



Cloud-Based industrial sensor monitoring and Predictive Fault Detection system

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Abstract- industrial monitoring systems have many problems when it comes to the lack of real-time data to access, limited remote control, and fault detection performance. Many industries rely on manual or stand-alone (isolated) monitor systems to monitor industrial machines and equipment. Because of this, operators cannot continuously track critical parameters, such as temperature, voltage, and equipment status. Consequently, this can cause delays in identifying abnormal conditions and result in a defective machine, lack of production time because of machine downtime, and higher maintenance costs. Additionally, traditional industrial monitoring systems do not use centralized (cloud) storage for data, thus there is no ability to integrated historical sensor data to continuously improve operational performance. Therefore, with out intelligent detection/analytics there are no means for an industry to predict potential failure(s) before an event occurs and therefore only use reactive maintenance/efforts rather than proactive. Finally, without an integrated dashboard to monitor/control multiple machines from one place, it's difficult to efficiently manage multiple machines by the operators. Thus, there is an urgent demand for new technological solutions that integrate iot, cloud-based data management, and ai technologies to enable real-time monitoring, remote control of equipment, predictive fault-detection, and increased productivity in industrial environments.

Keywords- Industrial Monitoring System, Internet of Things (IoT), Cloud Computing, Real-Time Monitoring, Predictive Maintenance, Artificial Intelligence (AI), Sensor Data Analysis, Fault Detection, Web-Based Dashboard, Remote Equipment Control, Industrial Automation, Smart Manufacturing.

I. INTRODUCTION

Continuous monitoring of machinery and equipment is crucial in industrial environments to maximize productivity, ensure safety, and enhance operational efficiency. Key parameters such as temperature, voltage levels, and overall equipment status must be tracked in real time to maintain operations within safe and optimal limits. Traditional monitoring methods, which are largely

manual or rely on standalone systems, often lack real-time data access and centralized control, limiting the ability of operators to respond promptly to abnormal conditions or equipment malfunctions. These limitations frequently lead to unplanned machine downtime, production delays, increased maintenance costs, and overall inefficiencies in industrial operations. With the rapid advancement of industrial automation and the rise of smart manufacturing, there is a growing need for sophisticated monitoring



solutions. Modern systems provide continuous, real-time insights and support intelligent decision-making, enabling proactive maintenance, enhanced safety, and more efficient management of industrial processes.

Modern industries are increasingly leveraging advanced technologies to tackle challenges such as machine failures and operational inefficiencies, with tools like the Internet of Things (IoT), cloud computing, and artificial intelligence (AI) playing a central role. IoT-enabled sensor networks allow for continuous collection of real-time machine data from multiple sources, which is then transmitted to cloud-based platforms for storage and analysis. These cloud systems enable industries to remotely monitor equipment, access historical sensor data, and visualize critical operational metrics through intuitive dashboard interfaces. Such capabilities enhance system transparency and provide operators with the ability to manage multiple machines from a centralized platform efficiently. Moreover, integrating intelligent analytics into these systems supports a shift from reactive maintenance practices to proactive maintenance strategies. By predicting potential faults before they lead to equipment failure, industries can optimize performance, reduce downtime, and improve overall operational reliability, ultimately driving productivity and cost-efficiency.

This project presents an advanced industrial monitoring system leveraging cloud technology and IoT connectivity to enhance machine management and operational efficiency. IoT sensors continuously capture real-time data from industrial equipment, which is then transmitted to the cloud for storage, analysis, and visualization. A user-friendly web interface allows operators of all levels to monitor machine performance, observe trends over time, and make informed decisions based on predictive analytics powered by AI. Beyond passive monitoring, the system enables operators to actively control machinery through the dashboard, including switching devices on or off, adjusting temperature settings, and setting high-voltage limits. By centralizing these functionalities, the platform provides a comprehensive overview of multiple machines, helping operators detect anomalies, optimize processes, and maintain safety standards. The integration of real-time data, predictive insights, and manual control ensures more precise and efficient management of industrial operations, reducing downtime and improving overall productivity across complex industrial environments.

