

Machine Learning for Personalized Medicine: Transforming Patient Care with Artificial Intelligence

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ABSTRACT

Machine learning revolutionizes personalized medicine by facilitating personalized healthcare based on data. This review goes through its contribution to the shift from conventional therapies to precision medicine, by incorporating genomic, clinical, imaging and wearable data. Supervised learning, Unsupervised learning, deep learning - all of these machine learning approaches are applicable to some of the medical uses of machine learning such as disease prediction, treatment optimization, drug discovery, and clinical decision support. While there has been considerable progress, there are still issues to be addressed, including data privacy, bias, interpretability, and regulatory compliance. Ethical and legal issues also are important to implementation. The future of health care delivery is anticipated to become even more accurate and individualized, with digital twins, federated learning, and real-time AI systems all on the horizon.

INTRODUCTION

Personalized medicine is a radical departure from the current one size fits all medicine by a shift from treatment plans to individual patients. In this way, variability in genes, environment, lifestyle, as well as clinical history is taken into account to make more effective prevention, diagnosis, and treatment



plans. The aim is to optimize the outcome of treatment, minimize adverse effects and avoid excessive health care costs [1]. The past few years have seen fast developments in personalized medicine, driven by the rapid progress of data science and computational technologies.

One of the most important technologies driving this change is machine learning (ML), a fundamental part of artificial intelligence (AI). The main advantage of using machine learning over traditional statistics is the ability to detect complex and non-linear patterns in large and diverse data sets [2]. In the healthcare sector, these data sets can encompass electronic health records, genomic sequences, medical imaging, lab results, and even information from wearable devices. The capability of machine learning models of learning from this huge and diversified information makes them a very good fit for supporting individualized medical decision-making [3].

AI is revolutionizing personalized medicine by enabling doctors to better understand diseases and patients. Predictive models can for instance evaluate the likelihood that a patient will develop chronic diseases, such as diabetes, heart disease or cancer, before the important events even take place. Likewise, classification algorithms can help to diagnose diseases more accurately, based on imaging data or genetic markers [4]. In oncology, ML is being utilized more and more to pinpoint the most appropriate treatments for particular tumor profiles, which allows for precision treatment strategies. A key part of this transformation is the increasing accessibility to big data in healthcare. The availability of electronic health records and the development of sequencing technology has created huge amounts of patient information [5].

Machine learning methods are key to gleaning useful information from this data and to transforming it into actionable clinical knowledge. Further, continuous data collection (using wearable devices and mobile health applications) enables real-time monitoring of patient health, further improving personalized care. With great potential, machine learning also poses challenges for personalized medicine [6]. These are issues of data privacy, model interpretability, algorithmic bias, and integration into clinical workflows that must be carefully considered. However, the ongoing research and technology developments are gradually breaking down these barriers.

Machine learning and the personalized medicine are revolutionizing the healthcare industry into a more predictive, preventive and patient-centric healthcare system. The article examines the principles, use cases, and difficulties of ML in personalized medicine and its eventual transformative impact on patient care. This article delves into the principles, use cases, and challenges of ML in personalized medicine and its future implications for patient care.

EVOLUTION OF PERSONALIZED MEDICINE

Personalized medicine has come a long way in history and has been a result of advances in biology, technology, and data science. Historically, treatment for disease has been general or “one size fits all”, meaning that patients with the same disease received a similar treatment despite individual variations. This strategy worked well for many acute diseases, but did not always take into account the individual variability of patients, particularly in the context of chronic and complex diseases [7]. The challenges of traditional medicine gave rise to a new approach, ultimately personalized medicine.

The roots of personalized medicine can be traced back to the very first recognition that people respond differently to drugs and diseases based on their genetic differences. Some of the earliest steps in this progression were the creation of pharmacogenomics, the study of how the genes of a person influence their responses to drugs [8]. This field has proven that there is variation in the genes that affect drug efficacy and toxicity, and could enable more personalized drug treatment. For instance, some genetic markers were determined to affect the metabolism of patient drugs like warfarin, and dosage adjustments made according to genetic profile [9].

Now, with the completion of the Human Genome Project in the early 2000s, personalized medicine has ushered in a new era. The capacity to sequence the whole human genome gave scientists a new way of understanding genetic differences that could be linked to different diseases. This finding enabled researchers to pinpoint genes that make people susceptible to disease and create therapies that target those genes [10]. With the advent of increasingly accessible and cheaper sequencing technologies, genomic information started to become more and more significant in clinical research and healthcare choices.

