Artificial Intelligence & Machine Learning Project Documentation

Introduction

Project Title:

Revolutionizing Liver Care : Predicting Liver Cirrhosis using Advanced Machine Learning Techniques

Members:

- Kuruva karthik–Data Collection and Cleaning
- Kammari Likhitha–Model Design and Training
- Kothapalli Subramanyam–Evaluation and Optimization
- Kopparapu Neha Sree–Deployment and Integration

Project Overview

Purpose:

The purpose of the this project is to leverage AI/ML to enable early detection and accurate prediction of liver cirrhosis. By analyzing patient health data, the model can identify patterns and risk factors that may not be easily visible through traditional diagnostic methods. This predictive approach helps healthcare professionals make timely decisions, improve treatment outcomes, and reduce the progression of liver damage. The project aims to support proactive liver care by integrating intelligent, data-driven tools into the medical workflow.

Goals:

- To build an accurate machine learning model for early prediction of liver cirrhosis.
- To identify and analyze key health features influencing cirrhosis risk.
- To support doctors with reliable, AI-driven diagnostic assistance.
- To improve patient outcomes through timely detection and intervention.

Key Features:

- **Machine Learning-Based Prediction** Accurately predicts liver cirrhosis using patient health data.
- **User-Friendly Web Interface** Simple Flask application for easy data input and result display.
- **Feature Importance Analysis** Highlights critical health indicators influencing the prediction.
- **Model Optimization** Includes performance evaluation and hyper parameter tuning for better accuracy.

Architecture

Frontend:

- Framework Used: HTML,CSS,JINJA2 templates
- Functionality:
- Collect patient data (e.g., age, bilirubin, albumin, etc.).
- Sends data to the Flask backend for prediction.
- Shows the prediction (Cirrhosis: Positive/Negative).
- Ensure all fields are filled correctly before submission

Backend:

Technology Used: Flask, Joblib , NumPy

- Responsibilities:
- Receives data from the frontend (via POST request).
- Normalizes/reshapes input to match the trained model format.
- Uses the saved .pkl model and normalizer to make predictions.
- Sends prediction result (e.g., "Cirrhosis: Positive") back to the frontend.

Model Integration: A trained machine learning model (Random Forest) and data normalizer are loaded into memory at runtime to enable fast and accurate liver cirrhosis prediction based on patient health parameters.

Database:

- **Type:** UCI Liver Cirrhosis Dataset
- Stored Data:

Patient clinical records including age, bilirubin, albumin, alkaline phosphate, SGOT, SGPT, platelet count, and liver disease status.

- Use Cases:
- 1. Early prediction of liver cirrhosis.
- 2. Assisting doctors in diagnosis with data-driven insights.
- 3. Reducing manual analysis time in clinical settings.

Setup Instructions

Prerequisites:

- Pvthon3.x
- Libraries: TensorFlow/Keras, NumPy, Pandas, OpenCV, Matplotlib, Streamlit
- GPU or Google Colab (recommended for training)

STEP 1: Install Python and Required Packages

Ensure Python 3.x is installed.

Install dependencies using pip:

• pip install flask joblib numpy

STEP 2: Folder Structure

LiverCirrhosisPrediction/
app.py
rf_acc_68.pkl
normalizer.pkl
templates/
index.html
result.html (or inner-page.html)
static/
(optional styles/images)
Data/
liver_dataset.xlsx
Documentation/
report.pdf, screenshots, etc.

STEP 3: Running the App

Navigate to your project directory and run:

- python app.py
- Access the web app at: http://127.0.0.1:5000

How to Use:

- 1. Open the app in your browser.
- 2. Enter required medical values (e.g., bilirubin, albumin).
- 3. Click **Predict**.
- 4. View the result: "Cirrhosis Detected" or "No Cirrhosis".

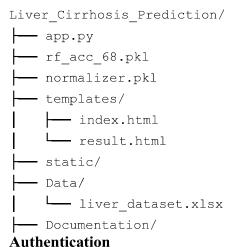
Offline Usage

Ensure the following files are present:

- rf_acc_68.pkl trained model
- normalizer.pkl normalization object
- templates/ and static/ folders
- Flask and required packages install

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Folder Structure



- - Not implemented in the current version.
 - Can be extended with login, session management, or token-based access.

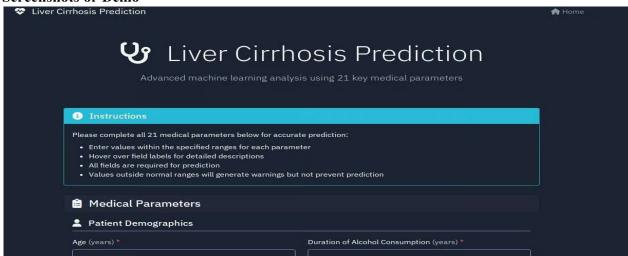
User Interface

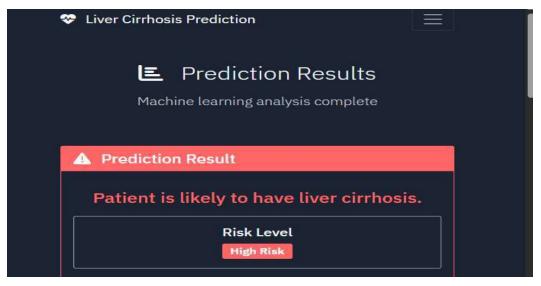
- Simple and clean form using HTML and Jinja2.
- Displays prediction result clearly.
- Can be extended with charts or model explanation.

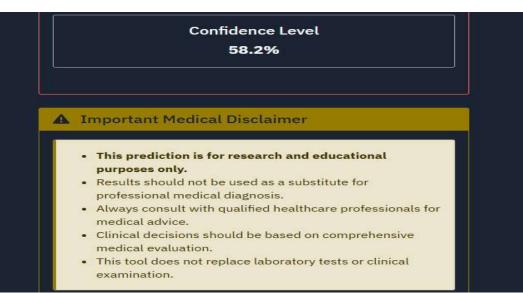
Testing

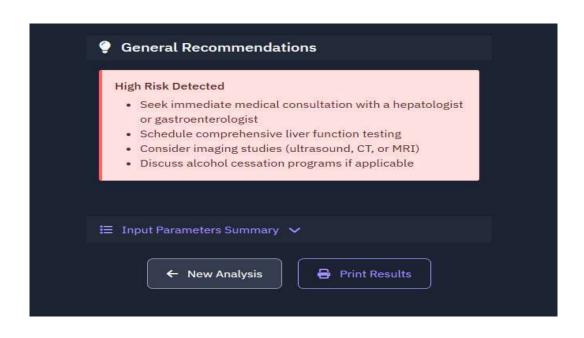
- Manual testing with sample inputs.
- Cross-validation during model training.
- Accuracy achieved: ~68% (Random Forest)

Screenshots or Demo









Known Issues

- Accuracy limited to ~68% due to dataset size.
- May misclassify borderline cases.
- UI is basic; no error handling for invalid input types.

Future Enhancements

- Improve model with more medical features and advanced algorithms.
- Add database for patient record storage.
- Deploy using cloud services (Heroku, AWS, etc.).
- Add feature explanation using SHAP or LIME.