The system includes an AI-based predictive model which will monitor patterns of incoming sensor data to detect possible equipment faults through the examination of anomalous trends and patterns in sensor

data readings. The system is designed to generate alerts for significant deviations or anomalies in sensor data to provide early warning signals of possible failures, allowing for timely maintenance. Predictive maintenance will greatly reduce equipment downtime; enhance equipment dependability; and, in many cases, improve production efficiency overall. Together with IoT connectivity, cloud-based data management, and AI, these capabilities provide an intelligent and scalable solution for the modern industrial sector's monitoring and predictive maintenance needs.

II. METHODOLOGY

This project focuses on developing a cloud-based predictive maintenance and industrial monitoring system that integrates IoT sensors, cloud storage, a web-based dashboard, and an AI-driven predictive model. The overall methodology involves collecting real-time sensor data, transmitting it securely to the cloud, visualizing the data through an interactive dashboard, controlling machine operations, and intelligently predicting potential faults. Industrial sensors are connected to the monitoring hardware, continuously tracking critical parameters such as temperature, voltage, and operational status. These sensors relay real-time measurements to the IoT communication module within their respective monitoring units. Once captured, the sensor data is transmitted over the Internet to the cloud platform, ThingSpeak, for secure storage and global accessibility. Leveraging this cloud-based infrastructure, operators and administrators can remotely monitor machine performance, detect anomalies early, and analyze historical data to plan predictive maintenance. This system enhances operational efficiency, reduces unexpected downtime, and supports informed decision-making in industrial environments.

Before being utilized in data analysis or visualizations, all incoming sensor readings undergo rigorous validation and pre-processing. This step ensures that the data received from each sensor is accurate, consistent, and free from transmission errors, providing a reliable foundation for further analysis. Pre-processing techniques, including normalization, filtering, and noise reduction, are applied to eliminate any distortions or invalid readings caused by environmental conditions or sensor malfunctions. Once validated, the cleaned data is stored in a continuously updated cloud database, allowing real-time access and secure long-term storage. Utilizing a cloud-based system enables the preservation of historical performance records, which are essential for detecting long-term patterns, identifying anomalies, and supporting data-driven operational decisions. By combining careful data validation with advanced pre-processing and cloud storage, the system ensures high-quality, reliable



information, enhancing overall decision-making efficiency, predictive insights, and the robustness of analytics outcomes in dynamic environments.

The web interface dashboard is designed to offer a highly user-friendly platform for the monitoring and management of industrial equipment. By integrating real-time sensor data from the cloud, it provides instantaneous graphical visualizations, precise numerical readouts, and clear indicators of machine status, enabling operators to make informed decisions efficiently. The dashboard allows direct manual control over equipment, including switching devices ON or OFF, adjusting operating temperatures, and defining safe voltage ranges with minimum and maximum thresholds. This centralized control system consolidates the management of multiple machines into a single interface, significantly improving operational efficiency. Operators can oversee and adjust equipment remotely, reducing the necessity for continuous on-site supervision while maintaining accurate records of operational parameters. By combining real-time monitoring, intuitive control features, and centralized management, the dashboard enhances convenience, safety, and productivity, ensuring smooth and reliable industrial operations without compromising performance or safety standards.

approach to predictive maintenance offers significant benefits, including reduced downtime, lower maintenance expenses, improved machine reliability, and extended equipment lifespan. The methodology integrates IoT connectivity for seamless data acquisition, cloud computing for scalable processing and storage, and advanced AI analytics for accurate pattern recognition and decision-making. Together, these technologies form a robust, intelligent industrial monitoring framework capable of supporting the operational demands of modern factories. By anticipating issues rather than reacting to them, industries can optimize performance, minimize losses, and ensure continuous, efficient production.

