



MINOR PROJECT

Smart Fall Detection and Daily Assistance System for Elderly People

Under the Guidance of Dr Krishna Pal Sharma Sir

Team Details

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Table of Contents

- Introduction
- Objectives
- Problem Statement
- Proposed Solutions
- Tools and Technologies Used
- System Architecture
- Project Development Phases
- Implementation Phases
- References

Problem

1

Falls are the leading cause of injury-related deaths among seniors aged above 65 and the second most common cause of unintentional injury mortality among adults of all ages.

2

In many cases, immediate medical help is not available, and delays in assistance can worsen the condition or lead to long-term complications.

3

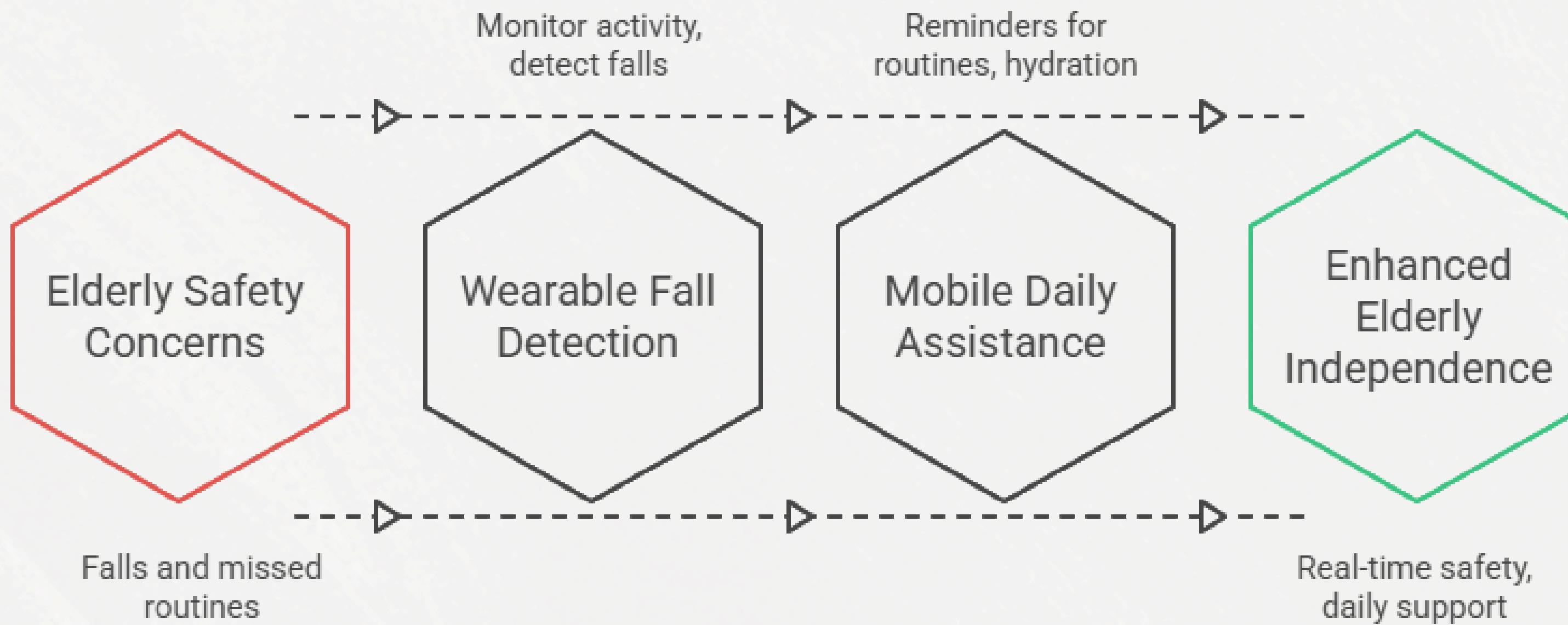
Additionally, many elderly people live alone and struggle with daily tasks or emergencies, increasing their dependence on others.

4

There is a strong need for a smart wearable system that can automatically detect falls, send real-time alerts to caregivers, and offer daily activity assistance — ensuring timely help, safety, and independent living for the elderly population

Solution

Comprehensive Elderly Care Solution



Tools and Technologies Used

Hardware Components

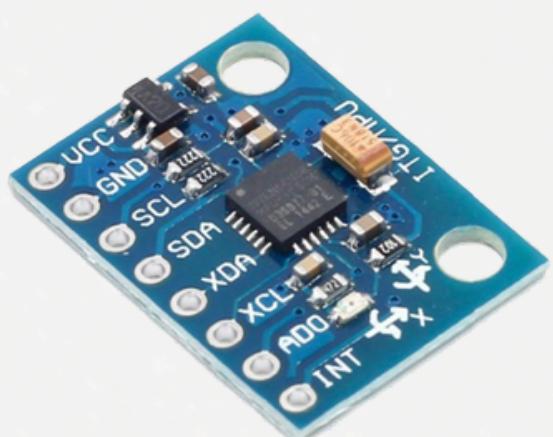
- **ESP32 Microcontroller** – Main processing unit with built-in Wi-Fi for data analysis and cloud communication.
- **MPU6050 Sensor** – 6-axis accelerometer and gyroscope for motion, orientation, and fall detection.
- **Battery (9V)** – Power source for the wearable device.
- **Jumper Wires** – Used to connect MPU6050 to ESP32 through the I²C interface (SDA, SCL, VCC, GND).

Components	Quantity
ESP32	1
MPU6050	1
Battery	1
Jumper Wires	7

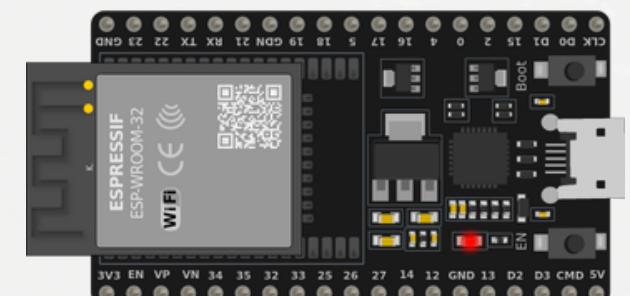
Software Tools

- **Arduino IDE** – Used for programming, compiling, and uploading Embedded C/C++ code to ESP32.
- **React Native Based Mobile App** – IoT platform for real-time monitoring, notifications, and device management.
- **Embedded C / C++** – Language used to implement the fall detection algorithm and handle sensor communication.
- **Dataset:** WEDA Dataset is used to train Decision tree model.

