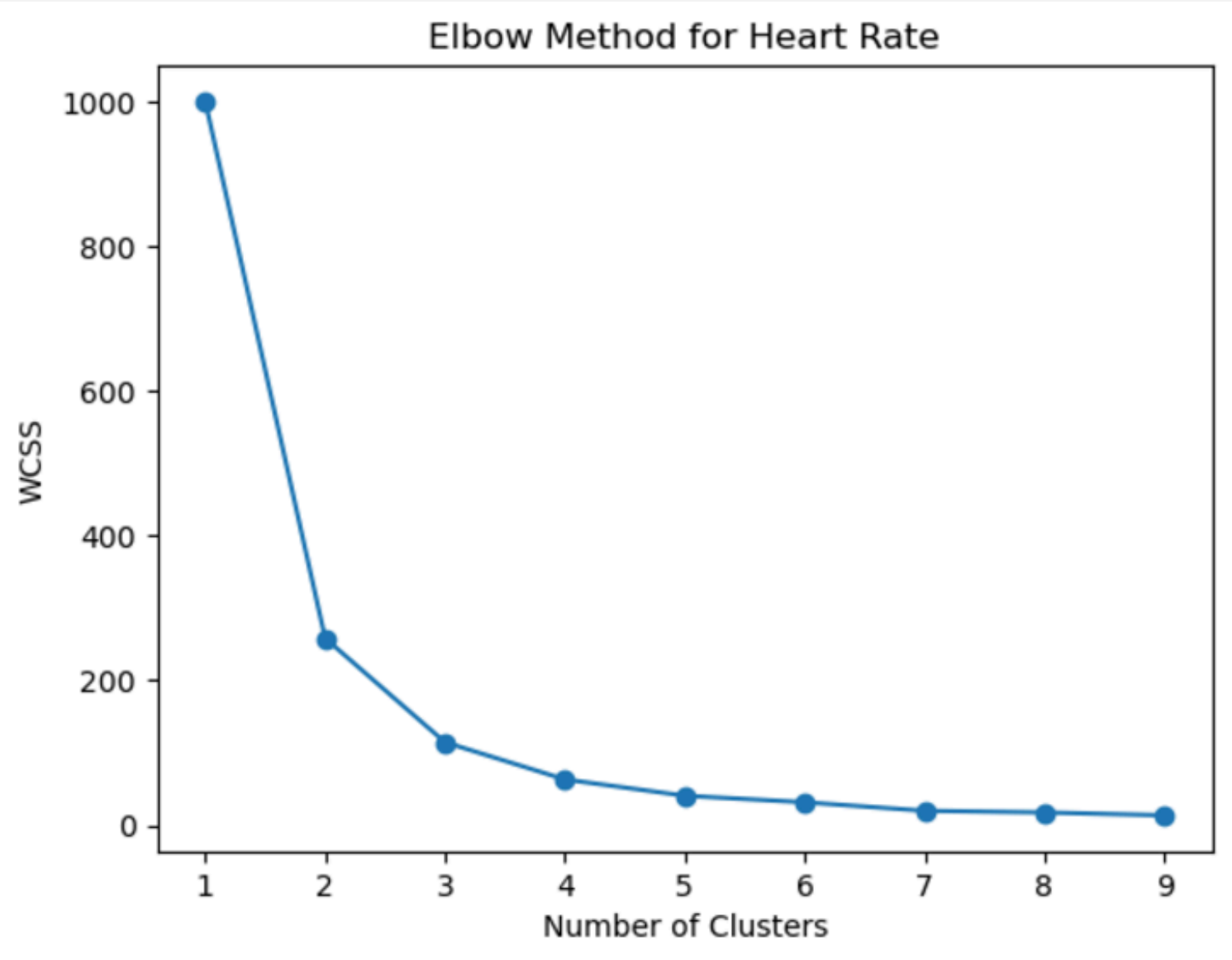


- Elbow Curve forms when number of clusters are 3 and 4
- Same trend was observed for almost all the variables affecting heart diseases

