



# CardioShield AI: A Smart Blockchain-Enabled System for Early Heart Disease Detection and Secure Telemedicine

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## Abstract:

The artificial intelligence and blockchain-powered intelligent healthcare system is designed to improve the quality of early detection of heart diseases, improve the safety of medical records, and allow remote access to healthcare. The model is based on a decision tree classifier to examine the health parameters of the patients and predict cardiovascular risk. A blockchain-inspired ledger secures the prediction output through SHA-256 hashing and chaining, providing tamper-evident medical record storage. The presence of a combined telemedicine unit will allow patients and doctors to communicate with each other remotely via report access and consultations. The experimental results demonstrate the model's satisfaction and the system's suitability for use in a multi-user environment. The system proposed is more secure, transparent, and accessible than the traditional healthcare platforms.

## Index Terms-

Smart Healthcare, Artificial Intelligence, Blockchain, Heart Disease Prediction, Decision Tree, Telemedicine, Secure Medical Records.

## INTRODUCTION:

Most of the developing areas have healthcare systems that are characterized by delays in diagnosing conditions, poor records and access to specialist care. Most hospital management systems are based on centralized databases, which can be hacked, coerced or hit by a single point of failure. Besides, Heart disease is one of the leading causes of mortality in the world, and it is therefore important to have a quick diagnosis. There is a good potential in predictive analytics of healthcare using Artificial Intelligence (AI). Machine learning models are made to examine the characteristics of the patients, including age, cholesterol level, blood pressure and the type of chest pains, to predict the risk of disease. Nevertheless, prediction will not be sufficient when medical records lack security and verifiability. The cryptographic hashing and distributed records offer blockchain technology immutability and transparency. This project will result in an integrated smart health care ecosystem by combining AI-based prediction, blockchain-style secure storage, and telemedicine functionality.

## LITERATURE SURVEY

Several studies use machine learning models, including logistic regression, Support Vector Machines, and Decision Trees, on data related to heart disease, like the UCI Cleveland data. The techniques are moderately predictive but, in most cases, may not be able to take into consideration medical record security. Healthcare blockchain applications are aimed at protected storage and decentralized access

control. Nonetheless, the majority of blockchain systems are achieved without AI-powered analytics. Telemedicine systems increase accessibility, although they mostly depend on central storage systems. The gap in the research is the integration of AI-based predictive analytics, a blockchain-ensured integrity of medical records, and an Internet-based telemedicine platform into a single architecture.

### SYSTEM ARCHITECTURE

The system follows a three-layer architecture.

#### A. Presentation Layer

It was built in React.js and TypeScript and offers patient, doctor, and administrator role-based dashboards.

#### B. Application Layer

Functioning on Flask back-end API controls of AI prediction processing, authentication, ledger hash creation, and telemedicine processes.

#### C. Data Layer

Patient information and prediction history are stored in PostgreSQL. A table of ledger is of the blockchain fashion, where the hash of the data is made using SHA-256, which builds a secure hash chain to maintain the data integrity.

