## **Solution Architecture**

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

**Date:** 28 June 2025

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**Project Name:** Liver Cirrhosis Prediction System

Maximum Marks: 4 Marks

#### 1. Solution Architecture Overview

The solution architecture bridges the gap between the clinical need for early liver cirrhosis detection and advanced machine learning technology. This system provides a comprehensive, non-invasive approach to predict liver cirrhosis using clinical and biochemical data.

#### Goals:

- **Primary Goal:** Develop an accurate ML-based prediction system for early liver cirrhosis detection
- **Technical Goal:** Create a scalable web application that integrates multiple ML models
- Clinical Goal: Provide healthcare professionals with a cost-effective diagnostic support tool
- **Business Goal:** Reduce healthcare costs by enabling early intervention and preventing advancedstage treatments

# 2. System Architecture Components

# 2.1 High-Level Architecture

```
PRESENTATION LAYER
Web Browser | Mobile Client | API Client |
     APPLICATION LAYER
      Flask Web Server
  Routing | Input | Output | |
  Handler | Validation | Formatter | |
     PROCESSING LAYER
   ML Pipeline Engine
 Data | Feature | Model |
  Preprocessor | Engineering | Inference |
     MODEL LAYER
XGBoost | Random Forest | SVM |
Model | Model | Model |
(Primary-90%) | | (Backup-88%) | | (Backup-85%) | |
```

```
DATA LAYER

| Training Data | Model Files | Configuration | |
| (CSV/Excel) | (.pkl files) | Files | |
```

## 2.2 Detailed Component Architecture

#### A. Data Input Layer

- Input Sources: Web forms, API endpoints, file uploads
- Data Types:
  - Patient demographics (Age, Gender)
  - Clinical parameters (Bilirubin, Albumin, Hemoglobin)
  - Laboratory results (Liver enzymes, Prothrombin time)
- Format Support: JSON, CSV, Excel, Form data

#### B. Data Processing Pipeline

```
Raw Input → Data Validation → Missing Value Handling → Feature Encoding → Normalization → Feature Selection → Model Input
```

#### **Processing Steps:**

- 1. Data Validation: Input format and range validation
- 2. **Missing Value Handling:** Imputation using median/mode
- 3. Categorical Encoding: Label encoding for gender and other categorical features
- 4. **Normalization:** Min-Max scaling or standardization
- 5. **Feature Selection:** Based on correlation analysis and domain knowledge

#### C. Machine Learning Engine

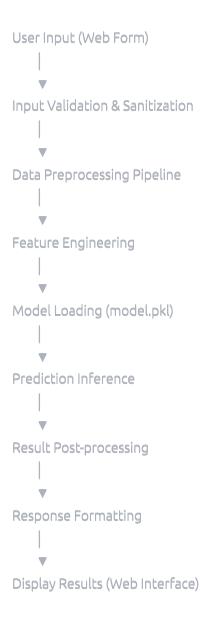
- **Primary Model:** XGBoost (90.1% accuracy)
- Backup Models: Random Forest (88.6%), SVM (85.2%)
- Model Selection Logic: Automatic fallback mechanism
- **Ensemble Option:** Voting classifier for critical cases

#### D. Web Application Framework

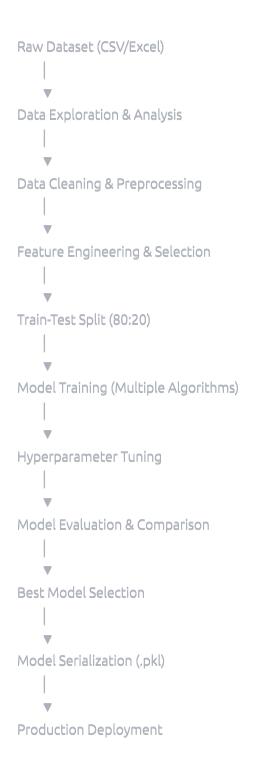
- **Backend:** Python Flask
- **Frontend:** HTML/CSS/JavaScript
- Templates: Jinja2 templating engine
- Static Assets: CSS, JavaScript, images

