Healthcare Predictive Analytics Project



Al-Powered Heart Disease Risk Prediction

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The Problem and Our Solution

- Heart disease is a leading cause of preventable deaths.
- Early risk detection is critical but often delayed.
- Our Solution: A machine learning-based platform that predicts patient heart disease risk in real-time.

How the Model Works

- Trained on 70,000+ patient records (age, vitals, lifestyle, labs).
- Inputs: Age, BP, cholesterol, glucose, etc.
- Output: Risk score (0 = low, 1 = high).
- Accessible via a user-friendly web interface.



Web Application Architecture

- app.py: Flask app initializer and route controller.
- model.py: handles model loading and predictions.
- retrain.py: enables model retraining from new CSV datasets.
- preprocessing.py: reusable data cleaning functions.
- templates/: HTML files for user interface.
- static/: CSS, JS, and design assets.



Application Features

- Heart Disease Prediction:
 - Manual input form for patient data.
 - Real-time prediction output from trained model.
- Visualization Tool:
 - Upload datasets.
 - Generate descriptive statistics and visualizations.
- Model Retraining:
 - Upload new labeled data.
 - Preprocessing + model retraining + evaluation.





Empowering Smarter Healthcare Decisions

Welcome to **Heart AI** — your intelligent assistant for early detection of heart disease risks.

Combining the power of advanced **Machine Learning** models with real-time patient data, we help healthcare professionals make faster, smarter, and more reliable decisions that can save lives. Together, we shape the future of predictive healthcare.

What You Can Do

- Predict: Instantly assess patient risk levels with Al-powered predictions.
- Visualize: Dive deep into your data through beautiful, insightful charts and analytics.
- @ Retrain: Continuously improve accuracy by training the model with your latest datasets.



Age (Years)		
Gender		
Select		~
Height (cm)		
Veight (kg)		
Systolic BP		



The patient is NOT at risk.

Entered Patient Data

Feature Featur	Value
Age (Years)	50.3
Gender	Female
Height (cm)	168.0
Weight (kg)	62.0
Systolic BP	110.0
Diastolic BP	0.08
Cholesterol Level	Normal
Glucose Level	Normal
Smoking	No



The patient is at risk!

Entered Patient Data

Feature Feature	Value Value	
Age (Years)	58.0	
Gender	Male	
Height (cm)	185.0	
Weight (kg)	100.0	
Systolic BP	140.0	
Diastolic BP	90.0	
Cholesterol Level	Well Above Normal	
Glucose Level	Well Above Normal	
Smoking	Yes	

Dataset and Features

- Dataset: 70,000+ records (cardiovascular.csv).
- Demographic features: Age, Gender, Height, Weight Males and older patients show higher prevalence.
- Clinical features: Blood Pressure, Cholesterol, Glucose. High cholesterol and systolic BP = strong risk indicators.
- Lifestyle indicators: Smoking, Alcohol, Physical Activity.

 Sedentary lifestyle and smoking correlate with elevated risk
- Target: Binary classification (at-risk_1 / not at-risk_0)



Data Preprocessing and Exploration

- Handled outliers using IQR and domain thresholds.
- Normalized and encoded features for consistency.
- Split into training (80%) and testing (20%) sets.
- EDA techniques used:
 - Correlation heatmaps.
 - Distribution plots and boxplots.
 - Target class balance visualized.