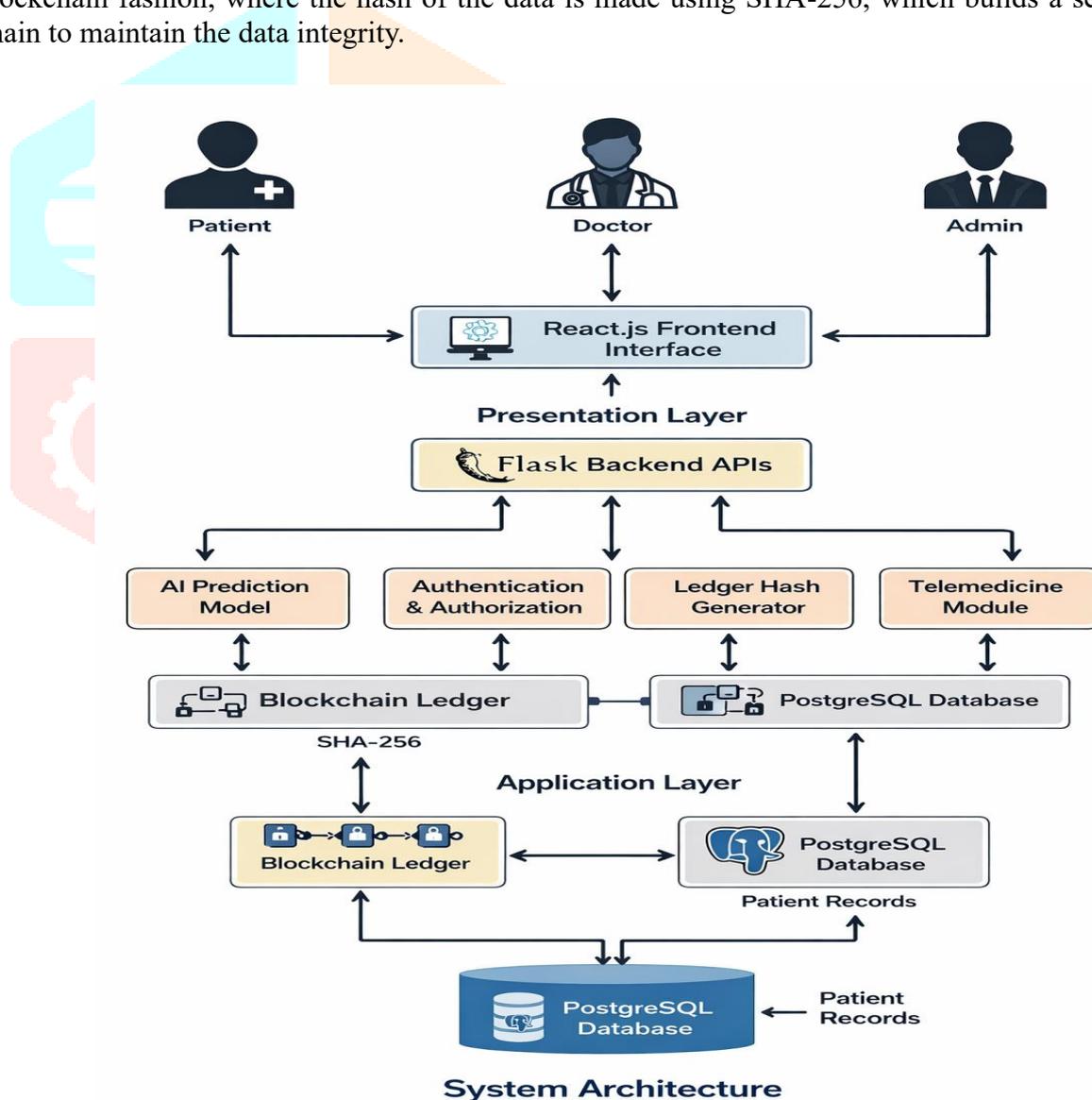


Fig. 1. Three-level system execution of clinical input to secure blockchain-based storage data flow

## METHODOLOGY

### A. Dataset and Preprocessing

The UCI Cleveland Heart Disease dataset is the dataset that will be used. Data preprocessing consists of managing the missing values, feature encoding, normalisation, and an 80:20 train-test split.

### B. Decision Tree Model

Scikit-learn is used to train a Decision Tree classifier. As the splitting criterion, the Gini index is added. The model is trained to learn decision rules that will be used to classify the patients as either disease or no-disease cases.

### C. Blockchain Ledger Mechanism

After prediction:

- 1) Report details are concatenated.
- 2) SHA-256 hash is generated.
- 3) The previous hash is appended.
- 4) Ledger entry is stored in the database.
- 5) This guarantees tamper-evident record storage.

### D. Telemedicine Workflow

The patients can review reports, download QR codes, and consult. Doctors are able to retrieve reports on the patient, check their integrity by comparing hashes, and add clinical notes.

## IMPLEMENTATION

### A. Technologies Used

The implementation uses Python with Flask for REST APIs, React.js for the dynamic frontend, PostgreSQL for persistent storage, Scikit-learn for machine learning, and SHA-256 cryptographic hashing for security.

