

Smart Infant Incubator Monitoring System Using Cloud Sync with IoT Integration

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Abstract. The infant mortality rate remains close to historical levels in developing countries due to the lack of incubation systems and their affordability in small hospitals. This research introduces a cost-effective and smart incubator equipped with an IoT-based monitoring system for infants. The smart incubator continuously monitors critical parameters using IoT sensors. These sensors track various factors, such as body temperature, heart rate, humidity, sound levels, vibration (movement), and the temperature inside the incubator. The system utilizes ESP8266 and Arduino UNO to gather data from sensors and upload it to the Blynk IoT cloud for real-time monitoring via smartphone while also sending automated SMS or email notifications to the designated doctor during abnormal conditions. This ensures that both medical professionals and caregivers can promptly intervene in case of anomalies, thereby reducing response time and enhancing infant safety. The system facilitates remote monitoring for the incubators, and in doing so, it offers a significant advantage in resource-constrained healthcare environments.

Keywords: Infant Mortality, Real-Time Monitoring, Cloud Sync, Smart Incubation.

1 Introduction

According to a study, the major reason for the 700000 neonatal deaths in India in 2015 were the low- birth weight and unseasonable birth. It's also set up that neonatal deaths due to the reasons mentioned over are around 29 per 1000 births each time. The rise in the deaths of unseasonable babies and low weight at birth is ununiform across India. It means that the death rate is high in pastoral areas and lower in civic areas. likewise, these babies demanded more investments in incubators and ferocious Care Units (ICU) for proper neonatal care. The recent advancement in technology leads to improvement of the medical assiduity and hence the mortality rate of unseasonable babies is also controlled. Due to the rate of the number of caregivers to the number of cases not matching i.e., further cases and smaller caregivers, the workload of the instrument- health caregiver is high, which leads to indecorous monitoring of the incubators [1]. Neonatal care is critical in reducing child mortality, especially for unseasonable or at- threat babe. Traditional Incubation Systems give only temperature and moisture regulation but

frequently warrant real- time monitoring. This design introduces a pull sync with the IOT Integrated Smart Infant Incubation system that captures the detector data and environmental data, transfers the data to the cloud storage, and live alert through SMS. In many developing nations, small and rural hospitals often struggle to provide proper neonatal care due to the high costs associated with installing and maintaining advanced incubator systems. These financial barriers prevent access to modern, automated monitoring solutions for premature or critically ill infants. Consequently, healthcare workers are forced to rely on manual methods to track vital parameters such as temperature, humidity, and respiratory rate. This manual approach not only demands constant attention but also increases the likelihood of human error, potentially leading to delayed responses and serious health risks for newborns. Additionally, the absence of integrated digital systems limits the ability to store, analyse, or remotely access data, making it difficult to ensure consistent and informed care. To address these challenges, there is a clear need for a low-cost, intelligent incubator monitoring system that incorporates IoT technology and cloud connectivity, enabling real-time data acquisition, automation, and remote supervision to enhance neonatal care in underserved healthcare environments.

2 Literature Review

Recent advances in IoT-predicated child incubator systems have significantly bettered neonatal care, particularly in low- resource settings. Early work concentrated on erecting cost-effective, portable systems integrated with microcontrollers analogous as Arduino and introductory environmental sensors. For case, one analogous system combined sensors for temperature, humidity, headwind, and oxygen situations with bedded IoT connectivity, enabling real- time monitoring and active environmental control. This setup also featured safety mechanisms including suckers, heaters, and power failure discovery systems. still, it demanded predictive analytics, advanced anomaly discovery, and integration with broader sanitorium structure, limiting its scalability and clinical robustness [2].

Expanding on this foundation, more recent sweats have incorporated machine knowledge and energy optimization into incubator design. A prototype using a crossbred CNN- LSTM model was introduced to cover and predict energy consumption within incubator surroundings. By integrating sensors like DHT11 and ACS712, and combining Arduino Mega 2560 with boo Pi using MQTT protocols, the system achieved estimable prophecy delicacy. Yet, the work concentrated solely on energy operation, overlooking clinical issues, health monitoring, and anomaly discovery. also, the lack of NICU trials and system scalability assessments left critical questions unanswered [3]. Security and remote vacuity have also come central to modern incubator results. An IoT-predicated system using Arduino Uno and ESP8266 was designed with sensors for body temperature, humidity, gas presence, and an RFID module for security access. Data transmission passed via Wi- Fi to a remote database, while cautions were accessible through a web interface. Despite its donation to monitoring and access control, the system demanded independent environmental control, backup alert mechanisms in case of Wi- Fi failure, and integration with sanitorium systems. also, long- term health data logging and analysis were not addressed [4]. Other designs have emphasized smart automation and real- time responsiveness. One analogous result employed ThingSpeak and

