

Gps Enabled Crack Detection Robot Using Esp32 Cam and Arduino

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Abstract

Ensuring the safety and durability of infrastructure is essential, particularly for roads, bridges, and industrial structures. This research introduces a Crack Detection Robot that leverages Arduino Uno, ESP32, and ultrasonic sensors to identify surface cracks efficiently. The system operates remotely via a Bluetooth module, navigates using a motor driver, and utilizes I2C communication for seamless data exchange between components. Crack detection is achieved through ultrasonic sensing, while the ESP32 manages wireless control. The software, developed in Arduino IDE, is optimized for precise motor control and real-time sensor data processing. This autonomous approach to crack detection reduces the reliance on manual inspections, making the process cost-effective, accurate, and time-efficient. To overcome challenges such as sensor calibration, inter-device communication, and power consumption, the system is fine-tuned for enhanced performance. Future enhancements may incorporate AI-based crack classification, Wi-Fi-enabled live monitoring, and solar-powered operation, further improving efficiency and automation. This research contributes to modern infrastructure monitoring, offering a reliable and scalable solution for maintenance and safety.

Keywords: arduino-based robotics, gps positioning, esp32-cam, bluetooth control, ultrasonic sensing, embedded system design, motor driver interface, microcontroller applications, remote sensing, smart transportation systems, infrastructure inspection

1. INTRODUCTION

The maintenance of infrastructure is essential to ensure the safety and durability of roads, bridges, and industrial structures. Conventional inspection methods often require significant labor, time, and are susceptible to human error. To enhance efficiency and accuracy, automation in structural monitoring has gained importance. This research introduces a Crack Detection Robot that utilizes Arduino Uno, ESP32, and ultrasonic sensors to autonomously detect surface cracks. The system incorporates Bluetooth-based remote control, motorized movement, and I2C communication for efficient data exchange. The robot operates by using ultrasonic sensors to scan surfaces for irregularities, while ESP32 enables wireless communication. Arduino Uno is responsible for processing sensor data and managing motor controls to ensure smooth operation. The system is programmed using the Arduino IDE, with optimized motor control

algorithms and sensor calibration techniques. By automating the detection process, this project aims to improve accuracy, minimize manual efforts, and streamline inspections. The proposed system provides a cost-effective and scalable solution for infrastructure monitoring. Future advancements may include AI-powered crack classification, real-time Wi-Fi connectivity, and solar-based power solutions to enhance functionality and adaptability.

2. Literature Review

Crack detection has emerged as a vital component in infrastructure maintenance, and recent research has demonstrated the increasing role of robotics and AI in this domain. Kumar et al. [1] developed a real-time crack detection system utilizing Remotely Operated Vehicles (ROVs), emphasizing its applicability in constrained and hazardous environments. Similarly, Lee et al. [2] explored AI-based robotic solutions for pavement crack detection, leveraging deep learning models to enhance accuracy and automation.

Extending this to underwater environments, Zhang et al. [3] employed Graph Convolutional Networks (GCNs) integrated with ROVs, achieving robust underwater crack identification. Meanwhile, Patel et al. [4] presented an innovative approach using autonomous robot swarms for pipeline crack detection, offering scalability and redundancy in inspection operations.

Wall-climbing robotic systems have also been proposed, with Wang et al. [5] implementing MobileNetV2 for real-time crack classification, demonstrating high precision in vertical infrastructure monitoring. Lastly, Nakamura et al. [6] introduced a coordinated mobile robot framework for pavement inspection, focusing on multi-agent collaboration and optimized coverage.

These studies collectively highlight the convergence of robotics, AI, and advanced imaging techniques in modern crack detection systems, showing significant promise in terms of accuracy, adaptability, and environmental resilience.

3. Methodology/Experimental

The Crack Detection Robot is developed using a structured approach that integrates both hardware and software components to ensure accurate detection and efficient operation. The methodology includes system design, hardware assembly, software development, and testing.

