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IoT-Based Real-Time Monitoring System for Laboratory Hazards

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Abstract: Laboratories are environments that often involve hazardous materials and sensitive equipment, making safety a top priority. Traditional hazard detection methods rely heavily on manual supervision and periodic checks, which can be slow, inefficient, and prone to human error. This project proposes the development of an IoT-based real-time monitoring system aimed at enhancing laboratory safety by continuously detecting and alerting users of potential hazards such as gas leaks, fire, smoke, and abnormal temperature or humidity levels.

The system integrates various environmental sensors (e.g., gas, temperature, humidity, smoke) with a microcontroller (such as ESP32 or NodeMCU) connected to a wireless network. Sensor data is transmitted in real time to an IoT platform where it is visualized and monitored. When critical thresholds are breached, the system instantly sends alerts via mobile applications, emails, or SMS, allowing for timely intervention. Additionally, the system supports remote access and historical data logging to improve preventive maintenance and safety analysis. The proposed solution is low-cost, scalable, and suitable for academic, research, and industrial laboratory environments, significantly improving hazard detection and response times.

Index Terms: ["Internet of Things (IoT)", "Real-Time Monitoring", "Laboratory Safety", "Hazard Detection", "Gas Sensor", "Temperature and Humidity Sensor", "Smoke and Fire Detection", "ESP32 / NodeMCU", "Remote Alert System", "Environmental Monitoring", "Wireless Sensor Network (WSN)", "Data Logging", "Safety Automation", "Smart Laboratory", "Blynk / ThingSpeak / Firebase"]

I. INTRODUCTION

Laboratories are places where important discoveries and innovations happen, but they can also be environments filled with hidden dangers. From toxic gases to flammable chemicals, and sometimes even high heat or electrical hazards, labs have many potential risks that need constant vigilance. Traditionally, keeping these spaces safe has depended heavily on people manually checking equipment and conditions or relying on isolated alarm systems. However, these methods are not always enough to catch problems early, especially when accidents can happen quickly and unexpectedly.

In recent years, the rise of the Internet of Things (IoT) has opened up new possibilities for improving safety in labs. IoT technology allows various sensors and devices to be connected and communicate over the internet, enabling continuous and real-time monitoring of important environmental factors. This means that dangerous situations, like gas leaks or sudden temperature spikes, can be detected immediately and alerts can be sent right away to the people responsible, no matter where they are. This kind of instant notification can make all the difference in preventing accidents and minimizing damage.

Beyond just sending alerts, IoT systems can collect and store large amounts of data over time, providing insights into patterns and trends. For example, if a particular piece of equipment tends to overheat at certain times, that information can help lab managers take preventive action before it turns into a serious hazard. This ongoing data collection also helps labs comply with safety regulations and improve their overall safety protocols, making the environment safer for everyone working there.

Despite these advantages, building an effective and affordable IoT-based hazard monitoring system for laboratories is not without its challenges. Different labs have different safety needs, layouts, and equipment, so any solution must be flexible and easy to adapt. The technology also needs to be user-friendly so that lab staff, who may not be tech experts, can operate and maintain it without difficulty. Finding the right balance between sophistication and simplicity is key.

In this paper, we propose an IoT-based real-time monitoring system designed specifically for laboratory hazard detection. Our system uses a combination of gas sensors, smoke detectors, and temperature and humidity sensors connected to a compact, internet-enabled microcontroller. The sensor data is sent continuously to a cloud platform where it can be accessed remotely through a web or mobile interface. When hazardous conditions are detected, instant alerts are triggered via SMS, email, or app notifications, ensuring that the right people are informed immediately.

We have tested this system in a controlled lab environment and found it to be reliable and effective at early hazard detection. By automating continuous monitoring and notifications, the system reduces reliance on manual checks and human error, ultimately creating a safer and more secure laboratory environment. This work highlights how IoT technologies can play a crucial role in transforming traditional safety practices, making labs smarter and safer places to work.

II. RELATED WORK

A. IoT-Based Fire Detection and Suppression System

Patil and colleagues proposed a fire detection and suppression system utilizing NodeMCU and MQ-3 gas sensors. The system detects fire and gas leaks in real time and automatically activates a submersible pump for fire control, while sending alerts via the Blynk app, showcasing a comprehensive hazard management approach.

B. Multi-Sensor Environmental Monitoring with Fire Suppression

Hussein et al. designed an environmental monitoring system combining gas, temperature, and pH sensors. The platform provides real-time data visualization and integrates an automatic fire suppression mechanism, illustrating how monitoring and response can be tightly coupled.

