

# IOT Enabled Railway Track Crack Detection Using Acoustic Waves

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**Abstract—** Keeping railway tracks in good condition is essential for the safe and smooth operation of trains. Cracks that remain undetected can cause major accidents, delays in service, and heavy financial losses. This project proposes an IoT-based railway track crack detection system that uses acoustic wave propagation to identify faults at an early stage with better accuracy. In this system, sensor nodes are installed along the railway track. Each node includes a piezoelectric sensor that sends acoustic waves through the rail and a microphone sensor that receives those waves after they travel through the structure. When there is a crack or any structural defect, the received signal shows noticeable changes compared to normal conditions. These signal variations are captured, processed, and transmitted by a microcontroller to a cloud platform. The data is then uploaded to the Blink IoT cloud, where it can be monitored, stored, and analyzed in real time. A user interface is developed to classify the condition of the track as minor, medium, or major cracks, helping maintenance staff understand the severity quickly and take action. An RFID module is also added to detect train arrival and departure, improving coordination and system awareness. Overall, the proposed system offers a cost-effective, non-destructive, and continuous monitoring solution that strengthens railway safety through early crack detection and remote access to information.

**Keywords—** Railway Track Monitoring, Crack Detection, Acoustic Waves, Piezoelectric Sensor, IoT, Blink IoT Cloud, RFID, Structural Health Monitoring.

## I. INTRODUCTION

Railway transportation is a widely used and affordable means of mass transit across the world. Maintaining the condition of railway tracks is critical for safe operation, as even small cracks can develop into major defects over time, which may result in derailments, service interruptions, and economic losses [1], [2].

However, the safety of railway transportation systems continues to be limited by shortcomings in current track inspection and monitoring practices. Most conventional methods depend largely on manual checks or the use of dedicated inspection vehicles. These approaches take considerable time, require significant manpower, and do not support continuous monitoring. As a result, early-stage cracks—particularly in remote or hard-to-reach areas—often go unnoticed, which raises the chances of accidents and sudden service interruptions.

Traditional railway track inspection methods involve human effort and fixed inspection schedules, which can delay the detection of faults and increase operational expenses. In addition, environmental conditions, heavy axle loads, and aging railway infrastructure speed up crack development and track deterioration. Without real-time monitoring systems, concerns about railway safety, maintenance effectiveness, and the long-term durability of infrastructure become more serious [3], [4].

To address these issues, crack detection techniques based on acoustic waves have attracted growing interest as a dependable and non-destructive method for monitoring railway tracks. In these systems, acoustic waves travel through the rail structure, and any crack or defect changes signal properties such as attenuation, time delay, and amplitude. This approach makes it possible to detect faults at an early stage without damaging the track and enables continuous monitoring of structural health [5], [6].

At the same time, incorporating Internet of Things (IoT) technology into railway systems provides a practical solution for real-time data collection, remote observation, and smarter decision-making. IoT-based

setups allow sensor data to be sent to cloud platforms for storage, processing, and visualization. This reduces the need for manual involvement and improves the speed and efficiency of maintenance actions. Cloud connectivity also increases the ease of access and scalability of monitoring solutions across extensive railway networks [7], [8].

Building on these developments, this project presents an IoT-enabled railway track crack detection system that uses acoustic wave propagation. The system utilizes a piezoelectric sensor to generate acoustic waves and a microphone sensor to receive them, both installed on the railway track. The obtained signals are processed through a microcontroller and transmitted to the Blink IoT cloud platform for real-time monitoring and evaluation. A user interface categorizes detected cracks as minor, medium, or major, and an RFID module is included to recognize train arrival and departure.

The proposed system offers a cost-effective, reliable, and scalable solution for enhancing railway safety through continuous and remote track condition monitoring [9], [10].

## II. LITERATURE SURVEY

### 2.1 Acoustic Wave-Based Railway Crack Detection Systems

The application of acoustic wave propagation techniques for railway track crack detection has gained significant attention due to their non-destructive nature and high sensitivity to structural defects. Acoustic wave-based systems utilize piezoelectric sensors to transmit mechanical waves through rail structures, where the presence of cracks alters wave propagation characteristics such as attenuation, delay, and signal distortion. These variations enable early detection of surface-level and internal defects without causing damage to the track. Several studies have demonstrated the effectiveness of acoustic sensing in identifying cracks under both laboratory and field conditions, making it a promising approach for railway safety applications.