Components



MPU6050

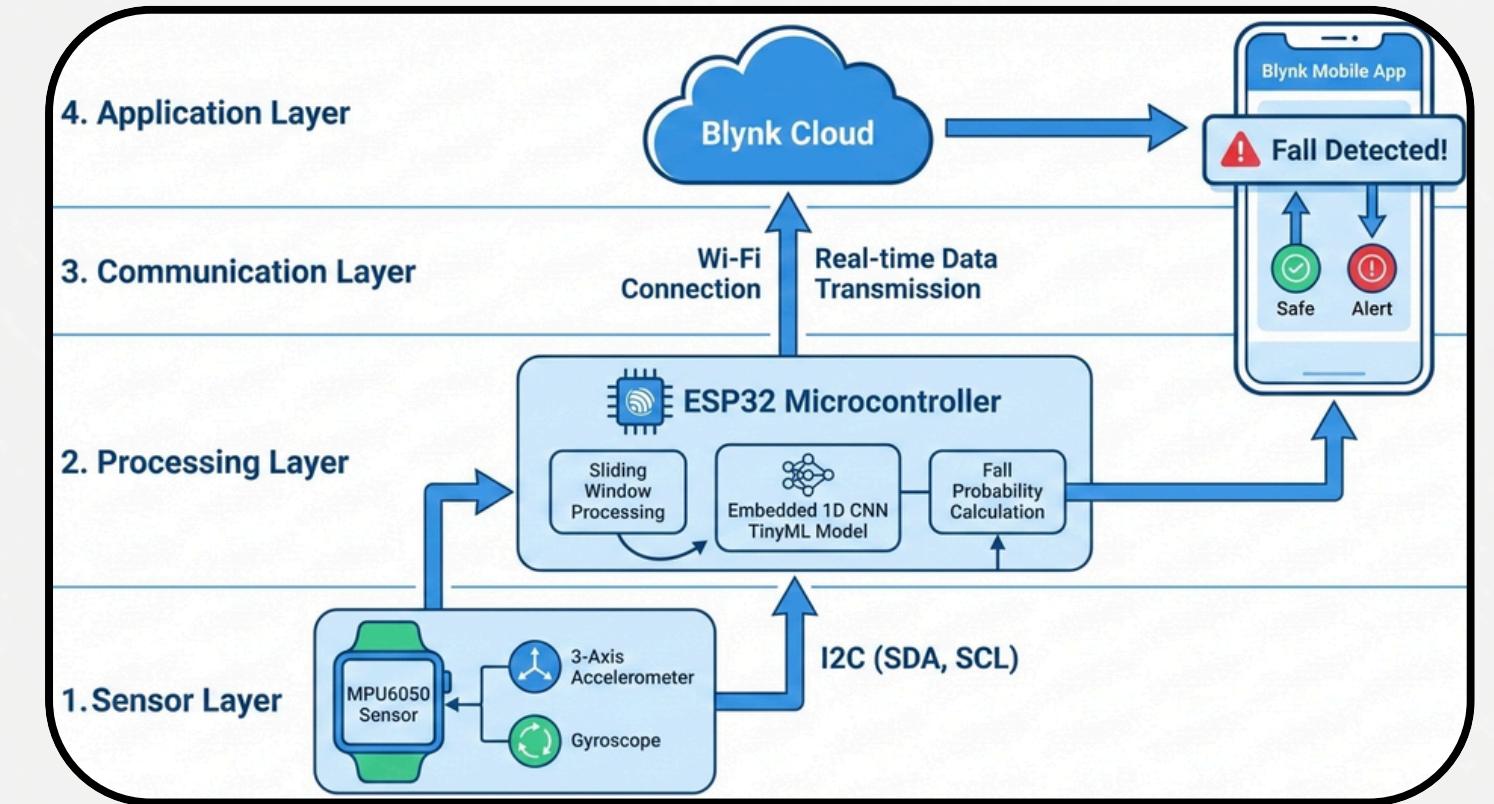


ESP32
Microcontroller

MODEL ARCHITECTURE

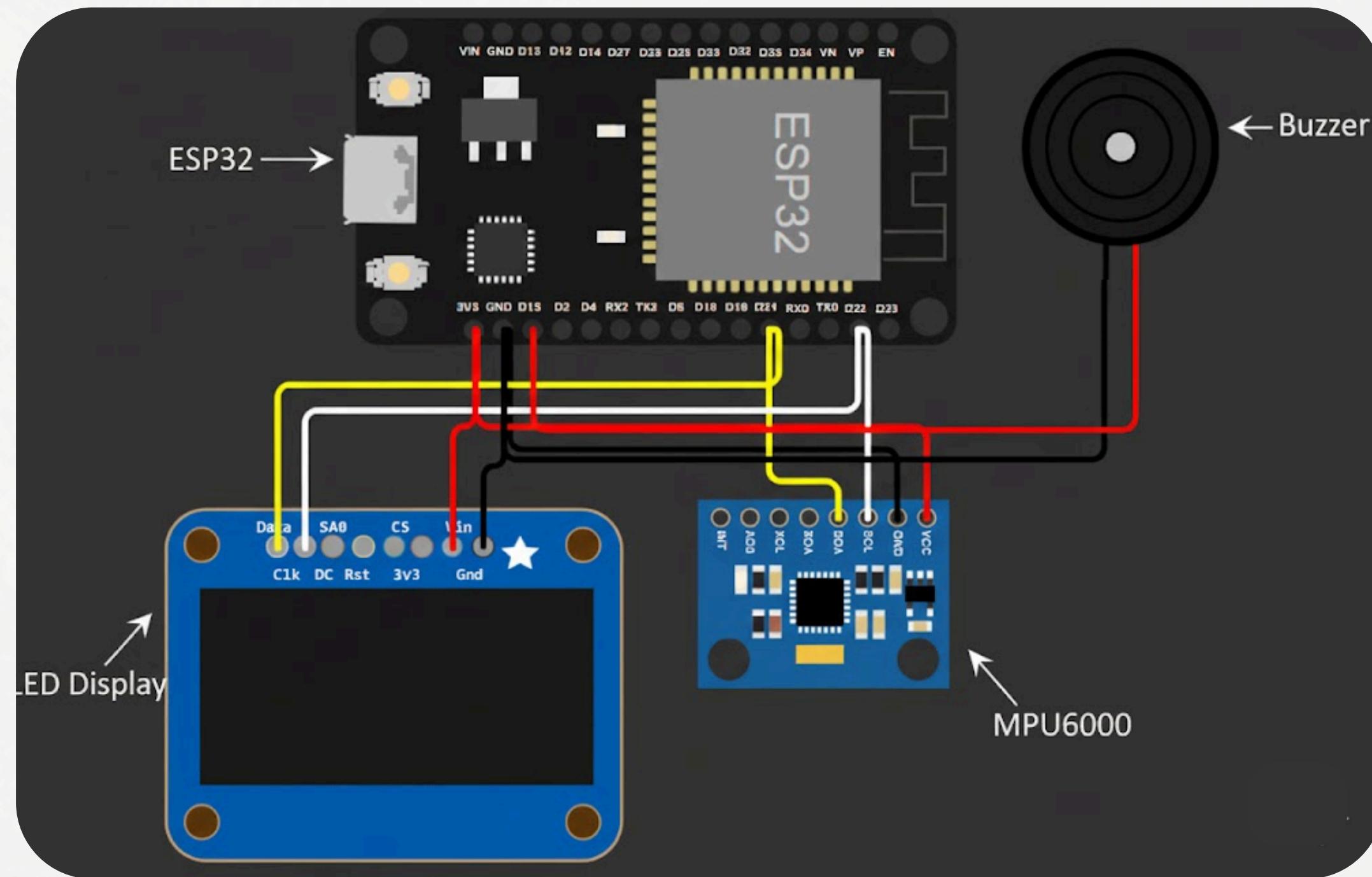
Fall Detection Model Architecture (1D CNN – TinyML)

