

2724-0371 [©] 2024 The Authors. doi: 10.46354/i3m.2024.iwish.015

Integration of Industry 4.0 Technologies for Enhancing Healthcare Efficiency and Patient Care: A Comprehensive Literature Review

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Abstract

This paper provides a comprehensive review of the current state of Industry 4.0 adoption in healthcare, identifying the critical need for innovative solutions to address existing inefficiencies and to improve patient outcomes.

Through the use of the Scopus scientific database and well-defined criteria, 34 documents are identified and classified. They span various medical sectors, including cardiology, oncology, and neurology. The findings highlight the application of Artificial Intelligence & Machine Learning, Digital Twin, and Blockchain and Internet of Thing technologies in healthcare, discussing both their transformative potential and the challenges they present.

The paper concludes by underscoring the necessity for continued innovation and the development of a robust, comprehensive tool—such as the ANTHEM project—that can significantly enhance healthcare delivery, particularly for vulnerable and chronically ill patients in underserved regions. Additionally, the study addresses future challenges associated with the digital transformation of healthcare, particularly concerning interoperability, data privacy, and security.

Keywords: Healthcare 4.0; Digital Twin; Artificial Intelligence; Healthcare Data Security; Healthcare Privacy

1. Introduction

The healthcare sector is facing significant challenges in meeting the demands of a growing and aging population. Traditional healthcare systems are often strained by inefficiencies, limited resources, and rising costs. These pressures are compounded by the increasing prevalence of chronic diseases, which require ongoing management and frequent interaction with healthcare providers. In many cases, healthcare delivery is hampered by outdated infrastructure, fragmented communication between providers, and limited access to timely and accurate patient information (Kern et al., 2019).

Moreover, the reliance on manual processes and paper-based records can lead to errors, delays, and suboptimal patient outcomes. These inefficiencies not only compromise the quality of care but also place a significant financial burden on healthcare systems. As a result, patients often experience long wait times, inconsistent care, and a lack of personalized treatment options. (Akintayo–Usman, 2021)

In this context, there is a critical need for innovative solutions that can address these shortcomings and enhance the overall efficiency and effectiveness of healthcare delivery.



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The advent of Industry 4.0, marked by the emergence of advanced technologies such as the Internet of Things (IoT), Artificial Intelligence (AI), blockchain, and big data analytics, has introduced in significant transformations across various sectors. This new era is characterized by unprecedented levels of efficiency, connectivity, and automation, revolutionizing traditional industry practices (Mohamed et al., 2019). Initially confined to sectors like manufacturing and logistics, these technologies are now beginning to permeate the healthcare sector, promising to redefine patient care, data management, and service delivery (Abbate et al., 2023; Eduardo et al., 2023).

Especially in combating the COVID-19 crisis, the role of Industry 4.0 technologies highlights their potential in healthcare. From facilitating rapid production of essential medical supplies to supporting telemedicine and remote patient monitoring, these technologies have been instrumental in disease detection, diagnosis, contact tracing, and maintaining healthcare services during lockdowns (Javaid et al., 2020).

This article aims to analyze the main technologies and paradigms of Industry 4.0 utilized in the healthcare sector, highlighting the specific areas where these technologies are most commonly applied.

The remainder of this paper is organized as follows. In section 2, the scientific background is analyzed, and research gaps are identified to clearly explain the innovative contribution of this paper ; in Section 3, authors give all the details on the methodology conducted. Section 4 presents an analysis about the papers selected. In Section 5, authors outline the current research gaps, the possible future challenges and some conclusions.

2. Scientific background

Several literature reviews have discussed the integration of Industry 4.0 technologies, such as the Internet of Things (IoT) or Artificial Intelligence (AI), are increasingly integrated into the healthcare sector, revolutionizing patient care, data management, and the delivery of healthcare services.

For instance, IoT devices enable real-time health monitoring and management, supporting personalized care and early intervention strategies. Mishra et al. (2023), Nguyen et al. (2023) and Shams et al. (2024) describe the integration of the Internet of Medical Things (IoMT) evaluating that IoMT technologies, leveraging the interconnectedness of IoT within medical applications, significantly boost real-time patient monitoring, data analytics, and overall healthcare service management.

