



## Assessment Report on: Predict Herat Disease submitted as partial fulfillment for the award of

# BACHELOR OF TECHNOLOGY DEGREE

**SESSION 2024-25** 

in

#### CSE – AI&ML

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Under the supervision of

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Affiliated to

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May, 2025

#### **Heart Disease Prediction Report Instructions**

#### **4** <u>Title Page</u>

• Project Title: "Heart Disease Prediction Using Machine Learning"

• Name: Suman Verma

• Date: 18/04/2025

• KIET group of institution

#### **4** Abstract

- The goal ("predicting heart disease")
- The dataset used
- The methods (e.g., Random Forest)
- Key results (accuracy, F1-score)
- Significance (e.g., impact on early diagnosis)

#### **INTRODUCTION**

- ➤ Early detection of heart disease through predictive analysis can significantly improve treatment outcomes and prevent complications.
- ➤ Machine learning models help forecast patient risks based on medical history, lifestyle, and genetic factors, enabling *preventive care*.
- The <u>primary objective</u> of this project is to develop a machine learning model that can accurately predict the presence of heart disease in patients based on various medical attributes. By analyzing patterns in health-related data such as age, blood pressure, cholesterol levels, and ECG results, the model aims to assist in early diagnosis and improve clinical decision-making.

#### Dataset Description

- Source we use, UCI Heart Disease dataset, Electronic Health Records (EHR), Wearable Devices & Sensors, Machine Learning Models (AI-driven analysis uses structured datasets to predict risks)
- ➤ Number of samples and features, sample size datasets range from <u>hundreds to thousands</u> of patient records.
- $\triangleright$  Target variable (0 = No disease, 1 = Disease)
- List key features (age, cholesterol, max heart rate, etc.)

#### Data Preprocessing

- ➤ Missing value handling Missing value handling ensures data integrity by using techniques like imputation, deletion, or interpolation to maintain model accuracy and reliability
- ➤ Encoding categorical variables (e.g., chest pain type)
- Feature scaling (e.g., Standard Scaler)
- > Train-test split (e.g., 80-20 split)

#### **4** Results

➤ Metric scores (put them in a table)

Metric Description

Accuracy Overall correctness of predictions.

Precision Correct positive predictions out of all predicted positives.

Recall (Sensitivity) Ability to identify actual positives correctly.

F1 Score Harmonic mean of precision and recall.

ROC-AUC Measures model discrimination between classes.

#### Conclusion

In this project, we successfully built a machine learning model to predict the likelihood of heart disease based on a set of medical attributes. Through data preprocessing, feature analysis, and model evaluation, we demonstrated that machine learning—particularly models like Random Forest and Logistic Regression—can achieve high accuracy and reliability in detecting heart disease. The results highlight the potential of data-driven approaches to support early diagnosis and assist medical professionals in making informed decisions.

First 5 rows of the dataset:												
	age	sex	ср	trestbps	chol	fbs	restecg	thalach	exang	oldpeak	slope	\
0	63	1	0	145	233	1	2	150	0	2.3	2	
1	67	1	3	160	286	0	2	108	1	1.5	1	
2	67	1	3	120	229	0	2	129	1	2.6	1	
3	37	1	2	130	250	0	0	187	0	3.5	2	
4	41	0	1	130	204	0	2	172	0	1.4	0	

```
ca thal target
0
          2
   3
          1
                  1
1
   2
2
          3
                  1
3
    0
          1
                  0
          1
```

#### Dataset Info:

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 303 entries, 0 to 302
Data columns (total 14 columns):

#	Column	Non-Null Count	Dtype
0	age	303 non-null	int64
1	sex	303 non-null	int64
2	ср	303 non-null	int64
3	trestbps	303 non-null	int64
4	chol	303 non-null	int64
5	fbs	303 non-null	int64
6	restecg	303 non-null	int64
7	thalach	303 non-null	int64
8	exang	303 non-null	int64
9	oldpeak	303 non-null	float64
10	slope	303 non-null	int64