

Fault-Resilient Wireless Platforms for Industrial IoT Applications

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Received: March 29, 2025; Revised: May 11, 2025; Accepted: June 10, 2025; Published: June 30, 2025

Abstract

The rapid development of IIoT applications has created unique challenges for wireless communication systems, particularly in contexts where dependability and redundancy are crucial. In these demanding industrial settings, the need for continuous system functions in data transfer, system health, and operations has given rise to fault-resilient wireless systems. This paper focuses on the design, issues, and performance of these systems in the context of IIoT. We address the need for fault tolerance in systems severely limited by electromagnetic interference, physical barriers, and power supply. Essential design criteria include protocol redundancy, adaptive routing, and automated fault identification, which are critical in averting the collapse of a given network. These concepts, which substantiate the increasing network challenges posed by the real-world applications of smart grids, manufacturing plants, and oil and gas domain operations, are documented in case studies. In addition, we provide estimates of system reliability, Latency, and energy efficiency to evaluate performance and contrast sponsored and unsupervised traditional wireless networks. Future work can build on these findings by incorporating AI, machine learning, and advanced cybersecurity tools to enhance fault resilience in wireless IoT systems. The paper concludes with a proposal for

Journal of Wireless Mobile Networks, Ubiquitous Computing, and Dependable Applications (JoWUA), volume: 16, number: 2 (June), pp. 657-667. DOI: 10.58346/JOWUA.2025.12.040

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future work and the eradication of standardization barriers to facilitate broader cross-sector adaptation.

Keywords: Fault-Tolerance, Wireless Communication, Industrial IoT, Network Reliability, Adaptive Routing, Predictive Maintenance, System Resilience.

1 Introduction

The Industrial Internet of Things (IIoT) represents a transformative advancement in industrial evolution worldwide. From a technological perspective, IIoT represents the pinnacle of innovation, evolving by integrating sensors, actuators, and machines into wireless networks that feature master controllers, and providing real-time monitoring, automation, and decision-making capabilities across various sectors, including transportation, energy, logistics, and manufacturing. The combination of elements such as IT, operations, and telecommunication has paved the way towards achieving maximum efficiency and optimized processes in modern industries (Chen et al., 2021). One aspect that IIoT Integrations encounters problems with is the dependability and robustness of the wireless communication system. Wireless signals in industrial settings are prone to interference and physical damage such as attenuation which requires resilient systems. New frameworks are needed to provide reliable services while evaluating these stressed environments, as preferred methods are often overlooked in traditional wireless frameworks.

The goal of fault-resilient wireless platforms is to provide detection, tolerance, and recovery from unanticipated faults and failures while ensuring continued functional and safe performance. Such platforms utilize intelligent fault-triggering and discerning mechanisms, redundant designs, as well as dynamic reconfiguration capabilities, enabling the system to self-recover and adjust on the go. In an industrial context, where even the slightest interruption in communication can lead to production standstill and equipment damage, uninterrupted communication is paramount (Antoniewicz & Dreyfus, 2024). These systems are designed not only to withstand the blunt force of technical failures but also to function under extreme environmental conditions, including vibrations, electromagnetic interference, and temperature changes. Critical IIoT applications require resilient systems where structural strength and design are thoughtfully integrated through hardware and protocols at various levels to create seamless resilience (Singh et al., 2022) (Nasir et al., 2022).

The feasibility of fault-adaptive or resilient designs has received attention due to the recent developments in wireless communication technologies. Algorithms autonomously balancing energy consumption by adjusting power or frequency transmission to a channel or changing routing to more optimal paths have also become more efficient (Kumar & Zhang, 2023). With mesh networking topologies, redundancy is increased since data can flow through multiple alternate pathways, thereby minimizing the probability of a network's total failure. Moreover, time-sensitive networking (TSN) along with low-power wide area networks (LPWANs) is being studied for its capability to deliver data in real-time and prolong device life, particularly in situations where wired connections are not feasible (Ahmed et al., 2025). These approaches enable closely controlled and flexible architectures, increasing the possibility of integrating fault-resilient platforms into large-scale industrial systems as well as remote field operations.