- Input: 6-axis motion data
- $(A_x, A_y, A_z + G_x, G_y, G_z)$
- Window Size: 100 samples (≈ 2 sec motion)
- Model Type: 1D Convolutional Neural Network
- Layers:
 - Conv1D + ReLU (extract motion spikes)
 - Conv1D + ReLU (detect fall patterns)
 - Flatten Layer
 - Dense Layer
- Sigmoid Output
- Output: Fall Probability (0–1)
- $0.90 \rightarrow \text{Fall}$
- $\leq 0.90 \rightarrow \text{Normal Activity}$
- Optimized using TensorFlow Lite (TinyML) for ESP32 deployment.



IMPLEMENTATION

SIMULATION OF PROPOSED HARDWARE SOLUTION:



WEDA-FALL DATASET

WEDA-FALL DATASET

Dataset Overview:

This comprehensive wrist-based dataset provides multi-sensor data specifically designed for fall detection research.

- Captured using Fitbit Sense Smartwatch
- Sampling Frequency Used: 50 Hz
- Sensors Included: Accelerometer (Ax, Ay, Az), Gyroscope (Gx, Gy, Gz), Orientation Sensor, Vertical Acceleration
- Multiple frequency versions available (50Hz, 40Hz, 25Hz, 10Hz, 5Hz)
- The file `fall_timestamps.csv` contains manually annotated start and end timestamps of actual fall events.

WHY WE CHOOSE 1D CNN + WEDA DATASET

- **Wrist-Based & Realistic**

Captured using a Fitbit Sense smartwatch, matching our device's wrist placement for practical and realistic deployment.

- **Includes Elderly Participants**

Age range 20 – 95 years, providing crucial real elderly movement data for accurate model training.

- **High-Quality Sensor Data**

50 Hz sampling frequency with 6-axis IMU data (Accelerometer + Gyroscope) provides precise fall phase timestamps.

- **Diverse Activities**

8 types of falls and 11 ADLs (Activities of Daily Living), designed to reduce false positives by mimicking fall-like motions.

PARTICIPANTS & DATA COLLECTION

- Two Groups: Young (age 20-40) and Elderly (age 60-95)
- Total Participants: 25 (with real elderly movement data)
- Note: Elderly participants performed only safe ADLs (no falls) for safety.
- Fall Signals (Young): 350
- ADL Signals (Young): 462
- ADL Signals (Elderly): 157
- Total ADL Signals: 619
- Data collected on mattress for fall safety simulation.

Weda dataset link:

<https://github.com/joaojtmarques/WEDA-FALL>

MODEL PIPELINE

Data Source – WEDA-FALL Dataset

- Wrist-based IMU data (50Hz)
- 6-axis input:
- Accelerometer (Ax, Ay, Az)
- Gyroscope (Gx, Gy, Gz)
- Window Size: 100 samples (2 seconds)
- Binary Labels:
- 0 → ADL
- 1 → Fall

Data Preprocessing

- Merge accelerometer + gyroscope
- Apply Sliding Window (100×6)
- Normalize using StandardScaler
- Train-Test Split (80:20)
- Handle imbalance using class weights

1D CNN Model Training

- Input Shape: (100, 6)
- Architecture:
 - Conv1D (Feature Extraction)
 - Conv1D (Pattern Learning)
 - MaxPooling
 - Global Average Pooling
 - Dense Layer
 - Sigmoid Output

Model Evaluation

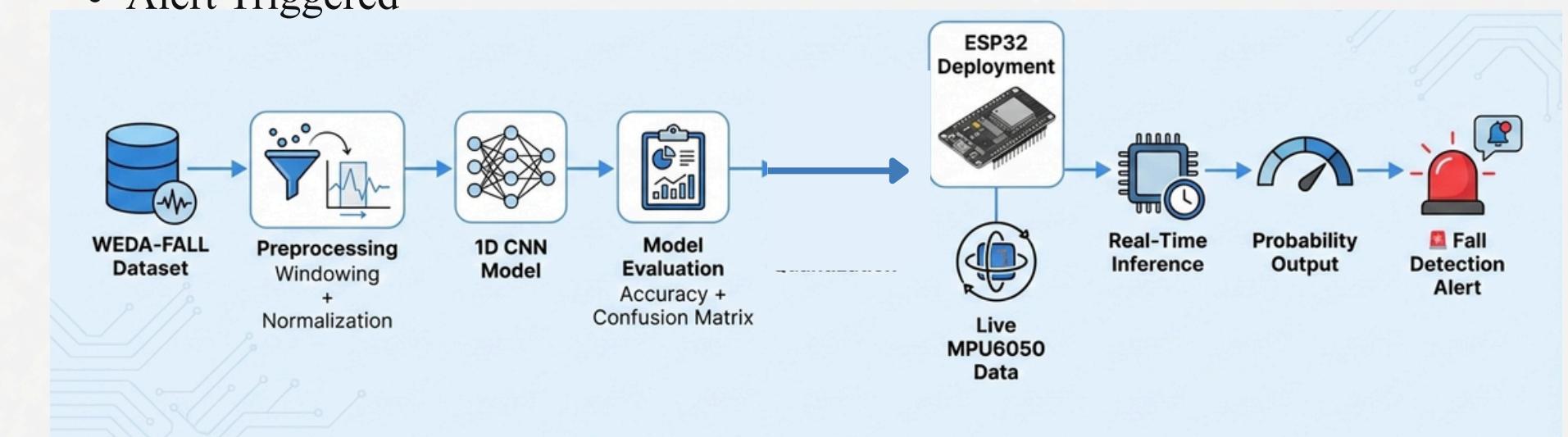
- Accuracy: 91%
- Fall Recall: 0.91
- F1 Score: 0.85
- Confusion Matrix Analysis
 - High recall ensures fewer missed falls
 - Balanced precision–recall tradeoff

TinyML Optimization

- Convert to TensorFlow Lite
- INT8 Quantization
- Final Model Size: ~8.8 KB
- Compatible with ESP32 memory constraints