GSM for remote monitoring and integrated multiple sensors including eyeblink, gas, sound, and stir. pickers like Peltier bias and humidifiers laboriously maintained environmental conditions, while cautions were delivered through buzzers and GSM dispatches. Data was displayed locally via television and uploaded to the pall via Arduino Wi- Fi modules. nevertheless, this system neglected predictive algorithms, EHR integration, and power redundancy. It also demanded security protocols analogous as encryption and authentication [5].

In low- cost, GSM- predicated systems, sweats have been made to produce introductory waking mechanisms using SMS. One prototype covered pivotal vitals analogous as temperature, eyeblink, weight, and sound, transferring continuous SMS cautions to healthcare providers until they were conceded. While this design achieved simplicity and affordability, it did not support pall connectivity, long- term data logging, automation, or scalability for sanitorium surroundings. It was also devoid of predictive analytics or smart alert escalation sense [6]. Ultimately, attempts have been made to develop real-time IoT- predicated incubator systems with mobile app integration. One analogous system, erected using NodeMCU ESP8266, incorporated sensors like DHT11, MQ2, KY-039, and LM393, and transmitted data to Firebase for mobile access. delicacy was vindicated using a t- test against a marketable OHMEDA incubator. still, limitations included shy testing with factual babes, unreliable humidity sensor estimation, and lack of AI integration — despite being mentioned as an ideal. Clinical evidence, durability, and hygiene considerations were also not addressed [7]. Ultimate of the systems reviewed also calculate on simple sensor data and leave advanced machine knowledge or artificial intelligence for anomaly discovery. Likewise in Table .1. they are constantly not scalable or adaptable to dynamic sanitorium surroundings.

Table .1. IoT- predicated child Incubator Systems

S.No	Key Focus Area	Technologies Used	Observed Limitations	Ref.
1	Portable, IoT-enabled incubator with sub-bag oxygen system	Arduino, sensors (temp, humidity, airflow, O ₂), fan, heater	No AI integration, no hospital system integration	[2]
2	Energy consumption prediction in smart incubator	CNN-LSTM, Arduino Mega, RPi, DHT11, ACS712, MQTT	Focused only on energy; no patient metrics or anomaly detection	[3]

S.No	Key Focus Area	Technologies Used	Observed Limitations	Ref.
3	Remote monitoring and security via RFID and IoT	Arduino Uno, ESP8266, Wi-Fi, sensors, RFID	No autonomous control, poor alert redundancy	[4]
4	Smart environmental control with GSM and ThingSpeak	Arduino Wi-Fi R3, Pel-tier, GSM, ThingSpeak, multiple sensors	No predictive analytics, no backup, limited security	[5]
5	GSM-based alert system for basic vitals monitoring	GSM, temp, heartbeat, sound, weight sensors	No cloud/web interface, no automation, manual reset needed	[6]
6	Real-time monitoring with mobile app and Firebase	NodeMCU ESP8266, Firebase, sensors (DHT11, MQ2, KY-039, LM393)	Limited clinical validation, no AI use despite objective	[7]

In particular, the systems described by [2] and [3] concentrate on introductory environmental control and energy consumption prophecy, but neither integrates advanced health monitoring or anomaly discovery, which are critical for neonatal care. Others, analogous as the system presented by [4], show pledge in enhancing security and remote access, but they are limited by the lack of automation and backup mechanisms, which could be vital in emergency situations. Thus, future IoT- enabled incubator systems need to not only meliorate energy effectiveness and security but also incorporate adaptive anomaly discovery and AI- predicated health monitoring systems. These advancements would enable real- time responses to abnormal conditions, thereby enhancing the quality of care for babes in both low- resource and advanced sanitorium settings.

