

REVIEW ARTICLE

Survey on Edge Computing and Internet of Things (IoT)-for Low-Latency Industrial Automation

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Abstract— *Edge Computing and Internet of Things (IoT) has significantly changed the face of industrial automation, which is now able to support low-latency communication, real-time decision-making, and generally enhanced operational efficiency in smart manufacturing environments. This review delivers an in-depth survey of the research trends, the architectures, and the applications which combine the two technologies - edge and IoT - in industrial settings. Based on the examination of numerous researches, the article demonstrates how the local data processing at the edge not only lessens the congestion of the network but also increases the reliability and thus makes possible predictive maintenance, smart manufacturing, and energy optimization main contributors of Industrial Internet of Things (IIoT) such as MQTT, CoAP, and fog computing frameworks, and points their significance in achieving deterministic and secure industrial operations, unveils issues that have been left unsolved, such as scalability, interoperability, energy efficiency, and real-time analytics under high data loads, amid substantial progress made so far. In the end, they have a roadmap with emphasis on AI-driven edge optimization, 6G-enabled connectivity, and environmentally friendly industrial IoT ecosystems as the key to unlocking not only the full potential of Industry 4.0 but also the beyond*

Keywords—*Edge Computing, Internet of Things (IoT), Industrial Automation, Low-Latency Communication, Industrial Internet of Things (IIoT).*

INTRODUCTION

Industrial automation has changed various industries, which means it has gone beyond the traditional manufacturing ones. Basically, it is the use of control systems and information technologies in order to reduce human intervention in the production of goods and services. Whereas mechanization gave us machines that helped the workers with their physical tasks, automation takes this idea further by drastically lessening the requirement of human beings for sensory and cognitive functions[1][2]. Currently, industrial automation is moving away the control of single devices to the continuous, integrated automation solutions throughout the whole production environments.

Internet of things (IoT) means interconnection of physical devices, which are embedded with sensors and software. Internet of Things (IoT) represents a general concept for the ability of network devices to sense and collect data from around the world. The numbers of Internet

connected devices are increasing at the rapid rate[3][4]. These devices include personal computers, laptops, tablets, smart phones, PDAs and other hand-held embedded devices. The interconnected devices are also known as smart devices over the internet. IOT allows control the physical devices remotely through existing internet connection[5][6]. IOT devices can be connected to the internet and controlled from anywhere in the world, as long as there is an internet connection[7][8] such as quality of service (QoS), environments.

Edge computing paradigms represent a significant shift in the way computation and analytics are performed by bringing them closer to the data sources. Edge environments are made up of resource-limited devices such as smartphones, wearables, single-board computers, routers, and switches, as well as moderately capable resources like base stations and local networks. These systems cut down the reliance on centralized cloud servers by executing the initial data processing at the network edge[9][10]. Cloud computing is still used as a complementary method to offer large-scale storage and advanced machine learning or optimization capabilities. Edge paradigms like Mobile Edge Computing (MEC) and Fog Computing (FC) have been lithium-ion pivotal enablers for IoT-based applications in the areas of smart homes, smart grids, precision agriculture, retail automation, and connected vehicles

Industrial automation, robotics, autonomous guided vehicles (AGVs), and motion-control systems require extremely reliable and deterministic communication to be able to operate correctly and in real-time[11][12]. Data transmission delays, even if they are slight, may result in loss of efficiency, production errors, or safety hazards. Hence, IoT networks in the manufacturing industry have to be equipped with strong communication protocols that ensure very low latency, high reliability, and stable performance[13][14]. Low-latency communication makes it possible for machines, sensors, controllers, and cloud services to interact without any interruption[15], which in turn leads to higher operational efficiency, less downtime, and better decision-making due to real-time analytics

A. Structure of the paper

This paper is structured as follows: Section II reviews Edge Computing and IoT architectures and their industrial applications. Section III discusses low-latency communication and trends in industrial automation, while Section IV examines the integration of Edge-IoT systems, highlighting benefits and key applications. Section V surveys

recent literature on Edge-IoT frameworks, and Section VI concludes with future research directions.