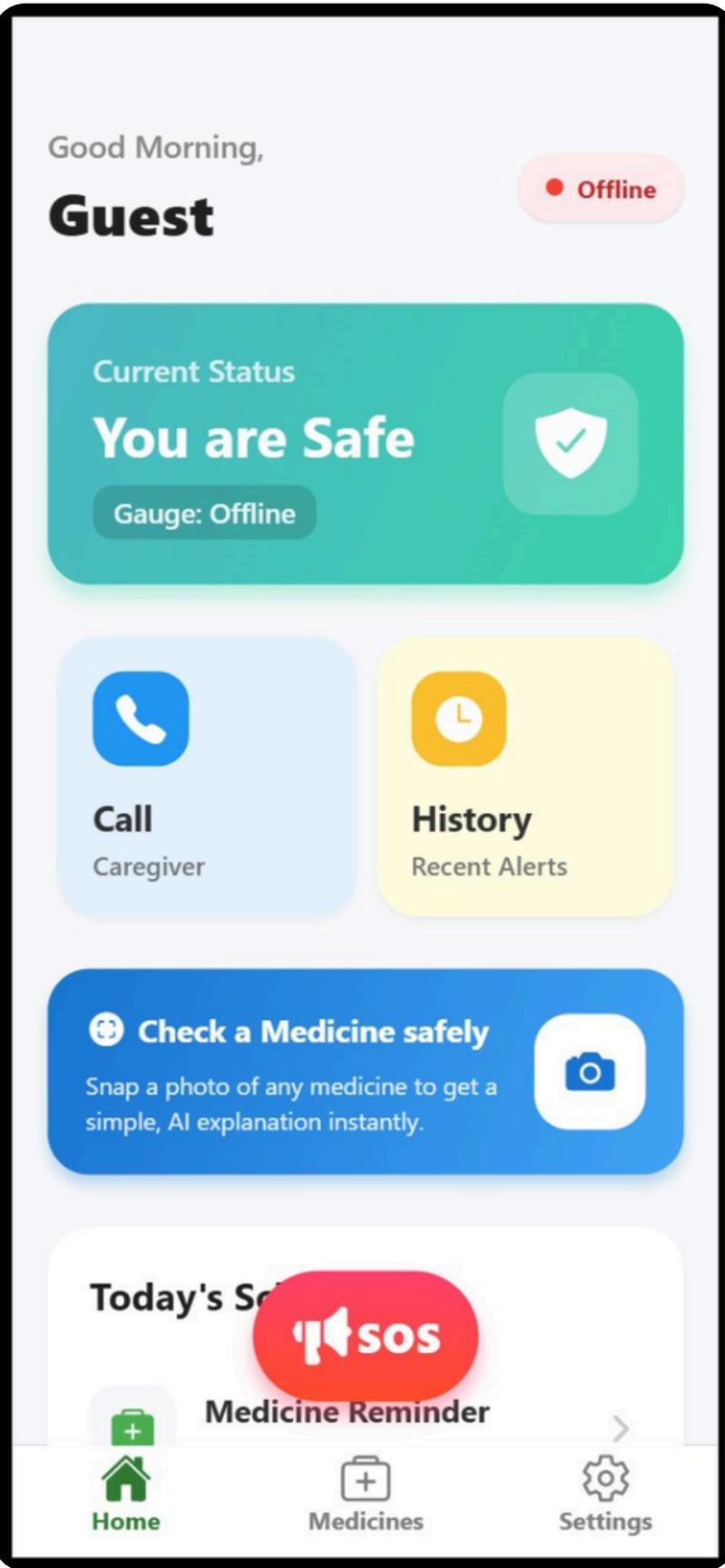
Real-Time Deployment

- MPU6050 Sensor → Collect 100 Samples →
- Normalize → Quantize → ESP32 Inference →
- Generate Fall Probability → Threshold Decision →
- Alert Triggered



Introducing..

CareGuard

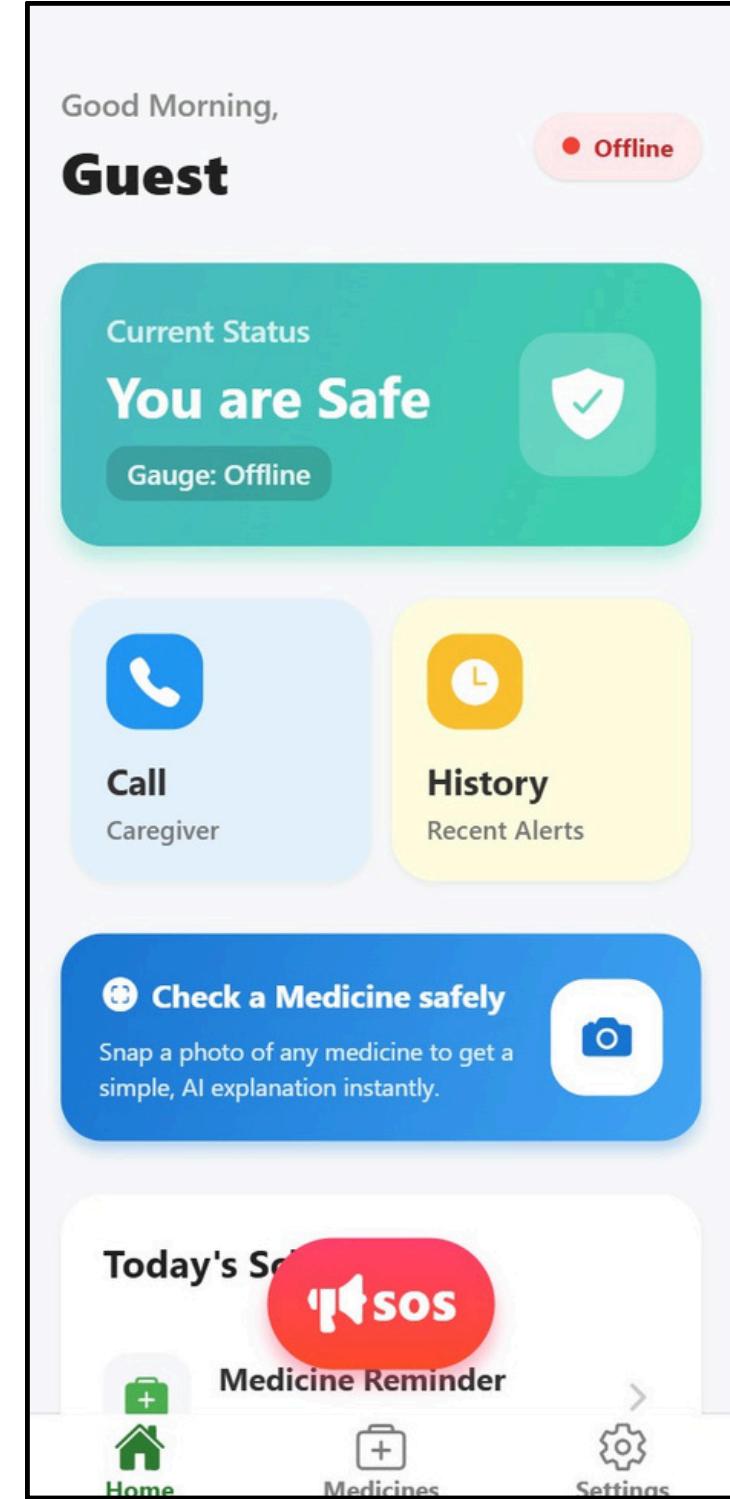
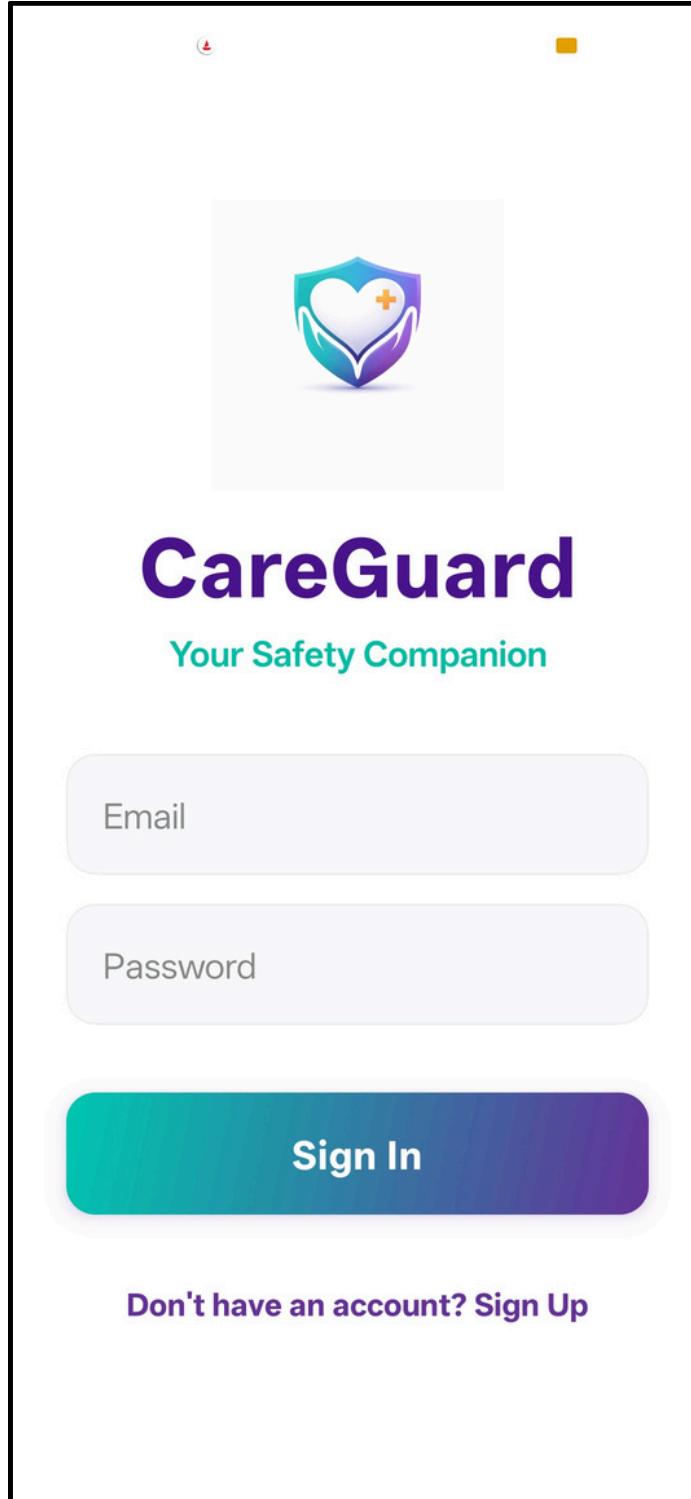