3 Proposed Model

This proposed model offers an integrated smart infant incubation system for neonatal cares by Integrating IOT and Cloud based Technology for enabling automated and real time monitoring of the sensors and the incubator conditions using the sensors, micro-controller (Arduino UNO) and a wireless communication Module (ESP8266). The data are stored in the cloud base Databases by the backend through the NodeJS and Triggers

the SMS alert system when the reading are abnormal. Display the data in the dashboard through visualizing the reading in BLYNK dashboard. This architecture ensures manual intervention while enhancing responsiveness to neonatal health events. The integration of data-driven decision-making empowers and medical action. This model includes multiple sensors connected to Arduino for data collection and ESP8266. The Arduino sender the data collected from the sensor to the ESP8266 through the Serial Communication (RX --> TX & TX ---> RX), which acts as an WI-FI enabled microcontroller. After that the Data is transmitted to the cloud-based MongoDB Database. The NodeMCU, equipped with Wi-Fi capabilities, acts as the gateway to the cloud, transmitting data to cloud via HTTP Post request. The Node.js based backend process and manages data. The workflow of the model as mentioned in Fig.1. the Mobile application, developed from the BLYNK IOT 2.0, fetch real time data via APIs and display it on a dashboard. Simultaneously, Alert Mechanisms via SMS and email are triggered through the Infobip when The Abnormal readings from the sensors are detected.

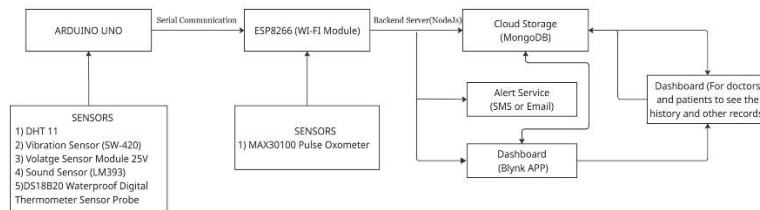


Fig. 1. Flow Diagram

Each sensor is connected to appropriate analog or digital pins on the Arduino UNO microcontroller. The DHT11 (temperature and humidity), DS18B20 (body temperature), MAX30100 (SpO2 and heart rate), sound sensor (LM393), vibration sensor (SW-420), and voltage sensor are physically integrated through breadboards and jumper wires. The Arduino is programmed to read each sensor's output, process or normalize the data where necessary, and transmit the formatted readings to the ESP8266 NodeMCU via UART (serial communication) mentioned in the Fig.2. The NodeMCU, equipped with built-in Wi-Fi, is configured as a client device that communicates with the cloud database. Power is supplied through USB and regulated connections to ensure consistent voltage to all components.

Hardware Components:

- *DHT11* – Measures ambient temperature and humidity inside the incubator to ensure a stable environment.
- *DS18B20 Thermistor Probe* – A waterproof digital sensor that measures the

baby's body temperature accurately.

- *MAX30100 Pulse oximeter* – A pulse oximeter and heart rate sensor that tracks the infant's SpO2 and heart rate.
- *Sound Sensor (LM393)* – Detects variations in sound levels, identifying if the baby is crying.
- *Vibration Sensor (SW-420)* – Monitors for abnormal vibrations or movement within the incubator structure.
- *Voltage Sensor (25V)* – Detects voltage levels to assess power supply stability.
- *Arduino UNO* – Acts as the central unit for analog and digital signal acquisition from sensors.
- *ESP8266 NodeMCU* – Serves as the communication bridge to the cloud via Wi-Fi

Software Stacks:

- *Arduino IDE* – Used to program the Arduino UNO and NodeMCU micro-controllers.
- *Node.js* – Powers the backend server, which handles sensor data, API requests, and alert logic.
- *MongoDB Atlas* – A cloud-hosted NoSQL database that stores sensor data with timestamps and alert flags.
- *Blynk IOT* – An open-source framework used to develop a cross-platform mobile app displaying real-time data.
- *Infobip API* – Facilitates automatic SMS alerts to medical personnel in case of critical readings.
- *Nodemailer* – Handles the automated email alerts as part of the backend service