II. EDGE COMPUTING AND IIoT: A CONCEPTUAL FRAMEWORK

Edge computing permits information from internet of things to be analyzed at the edge of the system before being sent to the data center or cloud. While a single device creating information can transmit it over a system effectively, issues emerge when the quantity of gadgets transmits information simultaneously increases. Not only will quality suffer because of idleness, yet the expenses in data transfer capacity can be gigantic. Edge-computing equipment and administrations help take care of this issue by being a local source for processing and storing a significant number of these frameworks. An edge gateway, for instance, can process information from an edge device, and afterward just send the important information back through the cloud, lessening transmission capacity needs[16].

A. Architectures of Edge computing

Edge computing is typically structured into a three-layered architecture, allowing seamless data processing and management between end devices and cloud infrastructure. The End Layer consists of devices that generate and collect data, such as IoT sensors (temperature, motion, and environmental sensors), smartphones, drones, smart cameras, autonomous vehicles, and industrial robots. These devices transmit data to higher layers for processing. Some high-performance devices, like smartphones and industrial robots, can perform limited local processing before sending data forward as shown in figure 1 .

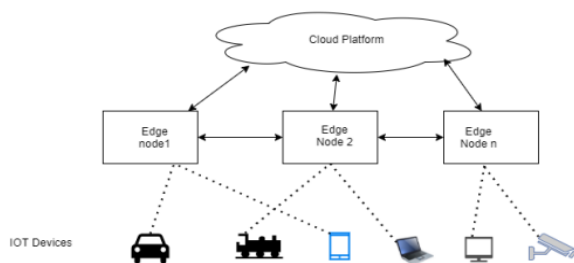


Fig.1. Edge Computing Architecture

The Edge Layer includes edge nodes responsible for processing data closer to the source. Common edge nodes include 5G base stations, which process mobile network data before sending it to cloud servers, gateways and routers that filter and analyze data, and local edge servers that store and process data for real-time applications[17]. This layer reduces latency, minimizes cloud dependence, and enhances security by keeping sensitive data closer to its source . In this Layer, data remains at the origin hence decreasing the probability of security risks. This also allows the architecture to increase the speed and response time. While edge computing minimizes cloud reliance, certain tasks, such as deep learning model training and big data analytics, still require high-performance computing (HPC) provided by cloud infrastructure.

B. Edge Computing In Real-Life Applications

The progress of IoT year by year, the merits of edge computing are attracting a great deal of attention and are already being used in many scenes.

- **Face Recognition System:** face recognition system that automatically recognizes a person from the digital image of the camera and has been introduced as a security enhancement such as immigration control, company entry/exit, and PC logon. Face recognition systems require high real-time performance, but the use of edge computing has made it possible to perform high-speed authentication.
- **Wearable Device:** A wearable device can be worn on the part of the body, such as the head, wrists, or ankles. Since it can be worn like glasses or a wristwatch, it is used in fields such as health management, medical fields, industry. Besides, smart watches can be used in conjunction with smartphones act as wearable devices and are becoming consumers as edge computing that utilizes mobile devices.
- **Human Behavior Analysis Service:** As the face recognition system, the wearable device is a mechanism that performs image processing of face recognition at the edge part and analyses it in cooperation with the data on the cloud. Wearable devices can predict and track the movement of a person extracted from camera images the analysis data of the tracking is processed in the temporary memory of the computer. for each camera, and besides a flow line of people it is used for the analysis of purchasing behavior at retail stores.
- **Edge Computing as the savior of the IoT Era:** In modern times, vast amounts of data are stored in cloud and processed on the cloud, such as the spread of IoT, digitization, diversification of data, and quality improvement[18]. However, the amount of data is increasing the processing on the cloud cannot be completed in time, and scalability problems (delay in data processing) will occur. Meanwhile, edge computing, which distributes communication traffic, come to be used, Therefore, edge computers can be said to be the savior of the IoT era and will increase the value of big data.