### 2.2 Ultrasonic and Guided Wave Techniques for Rail Defect Detection

Ultrasonic and guided wave methods have been extensively explored for detecting defects in railway tracks, particularly for identifying internal cracks.

These techniques rely on high-frequency waves that propagate over long distances along the rail, allowing inspection of large track sections from a single sensing point. Simulation-based and experimental studies have shown that guided waves are effective in detecting subsurface defects; however, the complexity of signal processing and high system cost limit their widespread deployment in real-time monitoring systems.

### 2.3 Acoustic Emission and Signal Processing Methods

Acoustic emission-based crack detection techniques focus on capturing transient elastic waves generated due to crack initiation and propagation under stress. Signal processing methods such as wavelet transform, entropy analysis, and frequency-domain analysis are employed to enhance crack detection accuracy. While these techniques improve sensitivity to early-stage defects, they often require sophisticated hardware and offline data analysis, restricting their applicability in continuous, real-time railway monitoring environments.

### 2.4 IoT-Based Railway Track Monitoring Systems

The integration of Internet of Things (IoT) technology into railway monitoring systems has enabled real-time data acquisition, cloud-based storage, and remote visualization of track conditions. IoT-enabled systems allow sensor data to be transmitted to centralized cloud platforms, reducing the need for manual inspections and enabling predictive maintenance strategies. However, many existing IoT-based solutions focus primarily on vibration or visual inspection methods and lack effective acoustic wave-based crack classification and severity assessment.

### 2.5 Research Gap and Motivation

From the literature survey, it is evident that while acoustic and guided wave techniques offer reliable crack detection capabilities, most existing systems lack real-time cloud connectivity, user-friendly visualization, and practical deployment features such as train movement awareness. Additionally, limited work has been carried out on integrating acoustic wave-based crack detection with IoT platforms for continuous monitoring and crack severity classification. These challenges highlight the need for an IoT-enabled railway track crack detection system that brings together acoustic sensing, cloud-based

monitoring, and clear data visualization to improve overall railway safety.

### III. EXISTING SYSTEM

Conventional railway track inspection systems primarily rely on manual visual inspections and periodic maintenance checks conducted by trained personnel or specialized inspection vehicles. In these systems, track conditions are assessed at fixed intervals, requiring human intervention and physical access to the railway infrastructure. Such inspection methods are time-consuming, labor-intensive, and often incapable of detecting early-stage cracks, making the process inefficient and unreliable.

Most existing railway monitoring techniques depend on offline analysis and do not support continuous or real-time crack detection. The inspection frequency is limited, particularly in remote and off-city regions, increasing the likelihood of undetected defects. Environmental factors, heavy axle loads, and aging railway infrastructure further accelerate crack propagation, raising serious safety concerns and maintenance costs.

Furthermore, traditional railway inspection systems offer limited automation and monitoring capabilities. The absence of real-time data transmission, cloud connectivity, and remote access restricts effective decision-making. Crack severity classification and immediate alerts are often unavailable, leading to delayed maintenance actions. These limitations highlight the need for an automated, real-time, and IoT-enabled railway track monitoring system to enhance operational safety and efficiency.

### IV. PROPOSED SYSTEM

The proposed system enables real-time railway track crack detection using acoustic wave propagation techniques. Acoustic signals are generated using a piezoelectric sensor mounted on the railway track and transmitted through the rail structure, where they are received by a microphone sensor. The captured data is processed using a microcontroller and transmitted to the Blink IoT cloud platform for storage and analysis. The system monitors variations in acoustic signal characteristics to detect and classify cracks as minor, medium, or major. An RFID module is incorporated to identify train arrival and departure, while a user

interface provides real-time crack status and monitoring information. This approach ensures reliable, automated, and cost-effective railway track monitoring with improved safety and reduced dependence on manual inspections.