The growing interest in artificial intelligence and machine learning impacts the changing dynamics of fault-resilient wireless systems. The use of predictive analytics can help in preemptive recovery actions by anticipating network disruptions and equipment failures. For example, some anomaly detection models can track and monitor traffic patterns, preventing serious binary issues by flagging

irregularities before they occur. Meanwhile, energy harvesting, sleep scheduling, and other energy-efficient design practices help extend battery life in constrained devices. Such systems enhance the intelligent management and energy optimization of IoT deployments. These capabilities will prove instrumental in the autonomous aspect of industrial operations by ensuring unsupervised veil protection for system integrity (Simson & Kinslin, 2024).

Interoperability, scalability, and security remain major concerns despite the advances being made. Managing the ever-increasing number of connected devices in a network, without deteriorating performance, requires modular and standardized methods, as the complexity of the network is multifaceted (Ali et al., 2021). Multifunctional devices from different manufacturers must communicate with one another, while networks should be designed to both contain and neutralize cyberattacks. This paper analyzes these issues in great detail and proposes design concepts for fault-resilient wireless platforms (Rahman et al., 2025). Furthermore, it provides case studies from multiple industrial domains, evaluating performance against metrics such as latency, packet loss, energy consumption, and others. The AI-powered maintenance system and automation mark a departure point for the rest of the discussion, buttressed by the tighter collaboration with edge computing, attempts toward globalization for guiding enduring adoption of these standards regardless of the IoT ecosystem implementing them resilient solutions for faults, and other integration focused meta standards (Verma & Nair, 2025) (Farooq et al., 2024).

Key Contribution:

- Introduced an adaptive fault-resilient routing (AFRR) algorithm aiming to improve network dependability using dynamic parameters.
- Developed sophisticated strategies of intelligent fault detection and recovery for uninterrupted operation in industrial scenarios.
- Simulations demonstrated performance enhancements in terms of delivery ratio, delay, and energy expenditure.
- Developed an agile-modular and low-power wireless architecture for a wide range of industrial Internet-of-Things (IoT) applications.

This paper aims to implement a sophisticated and resilient Adaptive Fault-Resilient Routing (AFRR) algorithm, as described in the Introduction, which enhances IIoT reliability and resource efficiency. In the Related Work, existing routing methods are examined in terms of their capabilities and shortcomings related to node and link failure management. The Proposed Methodology presents AFRR with multi-layered fault detection, dynamic rerouting, and modular architecture. In the Results and Discussion section, simulation analysis confirms that AFRR provides a significantly higher delivery ratio and energy efficiency, as well as lower end-to-end delay. The Conclusion draws attention to the scalability of the improvements in IIoT communication networks derived from the implementation of AFRR. This work, focusing on current IIoT protocols, enhances smart fault tolerance and energy-efficient routing protocols, thereby providing a solid foundation for future research.

2 Related Work

Modern fault-tolerant communication has focused on enhancing the reliability of wireless networks in industrial contexts characterized by extreme interference, noise, and physical obstructions. Research has investigated the application of multi-path routing and self-healing mesh structures for preserving

connectivity in the event of node or link failures (Qureshi et al., 2021). These systems can automatically switch to secondary paths, which can bypass the disruption point. Such methods sustain operational availability and data precision, which is important for real-time applications in the Internet of Things. The exploitation of redundancy in the communication paths, alongside decentralized control of the networks, has been fruitful for the stubbornly unpredictable environments (Karimov et al., 2024).