Handling Missing Values

Outlier Detection

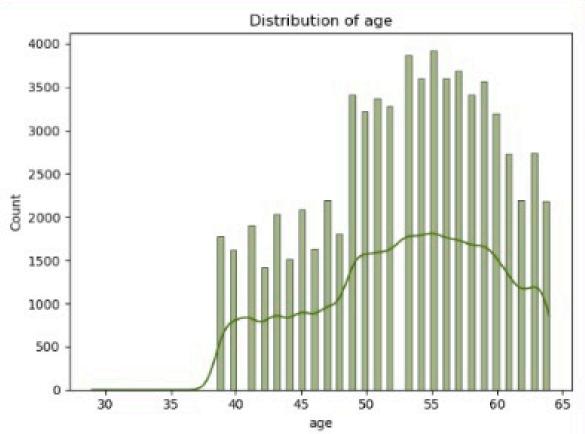
Feature Scaling

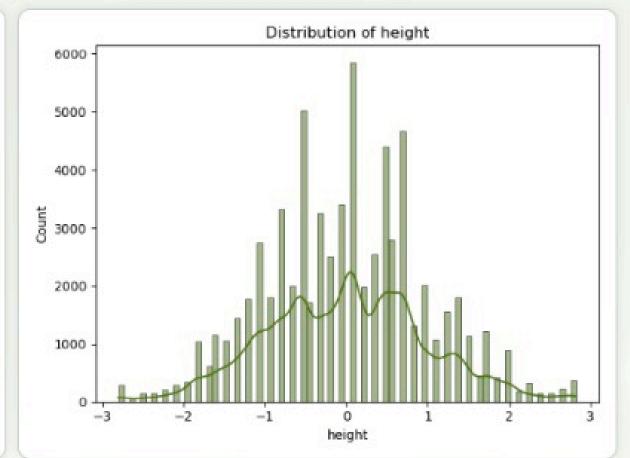
Feature Encoding

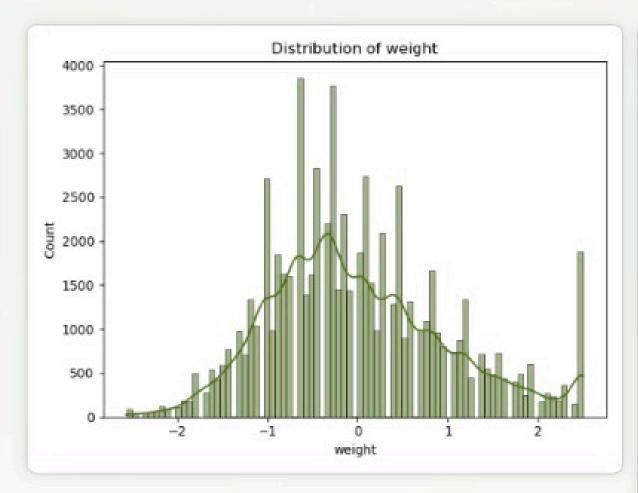


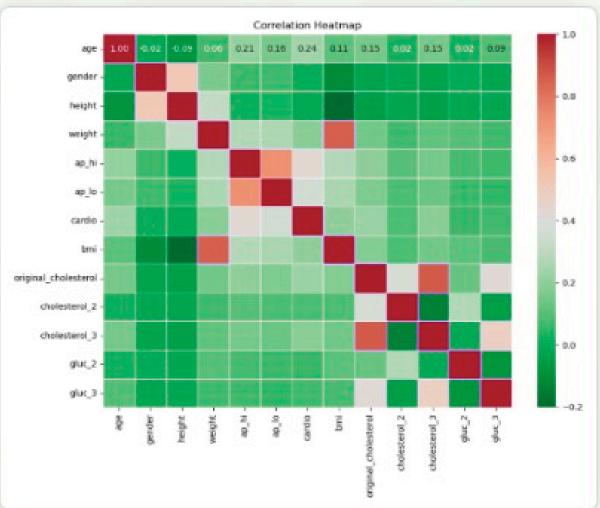


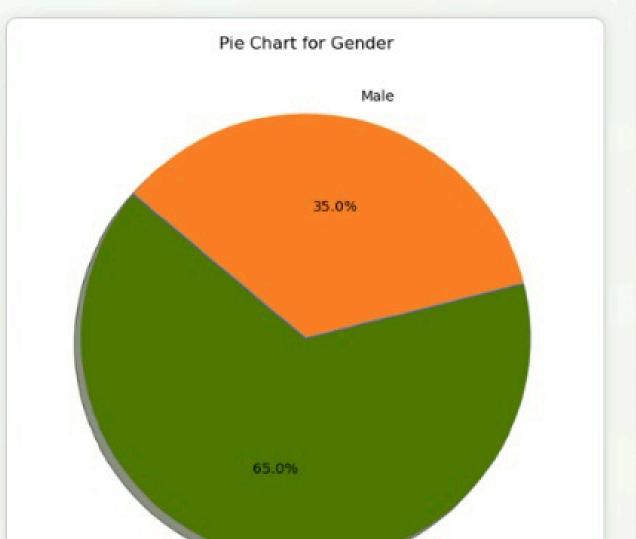












Model Development and Performance

- Models tested: Logistic Regression, Random Forest, Gradient Boosting
- Best model: Random Forest