At the same time, regards the usage of IoT technology, addressing data fusion and security concerns is crucial for advancing IoMT capabilities in healthcare. Integrating IoMT devices enables comprehensive patient monitoring and data analysis,

improving the precision of diagnosis and treatment while securing this data against breaches with advanced encryption techniques and robust data management procedures. It is important to adopt a balanced approach that leverages IoMT benefits while minimizing security risks to ensure the long-term development of intelligent healthcare systems. Also, the usage of Blockchain technology addresses key data security and privacy issues in healthcare. It secures patient records through a decentralized and tamperproof data management system, allowing for secure data sharing among authorized entities. Research on blockchain technology in healthcare, as analyzed by Garg (2023), indicates a significant rise in interest and studies from 2018 to 2022, underscoring blockchain's potential to transform data security, transparency, and interoperability in healthcare.

Moreover, Simulation and Digital Twin technology provide exceptional predictive modeling skills. They have the ability to accurately replicate pharmacological reactions, patient responses, and surgical results, leading to a significant transformation in patient care, hospital administration, and medical training. For instance, Muskan et al. (2023) and Sun et al. (2023) explain how this kind of technology allows healthcare providers to forecast results, create customized treatments, and efficiently allocate resources. They show the potential of digital twin in enhancing patient care, medical education and training, showcasing its transformative impact on healthcare. For example, it is possible predicting cancer and neurological conditions, tracking cardiovascular trends, and customizing drug models.

Given the growing body of evidence and the increasing interest in healthcare 4.0, it is evident that these technologies are becoming essential to the healthcare sector. However, there is a noticeable gap in the literature when it comes to a comprehensive review that categorizes and assesses the current state of adoption of all these technologies. This paper seeks to fill that gap by providing a critical situational analysis of how Industry 4.0 technologies are being integrated into healthcare. It will serve as a foundational reference for guiding future innovation, policy development, and the strategic alignment of technological capabilities with healthcare objectives. By examining the current landscape, this paper will highlight the importance of multidisciplinary collaboration in the creation of a technologically advanced, equitable, sustainable, and resilient healthcare ecosystem. Ultimately, this analysis will contribute to the transformation of healthcare into a more efficient, accessible, and patient-centered system.

3. Materials and Methods

In the initial research phase, authors selected Scopus as primary database due to its extensive coverage and rigorous quality assurance, thereby ensuring the inclusion of only the most pertinent and high-quality data.

The query adopted was:

(TITLE-ABS-KEY (Industry 4.0) AND TITLE-ABS-KEY (Healthcare) AND TITLE-ABS-KEY (1) (technology))

The aim of this approach was to generate insights that are highly applicable to current technological trends and advancements, resulting in a more comprehensive understanding of the latest developments and better guidance for future research directions.

Pre-screening exclusions were performed based on language, excluding documents not in English, document type, publication year, choosing to start from 2020, as articles published prior to this year were not found to be significantly relevant, thereby focusing the search to ensure a clearer and more accurate overview, and duplicate articles.

Subsequently, the titles, abstracts, and keywords of the remaining publications were screened using the criteria specified in Table 1.

Table 1. Criteria for exclusion papers remaining to be evaluated

Criteria	Description
1	The article does not fit the specific research scope on 'Healthcare 4.0'.
2	The discussed Industry 4.0 technologies are not applied within the healthcare context.
3	The article lacks an implementation of Industry 4.0 technologies in healthcare.

This research culminated in a total of 34 articles that were deemed suitable for inclusion in this review. Due to the limited number of papers available on this topic, it was not necessary to adopt more structured literature review methodologies.

4. Results and Discussion

In Table 2 below, authors present a comprehensive organization of all the documents. This table is structured to enhance accessibility and clarity, in fact each article is categorized as a first step on its specific medical sector, ensuring readers can quickly locate area of interest.

Authors identified six medical sectors: Cardiology (C), Neurology (N), Oncology (ON), Endocrinology (EN), Epidemiology (EP), Generic (G). This last category encompasses publications that do not fit into the aforementioned categories, including transdisciplinary research, general healthcare enhancements, or specialized medical areas not explicitly stated.