#### 4.1 Data Flow / Working Process

The data flow of the proposed system begins with the piezoelectric transducer generating acoustic waves that propagate through the railway track. These waves are received by the KY-037 microphone sensor and converted into electrical signals. The microcontroller processes the sensor data and transfers it to the Blink IoT cloud through the ESP8266 Wi-Fi module. The cloud platform stores and analyzes the data, and the processed results are displayed on the user interface, indicating crack presence and severity. Additionally, alert signals are generated locally through a buzzer to ensure immediate notification

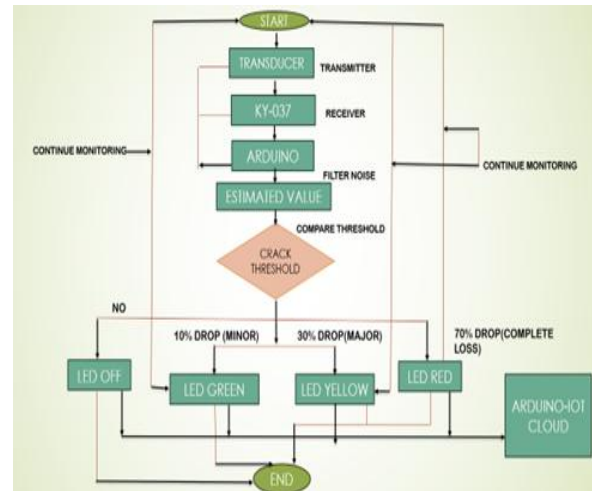


Fig.1.Data Flow

#### 4.2 System Architecture

The system architecture of the proposed model brings together acoustic wave sensing, embedded control, cloud connectivity, and monitoring units to support effective railway track crack detection. Acoustic signals are produced by a piezoelectric sensor and sent through the railway track, while a microphone sensor captures the signals after they propagate. The received data is then handled by a microcontroller and forwarded to the Blink IoT cloud platform for real-time monitoring. Changes in the signal are examined to determine the presence and seriousness of cracks, and the results are shown on the user interface.

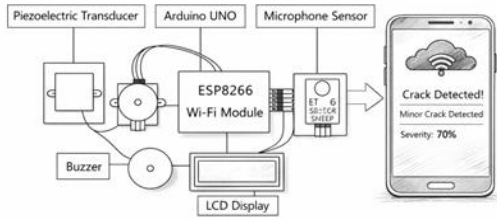


Fig.2.System architecture.

V.SYSTEM IMPLEMENTATION

5.1 SYSTEM PERFORMANCE EVALUATION

The system performance evaluation analyzes the practical behavior of the proposed IoT-enabled railway track crack detection system under different operating conditions. The evaluation focuses on the response of acoustic wave propagation, signal variation due to structural defects, and the reliability of crack classification. Experimental observations were recorded and represented graphically to demonstrate crack detection efficiency and severity distribution. The analysis confirms that the proposed system effectively identifies minor, medium, and major cracks based on variations in received acoustic signals. The graphical results highlight the stability, sensitivity, and real-time monitoring capability of the embedded sensing architecture.

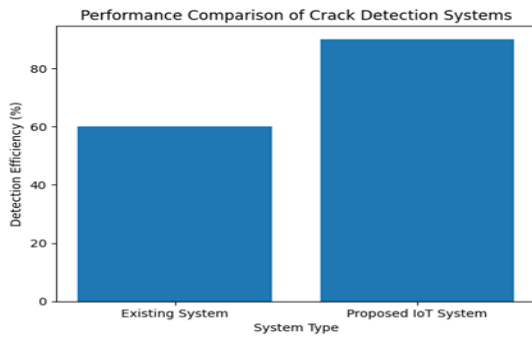


Fig.11.Hardware Implementation

5.2 Software Implementation

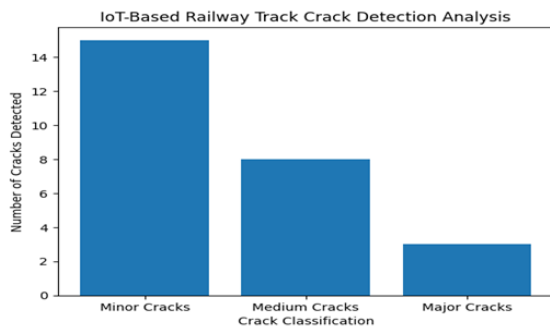


Fig.11.performance analysis

The software implementation of the proposed system is built around an IoT-enabled monitoring platform that connects with the microcontroller through the ESP8266 Wi-Fi module. The firmware is written in Embedded C/C++ using the Arduino IDE and handles the collection of acoustic sensor data from the piezoelectric transducer and the KY-037 microphone sensor. It processes the incoming signals and sends the data to the Blink IoT cloud platform over a wireless network for real-time monitoring and analysis.