Simultaneously, the ability of analyzing large-scale biological data became possible with the developments in bioinformatics and computational biology. They were key elements of making sense of raw genetic data into clinical information. As time went on, personalized medicine became more complex because genetics were not the only factors, and included proteomics and metabolomics, plus environmental factors, which gave greater insight into patient health [11]. This continued with the use of digital health technology. Patient data gathered from electronic health records, wearable sensors and mobile health applications started to flow. This transformation allowed for a more flexible and dynamic healthcare experience, with real-time patient monitoring and treatment adjustments made as information was analyzed [12].

Personalized medicine has been further evolved to precision medicine in recent years by the emergence of artificial intelligence and machine learning. These technologies enable the analysis of very complex data sets which contain genetic, clinical and lifestyle data. This has enabled healthcare

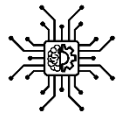
systems to make more accurate predictions of disease risk, more tailored treatment plans and better clinical outcomes [13]. The development of personalized medicine shifts the current view from blanket treatment strategies to highly individualized and data-driven health care systems.

FUNDAMENTALS OF MACHINE LEARNING IN HEALTHCARE

Machine learning (ML) is a subfield of AI where computers learn from data to make predictions and decisions without having to be specifically programmed for each situation. In medical field, machine learning algorithms can be used to analyze large and high-dimensional medical data sets, which can be helpful for diagnosis, prognosis and treatment planning. To gain insight into its applications in personalized medicine, you must grasp its basics [14].

Machine learning is based on algorithms trained with datasets in its core. These algorithms are able to find patterns, relationships, and structures in the data, which can then be used to make predictions on new or unseen data. In healthcare, this could involve patient data, clinical records, lab test results, imaging scans, and genomic data. This is a rich source of information that is valuable in clinical settings, where ML models can process and interpret this diversity. There are 3 main machine learning techniques: Supervised, Unsupervised, and Reinforcement [15]. Supervised learning is the most commonly used in healthcare. In this method, models are trained using labeled datasets, where the data is correlated with predetermined results. For instance, in a supervised learning model, the model can be trained to predict if a patient has a disease or not from the past diagnostic data. Examples of common algorithms are decision trees, support vector machines and neural networks [16].

Supervised learning, however, is used with labeled data. It is used to discover hidden patterns or groupings within datasets. Unsupervised learning in healthcare can be used to stratify patients, the process of assigning patients to groups based on their similarities in terms of symptoms, genetics, or treatment responsiveness [17]. In this context, clustering methods like k-means and hierarchical clustering are popular. Though less popular, technique used in healthcare is reinforcement learning. This is done by rewarding/penalizing students in a system and encouraging them to learn the best decision making process over time [18]. Such an approach is especially helpful in the treatment planning and adaptive therapy optimization process where decisions need to be made based on the patient's response.



Data Collection

- Clinical Records
- Imaging Scans
- Lab Results
- Genomic Data
- Patient Monitoring Data



Data Preprocessing

- Data Cleaning
- Normalization
- Feature Engineering



Machine Learning Techniques

Supervised Learning

- Uses labeled data
- Predicts outcomes (e.g., disease classification)

Unsupervised Learning

- Finds hidden patterns
- Patient grouping / clustering

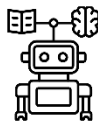
Reinforcement Learning

- Learns via reward and penalty
- Used for adaptive treatment decisions



Model Training & Evaluation

- Train the model (neural network icon)
- Cross-validation
- Evaluate performance
- Metrics:
 - Accuracy
 - Sensitivity
 - Specificity
 - AUC (graph shown)



Clinical Applications

- Disease Diagnosis
- Patient Stratification
- Treatment Planning
- Monitoring & Prognosis



Key Benefits

- Improved Diagnostic Accuracy
- Personalized Medicine
- Early Detection
- Cost Efficiency
- Better Patient Outcomes

Figure 1. Role of Machine learning in healthcare

Model training and validation is another critical factor of machine learning in healthcare. It is important for models to be trained on high-quality data and thoroughly tested to verify accuracy, reliability and generalizability. Techniques such as cross-validation and performance metrics like accuracy, sensitivity, specificity, and area under the curve (AUC) are commonly used to evaluate model performance [19]. Machine learning in healthcare has its own set of challenges, including data quality problems, the lack of interpretability in models, and the integration of machine learning into clinical practice. ML, however, is becoming more and more useful and practical with the constant improvements in algorithms and computing power [20]. The basics of machine learning are essential for its use in healthcare and allow systems to learn from data to make more efficient and accurate medical decisions with personalization.