C. Predictive Hazard Detection Using Machine Learning

Deshvena and Kulkarni explored machine learning integration in IoT systems for laboratory hazard prediction. Their system analyzes sensor data to forecast hazardous events, enabling early alerts that help prevent accidents.

D. Systematic Review of IoT in Clinical and Chemical Laboratories

Munir and co-authors reviewed IoT applications in clinical and chemical labs, highlighting real-time monitoring's role in safety enhancement. Their work emphasizes the need for scalable, flexible IoT frameworks to handle varied sensor data and improve lab safety.

E. Low-Cost Gas Leakage Detection System

Chaudhari and Bhamare developed a low-cost gas leakage detection system using MQ-series sensors and ESP8266 microcontrollers. The system sends instant alerts to smartphones, demonstrating effective hazard detection with minimal setup cost.

F. Multi-Sensor Industrial Safety Monitoring Platform

Rani and colleagues implemented an IoT platform combining gas, temperature, and smoke sensors with cloud-based analytics. Their system focuses on data visualization and remote notifications, facilitating rapid identification and response to industrial hazards.

Wireless Sensor Network for Chemical Lab Safety. Zhang et al. designed a wireless sensor network (WSN)-based safety system for chemical laboratories, focusing on real-time detection of toxic gases and fire hazards. The use of mesh networking improves reliability of data transmission in complex lab setups

G. Smart Laboratory Environment Control System

Kumar and Singh proposed an IoT-based smart environment control system that regulates temperature, humidity, and air quality in laboratories. Their system includes mobile alerts for abnormal conditions, promoting both safety and comfort.

H. Laboratory Reactor Monitoring System Based on IoT

Yang et al. designed an IoT-based monitoring system for laboratory reactors. Utilizing the STM32F407 microcontroller, the system collects temperature and motor parameters, transmitting data to the OneNET cloud platform via WiFi. This setup enables real-time remote monitoring of reactor operations, improving safety and control in laboratory processes.

I. IoT-Based Smart Laboratory System for Enhanced Safety and Efficiency

Oh and Kong developed a smart laboratory system integrating a robotic arm, various sensors, and a mobile application for remote monitoring and control. The system features contactless pipetting, real-time environmental monitoring, and automated adjustments to maintain optimal laboratory conditions. Employing an ESP-WROOM-32 microcontroller, MQ-2 gas sensor, DHT22 temperature and humidity sensor, and a high-torque robotic arm, the system enhances laboratory automation, reduces personnel risks, and improves resource utilization.

III. SYSTEM DESIGN & ARCHITECTURE

Designing a reliable real-time monitoring system for laboratory hazards requires a carefully layered architecture, ensuring that environmental data is accurately captured, processed intelligently, and delivered to users promptly. The proposed IoT-based system is structured into distinct components: the sensor network, local processing unit, communication infrastructure, cloud platform, and user interaction interface. These components work together to provide end-to-end hazard monitoring that is proactive, responsive, and scalable.

A. Sensor Network Design

The first layer of the system architecture revolves around environmental sensing. Laboratories often involve hazardous chemicals, high temperatures, and volatile equipment, making real-time data collection critical. The sensor suite is composed of multiple types of sensors, each selected based on its ability to detect specific threats. For instance, MQ-series gas sensors are deployed to identify the presence of combustible gases such as methane, butane, and carbon monoxide. To detect smoke and fire, flame sensors and smoke detectors are placed in high-risk zones. Additionally, DHT22 or similar sensors measure ambient temperature and humidity to capture overheating or fire-risk indicators, while PIR (Passive Infrared) sensors detect unauthorized movements or accidental human presence in restricted areas.

These sensors continuously operate in the background, collecting raw data without requiring manual intervention. Their strategic placement across the laboratory ensures comprehensive spatial coverage, reducing the likelihood of blind spots or unnoticed hazards.

B. Local Processing Unit

Sensor data is transmitted to a local microcontroller unit, which acts as the processing and control core of the system. Devices such as the ESP32 or Raspberry Pi Pico W are preferred due to their low power consumption, built-in wireless capabilities, and sufficient computational power to handle real-time data processing. The microcontroller performs initial data filtering, removes noise or redundant readings, and compares the values against predetermined safety thresholds.

When a reading exceeds a critical threshold—such as a gas concentration level indicating leakage—the controller triggers immediate responses. These responses may include activating audio-visual alerts, such as buzzers and warning lights, or triggering a preconfigured fail-safe mechanism like powering down non-essential equipment or activating ventilation systems. This localized decision-making allows the system to react quickly, even without relying on cloud connectivity.

C. Communication Infrastructure

The communication layer bridges the local system with remote services and users. Depending on the laboratory's layout and connectivity requirements, the microcontroller communicates via Wi-Fi or a long-range, low-power technology such as LoRaWAN. The system uses efficient protocols such as MQTT (Message Queuing Telemetry Transport) or HTTPS for lightweight, reliable, and secure data transmission.