C. Deploying Edge Computing in IIoT

The IIoT system consists of a large number of heterogeneous node devices, interconnected through heterogeneous include sensor networks, wireless Wi-Fi networks, mobile communication 3G/4G/LTE/5G networks and dedicated industrial buses. A large number of distributed heterogeneous industrial devices form an edge network, collecting industrial data in real time and transmitting it to the cloud server for computing and control[19]. With the development of IIoT, the scale of such networks is becoming larger, and traditional cloud data center networks struggle to satisfy the real-time, security, and reliability requirements of IIoT for massive data transmission and processing. For example, in a smart factory, a large number of production devices collect a large of sensory data, sometimes the data volume gets the level of GB per second.

III. LOW-LATENCY INDUSTRIAL AUTOMATION

Low-latency communication plays a vital role in ensuring the success of IoT applications, particularly in industrial automation, robotics, and control systems[20]. In applications such as robotic assembly lines, autonomous guided vehicles (AGVs), and industrial motion control, even minor delays in data transmission can lead to significant inefficiencies,

operational failures, or safety hazards. ~~Therefore~~, IoT networks require robust communication protocols that offer minimal delay, high reliability, and deterministic performance to meet the stringent requirements of industrial environments.

A. Emerging Trends in Industrial Automations

The industry has undergone three revolutions: mechanization, electrification, and information. The fourth industrial revolution (also referred to as “Industry 4.0”), currently underway, is marked by the pervasive deployment of IoT devices and services. In this revolution, a wide range of devices are being deployed in a self-organizing manner, typically relying on control and communication systems to manage their operation and interaction. For example, in Supervisory Control and Data Acquisition (SCADA) systems, proprietary communication systems have been mostly replaced by Sensorbus and fieldbus systems., they share similarities in terms of physical and logical organization complexity. Additionally, they share common requirements for determinism, reliability, interoperability, and traffic convergence.

- **Timing and determinism:** Industrial automation typically runs real-time applications with stringent requirements on their temporal behavior and accuracy when responding to internal and external events. Beyond network throughput, the commonly used performance metrics, packet transmission latency, and its time variations (jitter) are critical concerns for many industrial control systems. These factors accumulate non-deterministic delays in data transmission, which are unsuitable for real-time industrial applications. Therefore, to ensure correct operation, industrial automation systems require a certain degree of determinism
- **Reliability and availability:** Production losses in industrial automation due to unexpected stops caused by failure or deterioration of the communication environment are unacceptable. Thus, the reliability and availability of the system are critically important due to the need for accurate and continuous operation in any condition. Reliability can be quantified using appropriate measures such as mean time between failures or the probability of no failure within a specified period of time.
- **Interoperability:** Interoperability is crucial for industrial automation due to its many advantages. For example, by enabling seamless communication and coordination between various systems, businesses can experience enhanced accuracy and productivity. Real-time data exchange and coordinated control across the entire automation system also facilitate efficient decision-making, reducing errors and delays. Interoperability also improves scalability and flexibility, allowing for easier system expansion and modification.
- **Traffic convergence:** Industrial automation applications make use of different traffic types for different functionalities, e.g., sensing, control, alarming, and the like[21]. The diverse traffic types have different characteristics and thus impose varied QoS requirements. The traffic can generally be classified into critical traffic and best-effort traffic. Critical traffic typically has stringent QoS requirements, and different types of critical traffic may have particular QoS demands depending on the specific application scenarios.