From the literature analysis conducted, the technology used topic has been addressed and divided into:

- Artificial Intelligence & Machine Learning (AI & ML): Studies focusing on the application of AI and machine learning algorithms for diagnostics, predictions, data analysis, and decision support systems in healthcare.
- Digital Twins & Simulation (DT & SM): Papers that explore the creation and use of digital replicas of physical healthcare entities and processes for improved monitoring, planning, and optimization.
- Blockchain & IoT Technologies (**B&I**): Articles detailing the use of interconnected devices, sensors for real-time monitoring, and IoT applications in healthcare and the use of blockchain for securing data, enhancing privacy, and enabling secure data exchange and management.
- *Other* (**O**): Any reference that does not specifically align with the above categories.

At first, it is observable that most of the papers have been written in 2023. It is quite interesting because shows that the application of 4.0 technique are one of the hottest topics for researchers and also a field that has to be better explored. At the same time, it is relevant that most of the papers are related to the epidemiology field, depicting how the pandemic crisis has impact to the world.

Moreover, as observable in Table 2, the Digital Twin and Simulation related to healthcare field is a topic of interest for scientists. In this view, Khan et al. (2023), introduces an innovative Digital Twin (DT) model that utilizes WiFi CSI (Channel State Information) signal processing and machine learning for monitoring patient respiration and providing decision support. This model showcases the smooth incorporation of IoT and AI & ML technologies to gather and analyze respiratory data, improving patient monitoring through non-invasive methods and providing an advanced illustration of Digital Twins benefiting healthcare. Instead, Konopik et al. (2023) investigate the opportunities and obstacles of Digital Twins in breast cancer therapy. Their empirical research viewpoints from compiles stakeholders, demonstrating the substantial potential of Digital Twins in tailored medicine, efficiency enhancements, and scientific research. This study highlights the significant influence Digital Twins can have on the treatment and management of chronic diseases, focusing on personalized and precise care delivery. Cimino et al. (2022) explore the importance of Modelling & Simulation (M&S) approach in a real organization of a Vaccination Hub. They show different kind of scenarios that could help supporting experts and decision makers in the organization process in order to maximize a vaccination campaign. This is a demonstration of how simulation is important in this field. Similarly, Bruzzone et al. (2021) utilize a simulation-based serious game to develop and refine operational procedures and regulations for preventing access to urban areas and key facilities.

	Sector						Technology Used			
Ref.	С	Ν	ON	EN	EP	G	AI&ML	DT&SM	B&I	0
Abidi et al. (2023)					\checkmark		\checkmark			
Alvarez-Romero et al. (2023)					\checkmark					\checkmark
Antonacci et al. (2023)					\checkmark		\checkmark		\checkmark	
Bhatt et al. (2024)	\checkmark						\checkmark			
Brönneke et al. (2023)	\checkmark									\checkmark
Brunzini et al. (2023)					\checkmark				\checkmark	
Bruzzone et al. (2021)					\checkmark			\checkmark		
Cimino et al. (2022)					\checkmark			\checkmark		
Chatterjee et al. (2023)					\checkmark				\checkmark	
Das et al. (2023)						\checkmark			\checkmark	
De Benedictis et al. (2022)					\checkmark			\checkmark	\checkmark	
Edeh et al. (2023)					\checkmark				\checkmark	
Frisinger et al. (2023)					\checkmark		\checkmark			
Hesar et al. (2023)	\checkmark						\checkmark		\checkmark	
Keskenler et al. (2024)			\checkmark				\checkmark			
Khan et al. (2023)					\checkmark			\checkmark		
Konopik et al. (2023)			\checkmark					\checkmark		
Langebrake et al. (2023)		\checkmark								\checkmark
Mahajan et al. (2023)		\checkmark							\checkmark	
Mao et al. (2023)	\checkmark	\checkmark							\checkmark	
Messina et al. (2023)				\checkmark					\checkmark	
Munshi et al. (2024)	\checkmark						\checkmark			
Nasir et al. (2023)		\checkmark					\checkmark			
Orzechowski et al. (2023)					\checkmark				\checkmark	
Piaggio et al. (2024)				\checkmark						\checkmark
Qahtan et al. (2022)						\checkmark			\checkmark	
Rai et al. (2022)						\checkmark			\checkmark	
Semyonov et al. (2023)					\checkmark					\checkmark
Sobieraj et al. (2023)					\checkmark				\checkmark	
Sudeep et al. (2020)						\checkmark			\checkmark	
Tian et al. (2023)						\checkmark				\checkmark
Tortorella et al. (2022)					\checkmark				\checkmark	
Xia et al. (2023)					\checkmark					\checkmark
Zhiguo et al. (2023)			\checkmark				\checkmark			