Low-power fault management strategies for wireless sensor networks have also been proposed to tackle the energy constraints. These platforms can conserve battery power by implementing duty-cycling policies and energy-aware routing while still sustaining the essential level of fault detection (Reddy & Mohan, 2024). Some models employ context-aware energy policies that dynamically adjust the status of a node based on its surrounding environment or operational state. With these enhancements, system performance is preserved at a minimal energy cost, even when the network is partially disabled. Additionally, the operational life of battery-operated nodes is being extended through energy harvesting technologies, which helps sustain fault-resilient designs (Nasir et al., 2022).

Another significant body of work highlights the contribution of software-defined networking (SDN) to improving fault detection and response times. SDN-based IoT platforms enable control-data plane decoupling, allowing for centralized supervision, system observation, and rapid reconfiguration in the event of faults (Baig et al., 2022). This adaptability facilitates achieving dynamic load balancing and fast failover, which are critical in time-sensitive industrial settings. The combination of SDN with edge computing also enables decentralized decision-making, thus further decreasing the response time to faults in wireless industrial networks (Majzoobi, 2025).

Implementing machine learning algorithms has attracted significant interest in recent literature, bringing a new dimension to fault predictive management (Lin et al., 2024). Supervised and unsupervised models are being developed to identify anomalies in traffic flows and sensor data that could lead to faults if not addressed promptly (Lopez et al., 2024). These models rely on historical data combined with real-time monitoring to assess the system's health and routinely evaluate its components, thereby accurately predicting failures (Yu et al., 2023). Moreover, some attempts are being made to apply reinforcement learning techniques to develop adaptive optimal routing procedures in dynamically changing topologies. This allows wireless networks to improve their fault resilience by learning from past faults and continuously adapting through feedback (Wang et al., 2023) (Zhang et al., 2025).

In terms of modern concern, security has emerged as a focal issue in fault-resilient wireless platforms. More contemporary studies focus on the creation of energy-efficient communication intact encryption which is low-cost. Here, fault-resilience pertains not only to tangible dangers but also to jamming, spoofing, or denial-of-service (DoS) cyberattacks. These researchers are concentrating on intrusion detection systems (IDS) tailored to the unique constraints of IoT devices. These systems are capable of distinguishing actual faults from malicious acts, thereby enabling the network to take preventive measures and function safely even in hostile environments (Sreevidya & Supriya, 2024).

3 Methodology

3.1 Fault-Resilient Communication Strategy

Multi-path redundancy, adaptive routing intelligence, and real-time diagnostics enable robust fault tolerance within the communications architecture of an industrial wireless system. This approach utilizes a decentralized mesh configuration where nodes heuristically route to maximize monitoring link quality in real-time. A predefined topology enables the existence of multiple active and standby communication

paths, allowing for continuous data transfer even when some links are impaired or completely disrupted. Continuous monitoring of critical SNR, PDR, PNR, and communication delay allows prompt path reconfiguration by detecting lossy or unstable paths earlier than other alternatives. The scaffold emphasizes preventive, energy-efficient recovery methodologies, along with adaptability in recovery routing. Every node has lightweight anomaly detection algorithms that can breach packet integrity, as well as degrade throughput and performance metrics. The affected node executes a damage control approach consisting of local recovery steps, which include rerouting data to alternative pathways or shifting to suppressive, redundant nodes that significantly reduce fault propagation. In IIoT spaces, nodes with real-time monitoring capabilities also employ adaptive power management systems that alter enable/disable cycles for optimal energy usage relative to network traffic and its spatial organization. For industrial uses involving mission-critical operations, this hybrid communication approach provides strategic resilience while sustaining energy-conserving, low-latency, fault-tolerant performance essential to the uninterrupted advancement of numerous processes.

