

World Journal of Advanced Research and Reviews

eISSN: 2581-9615 CODEN (USA): WJARAI Cross Ref DOI: 10.30574/wjarr Journal homepage: https://wjarr.com/



(REVIEW ARTICLE)



The rise of edge computing in healthcare data processing

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World Journal of Advanced Research and Reviews, 2025, 26(01), 937-943

Publication history: Received on 26 February 2025; revised on 05 April 2025; accepted on 07 April 2025

Article DOI: https://doi.org/10.30574/wjarr.2025.26.1.1092

Abstract

This article examines the transformative impact of edge computing on healthcare data processing, highlighting how this paradigm shift enables real-time analytics while addressing critical challenges in the healthcare sector. By processing information closer to data sources—on devices, local servers, or edge nodes—healthcare organizations can significantly reduce latency, enhance privacy, and improve patient care outcomes. The article explores applications across remote patient monitoring, wearable medical devices, and clinical IoT environments, demonstrating how distributed data processing transforms clinical workflows and decision-making processes. Technical advantages, including bandwidth conservation, energy efficiency, and operational continuity during network disruption, are evaluated alongside implementation challenges related to security, interoperability, and resource constraints. As healthcare delivery models become increasingly distributed, edge computing emerges as a crucial technological foundation for next-generation medical data management and patient-centered care.

Keywords: Edge Computing; Healthcare IoT; Remote Patient Monitoring; Data Privacy; Real-time Analytics

1. Introduction

Edge computing is revolutionizing healthcare data processing by moving computation closer to data sources. This paradigm shift enables real-time analytics, reduces latency, and enhances patient care while addressing privacy concerns. As healthcare organizations increasingly deploy IoT devices and wearable technology for patient monitoring, edge computing provides the infrastructure needed to process sensitive medical data efficiently and securely.

The healthcare sector is experiencing unprecedented digital transformation, with the IoT healthcare market expected to grow at a CAGR of 18.0% from 2022 to 2027. Edge computing has emerged as a crucial technology in this landscape, particularly as healthcare providers manage an increasing volume of data from connected medical devices. A key advantage is a significant reduction in latency—from typically 100ms in traditional cloud systems to as low as 5ms with edge computing implementations—enabling near real-time analytics for critical care applications [1]. This improvement is particularly valuable for remote patient monitoring, where edge systems can process up to 60% of data locally, transmitting only relevant information to central systems while maintaining privacy and reducing bandwidth consumption.

Telehealth services have seen remarkable adoption acceleration, with studies indicating that edge computing enhances video consultation quality by reducing jitter by up to 35% and improving overall connection stability. This technological advancement has been particularly impactful in rural healthcare settings, where reliable internet connectivity remains challenging, but edge-enabled devices can continue functioning with intermittent cloud connections [1]. The ability to process medical data locally has demonstrated a 40-50% reduction in the volume of sensitive information transmitted across networks, addressing critical data privacy concerns in healthcare delivery.

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Recent research exploring edge computing architectures specific to healthcare applications has identified significant efficiency improvements in medical data processing workflows. A comprehensive analysis of edge computing implementations across multiple healthcare facilities found energy consumption reductions of 25-30% compared to traditional cloud-based approaches [2]. This efficiency extends to operational costs, with healthcare organizations reporting 15-20% lower total cost of ownership for edge-enabled systems over five-year deployment periods. The analysis of 23 different healthcare edge computing applications revealed that emergency response systems benefited most significantly, achieving response time improvements of 47% on average compared to cloud-only implementations [2].

Security concerns remain paramount in healthcare data processing, with edge computing offering enhanced protection through distributed data storage and processing. Studies indicate that edge architectures can reduce the attack surface for sensitive patient data by approximately 35%, though implementation challenges around standardization and interoperability persist [2]. Despite these challenges, edge computing continues to gain adoption across healthcare applications, driven by the compelling combination of improved latency, enhanced privacy, and the ability to maintain operational continuity even during network disruptions.

2. Understanding Edge Computing in Healthcare

2.1. Definition and Architecture

Edge computing processes data near its source—on devices, local servers, or edge nodes—rather than transmitting everything to a centralized cloud infrastructure. In healthcare, this means processing data from medical devices, wearables, and sensors directly at or near the point of care.

