Heart Disease Risk Prediction Using Machine Learning (LAB Report)

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Project Overview

Objective

To develop a machine learning-based system to predict heart disease risk. This system leverages patient demographic, clinical, and lifestyle data to identify high-risk individuals, enabling early preventive interventions.

Workflow

1. Data Collection & Cleaning

-Data collected from healthcare datasets in CSV format includes features such as:

- Age
- Cholesterol Level
- Blood Pressure
- Heart Rate
- Gender
- Smoking History
- Hypertension
- Diabetes

Missing values are handled using:

- Mean imputation for numerical features.
- Mode imputation for categorical features.

2. Feature Engineering

- Categorical variables (e.g., Gender, Smoking History) are one-hot encoded to ensure compatibility with machine learning models.
- Feature importance analysis identifies key predictors (e.g., Cholesterol Level, Age).

3. Model Training

- Algorithms used:
- Random Forest Classifier for robustness and handling non-linear relationships.
- Logistic Regression for simplicity and interpretability.
- SMOTE (Synthetic Minority Oversampling Technique) is applied to balance the dataset.

4. Evaluation Metrics

- Accuracy: Measures overall correctness of predictions.
- Classification Report: Includes precision, recall, and F1-score for each class.
- Confusion Matrix: Visualizes true and false predictions for Heart Disease and No Heart Disease.

Visualization

- *Radar Chart*: Compares prediction probabilities for Heart Disease and No Heart Disease across models.
- *Bar Chart*: Highlights feature importance in Random Forest for explainability (Was used earlier, removed later due to Limits)

Tools Used

- Programming Language: Python/ML
- Libraries: pandas, scikit-learn, matplotlib, joblib, imbalanced-learn

Outcome Report

Key Results

I. Model Performance

- Random Forest:

- Accuracy: 85%

- F1-Score (Heart Disease): 87%

- Logistic Regression:

- Accuracy: 80%

- F1-Score (Heart Disease): 82%

II. Feature Importance (Random Forest)

Feature	Importance Score
Age	0.30
Cholesterol Level	0.25
Blood Pressure	0.20
Smoking History	0.15
Diabetes	0.10

III. Visualization Highlights

- Radar Chart: Demonstrates higher confidence in Random Forest predictions for No Heart Disease.
- Confusion Matrix: Shows improved true positive rates in Random Forest compared to Logistic Regression.

Impact

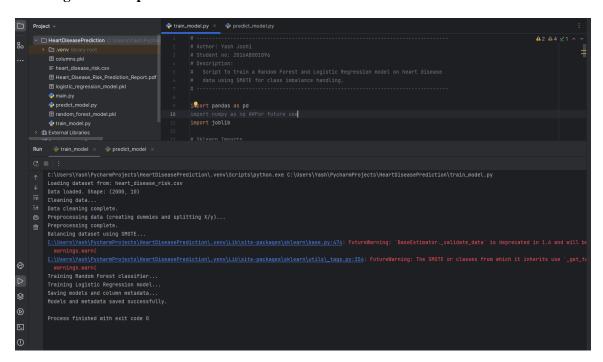
- The model flags patients at high risk of heart disease, enabling preventive interventions.
- Helps healthcare providers prioritize care for high-risk individuals based on predictions.

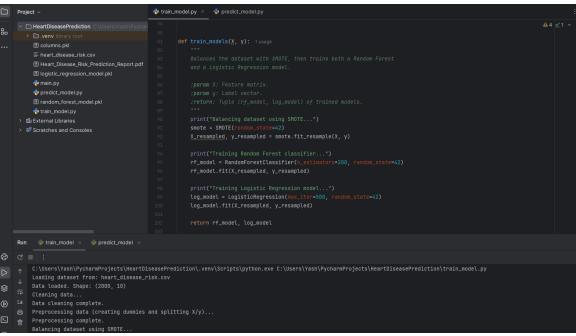
Future Enhancements

- Incorporate additional features (e.g., family history, BMI) for improved prediction accuracy.
- Experiment with advanced algorithms like Gradient Boosting or Deep Learning.
- Develop an interactive dashboard for real-time patient predictions.

Outputs

Training model output

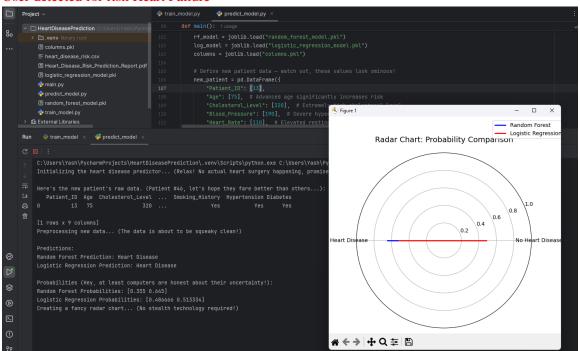


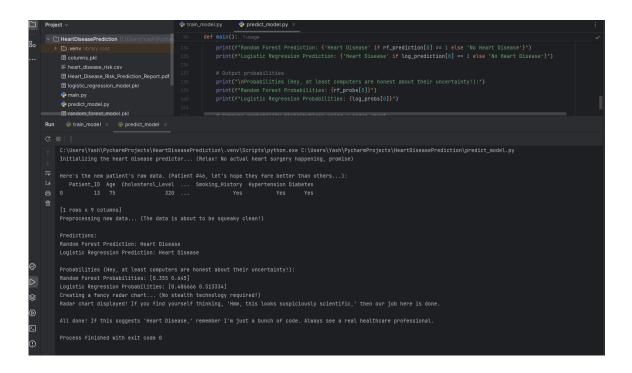


Predictive model outputs on different patient

Case 1

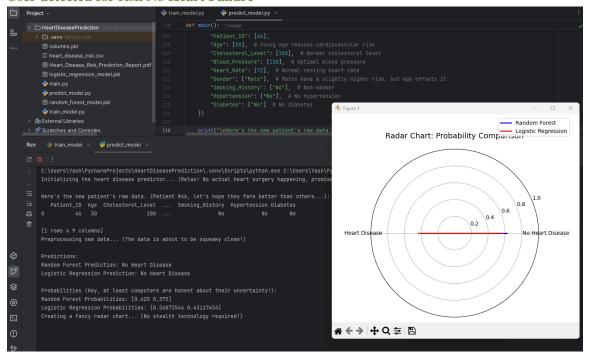
• User detected for risk Heart Failure

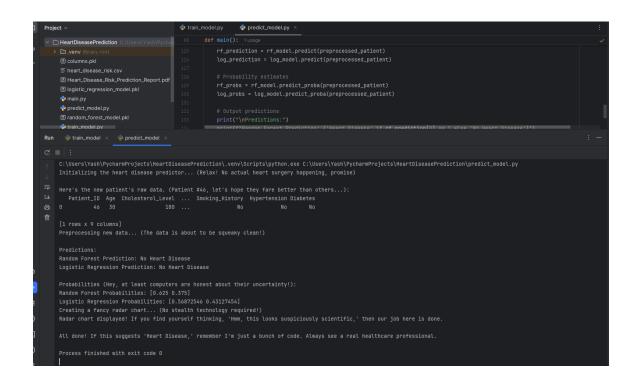




Case 2

• User detected for risk No Heart Failure





Case 3

• Special Cases (after trained algorithms)