The proposed cloud-based industrial remote monitoring system leverages advanced mathematical operations to analyze real-time sensor data, assess machine health, and enable predictive fault detection. Core calculations within the system include temperature monitoring, voltage threshold evaluation, and fault prediction through statistical analysis of the collected data. Sensor inputs are continuously processed and compared against predefined operational limits, allowing the system to detect deviations that may indicate early signs of equipment malfunction. By systematically evaluating these parameters, the system can identify potential abnormal conditions before they escalate into critical failures. This proactive approach ensures timely alerts are generated, providing maintenance personnel with actionable insights to prevent unplanned downtime. The integration of mathematical models with cloud-based data processing enables efficient monitoring of multiple machines simultaneously, enhancing operational reliability and safety. Overall, these calculations form the backbone of predictive maintenance, improving equipment longevity and minimizing industrial operational risks.

The first critical component in this system is voltage monitoring, which ensures that all industrial machines operate within a safe and stable electrical range. During installation, the operator sets the maximum and minimum allowable voltage limits according to the machine's specifications. The system then uses a dedicated voltage sensor to continuously measure the incoming voltage in real time. This sensor constantly compares the current voltage level against the predefined upper and lower thresholds stored in the installation configuration. If the measured voltage exceeds the maximum limit or drops below the minimum limit, the system immediately generates a warning, which is displayed on the operator's dashboard. This alert allows the operator to take corrective action before the machine experiences damage or operational inefficiency. By tracking these deviations, the system not only protects equipment from electrical stress but also provides essential data for maintenance planning

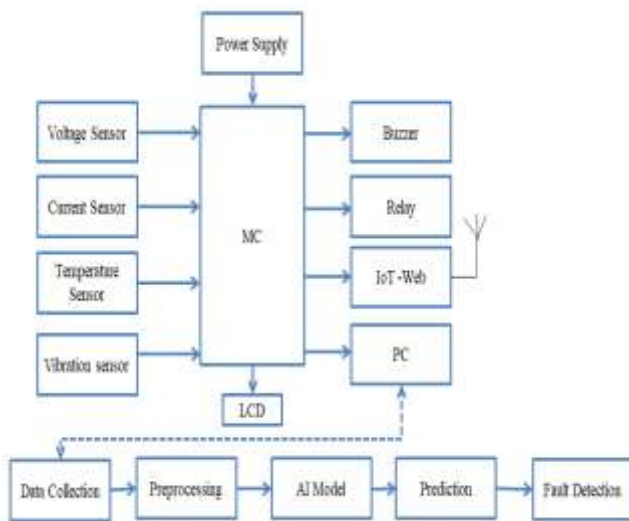


Figure 1 : Proposed Block diagram

To enhance equipment monitoring and fault detection, the described systems incorporate an AI-driven predictive model. Leveraging machine learning, this model analyzes both historical and real-time sensor data to identify deviations from normal operational patterns. By recognizing early signs of potential failures or anomalies, the system can generate timely alerts, enabling operators to take corrective actions before minor issues escalate into costly breakdowns. This proactive



and performance analysis, ensuring long-term operational safety and reliability.

The formula for the amount by which measured voltage differs from the reference voltage (i.e., the voltage deviation) is

$$\text{Voltage Deviation} = \text{Measured Voltage} - \text{Reference Voltage.}$$

If the amount by which measured voltage differs from reference voltage exceeds the specified tolerances, then the system will identify that condition as unsafe. By doing this, we protect the equipment from electrical damage due to overloading and also provide for consistent operation of the system.

Another important calculation used to evaluate machine performance is temperature monitoring. Machines that operate in the industrial environment generate heat during normal operation, and high operating temperatures can damage components or create less-than-optimal operating conditions. For this reason, the system calculates the amount that the sensor current temperature differs from the established potential for safe operation based on temperature. The amount that current temperature differs from safe temperature can be calculated as follows:

$$\text{Temperature Difference} = \text{Current Temperature} - \text{Safe Temperature}$$