CareGuard APP

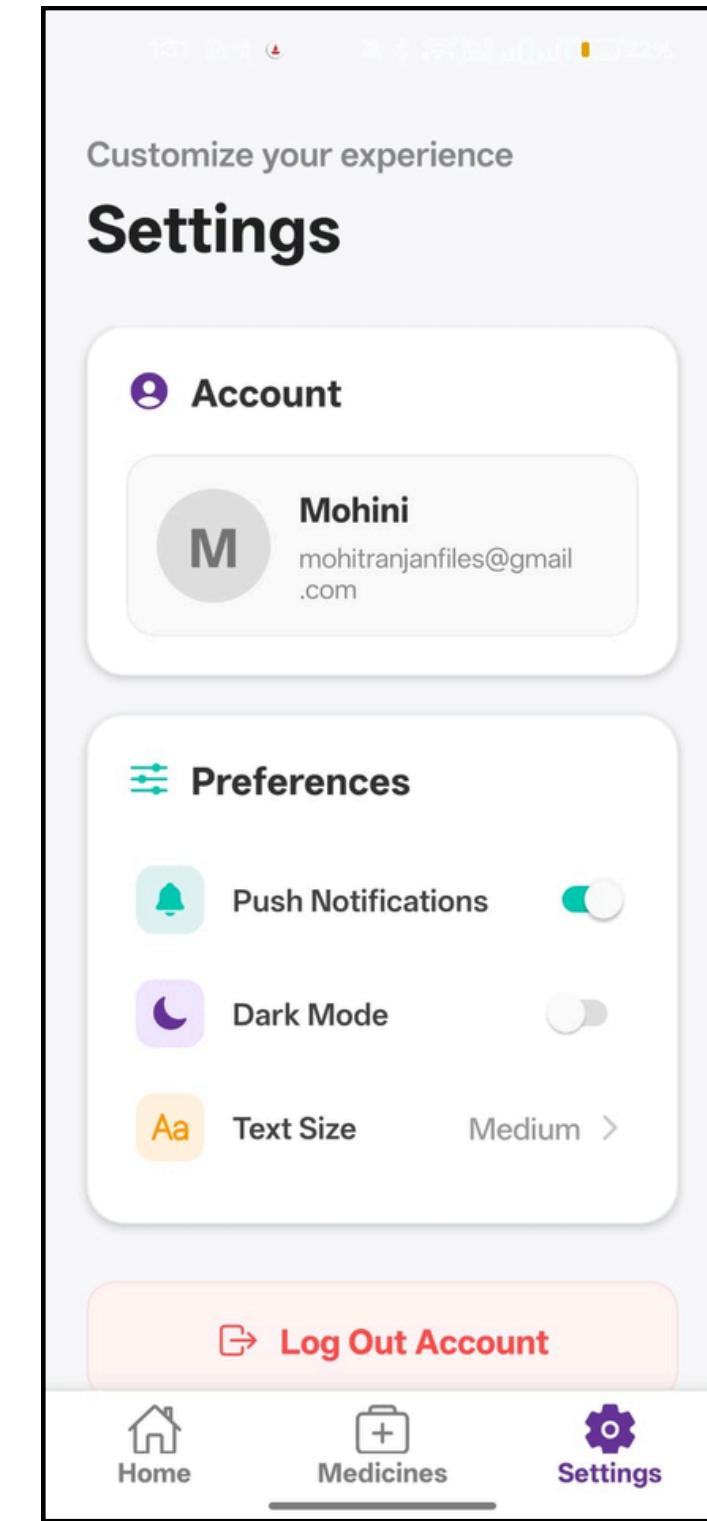
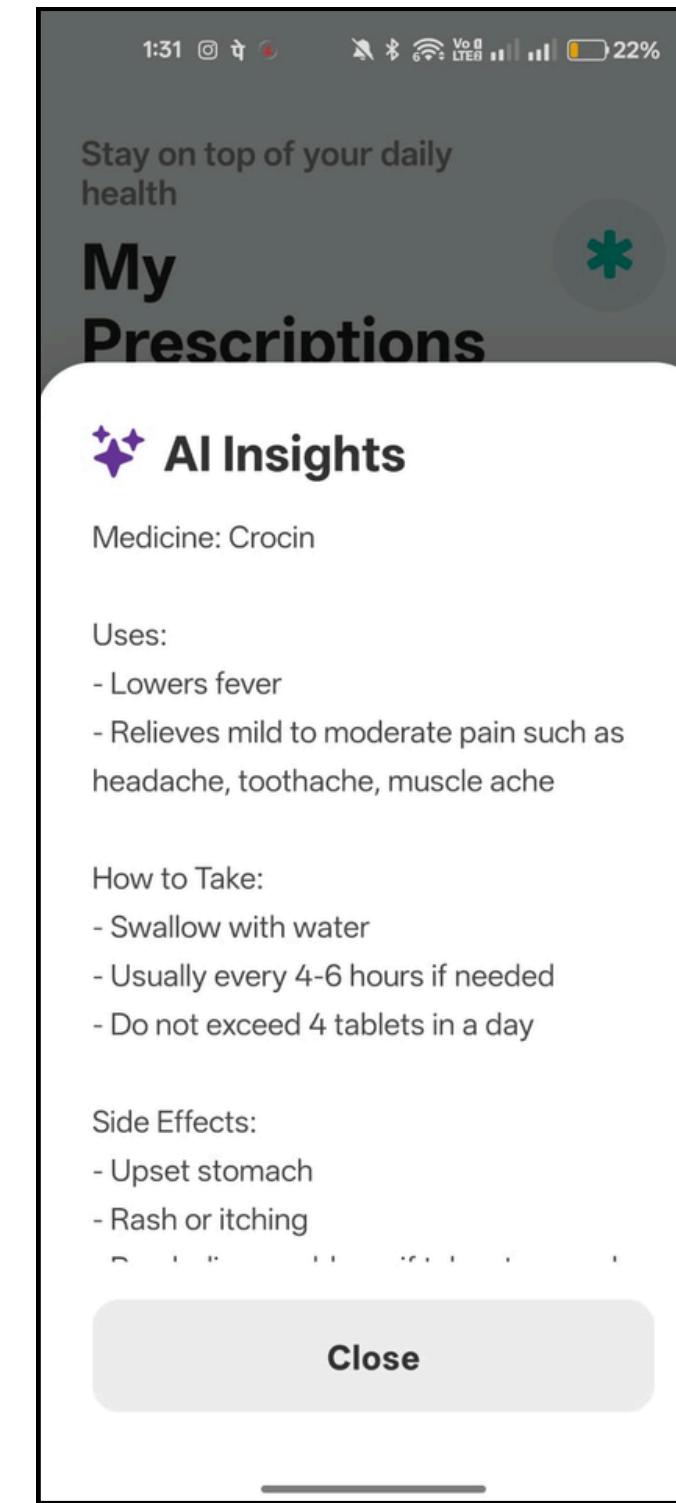
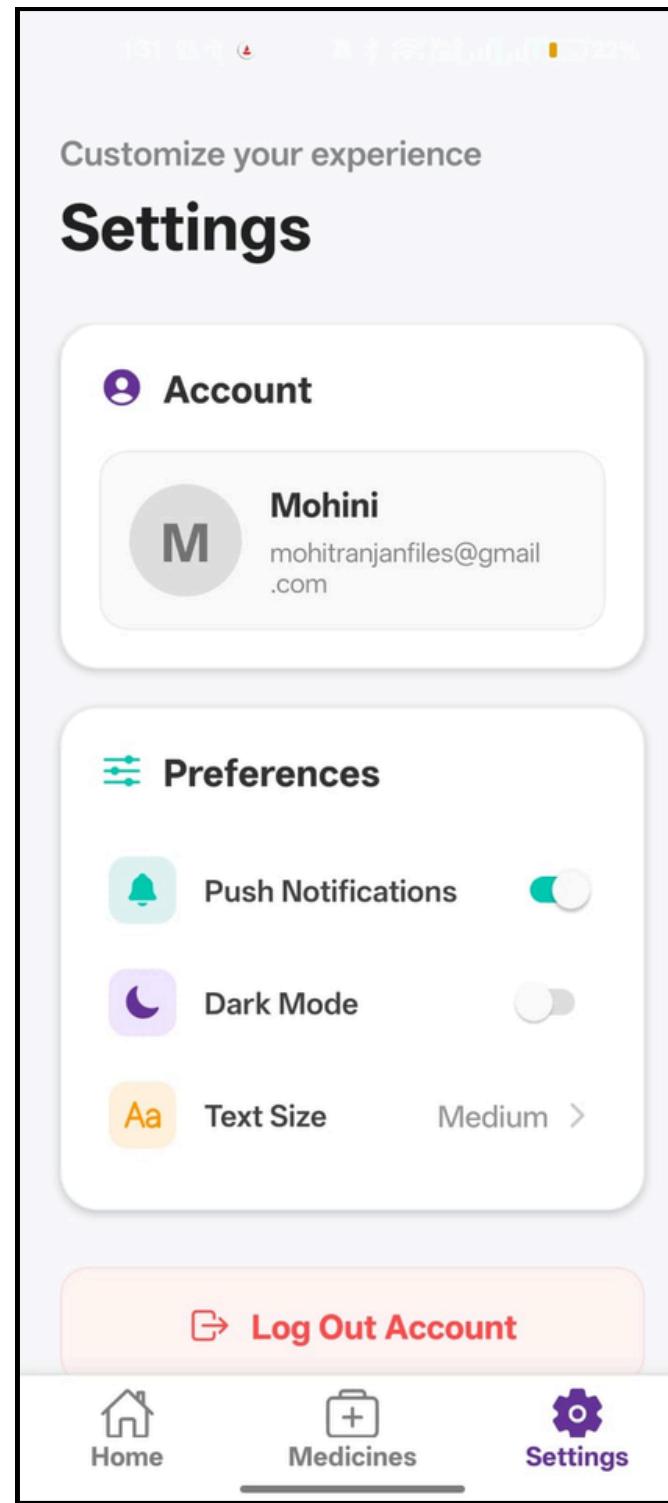