B. Communication Technologies Used In Industrial Automation Applications

Industrial automation applications heavily rely on wired fieldbus standards as the existing wireless technologies have still not been able to provide stringent guarantees on the quality of service (QoS) metrics. Fieldbus standards currently used in industrial automation include PROFINET, SERCOS, HART and CAN standards[22]. The use of wireless technologies in automation industries has mainly been limited to monitoring purposes. In recent years, automation industry has slowly started to investigate wireless technologies even for tight closed loop control applications. Several initiatives have been taken recently to pave the way for promoting wireless. Besides these wireless technologies, many proprietary radio stack implementations are also actively being used. Lack of globally acceptable standards and application programming interfaces (APIs) limit the interoperability and extensibility of currently used automation devices. Furthermore, wireless technologies currently used in industrial applications are unable to satisfy the stringent reliability and latency requirements. Coexistence features and regulatory constraints used by these technologies (e.g., LBT, radio duty cycle restrictions and transmit power limitations) have the downside of non-deterministic medium access and low coverage technologies for industrial automation.

C. System Architecture For Process industrial Automation

The automation industry has been centered on a five layer hierarchical architecture for years; A commonly used architecture model is the five level Purdue Reference Model, which later formed the basis for the ISA-95 standard. The model is typically expressed as levels from highest to the lowest one: Business Systems, Plant (ERP, MRP, MES), Operation Unit level, Machine/Process Automation, Controller level and Sensor/Actuator Level[23]. In many industrial visions, the traditional automation pyramid, structurally separating hierarchical levels, has come to its end. IIoT/Industry 4.0 in Process Industry environment can be considered to include the following main elements: The digital automation systems themselves, smart equipment and the Internet based cloud-borne technology. The platform connects the sensors, actuators, controllers, robots, etc. to computational capabilities. To enable handling of the dynamic engineering processes it is necessary to connect the existing factory automation systems with enterprise resource planning (ERP) and manufacturing execution systems (MES) over the IoT infrastructure.

IV. INTEGRATION OF EDGE COMPUTING AND IIOT FOR INDUSTRIAL AUTOMATION

The advanced development of the Internet of Things (IoT) technology, there are certain paradigm changes in various sectors of the world, particularly the sphere connected with automation and predictive maintenance. During the past ten years, the prospect of the IoT to simplify the industrial processes has become the object of broader attention since it is expected to enable the capacity to observe the processes, gather data, and make an informed decision in real-time. The IoT-enabled gadgets have established themselves as the agent of the realization of high productivity, operational cost reduction, and promoted efficiency of an entire system(see

figure2). IoT is giving the industries some of the authority to be concerned and proactive about how they manage and maintain their systems; because they will be able to know when the failures and the bottlenecks will come ahead of time before they can be actualized[24].



A. Advantages of industrial automation

- **Reduced latency:** Processing data locally at the edge eliminates the time lag associated with sending data to and from a distant cloud[25]. This is critical for time-sensitive applications like robotics, real-time control, and safety systems.
- **Improved operational efficiency:** Faster data processing allows for real-time decision-making and immediate automated responses, leading to less downtime and greater productivity.
- **Enhanced security and privacy:** Processing sensitive data locally reduces the need to transmit it over networks to the cloud, minimizing exposure to cyber risks. This is vital for protecting proprietary manufacturing data.
- **Optimized bandwidth utilization:** By filtering and processing raw data at the edge, only critical information or aggregated insights are sent to the cloud. This reduces network congestion and saves on data transmission costs.
- **Increased reliability:** Edge systems can operate autonomously even if the internet connection is interrupted, ensuring the continuity of critical production processes.
- **Information Gathering:** Using data gathering, data from many sources can be shared analyzed. The acquired data must be analyzed using smart functions to transmit alerts or triggers to other systems. IoT is used to provide fresh concepts for challenges and improve the effectiveness of the procedures.
- **Scalable and flexible growth:** Companies can add localized processing nodes without requiring major changes to their central infrastructure. This allows for flexible growth and the easy integration of new devices.