If the computed value exceeds a certain point then it shows that overheating has occurred and an auto alert will occur. This gives the operator a chance to initiate the necessary measures (turn machine off/adjust conditions etc.) before any damage occurs. In relation to predictive fault detection, the system calculates historical sensor data to obtain statistics, including average and deviation, to detect any inconsistencies in machine behaviour. To calculate the average sensor value the following calculation is used:

$$\text{Average Value} = (\text{Sum of Sensor Values}) / (\text{Total Number of Values})$$

Also, the system can determine deviation by looking at how far the actual sensor value strays from what would be considered normal operating conditions:

$$\text{Deviation} = \text{Current Sensor Value} - \text{Average Sensor Value}$$

When the calculated deviation exceeds a set threshold then the AI will classify it as a potential fault condition. These calculations will assist in allowing the predictive model to not only detect any abnormal behaviour of the machine early enough but also generate alerts for preventative maintenance. The predictive models will assist in providing greater reliability, reducing unexpected failures and assisting with intelligent monitoring of the Industrial environment.

1. The Voltage Deviation Formula

In order for machines to function correctly, voltage monitoring is essential to keep the equipment running within the allowable voltage range set by the manufacturer. The difference between the voltage in question and the specified reference voltage is then evaluated mathematically.

$$V_d = V_m - V_r$$

2. The Temperature Deviation Formula

Industrial companies can avoid costly downtime due to excessive heat by monitoring the operational temperature of their equipment. To achieve this objective Temperature Deviations can be evaluated by calculation.

$$T_d = T_c - T_s$$

3. The Average Sensor Value Formula

Average Sensor Value Calculation allows the user to determine the average of multiple sensors over a period of time for the purpose of evaluating what "normal" machine operation should be based on a longer time frame of data.

$$X = \frac{\sum X_i}{N}$$

4. The Deviation Detection Formula

By taking the difference between the current value of the measurement being taken vs the previously calculated average value, you will be able to get the deviation amount from the established average value.

$$D = X_c - X^-$$



III. RESULTS AND DISCUSSION

The implementation of the proposed cloud-enabled industrial monitoring system has led to a marked improvement in real-time equipment supervision and operational efficiency. This advanced system integrates IoT sensors, cloud-based communication, and a web-based dashboard to monitor critical industrial parameters such as temperature, voltage, and machine operational status. Data from all connected sensors is continuously transmitted to a cloud platform via ThingSpeak, enabling real-time visualization and secure storage of equipment information. The interactive web-based dashboard allows operators to access live sensor readings, system status, and historical data trends through intuitive graphical representations. By maintaining uninterrupted communication between the sensors and the cloud, the system ensures reliable monitoring, seamless data collection, and timely reporting. Experimental results highlight the effectiveness of this integrated approach, demonstrating its ability to enhance operational oversight, facilitate predictive maintenance, and improve decision-making processes within industrial environments, ultimately supporting safer and more efficient equipment management.

A key advantage of this system is its capability to enable operators to control large industrial machines remotely through an online dashboard. From a single interface, operators can manually power machines on or off, adjust temperatures, and configure voltage settings, ensuring equipment is not subjected to unsafe conditions or operational errors. This centralized control allows supervision of multiple machines from one location, eliminating the need for constant on-site presence and reducing the reliance on direct hands-on monitoring. Prior to full implementation, the system's functionality was rigorously tested, demonstrating immediate responsiveness: machines reflected status changes as soon as commands were issued via the dashboard. Additionally, the system incorporates preconfigured voltage and temperature thresholds, acting as a safeguard to prevent operations beyond manufacturers' recommended limits. When sensor readings approach or exceed these thresholds, alerts are generated automatically, enhancing safety and preventing potential damage while maintaining efficient and reliable industrial operations.

In addition, the system was able to store and analyse data in the cloud. All of the information collected by sensors is uploaded to the ThingSpeak cloud-based platform for long-term storage and trend analysis. Operators will be able to see a visual representation of historical performance data using graphs, which allows

them to identify patterns in machine operation and to detect small deviations in outgoing data. This visual display of data helps operators to make better decisions and allows them to schedule maintenance based upon more accurate data. The integration of the cloud into the overall system provides remote access to sensor information at almost any location, creating more options for monitoring the performance of industrial equipment and improving the overall monitoring capabilities of industrial operations.

