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RADemics

AI Based Structural Health Monitoring and Crack Detection in Concrete Infrastructure

An abstract graphic in the bottom left corner featuring several thin, curved lines in dark blue and light grey, resembling stylized grass or reeds.

Iyappan G, S. Ohmshankar , Varsha Dange

NPR COLLEGE OF ENGINEERING AND TECHNOLOGY,
AGNI COLLEGE OF TECHNOLOGY, VISHWAKARMA
INSTITUTE OF TECHNOLOGY

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¹Iyappan G, Assistant professor, Civil Engineering, NPR College Of Engineering And Technology, Natham, Dindigul-624401. guruiyappan5@gmail.com

²S. Ohmshankar, AP, ECE, Agni College of Technology, ohmshankar.ece@act.edu.in, ohmshankar.ece@gmail.com

³Varsha Dange, Assistant professor, CSE in AIML, Vishwakarma Institute of Technology, Pune-411037, varsha.dange1@vit.edu

Abstract

The increasing complexity and age of concrete infrastructure demand advanced solutions for monitoring and maintenance to ensure safety and longevity. Traditional Structural Health Monitoring (SHM) techniques, although effective, often face limitations in terms of efficiency, scalability, and real-time data analysis. The advent of Artificial Intelligence (AI) has revolutionized the way SHM systems are designed and implemented, offering enhanced capabilities in data processing, damage detection, and predictive maintenance. This chapter explores the integration of AI-driven methodologies in SHM for concrete infrastructure, focusing on innovative applications such as AI-based crack detection, data fusion, and decision support systems. By leveraging advanced machine learning (ML) algorithms, deep learning (DL) models, and real-time sensor networks, AI technologies provide more accurate, cost-effective, and scalable solutions for infrastructure health monitoring. The chapter highlights key advancements in sensor technologies, including low-cost, high-efficiency sensors, and the role of Internet of Things (IoT) and edge computing in facilitating seamless AI implementation. Additionally, the potential of AI in improving SHM reporting, visualization, and predictive analytics through AI-driven dashboards is examined. Despite the promising benefits, challenges related to data quality, sensor integration, and organizational adoption remain. This chapter also identifies critical research gaps and future directions for AI in SHM, underscoring the transformative impact of AI on infrastructure management and maintenance strategies.

Keywords: Artificial Intelligence, Structural Health Monitoring, Machine Learning, Crack Detection, Internet of Things, Predictive Maintenance.

Introduction

Concrete infrastructure is fundamental to modern urban development, providing the backbone for transportation, utilities, and housing [1]. Over time, however, concrete structures are subjected to various stresses and environmental factors that lead to deterioration, making their health monitoring a critical concern [2]. Traditional methods of Structural Health Monitoring (SHM), such as visual inspections and non-destructive testing, have been used for decades to assess the condition of these structures [3]. While these methods provide valuable insights, they are often

limited by factors such as human error, subjective interpretation, and time constraints [4]. these traditional techniques are not always scalable or effective in providing real-time data for proactive decision-making. As infrastructure continues to age and the need for maintenance increases, there is a growing demand for more advanced and efficient methods to monitor the health of concrete structures [5].

In Artificial Intelligence (AI) has emerged as a promising solution to enhance the capabilities of SHM systems [6]. AI, particularly machine learning (ML) and deep learning (DL) algorithms, can process vast amounts of data generated by sensor networks embedded in concrete infrastructure [7]. These algorithms are capable of identifying patterns and anomalies in the data that may indicate potential structural issues, such as cracks, corrosion, or material degradation [8]. AI's ability to analyze large datasets in real time allows for faster and more accurate assessments, enabling engineers to detect problems earlier and make informed decisions about maintenance and repairs [9]. By integrating AI with SHM systems, the monitoring process becomes more efficient, accurate, and predictive, improving both the safety and longevity of concrete structures [10].

One of the key advancements brought about by AI in SHM is the ability to detect cracks and other forms of damage that may not be visible to the naked eye [11]. Traditional methods rely heavily on human inspectors to identify these issues, which can result in missed or delayed diagnoses [12]. AI-driven systems, on the other hand, use advanced image recognition techniques, such as convolutional neural networks (CNNs), to automatically detect and classify cracks from images or videos captured by drones, cameras, or other sensors [13]. This automation reduces the risk of human error and provides a more consistent, objective approach to crack detection [14]. AI's ability to analyze the severity and progression of cracks over time allows for predictive maintenance, enabling interventions before critical failures occur [15].

Another area where AI is transforming SHM is in the integration of data from various sensor types [16]. Modern SHM systems utilize a combination of sensors, including strain gauges, accelerometers, temperature sensors, and acoustic emission sensors, to monitor the condition of concrete structures [17]. These sensors generate vast amounts of data, which can be difficult to manage and analyze using traditional methods [18]. AI-based data fusion techniques enable the combination of data from multiple sources into a single, cohesive analysis. This holistic approach provides a more comprehensive understanding of the structure's health, allowing for better decision-making [19]. By using AI to correlate data from different sensors, SHM systems can detect patterns that may not be apparent when analyzing data from a single sensor type [20].