### B. Algorithm Steps

The workflow begins with loading and preprocessing the dataset. The Decision Tree model is trained and deployed. Upon receiving patient input, the system predicts disease risk, generates a SHA-256 hash linked to the previous record, stores the ledger entry, and displays the result with QR verification.

## RESULTS AND PERFORMANCE ANALYSIS

### A. Performance Metrics

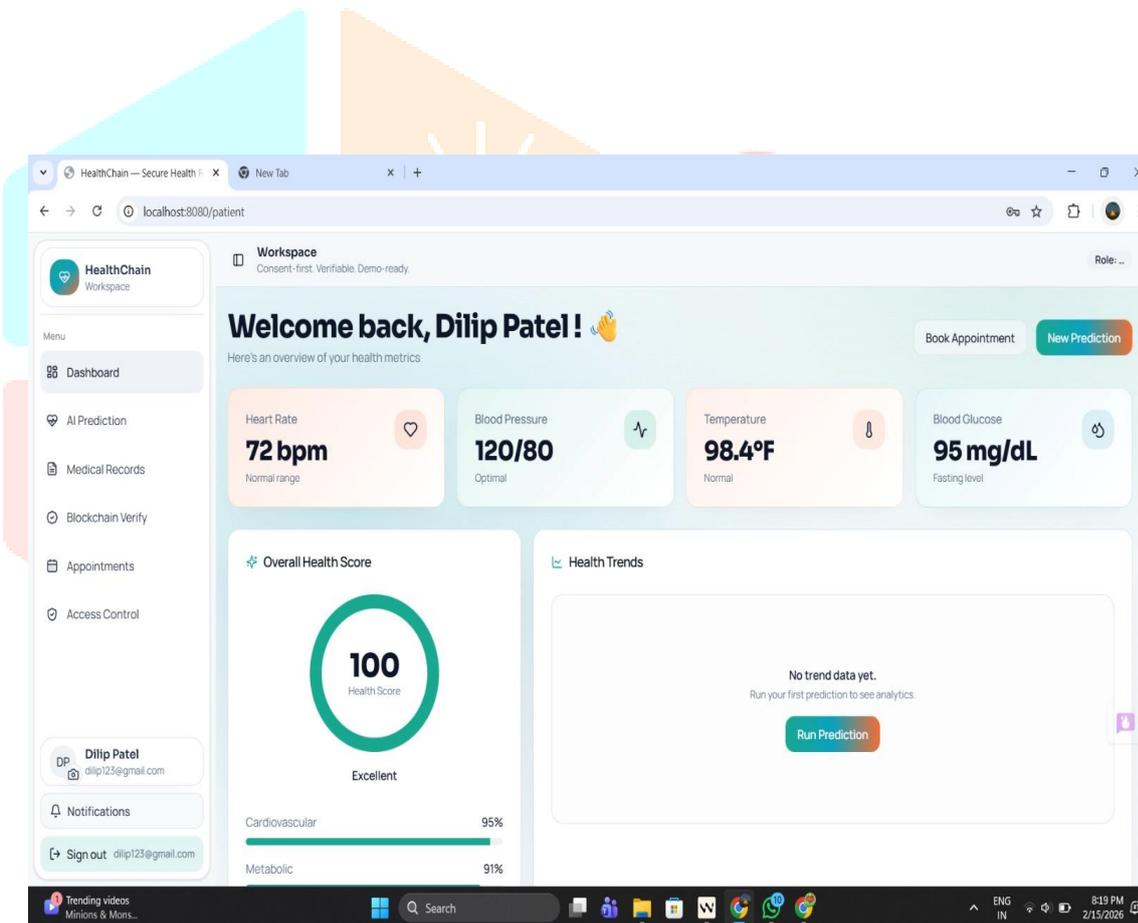
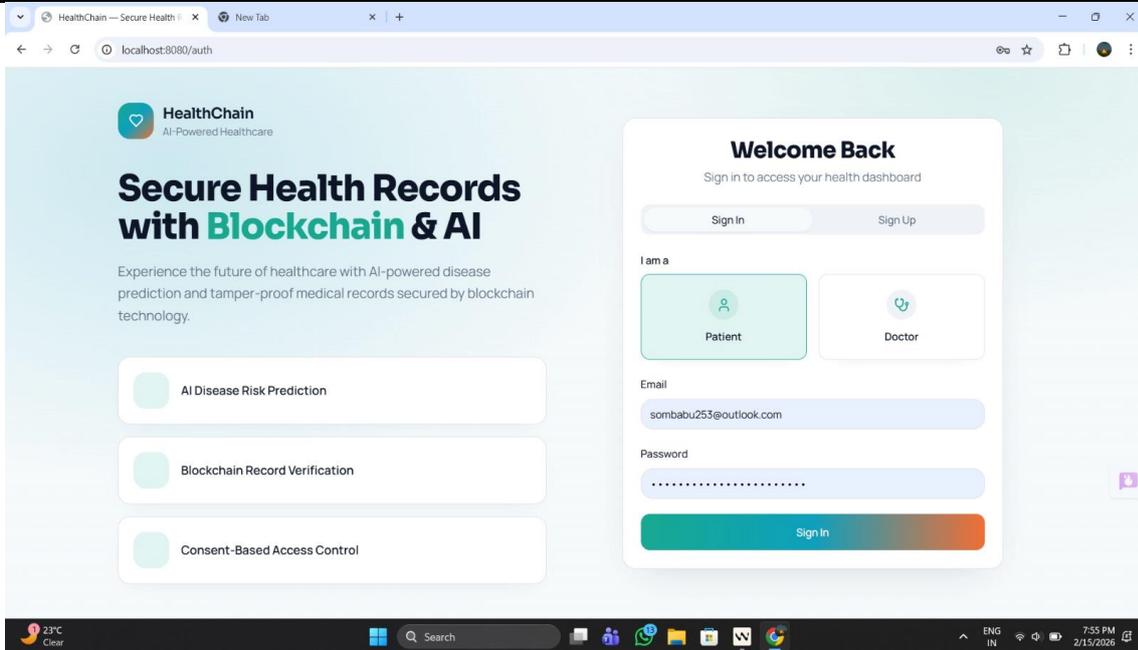
The Decision Tree model achieved acceptable accuracy, precision, recall, and F1-score on the test dataset.