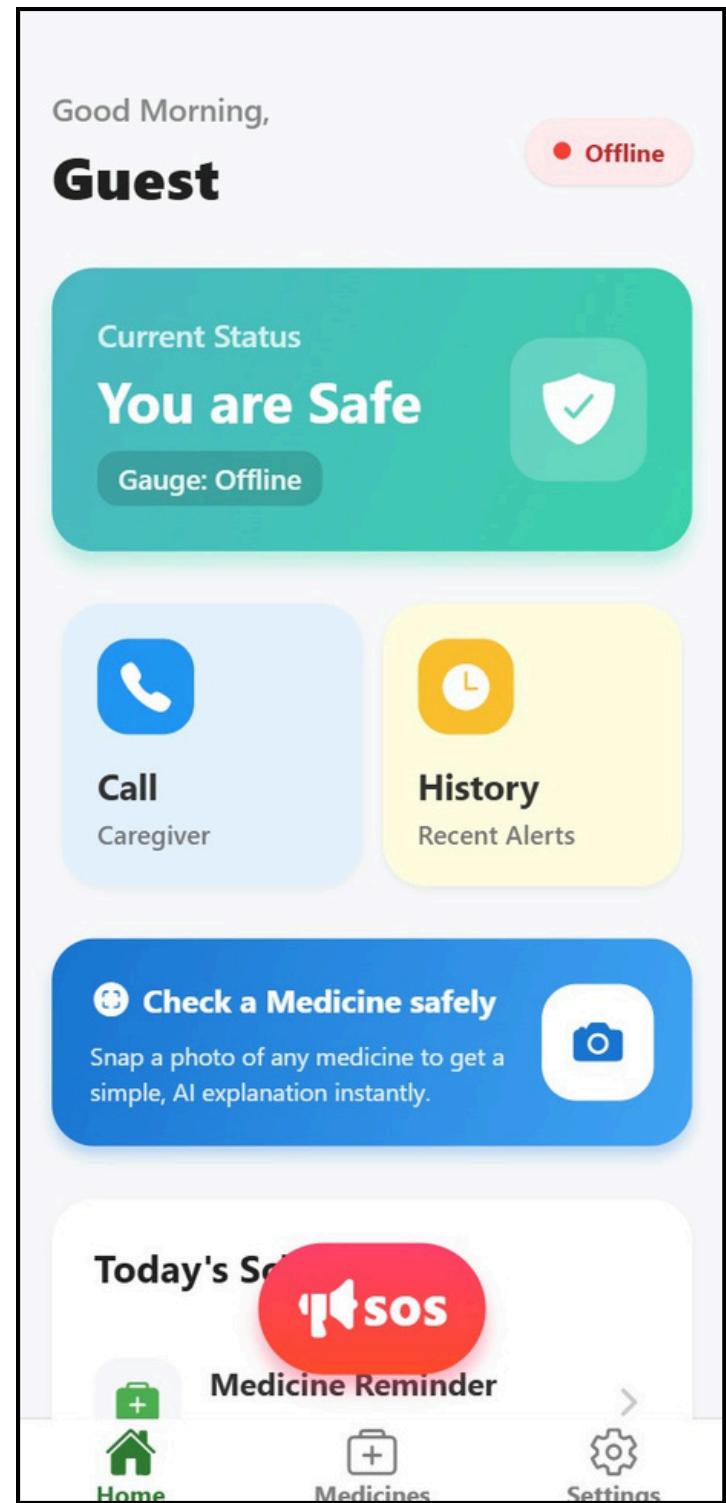
React Native app for real-time monitoring & elderly assistance

