## **Heart Disease Data Analysis & Prediction Model**

### **Overview**

- Heart disease is a leading cause of mortality globally, emphasizing the importance of early detection.
- The dataset includes features like age, sex, chest pain type, cholesterol levels, blood pressure, and other clinical indicators.
- Data analysis and machine learning techniques are used to identify patterns and predict heart disease risk.
- The project aims to improve healthcare decision-making and patient outcomes.

## Methodology

#### **Data Preprocessing:**

- Handle missing values, outliers, and inconsistent data.
- Normalize or standardize continuous variables.
- Encode categorical variables using label encoding or one-hot encoding.

#### **Exploratory Data Analysis (EDA):**

- Visualize data distributions and relationships.
- Identify trends and patterns in the dataset.
- Correlation analysis to determine feature importance.

#### **Feature Selection:**

 Use statistical methods or algorithms like Recursive Feature Elimination (RFE) to select the most relevant features.

#### Model Building:

 Train machine learning models like Logistic Regression, Random Forest, and XGBoost.

#### **Model Evaluation:**

 Assess models using metrics like accuracy, precision, recall, F1score, and ROC-AUC.

#### **Deployment:**

• Deploy the model as a web application or integrate it into a healthcare system for real-time predictions.

## **Heart Disease Distribution**

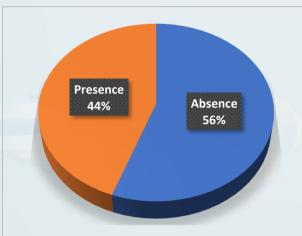
#### Insights:

- The number of individuals without heart disease (150) is higher than those with heart disease (120).
- This suggests a slightly lower prevalence of heart disease in the dataset.

#### **Distribution Summary:**

- Approximately 44.4%
   (120 out of 270) of the
   dataset represents
   individuals with heart
   disease.
- The remaining 55.6% (150 out of 270) represents individuals without heart disease.

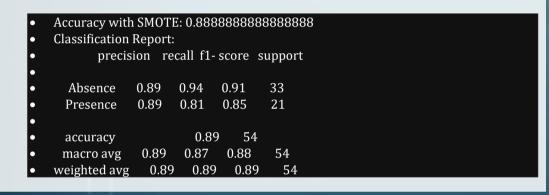
400



Heart Disease

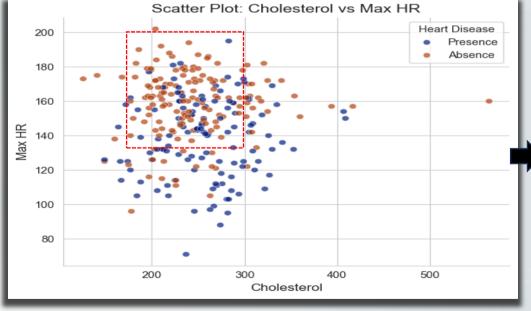
#### **SMOTE Classification Results**

- The model performs well overall, achieving high precision, recall, and F1-scores for both classes.
- SMOTE effectively balances the dataset, improving the model's ability to handle the minority class ("Presence").



## **Age vs Cholesterol Distribution**

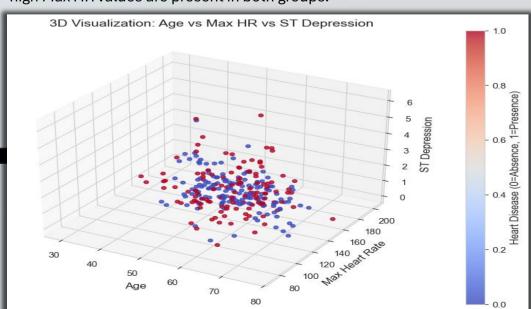
- Cholesterol levels vary widely across all age groups but appear to cluster between 200 and 300 mg/dL for most individuals.
- There are a few outliers with cholesterol levels exceeding 400 mg/dL.
- Most data points appear to be concentrated in the age range of 40 to 70 years.
- There is significant overlap between the cholesterol levels of those with and without heart disease. This suggests that cholesterol alone may not be a strong indicator of heart disease when age is considered.



#### 30 40 50 60 Age

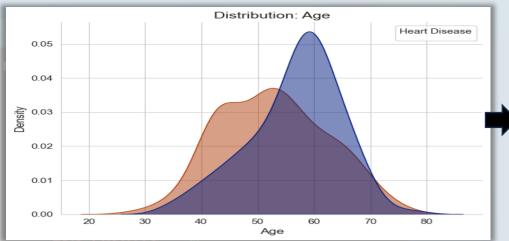
**Cholesterol vs Max HR Distribution** 

- Cholesterol levels for both groups are primarily concentrated between 200 and 300 mg/dL.
- There are a few individuals with cholesterol levels exceeding **400 mg/dL**, but these are rare outliers.
- Maximum heart rate (Max HR) values are generally between 80 and 200
   bpm for both groups.
- There is a noticeable cluster of data points around Max HR = 140–180 bpm, indicating that most individuals achieve their heart rate in this range.
- A few outliers exist where individuals with high cholesterol and relatively high Max HR values are present in both groups.



# Relationship between Age, Max Heart Rate, ST Depression, and Heart Disease

- Age vs. Max Heart Rate: As age increases, maximum heart rate generally decreases.
- ST Depression vs. Heart Disease: Higher ST depression values are strongly linked to a higher prevalence of heart disease.
- Age vs. ST Depression: ST depression slightly increases with age, but the trend is not very strong.
- Max Heart Rate vs. ST Depression: No clear relationship is observed between maximum heart rate and ST depression.