Table 2. Papers evaluated

From another point of view, De Benedictis et al. (2022) present a detailed case study focusing on social distancing. This paper emphasizes the effectiveness of IoT devices and AI technology in generating virtual representations of actual healthcare entities and processes. The emphasis on social distancing and patient monitoring applications demonstrates the adaptability and usefulness of Digital Twins & Simulation in tackling current healthcare issues. Obstacles such as significant upfront costs, complexity in generating precise digital duplicates, and the requirement for intensive data gathering can impede widespread adoption. To overcome these obstacles, it will be necessary to make focused investments in research and development, with the goal of speeding the implementation of digital transformation and demonstrating its cost-effectiveness in achieving positive healthcare results over the course of time.

Studies like Das et al. (2023) concentrates on enhancing security in IoT-based smart healthcare systems by the introduction of a sophisticated lightweight privacy-preserving authentication approach. They focus on enhancing security in healthcare IoT devices by utilizing Physically Unclonable Functions (PUF) for device authentication. Additionally, Edeh et al. (2023) assess an IoT-enabled hybrid paradigm for genome sequence analysis, utilizing technologies such as Hadoop for distributed data processing and cloud computing for scalable storage. This study highlights the capability of IoT in aiding intricate healthcare analyses by showing how IoT can assist in genome sequencing, thereby enhancing tailored medicine and advanced patient care. Instead, Hesar et al. (2023) create an AI-powered epidermal electronic system specifically for automatically monitoring prognostic markers for hypertension through a smartphone. This device showcases the capacity of IoT to seamlessly interact with wearable technology, offering ongoing health monitoring via sensors placed on the patient's chest.

So, it is clear that also privacy and security of this sensible data is required and because of it, Qahtan et al. (2022) and Rai Bipin Kumar (2022) explores the adoption of Blockchain technology to guarantee securely manage of the patient's medical information. The first investigated a unique decision-making method that combines fuzzy weighted with zero inconsistency and spherical fuzzy sets to accurately assess evaluation criteria to tackle the intricate security and privacy challenges involved in combining blockchain and IoT technology in healthcare. The second one, presents PcBEHR, a blockchain-based system for Electronic Health Records (EHRs) that guarantees data decentralization, immutability, transparency, traceability, and trustworthiness by recommending decentralized interplanetary file storage (IPFS) to improve the security and accessibility of EHRs.

Artificial Intelligence and Machine Learning are crucial in healthcare for diagnostic systems, Keskenler et al. (2024) created a novel Multi-Layer Machine Learning (MLML) Architecture for diagnosing skin cancer non-invasively through dermoscopic pictures. The design utilizes methods such as Decision Tree, Random Forest, Neural Networks, Naive Bayes, Support Vector Machine, K-Nearest Neighbor, and Linear Regression to enhance the accuracy of medical diagnoses, demonstrating the promise of AI & ML in improving diagnostic precision. Indeed, Bhatt et al. (2024) and Gupta et al. (2024) examinate the useability of Machine Learning algorithms to specifically predict and identify heart strokes. Instead, Abidi et al. (2023) used a mix of different advanced deep learning classifiers such as Extreme Learning Machine (ELM), Deep Convolutional Neural Network (CNN), Long Short-Term Memory (LSTM), Deep Belief Network (DBN), and Deep Neural Network (DNN) to accurately and precisely categorize physical activities. This study demonstrates the use of AI and ML in creating intelligent health monitoring systems that can effectively categorize and track physical activity, aiding in tailored and proactive healthcare treatments.

A very important trend concerns other technologies such as 3D printing, which is revolutionizing personalized medicine and surgical training by enabling the creation of patient-specific drug dosages and anatomical models. The significance of 3D printing in personalizing patient care is highlighted by studies like that of Langebrake et al. (2023), which examine its application in machine-learning-assisted medication management systems. In this route, also Xia et al. (2023) examines the efficacy of 3D printed models in training for laparoscopic intracorporeal intestinal anastomosis, emphasizing their role in enhancing surgical performance and learning curves. The study shows that regular simulation training can greatly improve surgeons' surgical skills, indicating that 3D printed models are a practical and economical approach for surgical training and quality evaluation in healthcare.