The predictive capability of the AI-based system plays a crucial role in maintaining the stability and efficiency of the overall operational environment. By continuously analyzing incoming sensor data against historical performance records and current trend patterns, the system can detect anomalies or deviations in real time. When such inconsistencies or potential exceptions are identified, the system generates early warning alerts, enabling operators to take proactive measures and prevent machine failures before they occur. This predictive approach has been shown to significantly reduce unexpected downtime, lower maintenance expenses, and extend the operational lifespan of industrial equipment. By leveraging advanced technologies such as the Internet of Things (IoT), cloud computing, and artificial intelligence (AI), the system provides a comprehensive framework for intelligent monitoring and predictive maintenance. Consequently, it represents an effective and modern solution for industrial environments seeking enhanced reliability, operational efficiency, and long-term asset optimization.

IV. CONCLUSION

Leveraging IoT (Internet of Things), cloud computing, and AI (Artificial Intelligence) technologies, this intelligent cloud-based industrial monitoring system enables more efficient management of industrial chemicals and machinery. The system provides real-time monitoring of critical machine parameters, including temperature, voltage levels, and operational status, accessible via a web-based dashboard. Sensor data is continuously collected, transmitted to the cloud platform through ThingSpeak, and stored securely, allowing operators to access both live and historical performance data from any location. This visibility empowers operators to manage multiple machines efficiently and respond promptly to abnormal conditions.

Manual control features add operational flexibility and safety, allowing operators to turn equipment on or off, adjust temperatures, and set safe voltage limits. In parallel, the system's AI predictive model analyzes sensor data trends to detect potential faults and identify early warning signs of abnormal operation. This predictive maintenance



capability helps prevent unplanned machine failures, reduce downtime, and minimize associated costs.

By combining IoT connectivity, cloud-based data management, and intelligent analytics, the system offers a comprehensive industrial monitoring solution. It enhances overall reliability, productivity, and safety, while supporting long-term performance planning and maintenance strategies, making it an essential tool for modern smart industrial environments.

V. FUTURE WORK

This cloud-based industrial monitoring solution effectively addresses challenges in real-time monitoring, remote operation, and predictive failure detection. However, several enhancements could further improve its performance, scalability, and intelligence. One potential enhancement is the integration of additional industrial sensors to measure parameters such as humidity, vibration, current draw, pressure, and energy consumption. Monitoring multiple parameters provides a comprehensive view of machine health and enables more precise fault detection.

Another improvement is the development of a mobile application that complements the web-based dashboard. Mobile access allows operators and maintenance engineers to receive immediate notifications, view equipment status, and control machinery directly from their smartphones or tablets. This enhances system accessibility and enables faster response times to abnormal conditions, especially in large industrial facilities where personnel may not be near monitoring stations.

Advancing the predictive maintenance AI model is also a key area for enhancement. Employing more sophisticated machine learning and deep learning algorithms on large volumes of sensor data can uncover complex failure patterns. As the AI model is trained on larger and more diverse datasets, its predictive accuracy improves, making it adaptable across different machinery types and industrial environments.

Moreover, as an extension of the core control and data management (CDM) system to other large-scale industrial networks, edge computing can facilitate further support for industrial applications, thus minimizing cloud latency when processing data through the Cloud and the corresponding time frame required for a particular function or activity, due to having a more timely response to critical industrial operations occurring in real-time. In addition, the establishment of adequate

security protocols, including, but not limited to, encrypted communication channels, appropriate authentication mechanisms and/or the use of secure Internet-of-Things (IoT) gateways, allows for improved security for sensitive industrial data against potential cyber threats and also provides for increased overall robustness, scalability and applicability of the system as it relates to implementing Industrial Internet of Things (IIoT) technologies in smart-factory/Industry 4.0 settings.

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