B. Smart Industrial Automation Systems Using Internet of Things

The Industrial Internet of Things (IIoT) is a term used to describe a linked system of interrelated detectors, equipment, and other components to enhance automated industrialization. This enabled system surveillance and remote control. The primary function of IIoT is to improve the industrial automation process. The IIoT concept is utilized to assess, turn on, and manage various machines in the shipping, automotive, textile, agricultural, food, and beverage industries[25]. Using data gathering, data from many sources can be shared and analyzed. The acquired data must be analyzed using smart functions to transmit alerts or triggers to other systems. IIoT is used to provide fresh concepts the effectiveness of the procedures. on maintaining efficient interfaces and interactions using controllers, robots, and monitors. Concepts from the Industrial Internet of Things were used to improve the effectiveness digital industrialization and its role in product design. Software and IIoT idea modules have been used in contemporary industrial processes. The adoption of IIoT across several industrial sectors has increased exponentially over the last few decades. there may be 24 billion IIoT gadgets. These gadgets create enormous amounts of data that require effective storage and processing. Increased machine-to-machine (M2M) and direct-to-device (D2D) connections are also a part of this process, which involves data sharing. A strong IIoT standards stack that can manage all problems with data transportation and analysis at different stages is required to address this massive data expansion.

C. Applications of IIoT in Industrial Automation

The deployment of IIoT technologies in industrial automation has resulted in substantial advancements across various sectors, transforming how processes are managed and optimized. The Internet of Things (IIoT) enables machines, devices, and sensors to communicate and interact in real time, creating a highly interconnected industrial ecosystem. This section examines the most significant applications of IIoT in industrial automation, with a focus on predictive maintenance, smart manufacturing, supply chain management, quality control, and energy management as shown in fig 3.

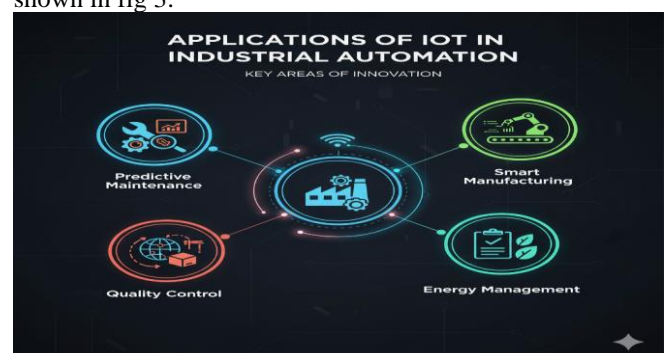


Fig.3.Applications of IIoT in industrial automation

- **Predictive Maintenance:** the most widely recognized applications of the Internet of Things (IIoT) in industrial automation is predictive maintenance. Traditional maintenance strategies often involve either reactive approaches where machinery is repaired after a failure or preventive maintenance, which typically

involves regular service intervals that may not align with the actual equipment needs.

- **Smart Manufacturing:** The Internet of Things (IoT) has also played a pivotal role in the development of smart manufacturing. In this approach, factories utilize IoT technologies to create highly automated and flexible production environments[26]. By connecting machines, sensors, and enterprise systems, the Internet of Things (IoT) enables real-time monitoring, optimization, and control of manufacturing processes.
- **Supply Chain Management:** Another major application of IoT in industrial automation is supply chain management, where IoT enhances visibility and efficiency throughout the supply chain. From tracking inventory levels to optimizing logistics, IoT solutions provide real-time data that can be used to streamline processes and reduce operational costs.
- **Quality Control:** The use of IoT in quality control has emerged as a key application for ensuring that products meet stringent quality standards. In industrial settings, IoT sensors are used to perform real-time quality inspections during production, providing immediate feedback on product quality and allowing manufacturers to correct defects before products reach the final stage. This reduces the number of defective goods and minimizes waste.
- **Energy Management:** The primary concern in industrial settings is energy consumption. IoT has solutions for energy management. Connecting energy meters and sensors to a centralized IoT platform enables companies to track the actual energy usage of various machinery and processes within an industry in real-time.