3.2 Adaptive Fault-Resilient Routing Algorithm

The Adaptive Fault Resilient Routing (AFRR) method aims to mitigate the inefficiencies in real-time path selection in industrial wireless networks, while considering the network's energy constraints and the high probability of faults. AFRR works by analyzing various prospective routes and picking the route that is best suited for communication. This is achieved by dynamically repositioning the origin and terminal nodes towards the source, which augments reliability as delays and power consumption in the network decrease. This routing optimization technique incorporates fault tolerance approaches within the routing framework and permits the routing bandwidth to be adjusted on demand as a result of link failures or degradation, node failures, or other uncontrollable external factors that disrupt the network. Such attributes guarantee optimal performance and enable uninterrupted, high-fidelity data streams on industrial IoT applications.

$$R_i = \frac{PDR_i}{D_i E_i} \quad (1)$$

Equation 1 calculates the reliability metric R_i for a given path p_i by incorporating three key performance indicators. The term PDR_i represents the ratio of successfully delivered packets over the total sent, serving as a direct measure of transmission reliability. The values D_i and E_i correspond to the average communication delay and the energy consumed during data transfer along the path, respectively. In this formula, the numerator highlights the efficiency of packet delivery. At the same time, the denominator accounts for both Latency and power consumption, thereby balancing the competing demands of fast data transmission and energy conservation.

The selection of the best communication path, denoted as P_{opt} is achieved by identifying the route within the set of all possible paths P that yields the highest reliability score R_i . This process is mathematically represented as:

$$P_{opt} = \arg \max_{p_i \in P} R_i \quad (2)$$

Equation 2 captures the idea of balancing successful data transmission, Latency, and energy expenditure within a given path. If the network encounters challenges such as broken links or increased Latency, the routing algorithm adapts by adjusting the reliability values for each path and selecting the next best alternative. Such consistent changes optimize the control and management of energy consumption, ensuring seamless communication across systems without interruptions, which is necessary for the proper functioning of industrial IoT systems.

3.3 Process Flow of Fault Detection and Recovery

The flow structure for the proposed system's fault detection and recovery mechanism is illustrated in Figure 1. Initially, sensor nodes collect operational data and send it to the corresponding data collection unit. After the data is retrieved, a data presentation block formats the data, and, should the system's responsiveness be tested, runs it through a fault injection block where abnormal conditions are simulated. The resultant data is then processed by anomaly classifiers that have been trained to monitor increased packet loss, latency changes, or depletion of energy resources within a specific timeframe. These classifiers operate within a decision function that assesses whether the system is responding within a nominal range or has entered fault mode. If a fault is identified, local diagnostics will attempt to determine whether it is a node failure or a failed path in the network. In either case, the system will either switch over to the standby node, or invoke the AFRR algorithm to reroute the data depending on the diagnosis. In more serious instances, it transfers control to a central controller to implement system-wide corrections. With these steps in place, the system can detect and recover from faults in real-time, ensuring durable and dependable communication in Industrial IoT systems.