In environments where internet access may be unreliable, the system is designed to store sensor data locally in onboard memory. Once the connection is restored, the buffered data is synchronized with the cloud server, ensuring no critical information is lost. This feature is particularly important for maintaining data integrity and providing consistent logs for post-incident analysis or audits.

D. Cloud Platform Integration

Once data reaches the cloud, it is stored, analyzed, and visualized using cloud computing services such as Firebase, AWS IoT, or open-source platforms like ThingSpeak. These services provide real-time dashboards that display current sensor readings, system status, and alert logs. Users can access this information through a secure login on their web browser or mobile app, providing visibility and control from virtually anywhere.

Beyond visualization, the cloud platform enables advanced features like trend analysis and anomaly detection. Historical data can be visualized to identify patterns—such as repeated overheating in a specific area or frequent gas leak alerts—offering insight into potential systemic issues. With additional integration, the system can also support predictive analytics powered by AI to forecast risk zones or equipment failures before they occur.

E. Alert Management and User Interaction

A critical aspect of the architecture is the real-time alert system. When a hazard is detected, users are notified immediately via push notifications, SMS, or email alerts. The alert messages include specific information such as the type of hazard, its severity, and the time of detection. This helps safety officers and laboratory managers respond appropriately and quickly, even when they are off-site. The user interface is designed with simplicity and clarity in mind. Dashboards display essential metrics like current gas levels, temperature, humidity, and system uptime, all updated in real-time. The interface also includes control features that allow users to reset alarm.

F. Scalability and Modularity

Scalability is a central design principle of the system. Laboratories may vary widely in size, layout, and hazard types, so the architecture must accommodate new sensors and features without major redesigns. The modular approach allows different laboratories to tailor the system according to their unique requirements. For example, a biochemistry lab may prioritize toxic gas detection, while an electronics lab may need thermal and fire monitoring. Additional sensor nodes can be added with minimal reconfiguration, and firmware updates can be pushed over-the-air (OTA) to ensure the system remains up to date. This flexibility ensures the system remains viable not only in current conditions but also as the laboratory evolves or as new safety requirements emerge. It can also be extended for compatibility with third-party platforms, such as national safety networks or institutional dashboards.

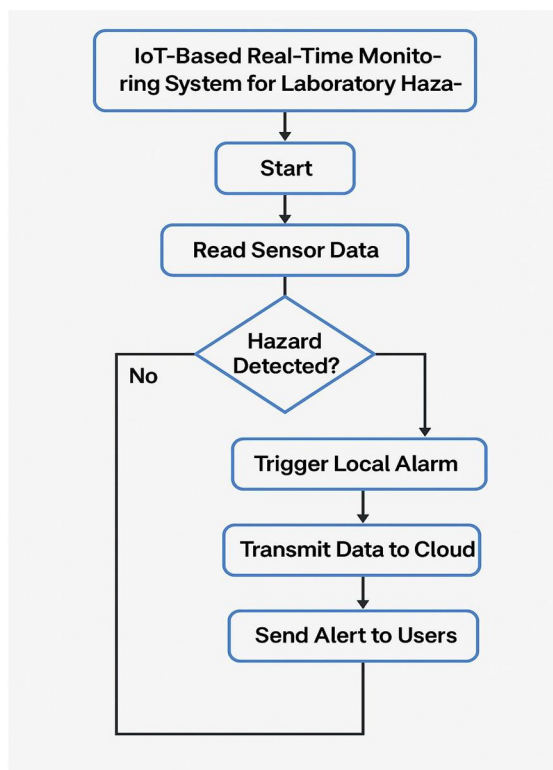


Fig : [1] System Workflow of IoT-Based Real-Time Laboratory Hazard Monitoring

IV. MATHEMATICAL CONCEPTS AND FORMULAS

A. Threshold-Based Decision Model

You can define hazard conditions using inequalities and logical expressions:

If

$$\text{Gas_level} > \text{Max_Gas_Threshold} \text{ OR } \text{Temperature} > \text{Max_Temperature_Threshold}$$

Then

$$\text{Trigger_Alert} = \text{TRUE}$$

This logic forms the core of your hazard detection system.