Key Features

- Fall Alert System** – Detects falls and sends instant alerts
- Medication Reminders** – Ensures timely medicine intake
- Hydration Alerts** – Promotes regular water consumption
- Health Dashboard** – Tracks activity, fall history & alerts
- WhatsApp API** – Enables scalable alert notifications
- AI Prescription Assistant** – Uses OCR + AI to simplify medicine instructions



Features



CareGuard AI

Medicine Scanner Workflow

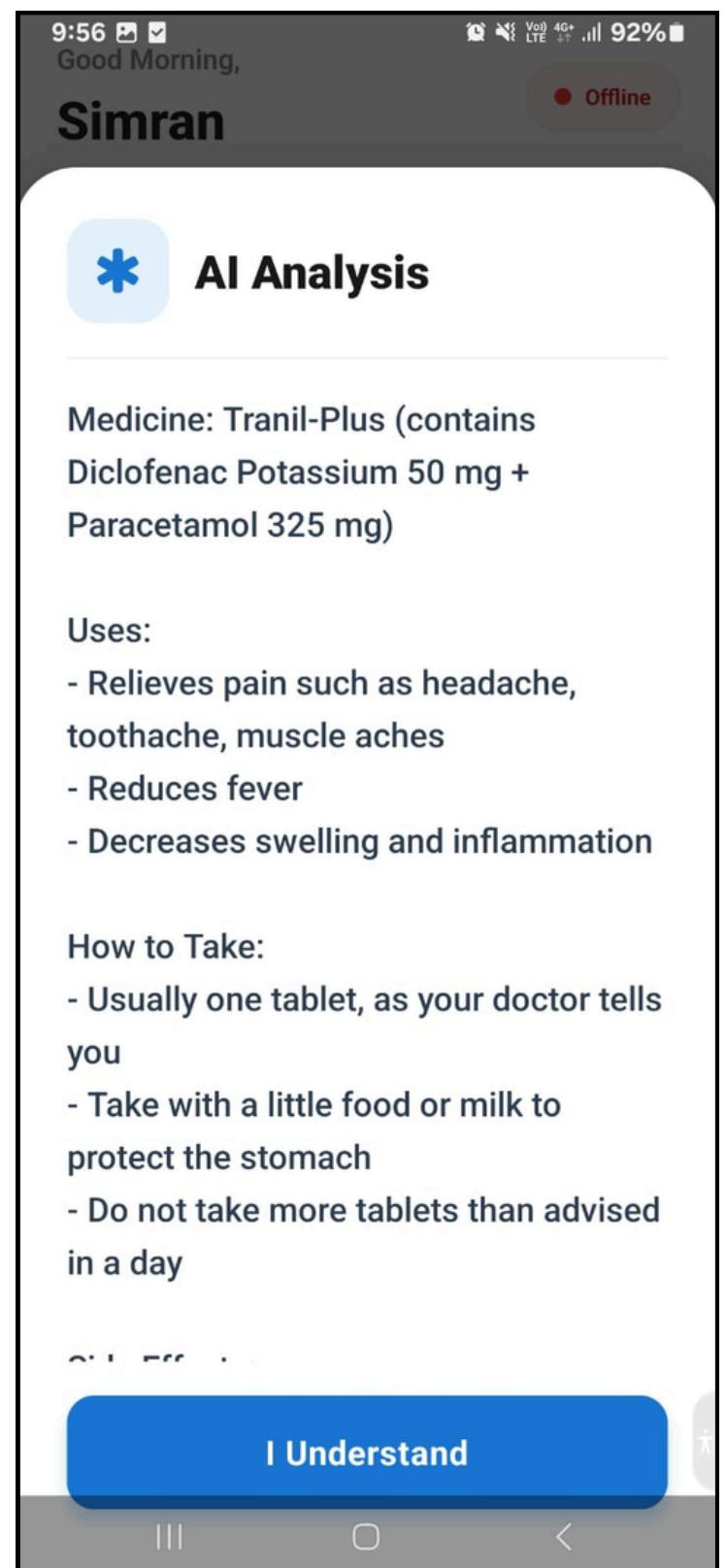