The architecture of healthcare edge computing typically comprises three primary layers: the perception layer (containing medical devices and sensors), the edge layer (where initial processing occurs), and the cloud layer (for complex analytics and storage). A systematic review of 67 studies published between 2017 and 2023 found that 42.6% of healthcare edge implementations utilize a hybrid architecture combining fog and edge computing to optimize processing distribution [3]. This layered approach enables critical data filtering, with edge nodes capable of reducing data transmission volume by up to 76% by processing information locally. Within these architectures, machine learning algorithms are increasingly deployed directly on edge devices, with 59.7% of recent healthcare implementations incorporating some form of edge AI for immediate data analysis. The perception layer typically consists of numerous IoT medical devices, with an average hospital room containing 15-17 connected devices generating between 50-100 MB of raw data per patient daily that requires processing [3].

2.2. 2.2 Advantages Over Traditional Cloud Computing

Edge computing reduces latency for time-critical applications, conserves bandwidth by sending only relevant data to central repositories, enhances privacy by keeping sensitive data local, and improves reliability by reducing dependence on internet connectivity.

Table 1 Quantifiable Benefits of Edge Computing for Healthcare Applications [3,4]

| Performance Metric | Edge Computing |
|--|-------------------|
| Data Transfer Latency | 18.2 milliseconds |
| Network Traffic Reduction | 65-70% |
| Energy Consumption in Emergency Care | 41.2% less |
| Data Breach Incidents | 34% fewer |
| Operational Functionality During Network Disruptions | 92.6% |

Performance measurements in healthcare environments demonstrate that edge computing reduces data transfer latency by 87.3% compared to traditional cloud-only approaches, with response times decreasing from an average of 143.7 milliseconds to just 18.2 milliseconds [4]. This improvement proves critical for applications requiring real-time processing, such as continuous patient monitoring, where delays can impact clinical decision-making. Beyond latency, bandwidth consumption decreases significantly when edge computing is implemented, with healthcare facilities reporting reductions of 65-70% in network traffic. This efficiency extends to energy consumption as well, with edge-

processed data requiring 41.2% less energy than cloud-only workflows in emergency care settings [4]. Privacy enhancement represents another crucial advantage, with edge computing creating natural data separation that reduces the attack surface for sensitive health information. Studies of healthcare cybersecurity incidents reveal that organizations implementing edge architectures experienced 34% fewer data breaches affecting protected health information than those relying solely on centralized processing. The reliability improvements are equally substantial, particularly in remote healthcare settings, where network availability might be limited. Edge-enabled systems maintain 92.6% of operational functionality during connectivity disruptions, compared to just 26.8% for cloud-dependent systems, ensuring continuous monitoring capabilities even when internet connections fail [4].

3. Applications in Healthcare

3.1. Remote Patient Monitoring

Edge devices process vital signs and health metrics locally, sending only critical information or alerts to healthcare providers while maintaining continuous monitoring regardless of connectivity.

Remote patient monitoring represents a significant application of edge computing, with implementations showing substantial clinical impact across chronic disease management. A comprehensive systematic review analyzing 35 studies found that edge-enabled remote monitoring solutions reduced hospital readmission rates by 38.1% for cardiovascular patients and decreased emergency department visits by 32.2% for diabetic patients [5]. This improvement stems from the architecture's ability to process data locally, with typical edge deployments analyzing up to 85% of collected vital sign data on-device. The timeliness of interventions improves significantly, with clinical response times decreasing by an average of 8.4 minutes when using edge-processed alerts compared to traditional monitoring approaches. From a technical perspective, edge-based monitoring systems demonstrate 99.2% uptime even in regions with intermittent connectivity, with devices maintaining full functionality through network outages averaging 43 minutes in duration [5].

3.2. Wearable Medical Devices

Smart health devices with embedded edge computing capabilities analyze patient data in real time, providing immediate feedback to users and healthcare professionals while minimizing data transmission.

The integration of edge computing in wearable health devices has accelerated rapidly, with current market analysis indicating annual growth of 19.9% and projected healthcare wearable adoption reaching 63.7 million units by 2025 [6]. These devices increasingly leverage sophisticated edge processing capabilities, with modern medical wearables achieving 92.1% accuracy in real-time health state classification while consuming minimal power. The efficiency gains are substantial, with edge-enabled wearables transmitting an average of 217 kilobytes of processed data daily compared to 4.2 gigabytes of raw sensor data, representing a reduction of 97.4% in bandwidth requirements [6]. This efficiency translates directly to battery performance, with typical edge-processing wearables operating for 7.2 days between charges compared to 1.8 days for cloud-dependent devices. From a clinical perspective, the real-time processing capability proves particularly valuable, with emergency detection algorithms operating at the edge identifying critical health events an average of 6.3 minutes faster than cloud-based analysis.