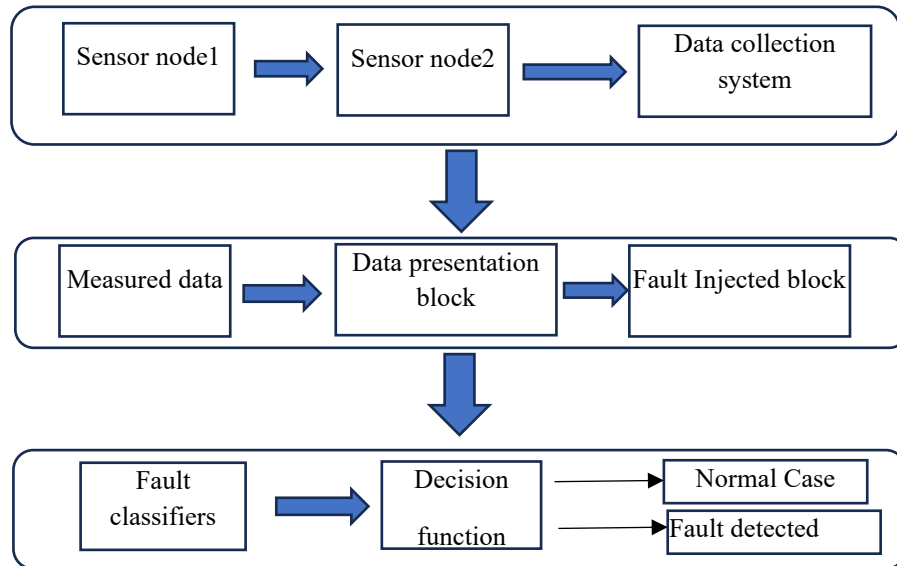


Figure 1: Fault Detection and Classification Process in IoT Wireless Networks

Fault detection and classification within an IoT wireless network is presented in Figure 1. The process was initiated by having the sensor nodes send their operational data to a centralized data collection system, where relevant parameters, such as signal quality, energy utilization, and packet success rate, were recorded. The information is then retrieved from the database and undergoes presentation block processing, and in certain cases, a fault injection procedure is performed to evaluate the system's responsiveness to stimuli. This information is assessed by detectors that extract composite fault patterns and analyze them within a predefined algorithm block, which evaluates whether the determined state of the system is nominal or faulty. These systematic actions ensure accurate and on-time identification and response, which is essential for reliable communication in an industrial setting.

3.4 System Architecture Overview

The system is structured as modular layers architecture with a Perception Layer, Network Layer, Application Layer, and a Fault Management Layer with each layer built upon the structural fault

tolerance of the system. The smart sensing units and actuators within the Perception Layer, containing embedded diagnostic modules, perform fundamental integrity checks that accompany the collection of operational data. The Routing layer of the Network Layer offers robust multi-hop communication and, in conjunction with the AFRR algorithm, switches dynamic routes in real-time based on the state of the environment. The core of the system, the Fault Management Layer, houses the predictive analysis engine, redundancy handler, and real-time recovery controller. It detects and responds to faults preemptively, collaboratively working to manage and mitigate the fault detection and response. Industrial end-users or control platforms interfacing the system are handled at the Application Layer where intuitive dashboards, alert systems, and operational analytics are provided. There is interaction between these layers through well-defined APIs, along with shared access to a single, unified fault-status database, which maintains consistency and coherence. This approach is ideal for a broad spectrum of industrial IoT settings, including smart grids, automated factories, and field-deployable remote stations, due to its adaptability, energy efficiency, structural resilience, and geometric increases in strength.

