

### G H Patel College of Engineering & Technology

(A Constituent College of CVM University) Vallabh Vidyanagar



#### **Heart Disease Prediction**

An AiMl Mini Project report submitted
In partial fulfilment of the requirements for the
Degree of Bachelor of
Engineering/Technology

In Computer Engineering (CP) Semester – VI

#### **Submitted By**

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A.Y. 2024-25 ODD TERM



# G H Patel College of Engineering & Technology (A Constituent College of CVM University)



## **G H Patel College of Engineering & Technology**

#### **CERTIFICATE**

| This is to certify that Meet Dadhaniya (12202040501038) & Param Dholakia                 |
|--|
| (12202040501049) has submitted the Aiml Mini Project report on " Heart Disease           |
| Prediction " for partial fulfilment of the degree of Bachelor of Engineering in Computer |
| Engineering, G H Patel College of Engineering and Technology, at The Charutar Vidya      |
| Mandal (CVM) University, Vallabh Vidyanagar, during the academic year 2024 – 25.         |

Dr. Priyang Bhatt

(Internal Faculty Guide)

#### **INTRODUCTION**

This project aims to predict the likelihood of heart disease in individuals based on various health metrics using machine learning models. The primary goal is to leverage data-driven insights to identify patterns and risk factors associated with heart disease, enabling early detection and intervention. The dataset includes a range of health-related features, and three machine learning models—

- Random Forest,
- K-Nearest Neighbors (KNN), and
- Gradient Boosting

—are employed to classify individuals as having heart disease or not.

#### **DATASET DESCRIPTION**

The dataset contains the following features:

| ID                           | Unique identifier for each individual               |  |  |
|------------------------------|---|--|--|
| Age                          | Age of the individual (in years)                    |  |  |
| Gender                       | Gender of the individual (Male/Female)              |  |  |
| Height_c                     | Height in centimeters                               |  |  |
| Weight_kg                    | Weight in kilograms                                 |  |  |
| BMI                          | Body Mass Index                                     |  |  |
| Daily_Steps                  | Number of steps taken daily                         |  |  |
| Calories_Intake              | Daily calorie intake (in calories)                  |  |  |
| Hours_of_Sleep               | Hours of sleep per day                              |  |  |
| Heart_Rate                   | Resting heart rate (in beats per minute)            |  |  |
| Blood_Pressure               | Blood pressure reading (in mmHg)                    |  |  |
| Exercise_Hours_per_Week      | Hours of exercise per week                          |  |  |
| Smoker                       | Smoking status (Yes/No)                             |  |  |
| Alcohol_Consumption_per_Week | Alcohol consumption per week (in units)             |  |  |
| Diabetic                     | Diabetic status (Yes/No)                            |  |  |
| Heart_Disease                | Presence of heart disease (Yes/No, target variable) |  |  |

The dataset provides a comprehensive view of each individual's health profile, with Heart\_Disease as the target variable.

#### DATA PREPROCESSING

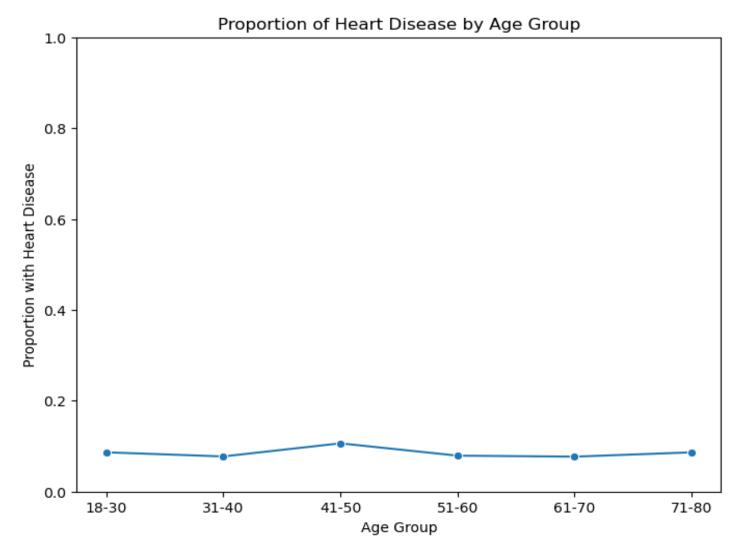
To prepare the dataset for modeling, the following preprocessing steps were applied:

- Categorical Variable Encoding: Categorical features such as
- Gender (Male=0, Female=1),
- Smoker (No=0, Yes=1),
- Diabetic (No=0, Yes=1), and
- Heart\_Disease (No=0, Yes=1) were mapped to numerical values.
- Blood Pressure Transformation: The Blood\_Pressure feature was split into two numerical features: Max\_BP (systolic) and Min\_BP (diastolic).

These steps ensured the dataset was numerical and ready for modeling.

#### **EXPLORATORY DATA ANALYSIS**

A line plot was created to show the proportion of individuals with heart disease across age groups, highlighting age as a potential risk factor.



#### MODEL TRAINING AND EVALUATION

Three models were trained:

- Random Forest: Ensemble of decision trees.
- KNN: Distance-based classifier.
- Gradient Boosting: Sequential tree ensemble.

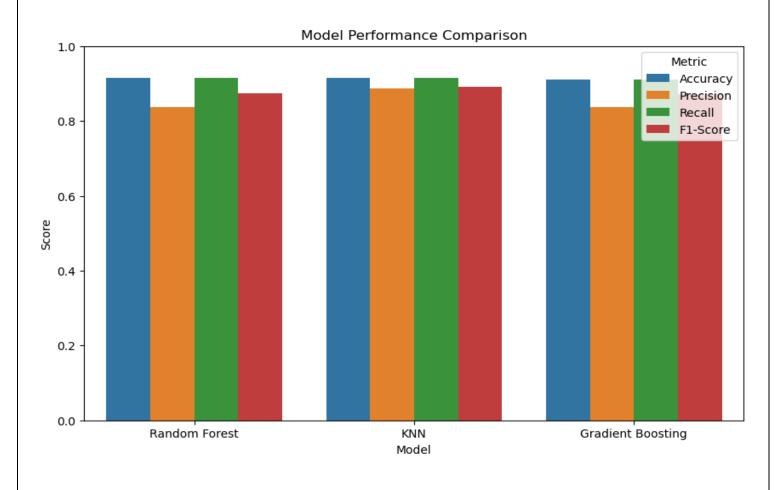
Performance was evaluated using accuracy, precision, recall, and F1-score.

#### **MODEL PERFORMANCE METRICS**

| Model    | Accuracy | Precision | Recall | F1-Score |
|----------|----------|-----------|--------|----------|
| Random   | 0.915    | 0.837     | 0.915  | 0.874    |
| Forest   |          |           |        |          |
| KNN      | 0.915    | 0.887     | 0.915  | 0.890    |
| Gradient | 0.910    | 0.837     | 0.910  | 0.872    |
| Boosting |          |           |        |          |

#### MODEL PERFORMANCE VISUALIZATION

A bar plot compares the models' performance across metrics.



# **CONCLUSION** • The project demonstrates the use of machine learning to predict heart disease, with GBC, KNN and RF showing strong performance. • Age was identified as a key risk factor. Additional features could further improve predictions. This work provides a foundation for early heart disease detection tools.