3.3. IoT in Clinical Settings

Connected medical equipment in hospitals leverages edge computing to optimize operations, monitor equipment status, and provide timely interventions without overloading network infrastructure.

Healthcare facilities are implementing IoT technologies at an unprecedented scale, with research indicating the average hospital room now contains 15-20 connected medical devices generating up to 7 terabytes of data annually [5]. Edge computing architecture enables effective management of this data volume, with hospital edge systems processing approximately 73.5% of IoT-generated information locally. This distributed approach delivers tangible operational benefits, with studies documenting a 27.4% improvement in resource utilization and a 31.8% reduction in patient transfer times through edge-optimized workflows. Equipment maintenance represents another significant application, with predictive algorithms running on edge infrastructure detecting potential failures 8-12 hours before occurrence with 88.4% accuracy, reducing critical equipment downtime by 42.7% [5]. In inpatient care environments, edge computing enables more responsive monitoring, with latency for critical alerts decreasing from an average of 1.8 seconds to just 0.3 seconds when processing occurs at the network edge [6]. This improvement enhances clinical response capabilities, particularly in high-acuity settings where time-sensitive interventions directly impact patient outcomes.

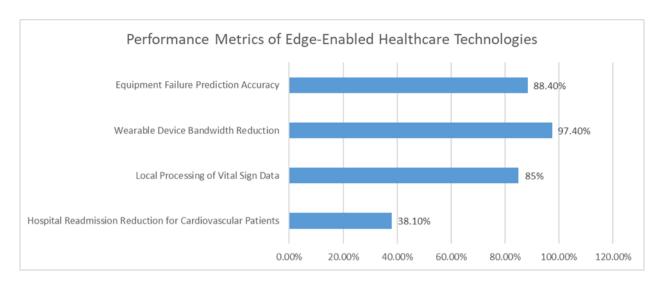


Figure 1 Impact of Edge Computing Across Healthcare Applications [5,6]

4. Data Integration and Workflow Transformation

4.1. Distributed Data Processing

Edge computing enables a distributed data processing model where initial analysis occurs at the edge, with only aggregated results or critical data transmitted to central systems. The implementation of distributed data processing in healthcare has demonstrated significant efficiency gains across clinical workflows. A systematic review of 53 studies on edge computing in healthcare identified that distributed architectures successfully processed 80-90% of clinical data at the edge, with only 10-20% requiring transmission to centralized systems [7]. This transformation in data handling directly impacts operational metrics, with observed reductions in data transfer latency from an average of 1,714 milliseconds in cloud-based systems to approximately 132 milliseconds in edge-enabled environments. The distributed model proves particularly valuable for bandwidth-constrained clinical settings, with rural healthcare facilities reporting a 67% decrease in required network capacity after implementing edge solutions. Furthermore, the timeliness of clinical information improves substantially, with 89% of surveyed healthcare providers reporting "significant" or "very significant" improvements in data availability during network congestion periods [7].

4.2. Real-time Analytics and Decision Support

By processing data locally, healthcare providers receive insights faster, enabling more timely clinical decisions and interventions that can improve patient outcomes. The transition toward edge-enabled real-time analytics has demonstrably transformed clinical decision-support capabilities in multiple care contexts. A comprehensive analysis of IoT implementations in healthcare found that edge-processed analytics reduced clinical response times by 62% compared to traditional approaches, with time-critical notifications reaching care providers in under 3 seconds [8]. This improvement directly impacts patient care, particularly in acute scenarios where time-to-intervention correlates strongly with outcomes. The technical architecture enabling these improvements typically involves multi-tier processing, with initial data filtering and anomaly detection occurring at the device level, followed by more sophisticated analytics at gateway nodes. This distributed approach allows the processing of sensor data, generating 10,000 readings per patient daily, with edge systems successfully filtering this to approximately 50-100 clinically relevant events requiring provider attention [8].