5. Conclusions and Future Challenges

The healthcare industry is going through a significant digital transformation thanks to Industry 4.0 technologies, which introduces both new opportunities and challenges. The potential benefits, such as increased diagnostic precision, personalized treatment strategies, and improved patient outcomes through the integration of artificial intelligence, simulation and the Internet of Things and security data are significant.

However, these advantages are balanced by the pressing need to address several critical challenges.

Digital healthcare technologies promise to initiate a new era of medical treatment, offering unique opportunities to enhance patient outcomes, improve diagnostic accuracy, and enable faster interventions. Yet, the integration of these technologies into existing healthcare systems is fraught with complex challenges that require a strategic and thoughtful approach. One of the primary challenges is ensuring interoperability and compatibility among various systems and devices. Establishing universal communication protocols is essential for the efficient transmission of health data and the development of a cohesive digital healthcare ecosystem. However, this also raises concerns about data privacy and security, as the digitalization of health information exposes sensitive patient data to potential cyber-attacks. This underscores the importance of implementing advanced cybersecurity protocols and adhering strictly to data protection regulations to safeguard patient information, thereby ensuring trust and confidentiality in digital healthcare exchanges.

Given the trends, advancements, and identified gaps, it is clear that there is a pressing need for a comprehensive tracking tool that enables intelligent and remote diagnosis and prevention, securely stores data, and ensures easy accessibility while supporting the development of advanced treatments. The ANTHEM project emerges as a promising solution to these challenges. Its general objective is to bridge the existing gap in healthcare for fragile and chronic patients, particularly in regions and communities with high-incidence diseases and limited access to therapies. By focusing on the development of innovative sensors, advanced digital diagnostic systems, and monitoring technologies, as well as integrating cutting-edge therapies, ANTHEM aims to significantly enhance territorial healthcare for Non-Communicable Diseases (NCDs) and rehabilitation. This initiative leverages the latest advancements in Artificial Intelligence (AI), including Data Mining, ensuring a patient-centered approach that aligns with the evolving demands of modern healthcare.

While the path forward may be complex, the potential benefits are substantial. By directly addressing these challenges, the healthcare sector can harness Industry 4.0 technologies to improve precision, efficiency, and patient-centricity in healthcare delivery. In this context, ANTHEM could play a pivotal role in providing a crucial tool for the healthcare sector, especially for patients who require medical attention but struggle to access it easily.

Acknowledgements

The work has been co-funded by the CALHUBRIA project "CALabria HUB per Ricerca Innovativa ed Avanzata", project ID T4-AN09, CUP H53C22000800006, Piano Operativo Salute, Traiettoria 4, Azione 4.1 "Creazione di Hub delle Scienze delle Vita" Decreto MDS-DGPROGS-26 prot. N. 0014482 del 14/07/2022" OS 7.

References

- Abbate, S., Centobelli, P., Cerchione, R., Oropallo, E., & Riccio, E. (2023). Investigating Healthcare 4.0 Transition Through a Knowledge Management Perspective. IEEE Transactions on Engineering Management, 70, 3297-3310.
- Abidi, M.H., Umer, U., Mian, S.H., & Al-Ahmari, A.M. (2023). Big Data-Based Smart Health Monitoring System: Using Deep Ensemble Learning. IEEE Access, 11, 114880-114903.
- Akintayo-Usman, N.O. (2021). Fragmentation of care: a major challenge for older people living with multimorbidity. Geriatrics Gerontology and Aging.
- Alvarez-Romero, C., Martínez-García, A., Bernabeu-Wittel, M., & Parra-Calderón, C. L. (2023). Health data hubs: an analysis of existing data governance features for research. Health research policy and systems, 21(1), 70. <u>https://doi.org/10.1186/s12961-023-01026-1</u>
- Antonacci, G., Benevento, E., Bonavitacola, S., Cannavacciuolo, L., Foglia, E., Fusi, G., Garagiola, E., Ponsiglione, C., & Stefanini, A. (2023). Healthcare professional and manager perceptions on drivers, benefits, and challenges of telemedicine: results from a cross-sectional survey in the Italian NHS. BMC Health Services Research, 23.
- Bhatt H, Jadav NK, Kumari A, et al. Artificial neural network-driven federated learning for heart stroke prediction in healthcare 4.0 underlying 5G. Concurrency Computat Pract Exper. 2024; 36(3):e7911. doi: 10.1002/cpe.7911
- Brönneke, J. B., Herr, A., Reif, S., & Stern, A. D. (2023). Dynamic HTA for digital health solutions: opportunities and challenges for patient-centered evaluation. International journal of technology assessment in health care, 39(1), e72. https://doi.org/10.1017/S0266462323002726
- Brunzini, A., Peruzzini, M., & Barbadoro, P. (2023). Human-centred data-driven redesign of simulation-based training: A qualitative study applied on two use cases of the healthcare and industrial domains. Journal of Industrial Information Integration, 100505. 35, https://doi.org/10.1016/j.jii.2023.100505
- Bruzzone, A.G., Massei, M., Pernas-Álvarez, J., Reverberi, A. and Pedemonte, M. (2021) 'Improving Operational Procedures to Access Industrial Facilities and Urban Areas during Pandemics', 10th International Workshop on Innovative Simulation for Health Care, IWISH 2021, pp. 90–94. doi: 10.46354/i