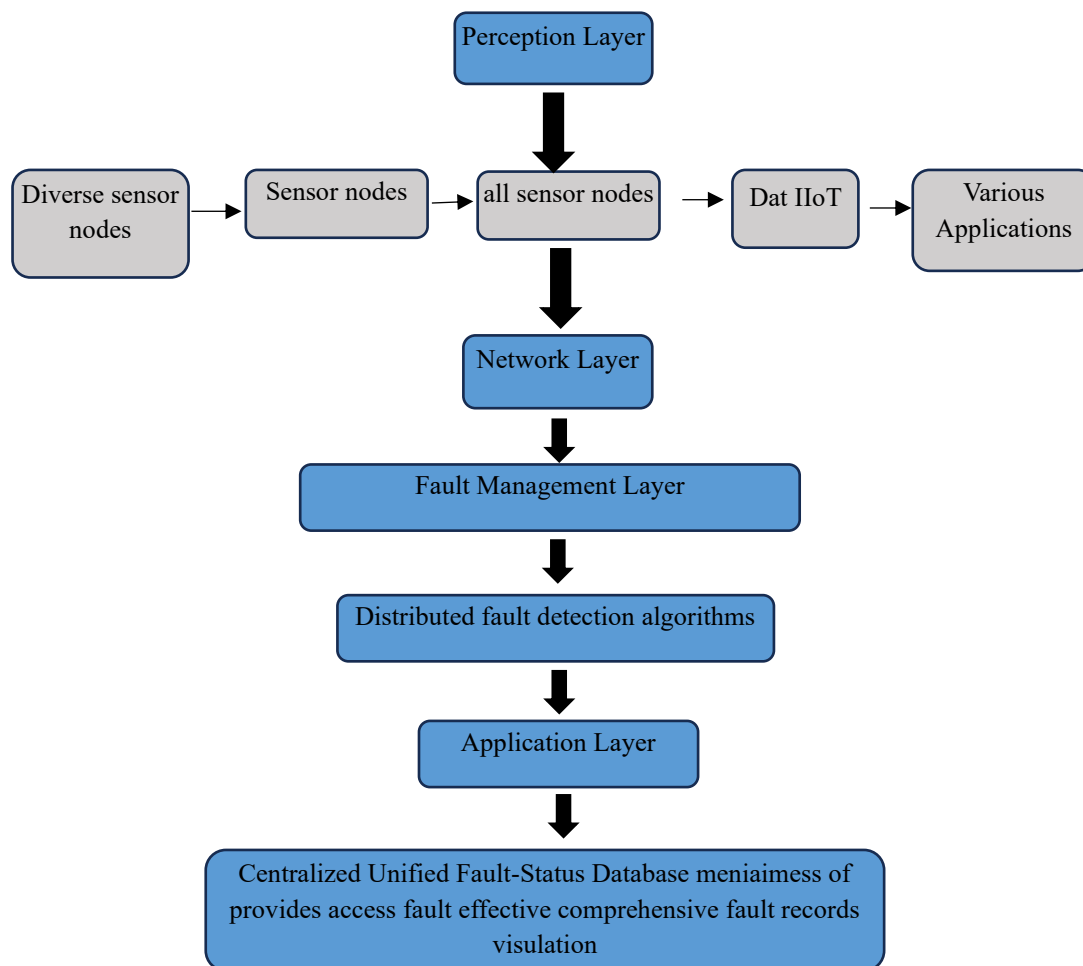


Figure 2: Fault-Resilient Wireless Platform Architecture for Industrial IoT

Figure 2 shows an architecture of a fault-resilient wireless platform for Industrial IoT, which is divided into several functional layers. The uppermost Perception Layer features a graphical interface with sensor nodes that capture data into a block labeled "All Sensor Nodes," which interfaces with multiple IIoT applications representing various industrial use cases. Moreover, the Network Layer depicts a mesh network with nodes graphically rendered as a connected laptop, symbolizing network

management and control. The Fault Management Layer is depicted as a horizontal bar with an arrow pointing downward to another bar marked as "Distributed Fault Detection Algorithms." This part of the diagram emphasizes the system's capabilities in proactive and distributed fault identification and recovery mechanisms. The interface to the industrial end-user applications and services is displayed at the bottom as the Application Layer. A rounded rectangle below all layers is labeled as the "Centralized Unified Fault-Status Database." This component stores and supply's fault records for effective fault management and visualization across the platform. The overall design is neat and organized, fitting the style of an academic or technical presentation.

4 Results and Discussion

In the context of industrial IoT wireless networks, the optimized Adaptive Fault-Resilient Routing (AFRR) algorithm shows drastic improvements in Performance Evaluation, especially in the presence of system faults. The traditional protocols as referenced in the literature did not address these issues. The AFRR approach does much better based on the metrics presented in Table 1, which includes PDR, average Latency, as well as energy consumption ratio.