## 3. Data Flow Architecture

#### 3.1 Prediction Workflow



# 3.2 Training Pipeline (Development Phase)



# 4. Technology Stack Specifications

## 4.1 Development Environment

• **Programming Language:** Python 3.8+

• Web Framework: Flask 2.0+

• Machine Learning: Scikit-learn, XGBoost

• **Data Processing:** Pandas, NumPy

• Visualization: Matplotlib, Seaborn

• Model Serialization: Pickle

# 4.2 Dependencies

```
python

Flask==2.0.1
scikit-learn==1.0.2
xgboost==1.5.0
pandas==1.3.3
numpy==1.21.2
matplotlib==3.4.3
seaborn==0.11.2
```

#### 4.3 File Structure

```
liver_cirrhosis_prediction/
  — app.py # Main Flask application
  --- model.pkl
                # Trained XGBoost model
  — requirements.txt # Python dependencies
  — templates/
  —— index.html # Input form
  result.html # Prediction results
  — static/
   ---- css/
    — js/
    — images/
  — data/
  ____ training_data.csv
 --- models/
  ---- random_forest_model.pkl
   --- svm_model.pkl
```

# 5. Deployment Architecture

## 5.1 Local Development

• **Server:** Flask development server

• **Port:** 5000 (default)

• Environment: Local machine

## **5.2 Production Deployment Options**

#### Option A: Cloud Deployment (Recommended)

• Platform: AWS/Google Cloud/Azure

• Service: Elastic Beanstalk/App Engine/App Service

• Load Balancer: Application Load Balancer

• Auto Scaling: Based on CPU/memory usage

#### Option B: Container Deployment

• Containerization: Docker

• Orchestration: Kubernetes

• Registry: Docker Hub/ECR

#### **Option C: Traditional Server Deployment**

• Web Server: Nginx/Apache

WSGI Server: Gunicorn/uWSGI

Operating System: Ubuntu/CentOS

## 6. Security Architecture

## 6.1 Data Security

• Input Validation: Prevent SQL injection and XSS

• Data Sanitization: Clean and validate all inputs

• HTTPS: SSL/TLS encryption for data transmission

• Session Management: Secure session handling

## 6.2 Model Security

• Model Protection: Pickle file integrity checks

• Access Control: Restricted model file access

Audit Logging: Track prediction requests

# 7. Performance Specifications

#### 7.1 Model Performance Metrics

Model	Ассигасу	Precision	Recall	F1-Score
XGBoost (Primary)	90.1%	89.0%	90.7%	89.8%
Random Forest	88.6%	87.2%	89.0%	88.1%
SVM	85.2%	84.0%	85.0%	84.5%
4	•			<b>&gt;</b>

## 7.2 System Performance Requirements

• **Response Time:** < 2 seconds for prediction

• **Throughput:** 100+ concurrent users

• Availability: 99.5% uptime

• Scalability: Horizontal scaling capability

# 8. Integration Points

#### 8.1 External Integrations

Healthcare Systems: HL7 FHIR compatibility

• Electronic Health Records (EHR): Integration APIs

Laboratory Information Systems: Data import capabilities

## 8.2 API Specifications

• REST API: JSON request/response format

• Authentication: API key-based authentication

• Rate Limiting: Prevent API abuse

# 9. Monitoring and Maintenance

## 9.1 Application Monitoring

• Health Checks: Endpoint monitoring

• **Performance Metrics:** Response time, error rates

• **Logging:** Comprehensive application logs

# 9.2 Model Monitoring

• Model Drift Detection: Monitor prediction accuracy over time

Retraining Schedule: Quarterly model updates

• A/B Testing: Compare model versions

#### 10. Future Enhancements

### **10.1 Technical Improvements**

- Deep Learning Models: Neural network implementation
- **Real-time Processing:** Streaming data support
- **Mobile Application:** Native iOS/Android apps

#### 10.2 Feature Enhancements

- Explainable AI: SHAP/LIME integration
- Multi-language Support: Internationalization
- Advanced Visualizations: Interactive dashboards

#### 11. Conclusion

This solution architecture provides a robust, scalable, and maintainable framework for the liver cirrhosis prediction system. The architecture supports the project's primary objectives of early disease detection, clinical decision support, and cost-effective screening while ensuring high performance and reliability.

The modular design allows for easy maintenance, future enhancements, and integration with existing healthcare systems, making it a comprehensive solution for revolutionizing liver care through advanced machine learning techniques.