DATA SOURCES IN PERSONALIZED MEDICINE

Personalized medicine is built on data, and the power of machine learning models is dependent on the quality, diversity and completeness of the data they are powered by. Multiple data sources are used in personalized medicine to provide a complete picture of an individual's patient. These sources of data provide biological, clinical, behavioral and environmental data, which allow better prediction, diagnosis and treatment planning [21].

One of the most important data sources is Electronic Health Records (EHRs). Structured and

unstructured clinical data, including patient demographics, medical history, diagnoses, prescriptions, lab test results and physician notes are entered in EHRs. EHRs are longitudinal, meaning they are able to capture a patient's health history over time [22]. These data can be used to train machine learning models that can detect patterns in the disease, forecast hospital readmission rates, and assist in clinical decision support. But data items in EHRs are frequently incomplete and may be erroneous, necessitating careful data preprocessing [23].

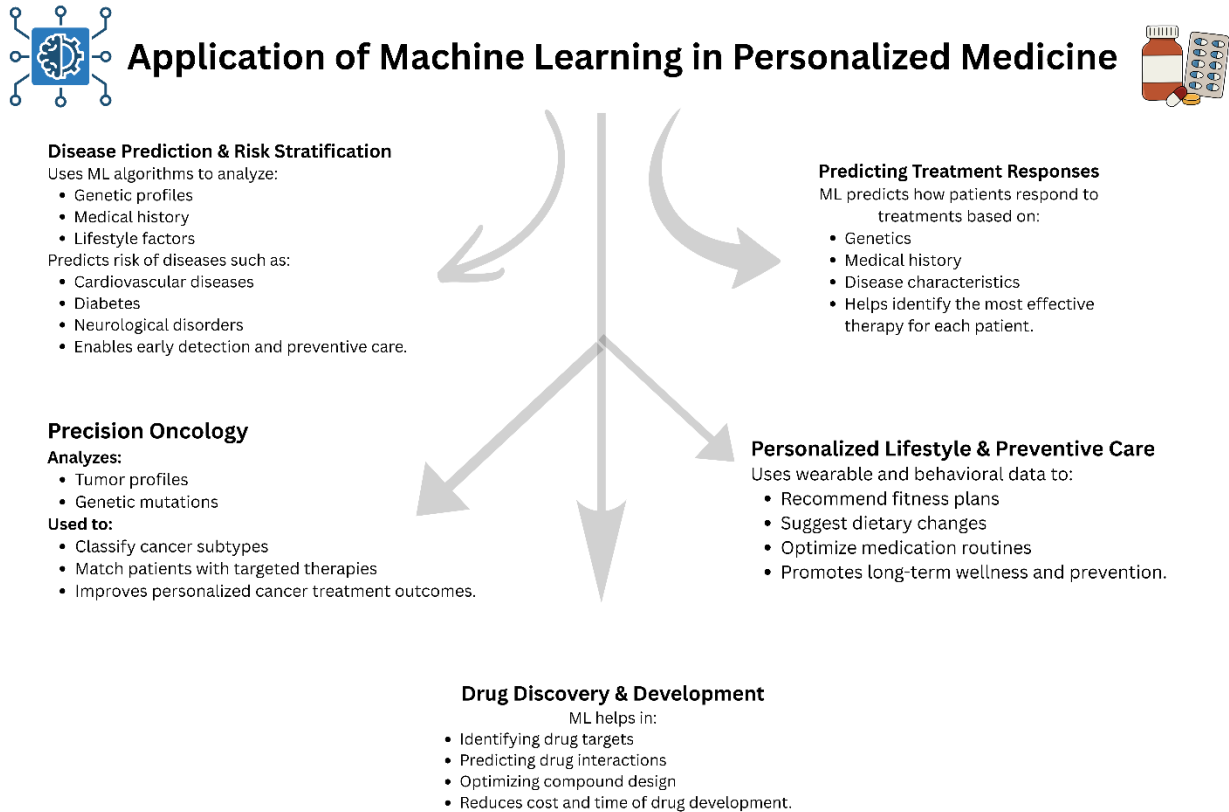


Figure 2. Application of Machine Learning in Personalized Medicine

Genomic and proteomic data are other critical information sources. Genomic data refers to the DNA sequence, gene expression and genetic variation of a person, while proteomic data is about the proteins that are produced in the body. These datasets will play a crucial role in elucidating disease mechanisms at a molecular level. In the field of personalized medicine, genomic information is being utilized in various sectors, including oncology, where information on the genetic mutations linked to cancer can direct customized treatment strategies [24]. Machine Learning models can assist in uncovering connections between genes and the likelihood of disease, facilitating tailored treatment approaches.