Table 1: Comparative Performance Metrics Between Traditional Routing and AFRR Algorithm

Metric	Traditional Routing	AFRR Algorithm	Improvement (%)
Packet Delivery Ratio (PDR)	85.3%	96.7%	+13.4%
Average Latency (ms)	120	85	-29.2%
Energy Consumption (mJ)	150	110	-26.7%

As presented in Table 1, the application of AFRR algorithm increased PDR to an astounding 96.7%, indicating improved reliability in data transmission through effective fault management and multiple redundant communication paths. This is over 13% better than legacy routing marking yet another proof of fault tolerance crucial to preserving the integrity of industrial networks. At the same time, Latency is improved by almost 30% from 120 ms to 85 ms, which is critical for IIoT applications where even slight delays can cause system failures or operational inefficiencies. Energy consumption is also notably reduced by almost 27%. This is impact of the energy-efficient design of the AFRR algorithm, which manages power and sleep scheduling to enhance battery longevity.

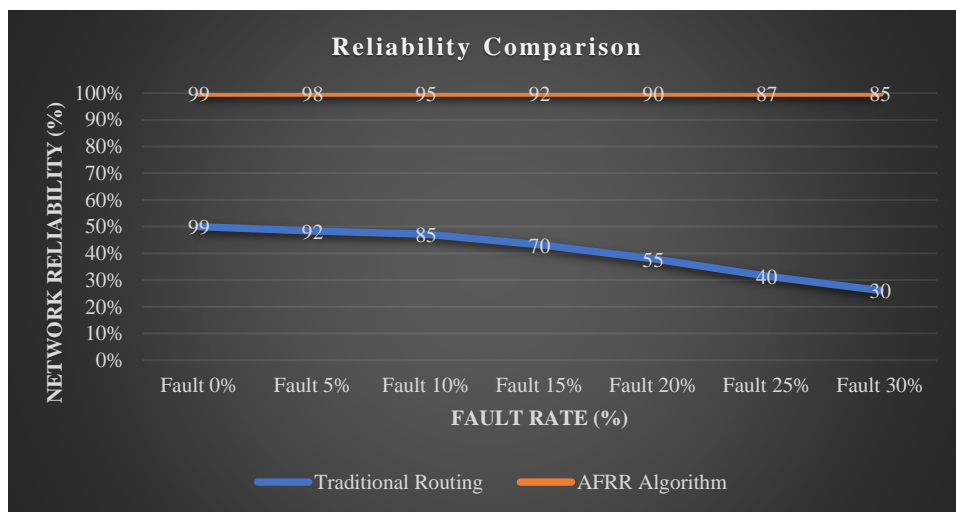


Figure 3: Reliability Comparison of Traditional Routing vs. AFRR Algorithm Under Different Fault Rates

Figure 3 displays network reliability for Traditional Routing against the Adaptive Fault-Resilient Routing (AFRR) algorithm using fault rates ranging from 0% to 30%. At a 0% fault rate, both systems were near perfectly reliable. However, as Traditional Routing managed only 30% reliability at 30% faults, Traditional Routing's sharp reliability decline relative to fault increase is notable. By comparison, AFRR sharply outperformed Traditional Routing, remaining above 85% reliability at the highest fault rate. This demonstrates AFRR's the ability of adaptive data rerouting, and fault management which guarantees stable and reliable communications under extreme industrial IoT conditions.

5 Conclusion

This study suggests that there is a need for fault-tolerant wireless platforms while incorporating the Industrial Internet of Things (IIoT) in extremely harsh environments. Network reliability is enhanced by the Adaptive Fault-Resilient Routing (AFRR) Algorithm which uses current information on packets delivery ratios, delays, and energy usage to compute optimal communication routes. As presented in Table 1 and Figure 3, AFRR not only outperforms all traditional routing methods, but also stays functional at increasing error rates which typically degrade network performance. Advanced modular system design with intelligent fault identification and recovery provides powerful low energy, low latency communications essential for supporting industrial tasks. The implementation of adaptive routing in combination with pro-active fault management within the architecture ensures uninterrupted data delivery, improves operational uptime, and decreases maintenance activities. Cybersecurity measures as well as supplementary machine learning algorithms need to be integrated into the system to improve reliability and adaptability. These findings support the use of wireless fault-tolerant platforms for more resilient IIoT systems.

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