V. LITERATURE REVIEW

This section related studies on the integration of Edge Computing and IoT in industrial automation and real-time systems, showcasing their roles in enhancing low-latency communication, operational efficiency, real-time decision-making, system reliability.

Sreedevi *et al.*,(2025), edge computing architecture for agricultural monitoring systems that addresses bandwidth constraints while maintaining real-time alerting capabilities. Traditional cloud-centric approaches to agricultural IoT often struggle with network, particularly in rural environments with unreliable connectivity. Node-RED as an edge processing platform to filter, aggregate, and selectively transmit sensor data from temperature, soil moisture, and humidity sensors the practical value of edge computing for making smart farming technologies more accessible and cost-effective, particularly in regions with connectivity options[27].

Ebadinezhad *et al.*,(2025), the design and performance MQTT and CoAP protocols of IoT that ensures low latency illustrate the performance, these protocols concerning key performance indicators that include latency, power consumption, and reliability makes theoretical analysis conducting hands-on experiments on the test-bed environment, including tools for the simulation of real-world scenario software. The results show that CoAP is more suitable for applications that require minimum latency, while MQTT provides better reliability in message delivery[28].

Kotha *et al.*,(2025), the integration of Dynamic Customer Relationship Management (CRM) and Power Automate within the framework of the Industrial Internet of Things (IIoT) and Data Analytics. We investigate how IIoT-driven data collection and analytics can enhance CRM systems through automation, enabling businesses to achieve personalized customer interactions, predictive analytics, and operational efficiency. By leveraging Power Automate, organizations can streamline workflows, automate repetitive tasks, and integrate IIoT data into CRM systems for real-time decision-making[29].

Mishra and Sealy,(2024), an integrated framework leveraging the latest advancements in industrial automation, particularly focusing on the synthesis of Programmable Logic Controllers (PLCs), Human-Machine Interfaces (HMIs), the Internet of Things (IoT), and the Industrial Internet of Things (IIoT) within a Docker container environment. It emphasizes the transformative impact of technologies such as the MQTT protocol, Docker containers, and digital twins, on enhancing efficiency, industrial automation, mechanical devices to sophisticated programmable logic controllers and the integration of cutting-edge IoT devices, a framework event-driven interactions, operational efficiency through the use of MQTT for efficient system communication and Docker for application scalability and portability[30].

Eleftherios, Papatsaroucha and Markakis,(2024), This paper examines the convergence of Mining 4.0 and mining digitization, with a specific emphasis on the pivotal role that remote sensing, edge computing, and decentralized communication systems play in data transmission and real-time monitoring in light of enhancing mining operations' safety and efficiency puts forward a European solution based on open standards and tools, revolving around an Industrial Internet of Things (IIoT) Gateway that is a core component of the proposed communication system. This gateway will function as an intermediary between the edge and cloud-based platforms and will facilitate real-time monitoring through the establishment of resilient wireless communication[31].

Raman and Halam,2023 embedded sensor network architecture, implementation, and prospective applications in smart industrial automation and control systems. It examines sensors, data processing, and communication protocols that enable industrial sensor integration network scalability, energy efficiency, data dependability, and current solutions and research directions. Additionally, integrated sensor networks are changing industrial automation and control systems. It enables continuous checking, foresight support, and flexible control. These networks use cutting-edge machine learning and analytics algorithms to maximize system performance, resource usage, and intelligent decision-making. This page presents embedded sensor network case studies and examples from many industrial domains[32].

Forsström and Lindqvist,(2023), This article explores a solution using fog computing and Internet of Things technologies to address some of the potential in the combination of scalable distributed systems technologies, for enabling a distributed cluster using typical lightweight IoT devices. Answering research questions related to the expected overhead, as well as how work and chunk size affect the computational time and scalability of the system. All in order to evaluate the performance and potential of this technology for Industry 5.0 and future industrial IoT applications. Our results show the great potential of the approach, but we also

identify necessary improvements that needs to be made in order to achieve performance superiority over the traditional cloud approaches and for the technology to truly proliferate[33].