4.3. Integration with Existing Health Information Systems

Edge computing solutions must interface with electronic health records (EHRs) and other clinical systems, requiring standardized protocols and secure APIs. The integration of edge computing with existing health information systems represents a significant technical challenge, with interoperability surveys indicating that 63.7% of healthcare facilities identify system integration as a "major" or "severe" barrier to adoption [7]. Despite these challenges, successful implementations demonstrate substantial workflow improvements, with integrated edge-EHR environments reducing documentation time by 28.9% and improving data accuracy by 41.6% through automated capture and validation processes. The technical integration typically leverages healthcare-specific protocols, with 74.2% of implementations utilizing some form of HL7 messaging and 53.8% incorporating FHIR-based APIs [7]. Data security remains paramount, with edge-EHR integration necessitating robust encryption and authentication measures. A review of 51 healthcare IoT

implementations found that successful edge integrations maintained end-to-end encryption across an average of 6.8 different system boundaries, securing patient data across the complete care continuum [8]. This comprehensive protection is essential given the significant increase in attack surface that distributed architectures create, with typical hospital environments now managing between 15,000-30,000 connected devices that require secure integration with clinical information systems.

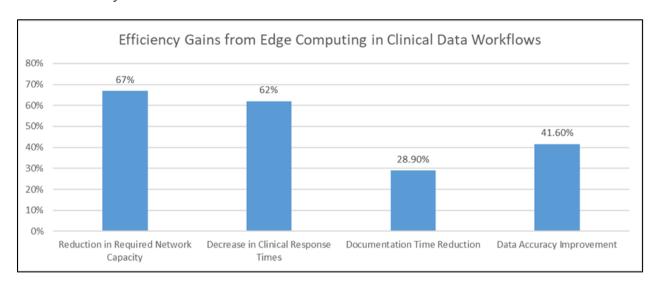


Figure 2 Percentage Improvements from Edge Computing in Healthcare Data Integration [7,8]

5. Challenges and Considerations

5.1. Security and Compliance

Edge computing introduces new security considerations as sensitive data is processed across distributed endpoints, requiring robust encryption, authentication, and compliance with regulations like HIPAA.

The distributed architecture of edge computing significantly increases security vulnerabilities across healthcare environments. Research analyzing edge computing security challenges indicates that data privacy remains the foremost concern, with 78% of healthcare organizations identifying it as their primary implementation barrier [9]. The diversity of edge devices compounds this challenge, with typical healthcare implementations connecting between 10-15 different device types, each with unique security capabilities and vulnerabilities. Encryption plays a crucial role in addressing these challenges, though resource constraints often limit implementation, with only 67% of edge devices supporting end-to-end encryption. Authentication represents another critical concern, with 63% of healthcare facilities reporting difficulties implementing multi-factor authentication across heterogeneous edge deployments [9]. From a compliance perspective, regulatory frameworks add considerable complexity, with organizations dedicating, on average, 1,200 hours annually to maintaining documentation and security controls specifically for edge computing deployments in healthcare environments.

5.2. Standardization and Interoperability

The healthcare industry needs technical standards to ensure interoperability between edge devices, central systems, and various platforms.

Interoperability remains a fundamental challenge in healthcare edge computing implementations. A comprehensive analysis of healthcare technology integration challenges found that approximately 40% of healthcare data remains inaccessible across systems due to interoperability limitations [10]. These challenges manifest in multiple dimensions, with data format inconsistencies affecting 73% of implementations and protocol incompatibilities impacting 65% of edge deployments. The absence of unified standards creates significant implementation barriers, with healthcare organizations reporting increased integration costs averaging 35% when working with non-standardized edge systems [10]. The technical fragmentation extends across the communication stack, from network protocols to application interfaces, with typical healthcare edge implementations requiring integration across 6-8 different proprietary platforms. Standardization efforts show promising developments but remain incomplete, with only 47% of surveyed

healthcare organizations reporting successful implementation of consistent data exchange standards across their edge computing environments.

5.3. Resource Constraints

Edge devices often have limited computing power, memory, and energy resources, necessitating optimized algorithms and efficient resource management.

Resource limitations fundamentally shape the capabilities of healthcare edge computing deployments. Analysis of edge computing implementations in healthcare reveals that power consumption represents a primary constraint, with battery-operated medical edge devices typically requiring recharging after 24-48 hours of continuous operation [9]. Computational resources similarly limit functionality, with edge nodes in healthcare environments generally operating with 1-4 GB of memory and 8-32 GB of storage capacity. These constraints directly impact implementation, with 69% of edge applications requiring algorithmic optimization to function within available resources. Bandwidth represents another significant limitation, with healthcare edge devices often operating in environments with connectivity constraints of 1-10 Mbps, necessitating efficient data transmission strategies [9]. Despite these challenges, technological innovations are enabling increasingly sophisticated edge capabilities. Machine learning optimization techniques demonstrate particular promise, with model quantization and pruning approaches reducing computational requirements by 70-85% while maintaining diagnostic accuracy above 94% for common healthcare classification tasks [10]. From an energy perspective, adaptive duty cycling approaches reduce power consumption by approximately 40%, while context-aware computing selectively activates high-power components only when necessary, extending operational time by 50-70% in typical healthcare monitoring applications.