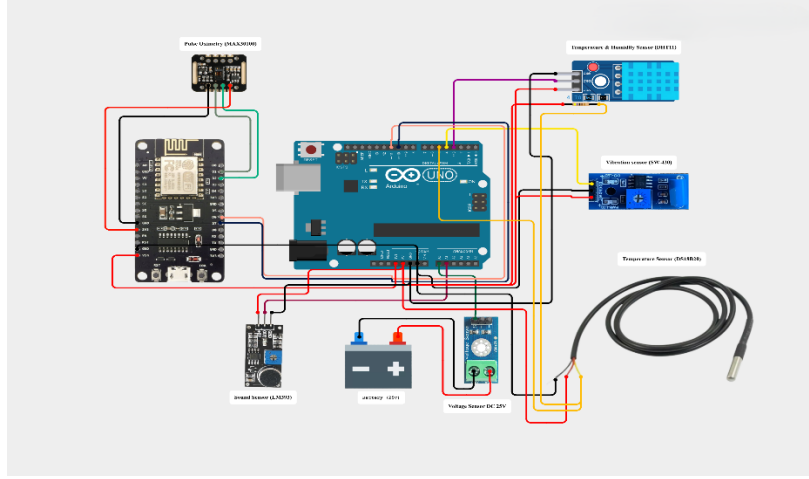


Fig. 2. Circuit implementation of the all the sensors

4 Implementation

The software stack consists of embedded code for microcontrollers and a cloud-based backend server. The Arduino is programmed using the Arduino IDE to read sensor inputs and send structured data in JSON format over the serial interface. The ESP8266 NodeMCU is also programmed via Arduino IDE to receive this data, parse it, and forward it to a web server using HTTP POST requests. The backend is developed in Node.js, where Express.js handles routing and API endpoints. The server parses incoming data, validates the structure, checks for threshold violations, and stores the results into a MongoDB Atlas cloud database. Custom API endpoints allow for retrieval of both real-time and historical data. MongoDB Atlas serves as the cloud database, offering a distributed and fault-tolerant infrastructure. Each incoming data packet is stored as a document in a structured collection, tagged with a unique timestamp and sensor identifier. Indexing ensures fast query responses for the mobile application and backend analytics. The database is designed to scale as more incubators or patients are added to the system, maintaining high availability and low latency. The dashboard utilizes the Blynk IoT platform. The NodeMCU communicates directly with Blynk's cloud server using the Blynk library and authentication token. In the Blynk mobile app, virtual pins are used to display sensor readings in widgets such as gauges, charts, and LEDs. Threshold breaches trigger notifications directly within the app, allowing instant awareness even without a custom application. This integration simplifies setup and ensures a rapid deployment for real-time monitoring. Blynk dashboards provide redundant, real-time access to patient vitals, making the system flexible, scalable, and accessible across different platforms. An integrated alert mechanism is employed to ensure timely notification to parents and caregivers in case of emergencies. When any sensor reading crosses a predefined threshold (e.g., high temperature, low SpO2, irregular heartbeat, loss of power, or crying detected), the Node.js backend processes this data, validates the critical status, and immediately triggers an alert. Using the Infobip API, an SMS is sent to the

registered mobile number associated with the infant's record. Simultaneously, an email is dispatched via Infobip to designated addresses.

5 Methodology

The Smart Infant Incubator Monitoring System integrates IoT technology with cloud services to enable continuous remote monitoring of both the infant's physiological parameters and the surrounding environmental conditions. By utilizing sensors connected to the cloud, the system ensures real-time access to critical data, allowing caregivers and healthcare professionals to track temperature, humidity, oxygen levels, and other vital metrics from anywhere. This integration provides a comprehensive solution for monitoring the infant's health and the incubator's environment, ensuring prompt detection of any abnormalities and enabling quick, informed responses to maintain optimal conditions for neonatal care.

5.1 Data Monitoring

The system includes six different sensors to collect data critical to infant health and incubator environment conditions. The sensors used are:

- *DS18B20 Digital Temperature Sensor:* Records the infant's body temperature with high accuracy to detect any temperature-related irregularities.
- *MAX30100 Pulse Oximeter Sensor:* Measures pulse rate and blood oxygen saturation, providing critical information for detecting respiratory or cardiac distress in newborns.
- *DHT22 Sensor:* Simultaneously monitors the ambient temperature and humidity inside the incubator, ensuring the infant is kept in a thermally regulated and humidified environment.
- *LM393 Sound Sensor:* Captures sudden sounds or infant cries, which may indicate discomfort or require immediate caregiver attention.
- *SW-420 Vibration Sensor:* Detects unusual movements or vibrations within the incubator that could point to equipment issues or physical restlessness.
- *DC Voltage Sensor (25V):* Monitors the incubator's power supply and detects any voltage irregularities that could affect system performance.

These sensors collect real-time data to support early detection of abnormal conditions and to help maintain a stable and safe environment for neonatal care.

5.2 Microcontroller Processing and Data Transmission

Sensor outputs are first processed by an Arduino Uno, which handles data acquisition and signal conditioning. The refined data is then sent via serial communication to the NodeMCU ESP8266, which provides Wi-Fi connectivity. The NodeMCU module serves as the bridge between the sensor network and the internet. It securely transmits the data to a cloud server where it can be stored, analyzed, and accessed remotely.