CLUSTER ANALYSIS



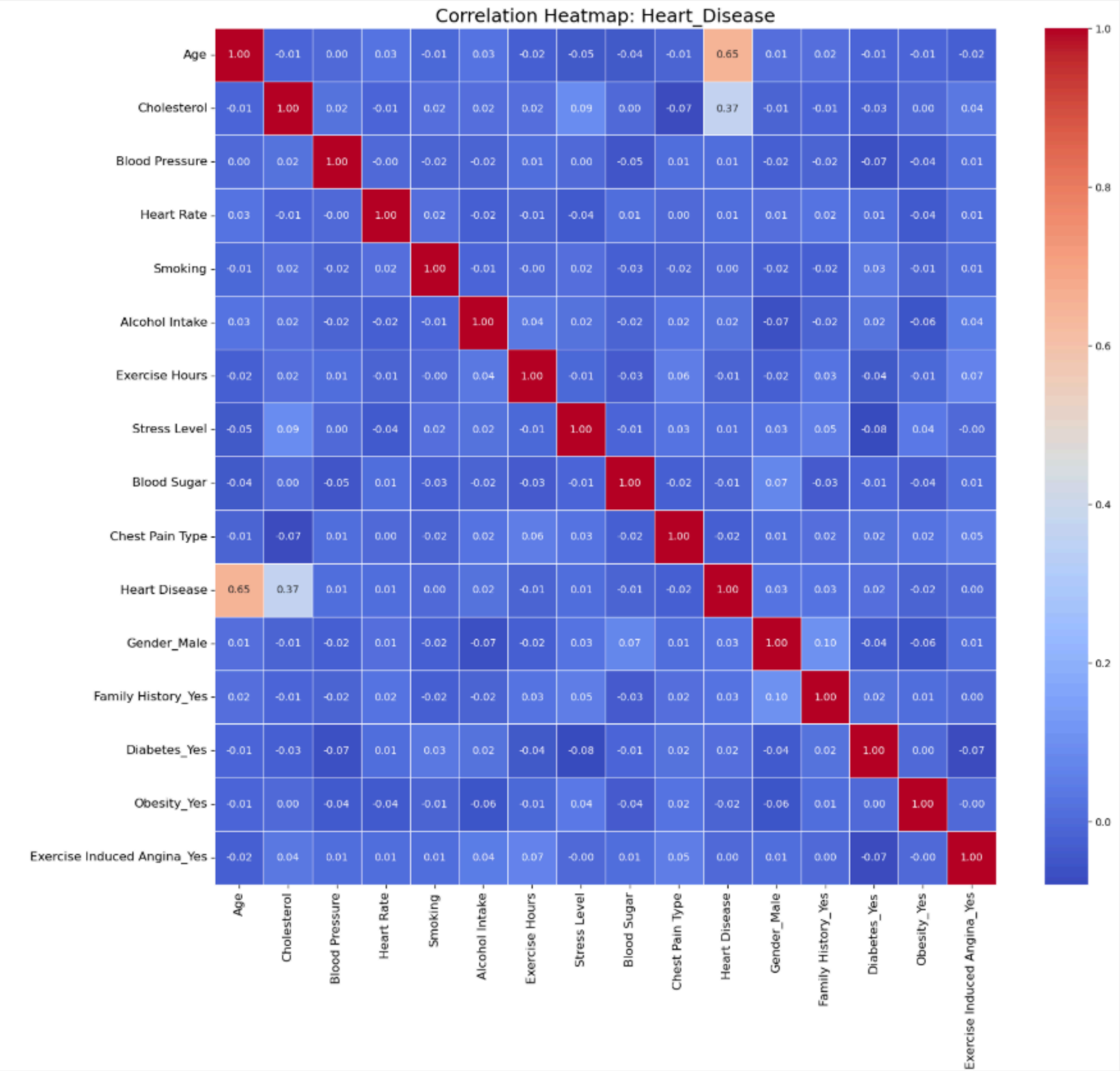
Cholesterol Levels	Heart disease patient count
150-218	47
219-285	171
286-349	174



Heart Rate Ranges	Heart disease patient count
60-73	138
74-87	141
88-99	113

HEATMAP

- Visual representation on how various features correlate with heart disease
- No Multicollinearity



WHY PREDICTIVE MODELING?

- To accurately assess heart disease risk based on multiple health and lifestyle factors, enabling early detection and prevention
- Enhances decision-making, supports personalized care, and improves public health by focusing on high-risk groups



Logistic regression

Ideal for binary classification tasks like predicting heart disease presence



Random Forest

Robust model that handles complex relationships.