The microcontroller interprets the sensor data and examines changes in acoustic signal features to determine the presence and severity of cracks. Based on these results, the software prepares suitable status updates and alert messages, which are forwarded to the cloud and shown on the user interface. At the same time, the software oversees system operation and activates the buzzer to provide immediate local alerts when needed. This software setup supports dependable data transfer, real-time observation, and easy operation of the IoT-based railway track crack detection system.

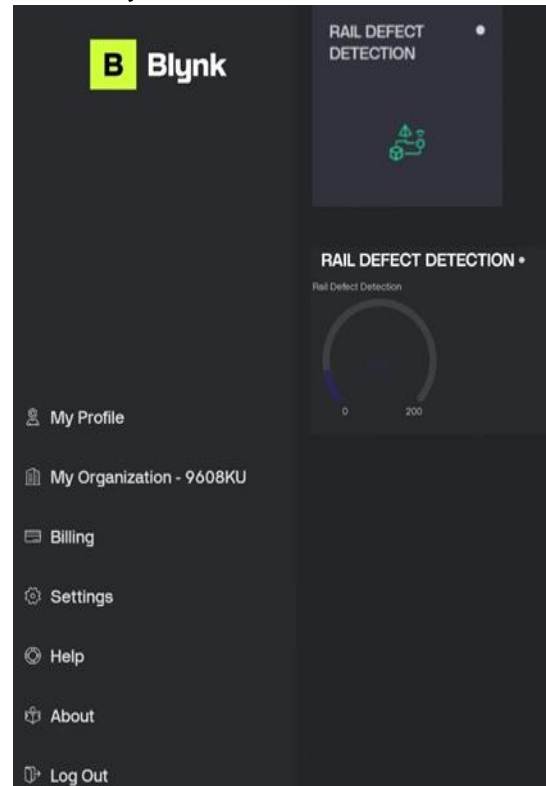


Fig.12.Software implementation

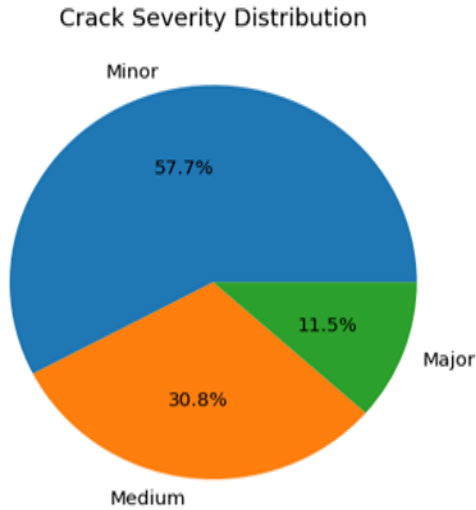


Fig.13.Crack Survey

## VII. CONCLUSION AND FUTURE SCOPE

### 7.1 Conclusion

This project clearly demonstrates an IoT-enabled railway track crack detection system that makes use of acoustic wave propagation techniques. By bringing together piezoelectric sensing, embedded control, cloud connectivity, and alert features, the system is able to detect structural defects in railway tracks at an early stage. It overcomes the drawbacks of manual inspection by offering continuous, automated, and real-time monitoring of track conditions.

### 7.2 RESULT

The proposed IoT-enabled railway track crack detection system successfully identified structural defects using acoustic wave propagation. The system accurately classified cracks into minor, medium, and major categories. Real-time cloud monitoring and alert mechanisms improved detection efficiency, reduced manual inspection dependency, and ensured reliable, continuous railway track safety monitoring.

### 7.2 Future Scope

The proposed system can be improved further by installing multiple sensor nodes along longer stretches of railway tracks, allowing monitoring over a wider area. Using more advanced signal processing methods and machine learning algorithms can help increase the accuracy of crack detection and provide better classification of defect severity.

Connecting the system with additional IoT platforms and predictive analytics tools can support real-time fault prediction, performance evaluation, and automated maintenance planning. In addition, adding GPS-based location tracking and adopting energy-efficient power management methods can make the system more practical and better suited for real-world railway infrastructure and smart transportation applications.

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