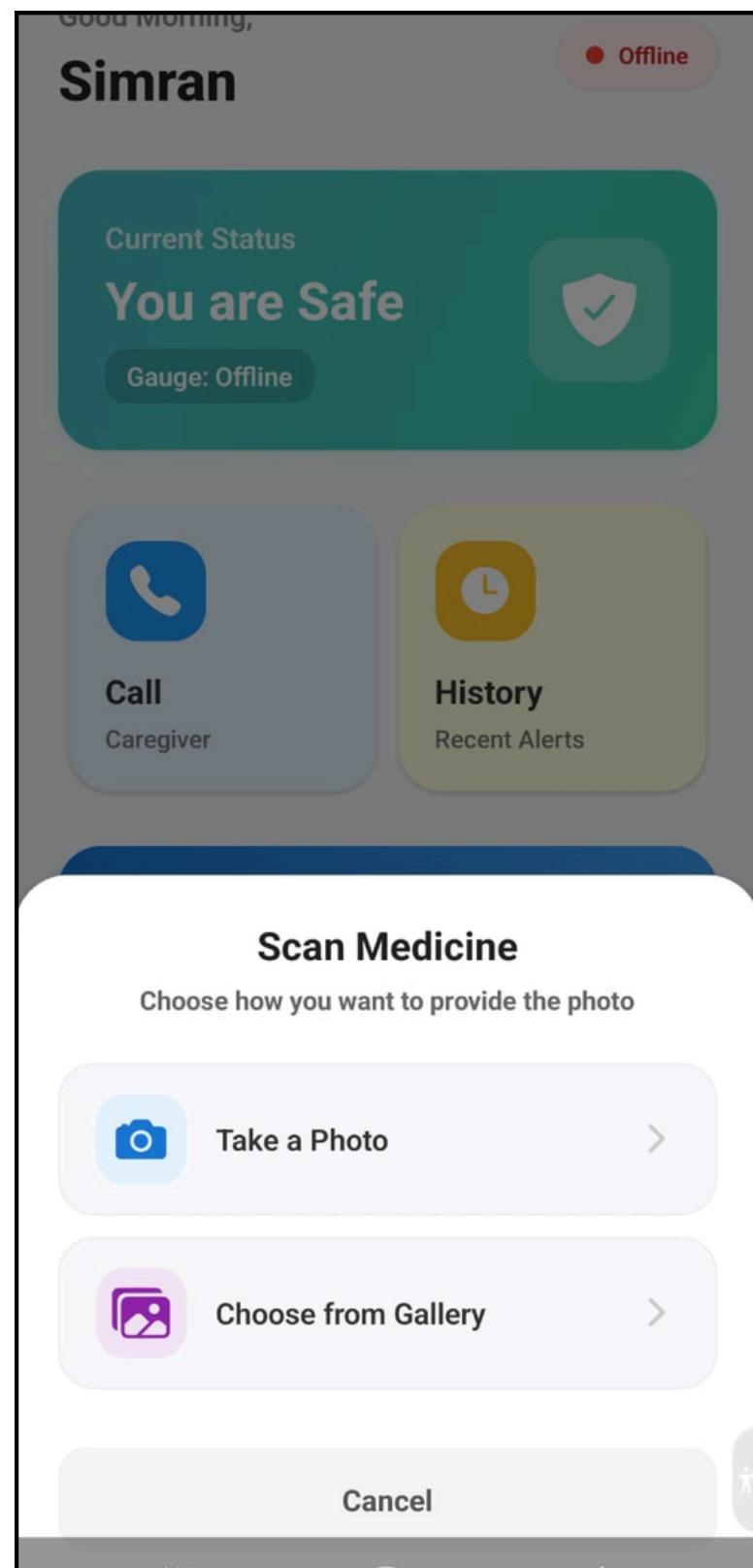
Image Capture – Photo taken & converted to Base64

OCR Processing – Extracts text using OCR.Space (Engine 2)

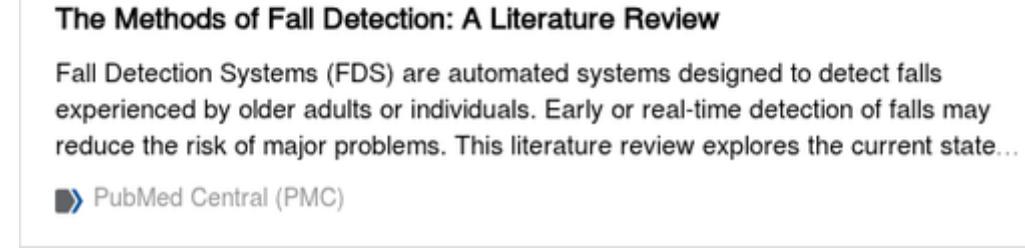
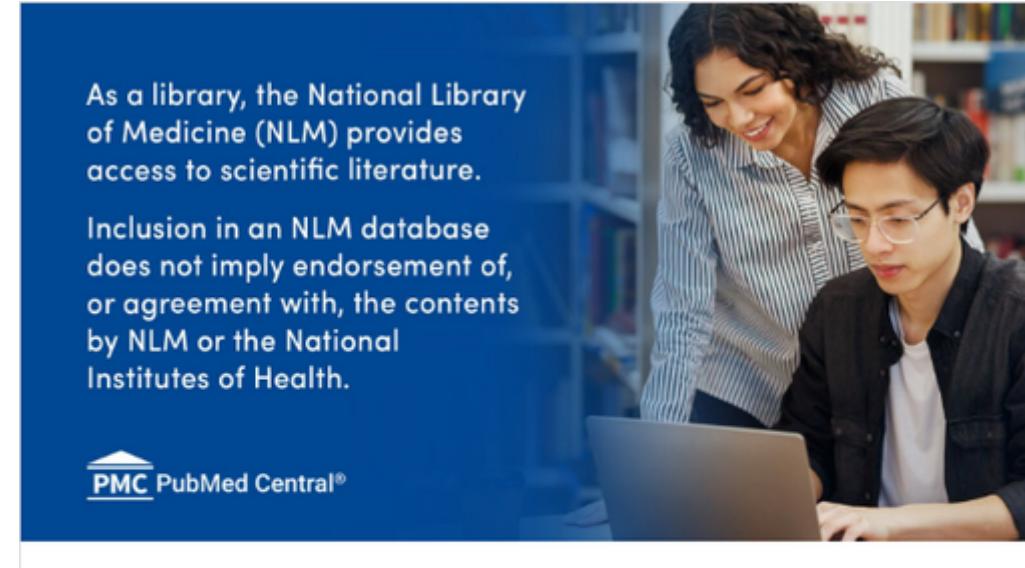
Text Cleaning – Removes noise from scanned text

AI Processing – Groq + GPT-OSS-120B interprets medicine details

Final Output – Simple, elderly-friendly explanation shown in app



REFERENCES



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THANK YOU
FOR YOUR ATTENTION