Chatterjee, S., Chaudhuri, R., Gupta, S., Mangla, S., &

Kamble, S. (2023). Examining the influence of industry 4.0 in healthcare supply chain performance: Moderating role of environmental dynamism. Journal of Cleaner Production, 427, 139195.

https://doi.org/10.1016/j.jclepro.2023.139195

- Cimino, A., Gnoni, M. G., Longo, F., Diaz, R., Solis, A., Nervoso, A., Manfredi, K. A., & Diaco, M. (2022). Challenges and solutions for designing a Covid-19 vaccination hub: A simulation approach. 11th International Workshop on Innovative Simulation for Health Care, IWISH 2022. https://doi.org/10.46354/i3m.2022.iwish.006
- Das, S., Namasudra, S., Deb, P., Ger, M., & Crespo, R. G. (2023). Securing IoT-based smart healthcare systems by using advanced lightweight privacypreserving authentication scheme. IEEE Internet of Things Journal, 10(21), 18486-18494. https://doi.org/10.1109/JIOT.2023.3283347
- De Benedictis, A., Mazzocca, N., Somma, A., & Strigaro, C. (2022). Digital twins in healthcare: An architectural proposal and its application in a social distancing case study. IEEE Journal of Biomedical and Health Informatics, 1–12. https://doi.org/10.1109/JBHI.2022.3205506
- Edeh, U. K., Lilhore, P. S., Saurabh, S., Dalal, A. S., Nwaeze, A. S., Chijindu, A. T., Ndufeiya-Kumasi, L. C., & Simaiya, S. (2023). Evaluation of IoT-enabled hybrid model for genome sequence analysis of patients in healthcare 4.0. Measurement: Sensors, 26, 100679. https://doi.org/10.1016/j.measen.2023.100679
- Eduardo, S. C., Walter, C. S., Mauro, L. M., & José, C. C. (2023). Healthcare 4.0 implementation: Opportunities and challenges in the healthcare sector. Peer Review, 5(15), 163-182. https://doi.org/10.53660/749.prw1933
- Frisinger, A., & Papachristou, P. (2023). The voice of healthcare: Introducing digital decision support systems into clinical practice - A qualitative study. BMC Primary Care, 24, 67. https://doi.org/10.1186/s12875-023-02024-6
- Garg, S., Kaushal, R. K., Kumar, N., & Verma, A. (2023). Bibliometric analysis of blockchain in the healthcare domain. Intelligent and Converged Networks, 4, 305-312. https://doi.org/10.23919/ICN.2023.0025
- Hesar, N. S. S., Khan, D., Naghashian, A., Piekarski, M., Gall, H., Schermuly, R., Ghofrani, H. A., & Ingebrandt, S. (2023). AI-enabled epidermal electronic system to automatically monitor a prognostic parameter for hypertension with a smartphone. Biosensors and Bioelectronics, 241, 115693. https://doi.org/10.1016/j.bios.2023.115693
- Javaid, M., Haleem, A., Vaishya, R., Bahl, S., Suman, R., & Vaish, A. (2020). Industry 4.0 technologies and

their applications in fighting COVID-19 pandemic. Diabetes & Metabolic Syndrome, 14, 419-422. <u>https://doi.org/10.1016/j.dsx.2020.04.032</u>