## **Distribution of Age and Heart Disease**

1.0

0.8

0.6

0.2

0.0

- Overall age distribution is right-skewed, with a peak around 60 years.
- Age distribution for individuals with heart disease (blue line) is also right-skewed, peaking slightly earlier, around 55-60 years.
- The blue line overlaps with the orange line at **younger ages**, indicating a **lower prevalence of heart disease** in this group.
- As age increases, the blue line becomes more prominent, showing an **increased risk of heart disease** in the middle-aged and older population.

Distribution: Chest pain type

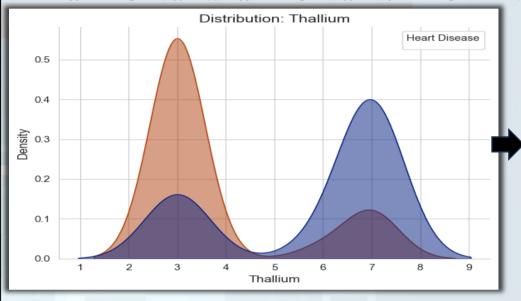
Heart Disease

Heart disease risk increases with age, particularly after 55 years.



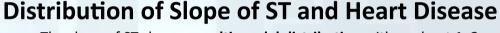
- The overall distribution is multi-modal, with peaks around chest pain types 2, 3, and 4. epression : 2.5,
- The distribution for heart disease cases is also multi-modal but shows a distinct peak at type 4, indicating a higher likelihood of heart disease.
- The blue line overlaps with the orange line for chest pain types 2 and 3, suggesting a lower prevalence of heart disease for these types.
- For chest pain type 4, the blue line is significantly higher than the orange line, indicating a strong correlation with heart disease.
- Chest pain type 4 is a strong indicator of heart disease compared to other types.

NOTE:- Typical Angina (Type 1) | Atypical Angina (Type 2) | Non-Anginal Pain (Type 3) | Asymptomatic (Type 4).

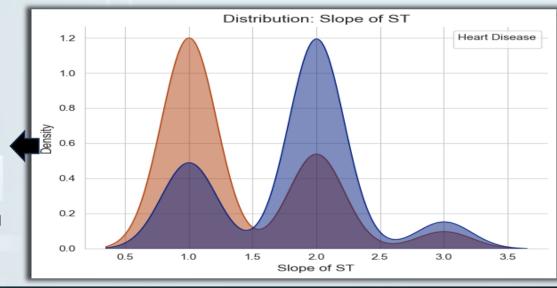


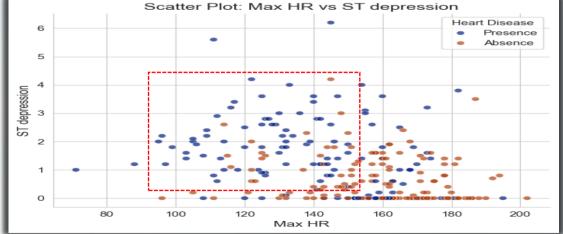
## **Distribution of Thallium and Heart Disease**

- The distribution is bimodal, with peaks around 3 and 7.
- For individuals with heart disease (blue line), the distribution is also bimodal.
- A more pronounced peak is observed around 7, suggesting a higher likelihood of heart disease.
- The blue line overlaps with the orange line at **lower thallium values**, indicating a **lower prevalence** of heart disease in this range.
- For higher thallium values (around 7), the blue line rises significantly above the orange line.
- This highlights a strong association between higher thallium values and heart disease.
- Thallium values closer to 7 are indicative of a higher risk of heart disease, while lower values show less association.



- The slope of ST shows a multi-modal distribution with peaks at 1, 2, and 3.
- For individuals with **heart disease** (blue line), the distribution is also multimodal but has a **distinct peak at 2**.
- A **slope of ST = 2** is strongly associated with heart disease, as indicated by the higher blue line at this value.
- For **slope values = 1 and 3**, the blue line overlaps with the orange line, indicating a **lower likelihood** of heart disease.
- A **slope of 2** is a potential risk factor for heart disease, while slopes of 1 and 3 show less association.





## **Max HR vs ST Depression Distribution**

- Max HR values range from **80 to 200 bpm**, with most data points clustered between **120 and 160 bpm**.
- ST depression values range from **0 to 6** but are concentrated below **2** for most individuals.
- There is a visible trend indicating that individuals with lower Max HR and higher ST depression are more likely to have heart disease.
- While there is overlap in the data, the combination of low Max HR (<140 bpm) and ST depression (>1) shows a stronger association with heart disease.
- Individuals with Max HR >160 bpm and ST depression close to 0 are mostly without heart disease.

#### **Heart Disease Prediction Model**

joblib.dump(model, 'heart\_disease\_model.pkl')
print("Model saved as 'heart\_disease\_model.pkl'")
Model saved as 'heart\_disease\_model.pkl'

## **Model Testing**

```
_input = {
     Chest pain type': 3,
    BP': 140,
     Cholesterol': 250,
     FBS over 120': 1,
    'EKG results': 0,
    'Max HR': 150,
     Exercise angina': 1,
    'ST depression': 2.5,
    'Slope of ST': 2,
    'Number of vessels fluro': 0,
    Thallium': 5
result = predict heart disease(new input)
print("Predicted Heart Disease:",
            if result == 1 else
Predicted Heart Disease: No
```