Logistic Regression 0.8 0.85 0.84 0.93 0.77 Random Forest 0.8 Similar 0.84 0.93 0.77 0.77 Gradient Boosting 0.77 Lower 0.8 0.82 0.79	Model	Accuracy	AUC-ROC	F1	Recall	Precision
	Logistic Regression	0.8	0.85	0.84	0.93	0.77
Gradient Boosting 0.77 Lower 0.8 0.82 0.79	Random Forest	0.8	Similar	0.84	0.93	0.77
	Gradient Boosting	0.77	Lower	0.8	0.82	0.79

Model Retraining Module

• The retraining module supports continuous learning:

Upload new CSV data via UI.

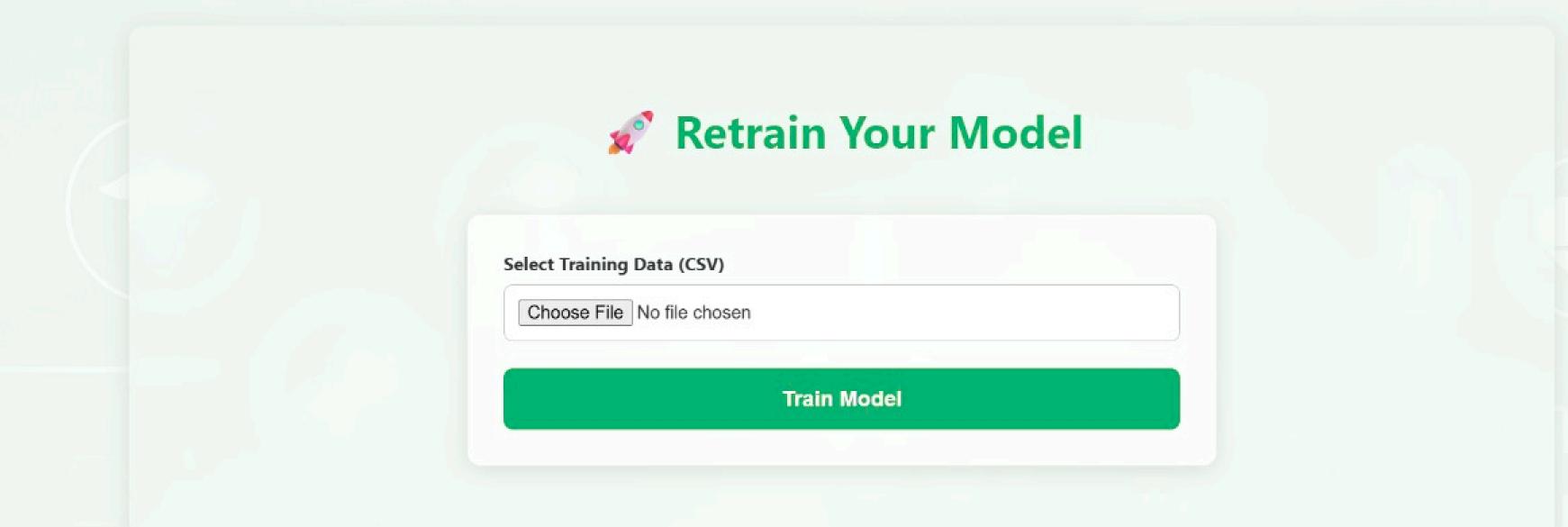
Automatically preprocess and merge with historical data.

Retrain Random Forest model.

Evaluate and overwrite previous model if new one performs better.

This supports MLOps practices and ensures model freshness.





Integration Recommendations

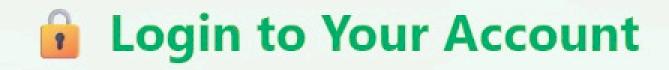
- Embed in EHR systems for automated risk scoring.
- Use in routine screenings to flag high-risk patients.
- Train clinical staff to interpret model results.
- Establish feedback loops for continuous improvement.

Future Improvements

- Add SHAP-based explainability (XAI).
- Support multilingual and mobile-friendly interface.
- Integrate with hospital APIs (HL7/FHIR standards).
- Conduct real-world validation in clinical settings.







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THANK YOU!