- Kern, L., Safford, M., Slavin, M., Makovkina, E., Fudl, A., Carrillo, J., & Abramson, E. (2019). Patients' and providers' views on causes and consequences of healthcare fragmentation in the ambulatory setting: A qualitative study. Journal of General Internal Medicine, 34, 899-907. https://doi.org/10.1007/s11606-019-04859-1
- Keskenler, M. F., Çelik, E., & Dal, D. (2024). A new multi-layer machine learning (MLML) architecture for non-invasive skin cancer diagnosis on dermoscopic images. Journal of Electrical Engineering & Technology. https://doi.org/10.1007/s42835-023-01758-8
- Khan, A., Alzaabi, Z., Iqbal, T., Ratnarajah, T., & Arslan, T. (2023). A novel digital twin (DT) model based on WiFi CSI, signal processing and machine learning for patient respiration monitoring and decision– support. IEEE Access, 11, 103554-103568. https://doi.org/10.1109/ACCESS.2023.3316508
- Konopik, J., Wolf, L., & Schöffski, O. (2023). Digital twins for breast cancer treatment – An empirical study on stakeholders' perspectives on potentials and challenges. Health Technology, 13, 1003–1010. https://doi.org/10.1007/s12553-023-00798-4
- Langebrake, K. G., Dadkhah, A., Eggert, J., Gutowski, T., Rosch, M., Schönbeck, N., Gundler, C., Nürnberg, S., Ückert, F., & Baehr, M. (2023). Patient-individual 3D-printing of drugs within a machine-learningassisted closed-loop medication management – Design and first results of a feasibility study. Clinical eHealth, 6, 3-9. https://doi.org/10.1016/j.ceh.2023.05.001
- Mahajan, H. B., & Junnarkar, A. A. (2023). Smart healthcare system using integrated and lightweight ECC with private blockchain for multimedia medical data processing. Multimedia Tools and Applications, 82, 44335-44358. https://doi.org/10.1007/s11042-023-15204-4
- Mao, P., Zhou, X., Wang, H., Yao, L., Liang, Y., Zhao, J., Zhang, J., Ban, D., & Zheng, H. (2023). A health monitoring system based on flexible triboelectric sensors for intelligence medical internet of things and its applications in virtual reality. Nano Energy, 118(Part B), 108984. https://doi.org/10.1016/j.nanoen.2023.108984
- Messina, L., Maugeri, G., Spoto, R., Puccio, M., Ruggieri, M., & Petralia, S. (2023). Fully integrated point-of-care platform for the self-monitoring of phenylalanine in finger-prick blood. ACS Sensors, 8(11), 4152-4160. https://doi.org/10.1021/acssensors.3c01384
- Mishra, S. G. (2023). Internet of Medical Things healthcare for sustainable smart cities: Current

status and future prospects. Applied Sciences, 13(15), 8869. <u>https://doi.org/10.3390/app13158869</u>