5.3 Cloud-Based Data Management and Visualization

The system is integrated with a cloud service such as Firebase, ThingSpeak, or AWS IoT, which performs several essential functions:

- *Live Data Streaming:* Real-time visualization of all monitored parameters.
- *Remote Accessibility:* Caregivers and medical staff can view the data via mobile apps or web dashboards from any location.
- *Automated Alerts:* In the event of abnormal readings, the system sends instant notifications through emails, SMS, or push alerts.
- *Data Archiving:* Historical sensor data is stored for later analysis, allowing trends and health patterns.

5.4 Power Backup System

To address the issue of unreliable power supplies in some facilities, a 25V backup battery is incorporated into the design. This backup source ensures continuous system operation during power outages, preserving real-time monitoring and data integrity.

6 Results and Analysis

The system has demonstrated consistent and accurate sensor readings under controlled laboratory and simulation conditions. Each sensor—ranging from physiological (MAX30100) to environmental (DHT11, DS18B20)—provides reliable data without interference or overlap. Data acquisition intervals can be dynamically adjusted for performance tuning and power efficiency. The ESP8266 NodeMCU transmits collected data to a cloud-based MongoDB database with an average latency of less than one second. This ensures near real-time data availability for visualization on both the Flutter and Blynk mobile dashboards, providing caregivers with instant insight into the infant's condition. The system incorporates an automated alert mechanism powered by Infobip for both SMS and email notifications. When a sensor reading crosses a predefined threshold (e.g., low SpO₂, high temperature, or noise/vibration detection), the system validates the anomaly and sends out alerts within 3 to 5 seconds. This rapid notification loop is essential in neonatal care, where response time can be critical to outcomes. Alerts include context-specific data and are delivered with minimal latency, ensuring timely interventions. Traditional neonatal incubator monitoring systems typically rely on manual observation and localized displays that offer limited scalability and lack remote connectivity. In contrast, the proposed smart system provides continuous monitoring, automatic threshold-based alerts, secure cloud synchronization, and multi-platform accessibility. To validate real-time monitoring capabilities, the system was tested using the Blynk IoT

platform. Below are three representative screenshots from the Blynk dashboard in the Fig. 3 and Fig. 4.



Fig. 3. Live visualization of temperature, humidity, and baby body temperature

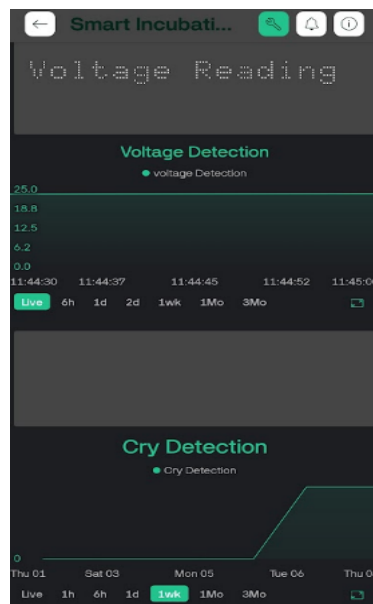
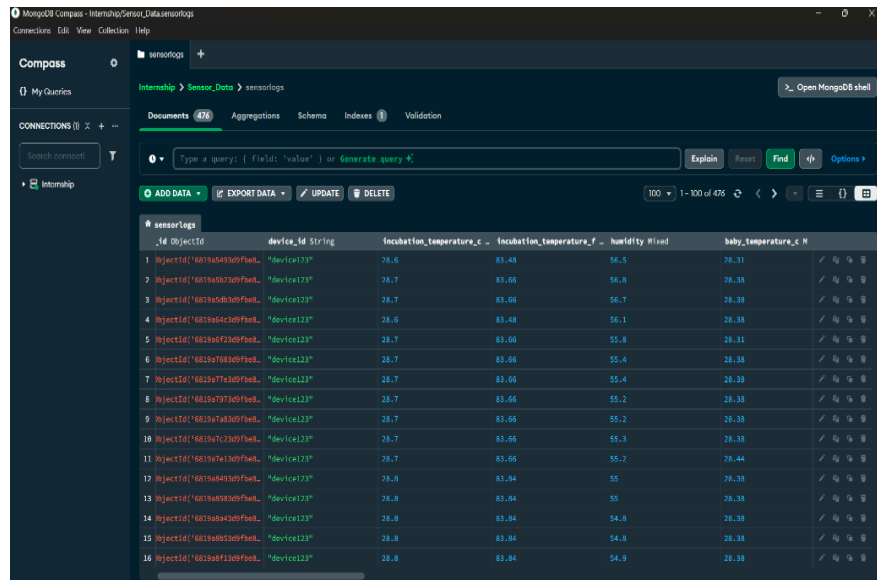