### B. Confusion Matrix Analysis

The confusion matrix indicates that the results are correctly classified in positive and negative cases. A prototype healthcare system can tolerate false positives and false negatives.

### C. System Performance

They had system benchmarks of an average prediction and ledger generation time of less than 2 seconds and less than 1 seconds respectively. The system was stable when tested under the conditions of multiple users.

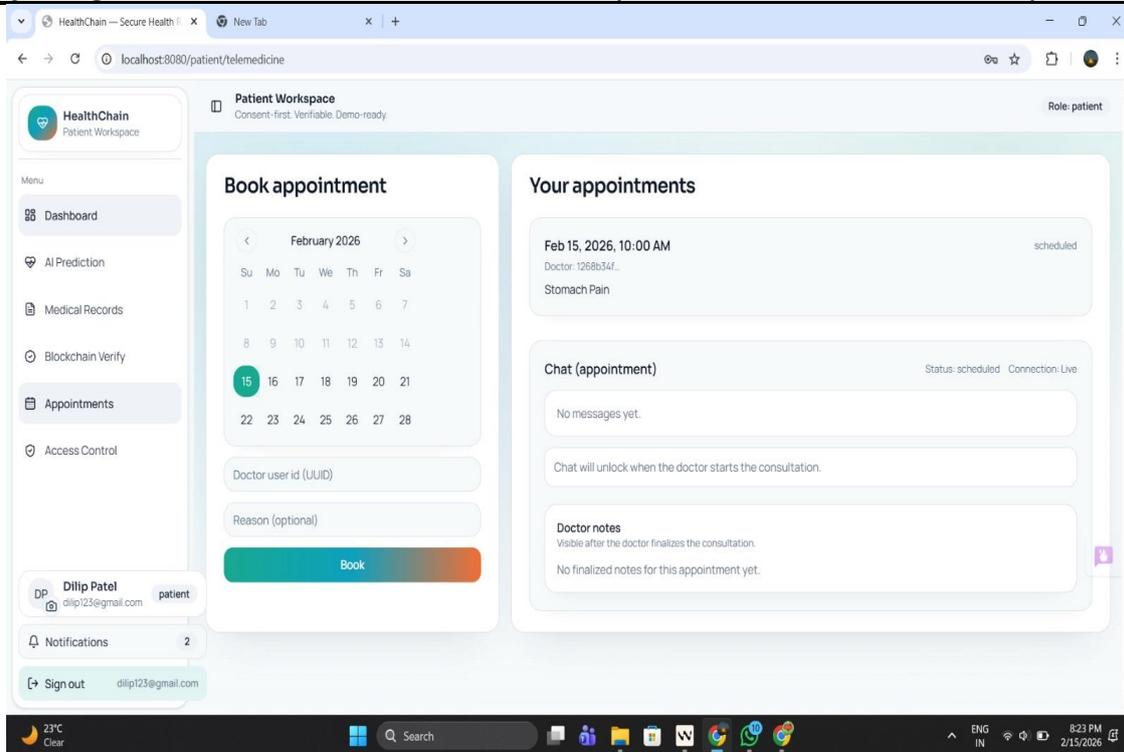


The screenshot shows a web browser window with the URL localhost:8080/patient/predict. The page is titled "Patient Workspace" and "AI Disease Risk Prediction". It features a sidebar menu with options like Dashboard, AI Prediction, Medical Records, Blockchain Verify, Appointments, and Access Control. The main content area displays various health metrics in input fields: Age (35), Blood Pressure (Systolic: 120, Diastolic: 80), Heart Rate (72), Blood Tests (Glucose: 95, Cholesterol: 180, BMI: 24.5), Temperature (36.8), and Lifestyle Factors (Physical Activity Level: Moderate, Smoking: On, Regular Alcohol: Off). On the right, a "Risk" section shows a 10% risk level (Category: Low), a Health score of 90/100, and a "Saved?" status of "No". A "Blockchain-style verification" section at the bottom right indicates that the report hash is anchored into a simulated ledger.

The screenshot shows a web browser window with the URL localhost:8080/verify/tx\_kyIEzW-dgY. The page is titled "Blockchain Record Verification" and displays a "Verified" status with the message "The record hash matches the ledger transaction." Below this, a table provides details about the verification:

Patient		Verification
diip123@gmail.com		Verified
Report type	Timestamp	Transaction ID
AI prediction	2/15/2026, 8:21:26 PM	tx_kyIEzW-dgY
Payload hash	Previous hash	
99b2c9f85730ca0feaa4928070f2436faaff8e0e597de1d3a0889a4f3f2f1a37c	(genesis)	

A "Report snapshot" section at the bottom shows: Risk: 10% (Low), Score: 90/100, and Created: 2/15/2026, 8:21:26 PM.



### COMPARATIVE ANALYSIS

This section compares the proposed Smart Healthcare System with traditional healthcare systems in terms of prediction capability, data security, verification mechanism, accessibility, and transparency.

### COMPARISON OF EXISTING AND PROPOSED SYSTEMS

Feature	Existing	Proposed
Prediction	Basic ML	Decision Tree
Data Security	Central DB	Hash-chained Ledger
Verification	Limited	QR-based
Accessibility	Physical Visits	Telemedicine
Transparency	Low	Ledger Tracking

### FUTURE SCOPE

Among the future enhancements, it is proposed to use larger datasets and deep learning models, full permissioned blockchain integration, live video consultation, and hospital management system integration.

### CONCLUSION

Smart Healthcare System is an effective implementation of AI-based prediction of heart diseases and blockchain-style management of medical records. The classifier named Decision Tree is useful for early diagnosis; a hash-chained ledger helps to ensure tamper-evident storage. Telemedicine is more accessible and offers a safe, transparent alternative to traditional platforms.

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