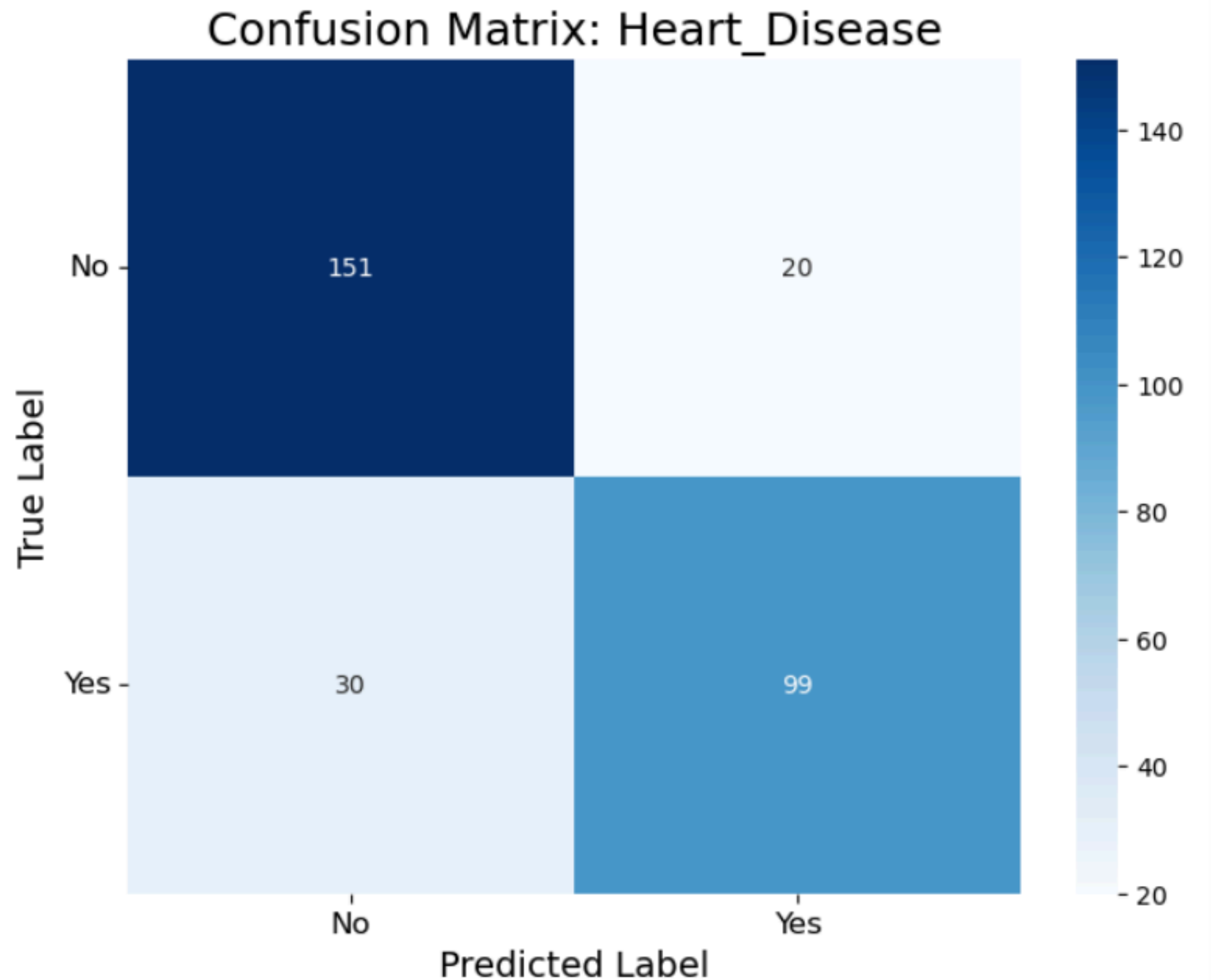


K-Nearest Neighbors

Effective for making predictions based on the similarity between data points

CONFUSION MATRIX

- **True Positives:** 76.4% of the actual heart disease cases were correctly identified
- **True Negatives:** 83.8% of the healthy individuals were correctly identified as not having heart disease
- **False Positives:** 13.2% of individuals were incorrectly predicted to have heart disease
- **False Negatives:** 19.1% of heart disease cases were missed by the model, indicating potential areas for improvement