B. Sensor Fusion Using Weighted Average

If multiple sensors are used to measure the same parameter (e.g., temperature), you can calculate a weighted average as follows:

$$T_{avg} = (w1*T1 + w2*T2 + \dots + wn*Tn) / (w1 + w2 + \dots + wn)$$

Where:

- $T1, T2, \dots, Tn$ = individual sensor readings
- $w1, w2, \dots, wn$ = weights (based on reliability or calibration)

C. Anomaly Detection Using Z-Score

To detect abnormal readings statistically:

$$z = (X - \mu) / \sigma$$

Where:

- X = current sensor reading
- μ = mean of previous readings
- σ = standard deviation of previous readings

If $|z| >$ threshold (e.g., 2 or 3), the value is considered an anomaly.

D. Probability of Hazard from Multiple Sensors

To estimate the likelihood of a hazard being real (based on multiple sensors):

$$P(\text{Hazard}) = 1 - (1 - P(S1)) * (1 - P(S2)) * \dots * (1 - P(Sn))$$

Where $P(Si)$ is the probability of hazard detection from the i -th sensor.

5. LoRaWAN Path Loss Model (for Communication Range)

Used to estimate signal degradation over distance in indoor/outdoor settings:

$$PL(d) = PL0 + 10 * n * \log_{10}(d / d0)$$

Where:

- $PL(d)$ = Path loss at distance d
- $PL0$ = Path loss at reference distance $d0$
- n = Path loss exponent (typically 2–4 depending on the environment)

V. CONCLUSION

Ensuring safety in laboratories is no longer just about routine checks or relying on manual supervision—it demands intelligent, responsive systems that can monitor and react in real time. This paper presented the development of an IoT-based solution aimed at addressing that very need by continuously tracking environmental conditions and responding to potential hazards such as gas leaks, fires, and abnormal temperature levels. The system we designed integrates reliable sensors, a microcontroller for local processing, and cloud-based services for data storage, real-time updates, and user notifications. This layered approach ensures that alerts are generated instantly when danger is detected and allows users to monitor laboratory conditions remotely through a simple web or mobile interface. The result is not just a smart alert system, but a proactive safety network that reduces risks and helps safeguard both people and infrastructure. What makes this approach especially effective is its scalability and flexibility. Whether it's a small academic lab or a larger industrial research facility, the system can be customized with different sensors or communication protocols to suit the specific risks present in each environment. Its modular structure also opens up opportunities for future enhancements, such as integrating AI for predictive analysis or blockchain for secure data records. Overall, this project demonstrates how accessible IoT technologies can be leveraged to create meaningful, real-world impact. By shifting from reactive safety practices to proactive, data-driven monitoring, we move one step closer to safer, smarter laboratory environments. This work serves as a foundation for further research and practical implementation in various sectors where real-time hazard detection is critical.

REFERENCES

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- [2] Hazardous Gas Detection Using Wireless Sensor Networks (2020) Kumar and Sinha developed a distributed network of gas sensors that communicated via ZigBee to detect toxic gas leaks in industrial labs. The paper emphasized sensor calibration and fault tolerance but did not focus on integrating other hazard types or remote monitoring through cloud-based dashboards.



- [3] LoRa-Based Environmental Monitoring System (2022) A study by Nair et al. utilized LoRaWAN for long-range, low-power communication in environmental monitoring systems. Their findings were useful in selecting communication protocols for areas with limited connectivity, such as remote labs. The study supports adopting LoRa as a reliable alternative to Wi-Fi in some laboratory environments.
- [4] Design and Implementation of a Fire and Gas Detection System (2021) Singh and Reddy presented a fire and gas detection system using Arduino and MQTT protocols. Alerts were sent via SMS and email. The research highlighted the importance of threshold tuning and user notification in real time, which is a critical aspect adopted in our system as well.
- [5] Real-Time IoT Framework for Indoor Air Quality Monitoring (2023) Chakraborty et al. developed an IoT-based platform that used air quality sensors and a web dashboard to monitor CO₂, PM2.5, and temperature in indoor workspaces. Their system emphasized user interaction through dashboards, an idea we expanded by integrating it with hazard alerts in laboratories.
- [6] A Smart Safety System for Industrial Workplaces (2022) Gupta et al. proposed a wearable sensor system for industrial environments that tracked gas exposure and worker movement. Though not focused on laboratories, their framework provided insights into combining safety protocols and automation, which our system incorporates using motion sensors and alarms.
- [7] IoT-Based Fire Detection and Suppression System (2020) Deshmukh and Patel designed a fire detection system integrated with IoT and automated extinguishing units. Though our system does not include suppression, the architecture for early detection and cloud alerts serves as a strong foundation for our design.
- [8] Integration of Cloud and Edge Computing in IoT Safety Systems (2023) Recent work by Liu et al. explored a hybrid cloud-edge model to reduce latency in safety-critical IoT systems. Their results supported the idea of processing sensor data locally while still enabling cloud-based visualization, aligning well with the hybrid design of our monitoring system.



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