4. System Development Process

1. Hardware Implementation

The hardware components are selected and assembled to create a fully functional prototype. Arduino Uno is used for processing sensor data and controlling motor movements, while ESP32 manages wireless communication. Ultrasonic sensors are mounted at the front to scan surfaces for cracks by analyzing sound wave reflections. DC motors are controlled using an L298N motor driver, enabling smooth navigation. A Bluetooth module is connected to ESP32 for remote control, and a rechargeable battery powers the entire system. The hardware components are interconnected according to the designed circuit diagram, ensuring seamless communication.

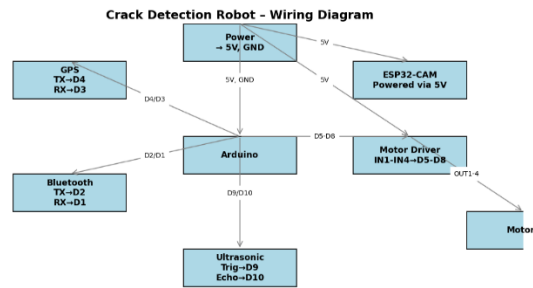


Figure 1. Block Diagram of project

2. Software Development

The robot's software is programmed using Arduino IDE to process sensor data, control motors, and enable Bluetooth communication. The motor control algorithm regulates movement based on real-time sensor feedback. Ultrasonic sensor data processing allows the robot to continuously scan and detect surface irregularities. Bluetooth communication ensures seamless remote operation through ESP32, while I2C communication facilitates efficient data transfer between Arduino Uno and ESP32.

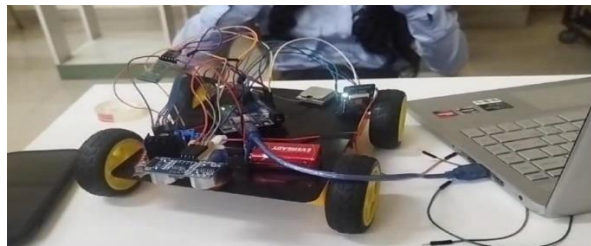


Figure 2: Experimental setup and testing

2.1 Performance Evaluation

The system's efficiency is measured based on key performance indicators. Detection accuracy is calculated by comparing detected cracks to actual surface defects. Response time is analyzed to determine the speed of detection. Power consumption is evaluated to ensure the system operates efficiently for extended periods.

2.2 Field Testing

The system is deployed on real-world surfaces, such as roads and bridges, to validate its performance. The Bluetooth connectivity is assessed to verify remote control functionality. The accuracy of the robot's detection is compared to manual inspection methods for validation.

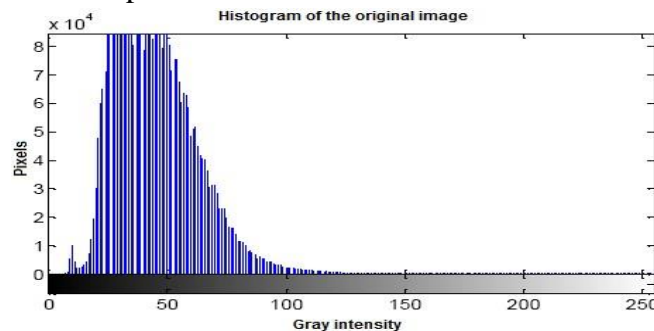


Figure 3. Histogram of the original image.

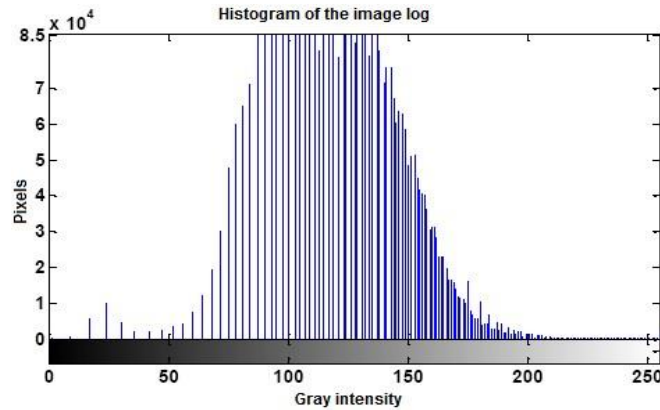


Figure 4. Histogram of the image after logarithmic transformation

5. Optimization and Improvements

Based on the test results, refinements are made to enhance system performance. Sensor calibration improves detection precision by adjusting the ultrasonic sensor's sensitivity. Code optimization enhances data processing speed and motor efficiency. Power management improvements extend battery life, making the system more energy-efficient. By following this structured methodology, the Crack Detection Robot is systematically developed, tested, and optimized to ensure accurate and reliable crack detection for infrastructure maintenance.