LOGISTIC REGRESSION

Accuracy Score: 83.33

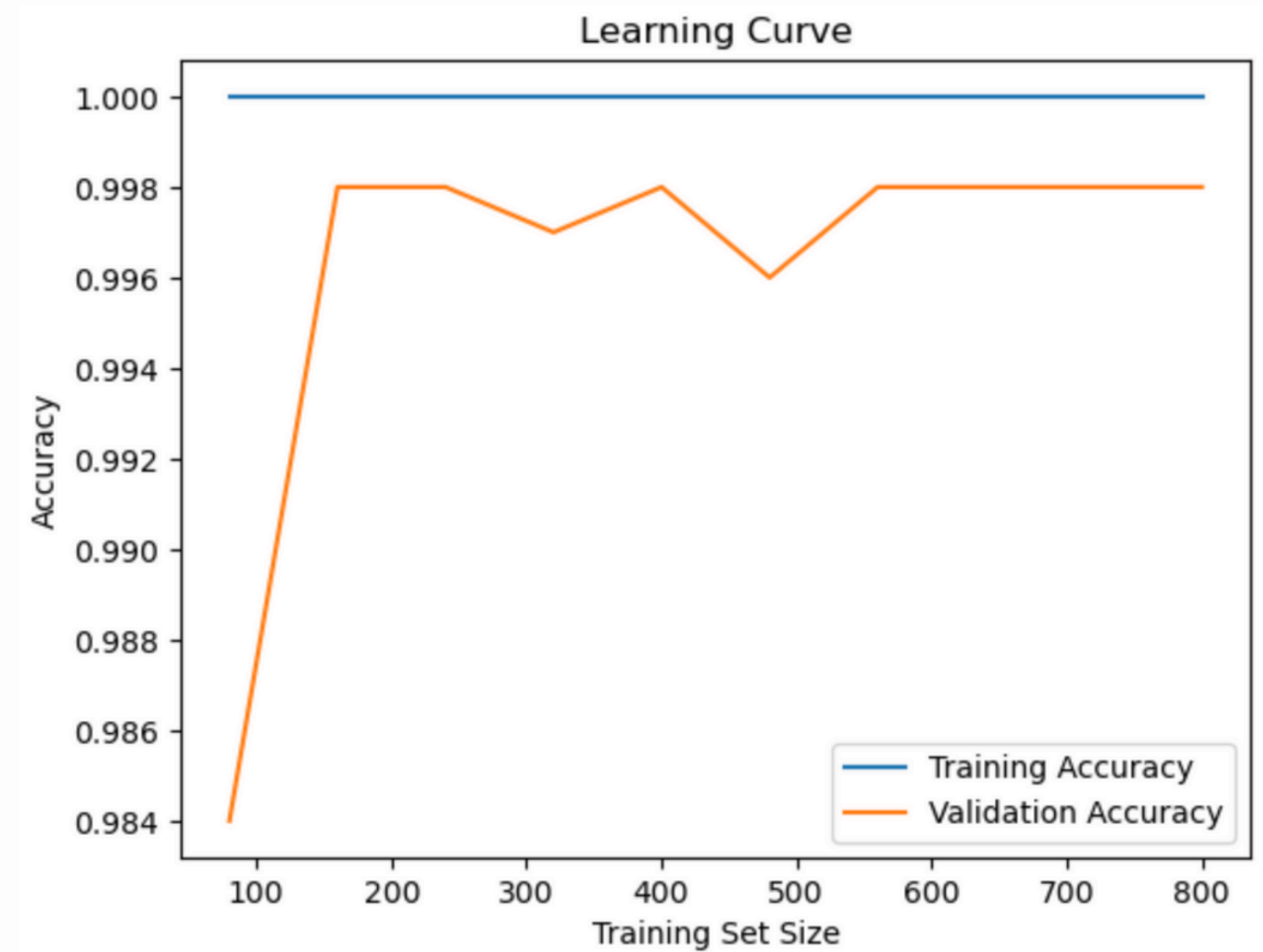
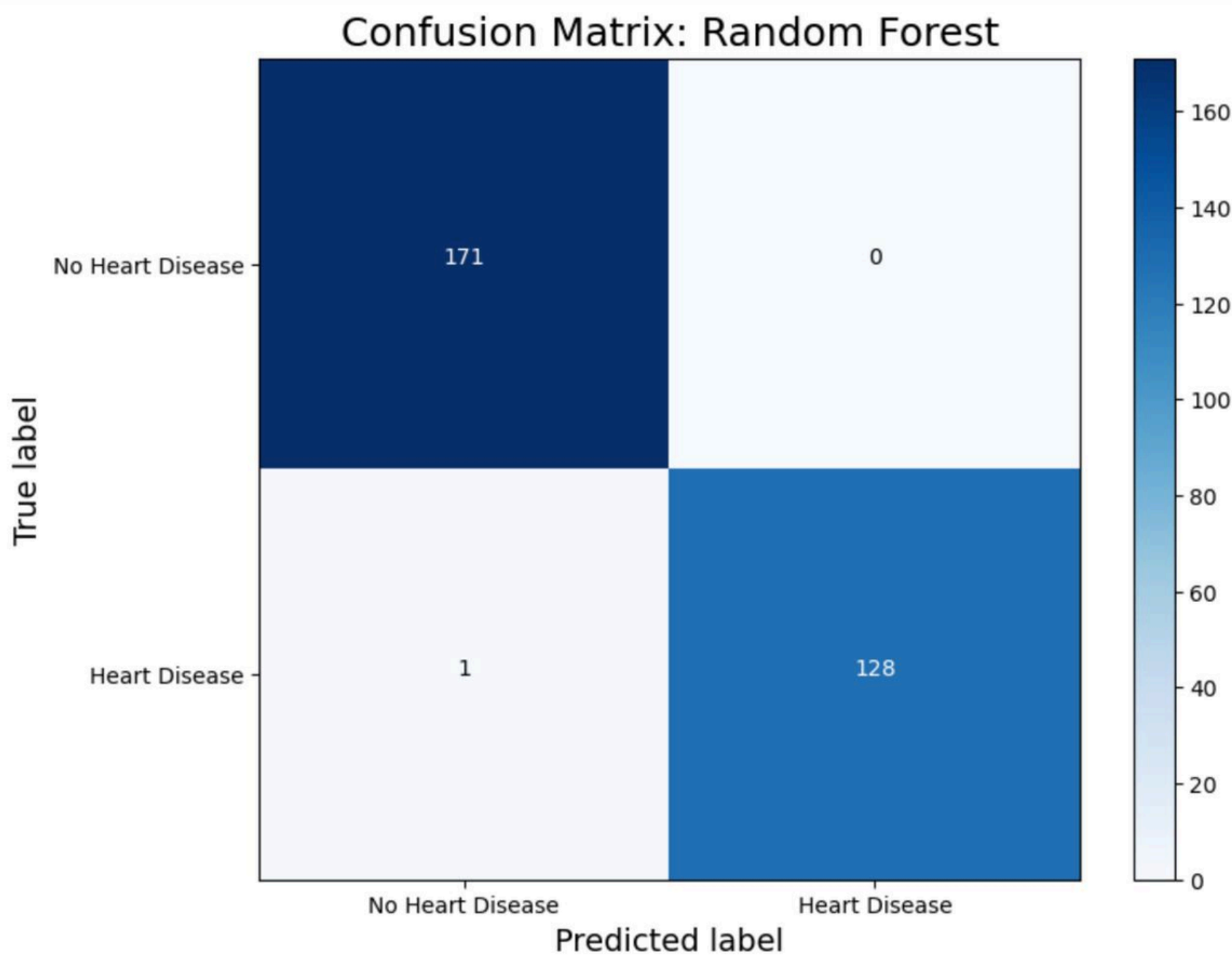
Logistic Regression	Precision	Recall	F1-score	Support
No Heart Disease	0.83	0.88	0.86	171
Heart Disease	0.83	0.77	0.70	129