Table I provides a summary of Edge Computing and IoT-enabled frameworks for Industrial Automation and Real-Time Applications, highlighting the study focus, proposed approaches, key findings, identified challenges or limitations, and suggested future research directions.

TABLE I. COMPARATIVE ANALYSIS OF EDGE COMPUTING AND IOT-ENABLED FRAMEWORKS FOR INDUSTRIAL AUTOMATION AND REAL-TIME SYSTEMS

Reference	Study On	Approach	Key Findings	Challenges / Limitations	Future Directions
Sreedevi et al., (2025)	Edge computing architecture for agricultural monitoring	Node-RED-based edge processing to filter and transmit IoT sensor data	Reduces bandwidth use and enables real-time alerts even in low-connectivity areas	Limited scalability for large-scale IoT networks	Extend framework to multi-node edge networks with adaptive load balancing
Ebadinezhad et al., (2025)	Low-latency communication using MQTT and CoAP protocols	Experimental comparison on test-bed with latency and power metrics	CoAP offers lower latency; MQTT offers higher reliability	Protocol efficiency may vary with hardware and topology	Hybrid protocol frameworks integrating adaptive switching
Kotha et al., (2025)	IIoT-driven CRM integration with Power Automate	Real-time analytics and automation through IIoT data pipelines	Enhances decision-making and customer personalization	Dependent on cloud-edge coordination	Apply ML-based orchestration for predictive automation
Mishra & Sealy, (2024)	Integrated IIoT framework with Docker and MQTT	Container-based IIoT deployment for real-time automation	Improves scalability, portability, and real-time data exchange	Requires skilled setup and orchestration	Edge containerization with AI-based optimization
Eleftherios, Papatsaroucha & Markakis, (2024)	Edge computing in Mining 4.0	IIoT Gateway for decentralized communication	Ensures real-time monitoring and system resilience	High energy consumption in continuous edge operations	Incorporate renewable energy-aware edge systems
Raman & Halam, (2023)	Embedded sensor networks for industrial control	Integrated sensors and ML for predictive automation	Improves monitoring, resource use, and decision support	High computational demand for complex ML models	Edge-AI model compression and lightweight inference
Forsström & Lindqvist, (2023)	Fog computing framework for distributed IoT systems	Distributed clusters with lightweight IoT devices	Achieves scalability and reduced latency	Requires performance tuning for high-load systems	Optimize fog-edge collaboration with 6G integration

VI. CONCLUSION AND FUTURE WORK

Edge Computing and the Internet of Things (IoT) is a major step towards achieving fast, smart, and self-sufficient industrial automation. When computational tasks are decentralized and data is processed closer to the source, edge-IoT systems have a great impact on the reduction of network congestion, enhancement of response time, and enabling of real-time decision-making which is absolutely necessary for any kind of industrial operations nowadays. This review has been focused on the current frameworks, architectures, and applications that exemplify how industrial systems powered by the edge can become not only more efficient and scalable but also secure in a wide range of sectors like manufacturing, energy, and logistics. However, on top of these benefits, there are still issues to be solved such as interoperability problems, data privacy concerns, insufficient computing power at the edge, and the requirement for standardized frameworks which can facilitate the deployment of the edge at a large scale. Further studies may consider embedding AI and ML capabilities at the edge to enable predictive maintenance, adaptive control, and self-governing operations. Besides, the fusion of 6G connectivity and federated learning could speed up local analytics and collaborative intelligence among distributed nodes. Moreover, it would be vital to create energy-saving and eco-friendly edge architectures to cohere with the environment while ensuring performance. Additionally, there should be a focus on facilitating secure data exchange using blockchain and privacy-enhancing technologies. Tackling these issues will be the stepping stones towards Industry 5.0, which will be characterized by smart, resilient, and human-centric industrial ecosystems.

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