Fig. 4. Alert indicators and system control panel including status flags

The MongoDB Atlas cloud database is used to store real-time and historical data from all sensors. Below is a snapshot of the live database in operation mentioned in the Fig. 5. The database structure supports indexed querying and efficient API access for mobile dashboards and analytics tools. The system enables doctors and nurses to monitor patients remotely via mobile dashboards. This capability is particularly valuable in rural or resource-limited areas where specialized staff may not be readily available on-site. The platform employs budget-friendly hardware and free or open-source software, making it suitable for deployment in both public hospitals and private clinics. It is scalable to manage multiple incubators across. The proposed model serves as an effective demonstration model in academic environments, illustrating key concepts in IoT, embedded systems, cloud computing, and healthcare informatics. It provides a hands-on learning experience for students in engineering, medical, and IT disciplines.



_id	device_id	incubation_temperature_c	incubation_temperature_f	humidity	baby_temperature_c
1	6019a543d9f5ad...	28.6	83.48	56.5	28.31
2	6019a543d9f5ad...	28.7	83.66	56.8	28.38
3	6019a543d9f5ad...	28.7	83.66	56.7	28.38
4	6019a543d9f5ad...	28.6	83.48	56.1	28.38
5	6019a543d9f5ad...	28.7	83.66	55.8	28.31
6	6019a543d9f5ad...	28.7	83.66	55.4	28.38
7	6019a543d9f5ad...	28.7	83.66	55.4	28.38
8	6019a543d9f5ad...	28.7	83.66	55.2	28.38
9	6019a543d9f5ad...	28.7	83.66	55.2	28.38
10	6019a543d9f5ad...	28.7	83.66	55.3	28.38
11	6019a543d9f5ad...	28.7	83.66	55.2	28.44
12	6019a543d9f5ad...	28.8	83.84	55	28.38
13	6019a543d9f5ad...	28.8	83.84	55	28.38
14	6019a543d9f5ad...	28.8	83.84	54.8	28.38
15	6019a543d9f5ad...	28.8	83.84	54.8	28.38
16	6019a543d9f5ad...	28.8	83.84	54.9	28.38

Fig. 5. Time-stamped data entries for SpO2, temperature, vibration, and power readings.

The reliance on Wi-Fi connectivity can introduce downtime in data synchronization if the network is unstable. Sensor drift and aging may also affect measurement accuracy over time, necessitating regular recalibration and maintenance. High-frequency data collection, especially for parameters like heartbeat or oxygen saturation, can strain cloud storage and bandwidth. Techniques such as data compression, edge filtering, and adaptive logging intervals are recommended for long-term efficiency. The handling of sensitive health data imposes a responsibility to enforce strong security protocols. Encryption, secure APIs, and access control mechanisms are in place, but further compliance with regulations such as HIPAA (USA) or GDPR (Europe) is necessary for clinical deployment. Incorporating artificial intelligence algorithms can enable the system to learn from

historical trends, detect anomalies more precisely, and generate predictive alerts. This could allow proactive interventions before critical events occur. A future enhancement may include integrating camera modules for visual monitoring of the infant's posture, facial expressions, or movement patterns. Combined with image processing techniques, this could add another layer of assessment. With modular design and cloud-based backend, the system can be deployed across multiple wards, clinics, and even home care setups. It can be customized for different patient needs and scaled as required.

Conclusion

The Smart Infant Incubator Monitoring System addresses several longstanding challenges in neonatal care by integrating modern technologies such as IoT, cloud computing, and mobile dashboards. It provides a comprehensive, scalable, and cost-effective solution that enables real-time monitoring, rapid alerts, and remote access to critical health data. These capabilities significantly reduce the risk of delayed interventions and improve the likelihood of positive outcomes for neonates. The ability to remotely monitor multiple incubators through a centralized dashboard enhances efficiency, particularly in settings with limited staff. As healthcare continues to evolve with technological innovation, this system offers a robust platform for future advancements. The potential for integration with artificial intelligence, visual monitoring, and smart analytics opens new frontiers in predictive and preventive neonatal care. With continued development, testing, and adherence to clinical standards, this system can become an indispensable tool in the global effort to reduce neonatal mortality and improve infant health outcomes.

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