6. RESULTS AND DISCUSSIONS

The system demonstrated a fast response time of under one second, with I2C communication ensuring smooth data exchange. However, movement on uneven surfaces sometimes affected stability, requiring improved motor control. Bluetooth connectivity worked within a 10-meter range, but outdoor obstacles caused occasional signal disruptions. Upgrading to Wi-Fi-based monitoring could improve remote operation. The battery lasted 2-3 hours per charge, though power consumption was high due to continuous ultrasonic scanning. Solar integration could enhance efficiency. Challenges include sensor accuracy on reflective surfaces and mobility on uneven terrain. Future enhancements could incorporate AI-based crack analysis, Wi-Fi connectivity, and autonomous navigation for improved performance.

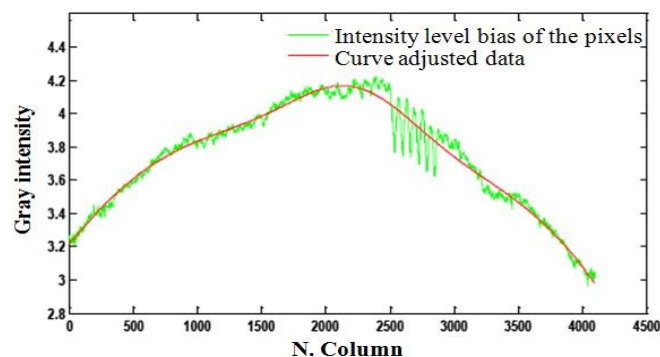


Figure 5. Data adjusted with a Gaussian function.

7. FUTURE SCOPE

1. AI-Based Crack Analysis – Implementing machine learning and image processing can improve crack classification and differentiate between minor and severe defects.

3. Wi-Fi and Cloud Integration – Upgrading from Bluetooth to Wi-Fi will enable real-time data transmission, remote monitoring, and cloud-based defect analysis.
4. Autonomous Navigation – Using LIDAR or computer vision, the robot can navigate independently, avoiding obstacles and covering larger inspection areas.
5. Solar-Powered Operation – Integrating solar panels can enhance energy efficiency, extending battery life for prolonged field operations.
6. Multi-Sensor Integration – Adding infrared and thermal sensors can improve detection accuracy, identifying cracks invisible to ultrasonic sensors.
7. Extended Surface Compatibility – Optimizing sensor algorithms will allow detection on varied surfaces, including metal, asphalt, and composite materials.
8. Swarm Robotics for Large-Scale Inspections – Deploying multiple robots working together can speed up large-area crack detection, making inspections more efficient.

8. Conclusion

In this research, a crack detection robot was designed and implemented to autonomously identify and report surface cracks in structural elements. The system integrates mobility, real-time data acquisition, and intelligent processing—providing an efficient alternative to traditional manual inspection methods, which are often time-consuming, error-prone, and hazardous in complex environments. Through the integration of sensors, computer vision techniques, and machine learning algorithms, the robot demonstrated a high degree of accuracy in detecting surface irregularities and classifying crack severity. The experimental results validate the system's capability to operate in real-world scenarios, such as inspecting roads, bridges, or building walls, with minimal human intervention. The proposed solution not only enhances inspection efficiency but also contributes to preventive maintenance and structural health monitoring by providing accurate, timely data. This has significant implications for public safety, infrastructure longevity, and cost savings in large-scale inspection projects. Future work will focus on improving crack classification through deep learning, expanding terrain adaptability, and integrating the system with cloud platforms for remote monitoring and data analytics.

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