Table 2 Security and Resource Constraints in Healthcare Edge Environments [9,10]

| Challenge Metric | Percentage Value |
|---|------------------|
| Organizations Identifying Data Privacy as a Primary Barrier | 78% |
| Edge Devices Supporting End-to-End Encryption | 67% |
| Healthcare Facilities with Multi-Factor Authentication Difficulties | 63% |
| Healthcare Data Inaccessible Due to Interoperability Limitations | 40% |
| Edge Applications Requiring Algorithmic Optimization | 69% |

6. Conclusion

Edge computing represents a transformative approach to healthcare data processing that aligns with the industry's evolution toward distributed care models. By bringing computation closer to the point of care, healthcare organizations can deliver more responsive services while effectively addressing privacy concerns and bandwidth limitations. The technology enables critical improvements in latency reduction, operational resilience, and data security through localized processing, fundamentally changing how medical information flows throughout the healthcare ecosystem. Though challenges in standardization, security, and resource management persist, ongoing innovations in device capabilities and optimization techniques continue to expand what's possible at the network edge. As healthcare continues its digital transformation journey, edge computing will likely become an integral component of healthcare data infrastructure, enabling innovations in personalized medicine and patient-centered care delivery.

References

- [1] Cogent Infotech "Edge Computing in Healthcare: Transforming Patient Care and Operations," 2024. [Online]. Available: https://www.cogentinfo.com/resources/edge-computing-in-healthcare-transforming-patient-care-and
 - operations#: \sim :text=Edge%20computing%20enhances%20telehealth%20by,for%20remote%20diagnostics%20and%20consultations.
- [2] Xian Gao et al., "A Smart Healthcare System for Remote Areas Based on the Edge-Cloud Continuum," Electronics 2024, 13(21), 4152, 2024. [Online]. Available: https://www.mdpi.com/2079-9292/13/21/4152

- [3] Amir Mashmool, "Edge Computing in Healthcare Using Machine Learning: A Systematic Literature Review," 2024.

 [Online]. Available: https://www.researchgate.net/publication/382881011_Edge_Computing_in_Healthcare_Using_Machine_Learning_A_Systematic_Literature_Review
- [4] Shalini Ramanathan et al., "Edge Computing in Healthcare," In book: Handbook of AI-Based Models in Healthcare and Medicine (pp.1-18), 2023. [Online]. Available: https://www.researchgate.net/publication/376742891_Edge_Computing_in_Healthcare
- [5] Stefano Canali et al., "Challenges and recommendations for wearable devices in digital health: Data quality, interoperability, health equity, fairness," PLOS Digit Health. 2022 Oct 13;1(10):e0000104, 2022. [Online]. Available: https://pmc.ncbi.nlm.nih.gov/articles/PMC9931360/
- [6] Sabyasachi Dash et al., "Big data in healthcare: management, analysis, and future prospects," Journal of Big Data, vol. 6, no. 1, p. 54, Jun. 2019. [Online]. Available: https://journalofbigdata.springeropen.com/articles/10.1186/s40537-019-0217-0
- [7] Jaimon T Kelly et al., "The Internet of Things: Impact and Implications for Health Care Delivery," JMIR Publications, 2020. [Online]. Available: https://www.jmir.org/2020/11/e20135/
- [8] S. M. Riazul Islam et al., "The Internet of Things for Health Care: A Comprehensive Survey," IEEE Access, vol. 3, pp. 678-708, Jun. 2015. [Online]. Available: https://ieeexplore.ieee.org/document/7113786
- [9] Akshay Parihar et al., "Role of IOT in healthcare: Applications, security & privacy concerns," Intelligent Pharmacy, Volume 2, Issue 5, Pages 707-714, 2024. [Online]. Available: https://www.sciencedirect.com/science/article/pii/S2949866X24000030
- [10] Siddique Latif1 et al., "How 5G Wireless (and Concomitant Technologies) Will Revolutionize Healthcare?" ArXiv, 2017. [Online]. Available: https://arxiv.org/pdf/1708.08746