

Action plan for making Heart attack predictions Model

1. Problem Definition

- Build a machine learning-based **Heart Attack Prediction System** to assess the likelihood of a heart attack based on user input.
 - Provide a user-friendly interface for users to input their details and get predictions in real-time.
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2. Steps to Build the System

Step 1: Data Preprocessing

1. Import Libraries:

- Use pandas for data manipulation.
- Use scikit-learn for preprocessing and model training.
- Use matplotlib and seaborn for visualization.

2. Data Cleaning:

- Handle missing values (if any) using mean/mode imputation.
- Normalize numeric features like age, cholesterol levels, etc.
- Encode categorical variables using one-hot encoding.

3. All the dependent and independent variable related information including graphical representation in uploaded in Github you can access the report via this link

Link: https://github.com/Guptaji-0/Health_Prediction-/blob/main/Proceesed_Report.html

Download the file in your machine to view the report

4. Data Splitting:

- Divide the preprocessed data into a **70-30 train-test split**.
- Use stratified sampling if the target variable is imbalanced.

Step 2: Feature Selection

1. Analyze feature importance using:

- Correlation matrix.
 - Feature importance from tree-based models (e.g., Random Forest).
 - Recursive feature elimination.
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3. Machine Learning Model Development

Step 1: Model Selection

- Train and evaluate the following models:
 1. **Logistic Regression**: Suitable for binary classification problems.
 2. **Naive Bayes**: Works well with categorical and continuous data.
 3. **Random Forest**: Robust model to handle non-linear relationships.
 4. **Decision Tree**: Easy to interpret but prone to overfitting.
 5. **Support Vector Machine (SVM)**: Effective for high-dimensional spaces.

Step 2: Model Training

- Train all selected models using the **training data**.
- Use appropriate hyperparameter tuning methods such as Grid Search or Randomized Search to optimize each model.

Step 3: Model Evaluation

- Evaluate models using:
 - **Accuracy**: Percentage of correct predictions.
 - **Precision, Recall, and F1 Score**: To assess model balance.
 - Choose the **best-performing model** based on the evaluation metrics.
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4. Deployment

Step 1: Create Web Application

- Use **Streamlit** to develop an interactive UI for the prediction system.

Step 2: Frontend Features

1. Input fields for user details:
 - Age, Gender, Cholesterol, Resting BP, etc.
2. Predict Button:
 - On click, fetch data, pass to the trained model, and display results.
3. Visualization:
 - Show insights such as risk factor distribution.

Step 3: Backend Implementation

1. Load the trained model as pickle file
 2. Process user input, make predictions, and return results in real-time.
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5. Testing and Validation

- Test the web application with various input scenarios.
 - Perform load testing to ensure scalability.
 - Validate predictions with domain experts to ensure real-world applicability.
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6. Deployment

- Deploy the application using cloud platforms such as:
 - **Heroku**
 - **Stramlit cloud**
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7. Deliverables

1. **ML Model:** Pretrained best-performing model.
2. **Web Application:** Streamlit-based user interface.