

SMART HEALTH ANALYSIS SYSTEM USING MACHINE LEARNING AND DEEP LEARNING

Project Reference No.: 49S_BE_7001

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Keywords:

Smart Health Analysis, Machine Learning, Deep Learning, Multi-modal Medical Data, Medical Image Classification, ECG Signal Analysis, Automated Diagnostic Prediction.

Introduction:

The healthcare sector produces large amounts of data such as medical images, blood reports, and ECG signals. Traditional diagnosis depends on manual analysis, which can be time-consuming and may lead to errors.

To address this, AyuSense is developed as a smart health analysis system using Machine Learning and Deep Learning techniques. It analyzes different types of medical data including skin images, X-rays, CT scans, ECG signals, and blood reports.

Deep learning models such as CNN, MobileNetV2, and ResNet50 are used for image analysis, while 1D-CNN is used for ECG signal classification. Machine learning models like Random Forest are used for structured data analysis. The system provides accurate and real-time predictions, helping in early disease detection and improving healthcare efficiency.

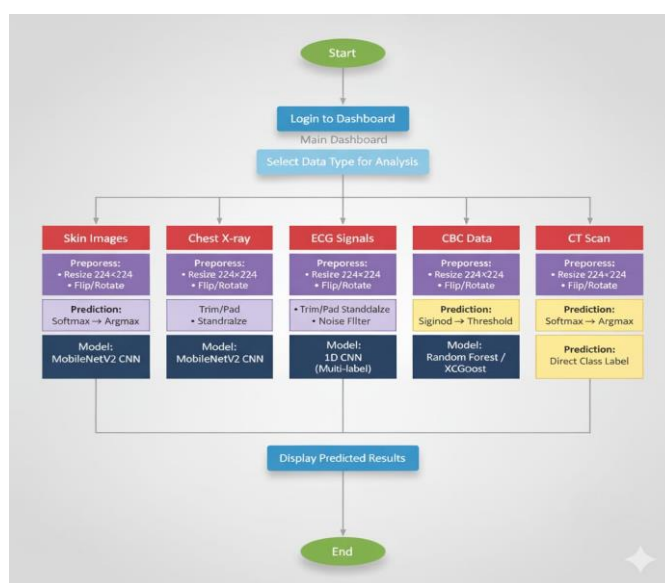


Figure 1: System Architecture

Objectives:

1. **Data Collection** – Gather diverse medical datasets, including blood tests, skin images, chest X-rays, and ECG signals, and preprocess them for use in model training.
2. **Model Training** – Train dataset-specific models using Random Forest for blood data, CNNs for images, and multi-label CNNs for ECG signals to accurately classify health conditions.
3. **Prediction** – Enable real-time, single-report prediction by allowing users to upload reports through a Tkinter interface and generate disease predictions with probabilities.
4. **Dashboard** – Provide a secure Streamlit-based dashboard for users to log in, upload reports, and visualize predicted health conditions along with interactive charts and insights.

Methodology:

1. The project begins with collecting datasets such as medical images, ECG signals, and blood reports. Preprocessing techniques like normalization, noise removal, and data augmentation are applied.
2. Deep learning models are used for feature extraction, where CNN-based models process images and 1D-CNN processes ECG signals. Structured data is analyzed using machine learning models.
3. MobileNetV2 and ResNet50 are used for image classification, while Random Forest is used for blood report analysis.
4. The models are trained and evaluated using performance metrics such as accuracy and F1-score. Finally, the system is implemented using a Streamlit interface where users can upload data and obtain predictions with confidence scores.

Result and Conclusion:

The system achieved good accuracy in analyzing different medical data types. Image-based models successfully detected diseases, and ECG models identified heart-related abnormalities. Blood report analysis also produced reliable predictions.

In conclusion, AyuSense effectively combines machine learning and deep learning techniques to provide accurate and real-time health analysis, reducing manual effort and supporting early diagnosis.

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Future Scope:

The future scope of this project includes:

1. Integration of additional diseases such as brain tumor and diabetic retinopathy detection.
2. Deployment as a mobile or cloud-based healthcare application.
3. Integration with IoT devices for real-time health monitoring.