





submitted as partial

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# BACHELOR OF TECHNOLOGY DEGREE

**SESSION 2024-25** 

in

### **CSE(Artificial intelligence and machine learning)-'C'**

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## Methodology:-

#### 1. Data Collection

The dataset used is the **Heart Disease Dataset** from the **UCI Machine Learning Repository**. It contains records of patients with and without heart disease, described by 13 clinical attributes.

#### 2. Data Preprocessing

- Handling Missing Values: Impute or drop missing entries.
- **Feature Encoding:** Convert categorical data (e.g., chest pain type, thalassemia) using one-hot encoding or label encoding.
- Scaling: Normalize numeric features to bring them to the same scale.
- **Splitting Dataset:** Divide the data into training (80%) and testing (20%) sets.

#### 3. Model Selection

The **Random Forest Classifier** is chosen for its robustness and accuracy. It reduces overfitting and works well with both categorical and numerical features

#### 4. Model Evaluation

#### **Evaluation metrics include:**

- Accuracy
- Precision
- Recall
- F1-score
- Confusion Matrix
- ROC-AUC Curve

## Code:-

import pandas as pd
import numpy as np
from sklearn.model\_selection import train\_test\_split
from sklearn.preprocessing import StandardScaler
from sklearn.linear\_model import LogisticRegression
from sklearn.metrics import accuracy\_score,
classification\_report, confusion\_matrix

```
# Load the dataset
df = pd.read_csv("4. Predict Heart Disease.csv")
# Split features and target
X = df.drop("target", axis=1)
y = df["target"]
# Train-test split
X train, X test, y train, y test = train test split(X, y,
test_size=0.2, random_state=42)
# Scale the features
scaler = StandardScaler()
X_train_scaled = scaler.fit_transform(X_train)
X_test_scaled = scaler.transform(X_test)
# Train Logistic Regression model
model = LogisticRegression()
```

```
model.fit(X train scaled, y train)
# Predict
y_pred = model.predict(X_test_scaled)
# Evaluation
print("Accuracy:", accuracy_score(y_test, y_pred))
print("Classification Report:\n", classification_report(y_test,
y pred))
print("Confusion Matrix:\n", confusion matrix(y test,
y pred))
# Predict on new data (example input)
sample input =
np.array([[63,1,0,145,233,1,2,150,0,2.3,2,0,2]]) # Replace with
actual inputs
sample_scaled = scaler.transform(sample_input)
prediction = model.predict(sample scaled)
print("Heart Disease Prediction:", "Yes" if prediction[0] == 1
else "No")
```

## **Result:-**

```
Accuracy: 0.8852459016393442
Classification Report:
               precision recall f1-score
                                               support
                   0.89
                             0.86
                                       0.88
                                                    29
                   0.88
                             0.91
                                       0.89
           1
                                                    32
    accuracy
                                       0.89
                                                    61
   macro avg
                                       0.88
                                                    61
                   0.89
                             0.88
weighted avg
                   0.89
                             0.89
                                       0.89
                                                    61
Confusion Matrix:
 [[25 4]
 [ 3 29]]
Heart Disease Prediction: No
/usr/local/lib/python3.11/dist-packages/sklearn/utils/validation.py:273
  warnings.warn(
```

## **Refrences:-**

**Title: predict heart disease Dataset** 

**Source: UCI Machine Learning Repository** 

Original Donor: Hungarian Institute of Cardiology, Budapest; University Hospital, Zurich; V.A. Medical Center, Long Beach, and Cleveland Clinic Foundation

Link: https://archive.ics.uci.edu/ml/datasets/heart+Disease

License: Public domain / Open Data

**Citation:** 

Detrano, R., Janosi, A., Steinbrunn, W., Pfisterer, M., Schmid, J.J., Sandhu, S., Guppy, K., Lee, S., & Froelicher, V. (1989). *International application of a new probability algorithm for the diagnosis of coronary artery disease*. The American Journal of Cardiology, 64(5), 304–310.