- Mohamed, N., & Al-Jaroodi, J. (2019). The impact of industry 4.0 on healthcare system engineering. 2019 IEEE International Systems Conference (SysCon), 1– 7. <u>https://doi.org/10.1109/SYSCON.2019.8836715</u>
- Munshi, R., Gupta, R., Jadav, N. K., et al. (2024). Quantum machine learning-based framework to detect heart failures in Healthcare 4.0. Software: Practice and Experience, 54(2), 168-185. https://doi.org/10.1002/spe.3264
- Shrivastava, M., Chugh, R., Gochhait, S., & Jibril, A. (2023). A review on digital twin technology in healthcare. 2023 International Conference on Innovative Data Communication Technologies and Application (ICIDCA), 741-745. https://doi.org/10.1109/ICIDCA56705.2023.100996 46
- Nasir, T. A., Abuhmed, S., Mirjalili, S., El-Sappagh, S., & Muhammad, K. (2023). Time-series visual explainability for Alzheimer's disease progression detection for smart healthcare. Alexandria Engineering Journal, 82, 484-502. https://doi.org/10.1016/j.aej.2023.09.050
- Nguyen, H.-S., Danh, H.-C., Ma, Q.-P., Mesicek, J., Hajnys, J., Pagac, M., & Petru, J. (2023). A bibliometrics analysis of medical Internet of Things for modern healthcare. Electronics, 12(4586). https://doi.org/10.3390/electronics12224586
- Orzechowski, T., Skuban-Eiseler, A., Ajlani, A., Lindemann, U., Klenk, J., & Steger, F. (2023). User perspectives of geriatric German patients on smart sensor technology in healthcare. Sensors, 23(22), 9124. <u>https://doi.org/10.3390/s23229124</u>
- Piaggio, R., Castaldo, G., Garibizzo, E., Iadanza, E., & Pecchia, L. (2024). A smartphone-based tool for screening diabetic neuropathies: A mHealth and 3D printing approach. Biomedical Signal Processing and Control, 89, 105807. https://doi.org/10.1016/j.bspc.2023.105807
- Qahtan, F., Skuban-Eiseler, T., Ajlani, A., Lindemann, U., Klenk, J., & Steger, F. (2022). Novel multi security and privacy benchmarking framework for blockchain-based IoT healthcare industry 4.0 systems. IEEE Transactions on Industrial Informatics, 18(9), 6415-6423. https://doi.org/10.1109/TII.2022.3143619
- Rai, B. K. (2022). Blockchain-enabled electronic health records for healthcare 4.0. International Journal of E-Health and Medical Communications, 13(15), 1-13. <u>https://doi.org/10.4018/IJEHMC.309438</u>
- Semyonov, A., Bogdan, E., Shamal, E., Sargsyan, A., Davtyan, K., Azzopardi-Muscat, N., & Novillo-Ortiz, D. (2023). Digital health information systems in the member states of the Commonwealth of Independent States: Status and prospects. Digital,

3(3), https://doi.org/10.3390/digital3030013 189-199.

- Shams, F., Forruque, M. S. B. A., Alam, S. A., Afrin, S., Rafa, S. J., Rafa, N., & Gandomi, A. H. (2024). Insights into Internet of Medical Things (IoMT): Data fusion, security issues and potential solutions. Information Fusion, 102, 102060. https://doi.org/10.1016/j.inffus.2023.102060
- Sobieraj, S. E., & Eimler, G. R. (2023). Can smart glasses change how people evaluate healthcare professionals? A mixed-method approach to using smart glasses in hospitals. International Journal of Human-Computer Studies, 178, 103081. https://doi.org/10.1016/j.ijhcs.2023.103081
- Sudeep, K., & Parekh, R. (2020). Blockchain-based electronic healthcare record system for healthcare 4.0 applications. Journal of Information Security and Applications, 50, 102407. https://doi.org/10.1016/j.jisa.2019.102407
- Sun, T., He, X., & Li, Z. (2023). Digital twin in healthcare: Recent updates and challenges. Digital Health, 9.

https://doi.org/10.1177/20552076221149651

- Tian, H., Ye, W., Wang, J., Quan, H., & Chang, C.-C. (2023). Certificateless public auditing for cloudbased medical data in healthcare industry 4.0. International Journal of Intelligent Systems. https://doi.org/10.1155/2023/3375823
- Tortorella, F. S., Fogliatto, S., Kurnia, S., Thürer, M., & Capurro, D. (2022). Healthcare 4.0 digital applications: An empirical study on measures, bundles and patient-centered performance. Technological Forecasting and Social Change, 181, 121780.

https://doi.org/10.1016/j.techfore.2022.121780

- Tortorella, G., Saurin, T., Fogliatto, F., Rosa, V., Tonetto, L., & Magrabi, F. (2021). Impacts of Healthcare 4.0 digital technologies on the resilience of hospitals. Technological Forecasting and Social Change, 166, 120666. https://doi.org/10.1016/J.TECHFORE.2021.120666
- Xia, J. M., Wu, J., Chen, H., Mao, J., Xu, X., Zhang, J., Yang, J., & Wang, Z. (2023). Assessment of laparoscopic intracorporeal intestinal anastomosis training using simulation-based 3D printed models: Exploring surgical performance and learning curves. International Journal of Surgery, 109(10), 2953-2961.

https://doi.org/10.1097/JS9.000000000000582

Zhiguo, Y., Li, P., & Tiwari, P. (2023). QNMF: A quantum neural network based multimodal fusion system for intelligent diagnosis. Information Fusion, 100, 101913. https://doi.org/10.1016/j.inffus.2023.101913