K-NEAREST NEIGHBORS

Accuracy Score: 87.33

KNN	Precision	Recall	F1-score	Support
No Heart Disease	0.87	0.92	0.89	171
Heart Disease	0.88	0.81	0.85	129

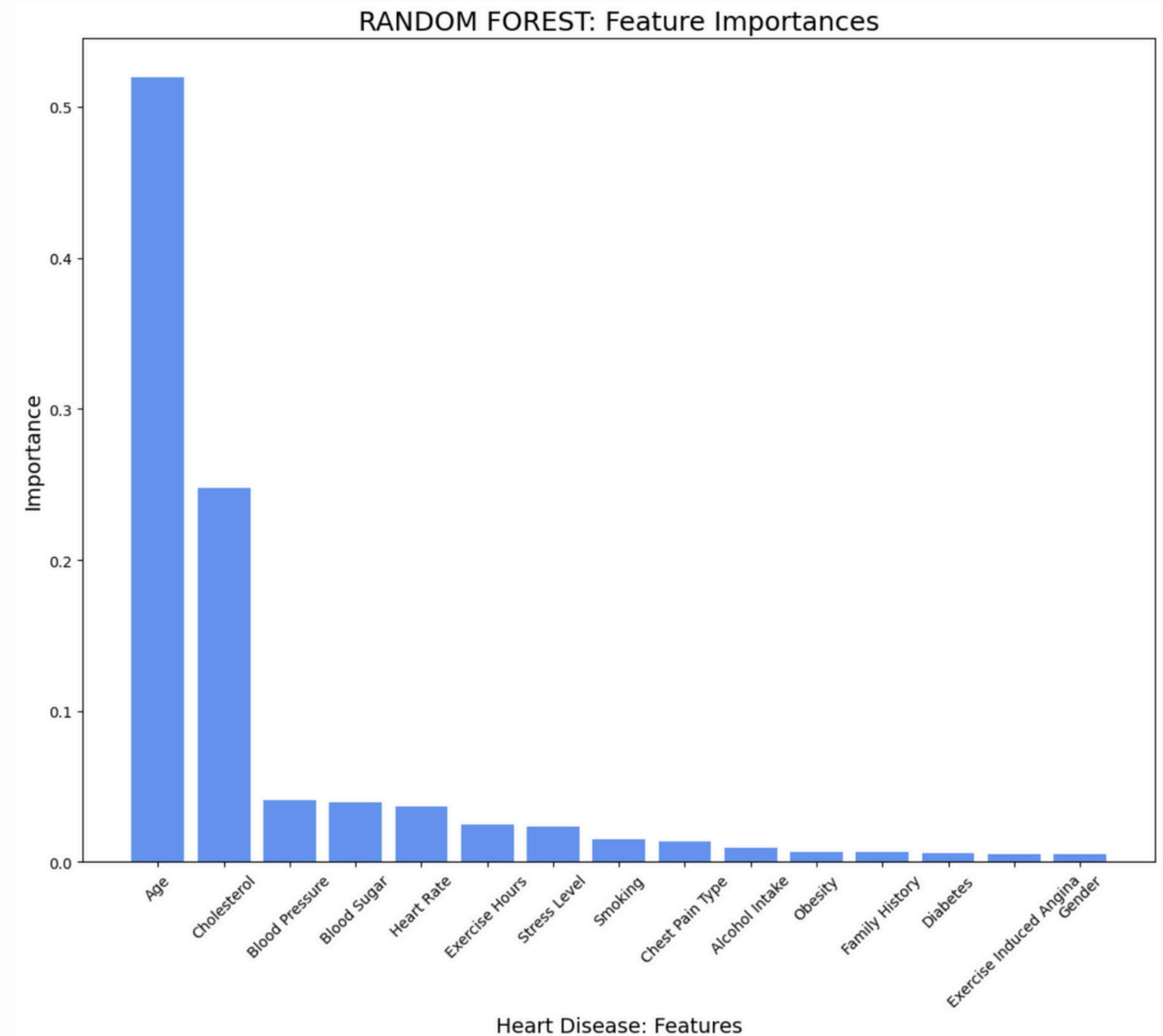
RANDOM FOREST



Cross Validation Accuracy: 99.8%

RANDOM FOREST

- *Accuracy: 99.9%*
- Age and Cholesterol are the most influential factors in predicting heart disease, significantly outweighing all other features.
- Lifestyle and demographic factors like Smoking, Stress Level, and Gender have minimal impact in this model.



INSIGHTS

- People above the age of 50 are more susceptible to heart disease
- Cholesterol levels above 218 may significantly increase chances of developing a heart disease.
- Typical Angina did not show any strong correlations in this data suggesting potential under diagnosis of other pain symptoms
- Moderate alcohol intake increases the possibility of heart disease

RECOMMENDATIONS

- Increase sample size of data to improve representation of people less than the age of 40
- Healthcare providers should evaluate atypical/typical angina's and other chest pain types to prevent misdiagnosis
- Insurance companies can target a better coverage plan for people above the age of 50
- Individuals should understand the risks of cholesterol and intake of alcohol
- Startup's can leverage this data to build and leverage products that track cholesterol