Another crucial aspect of personalized healthcare is medical imaging data. Methods like X-rays, CT scans, MRI and ultrasound produce a tremendous amount of visual data. These images are well suited for analysis by deep learning models, especially convolutional neural networks, which are highly

effective for detecting tumors, segmenting organs and classifying diseases. Imaging is an important tool in the early diagnosis and monitoring of treatment [25]. Wearable devices and mobile health technologies have become more and more significant sources of data. Smart watches and fitness trackers gather real-time physiological information, such as heart rate, activity, sleep quality and oxygen levels, on a constant basis [26]. This constant flow of information can be used to monitor a patient's health in real-time and detect potential health problems.

Real-world patient-generated data, such as lifestyle information, diet, environmental exposure, and social determinants of health, further add to personalized medicine. This kind of information can give a broader picture beyond the clinical context, leading to a more comprehensive understanding of the patient's health [27]. The incorporation of various data sources, including EHRs, genomic data, medical imaging, wearable devices, and lifestyle information, is crucial for the effective application of machine learning in personalized medicine. These data streams help create more accurate, predictive and personalized healthcare [28].

APPLICATION OF MACHINE LEARNING IN PERSONALIZED MEDICINE

In personalized medicine, machine learning is revolutionizing the way healthcare systems deliver individualized care strategies from a one-size-fits-all approach. Machine learning models can process large and complex datasets to identify patterns and make more accurate predictions of health outcomes, which can help inform clinical decisions [29]. They are utilized in several aspects of healthcare, such as disease forecasting, optimizing treatment, drug development, and specialized fields like oncology.

Disease prediction and risk stratification is one of the most important applications. Using machine learning algorithms to analyze patient data like lifestyle habits, genetic profile and medical history to predict the risk of developing some conditions. For instance, a model can forecast the likelihood of cardiovascular diseases, diabetes or neurological disorders before they cause symptoms [30]. This enables health care providers to take preventive actions early, which will enhance the health of their patients and save on future health care expenses. Risk stratification also aids in prioritizing high-risk patients for closer monitoring and timely intervention [31].

One of the other important applications is predicting treatment responses. Biological differences mean that not every patient will respond to the treatment in the same way. Machine learning models are used to find the most effective therapies for individual patients, depending on their genes, medical history, and nature of the disease [32]. This is particularly relevant in chronic diseases and cancer treatment, where the choice of therapy can make a big difference in the likelihood of survival and quality of life.

One of the most significant applications is in the field of precision oncology, where machine learning algorithms can interpret genetic mutations and tumor profiles to create personalized cancer treatments. However, in certain diseases like Cancer, the genetic changes can be quite different from one patient to another, even among those with the same type of cancer [33]. Machine learning plays a role in the classification of tumor subtypes and matching patients with therapies most likely to be effective for a specific patient's condition. Machine learning can aid in making personalized recommendations for lifestyle and preventive care. Algorithms can process the data from these wearable devices and patient behaviors to make personalized fitness recommendations, dietary adjustments, and medication routines [34].

Applications of machine learning in personalized medicine are abundant and rapidly growing. By enabling earlier disease detection, more targeted drug discovery, and personalized cancer therapy, all of these technologies are contributing to a shift in healthcare toward a more proactive, precise, and patient-friendly approach [35].

DEEP LEARNING IN CLINICAL DECISION SUPPORT

Deep learning, one of the more specialized forms of machine learning, has greatly contributed to the development of the clinical decision support system. Deep learning models, particularly neural networks, can automatically extract hierarchical representations from raw data, in contrast to the regular machine learning techniques that depend on handcrafted features [36]. Their ability to handle such large-scale, unstructured data makes them especially valuable in the medical field, where data types like physiological signals, clinical notes, and medical images are prevalent [37].

Medical imaging analysis is one of the most promising areas of deep learning for clinical decision making. Many radiological images, including X-rays, computed tomography (CT) images and magnetic resonance imaging (MRI) images, are interpreted using convolutional neural networks (CNNs). They can also identify subtle changes that might not be visible to the naked eye, like the onset of tumors or microfractures, or early signs of neurological disorders [38]. Deep learning has demonstrated comparable or superior performance to human experts in numerous medical applications, such as those in radiology and pathology fields, enhancing the efficiency and accuracy of clinical workflows.

The use of deep learning in the analysis of electronic health records (EHRs) is also widespread. Physician notes, lab results and patient histories are often included in clinical data within EHRs and are largely unstructured. Natural language processing (NLP), which is a subcategory of deep learning, can be used to help uncover insights from this textual data [39]. For instance, symptoms can be modeled and the progression of the disease can be detected using these NLP models, which can also

help to generate clinical summaries, thus saving the time of healthcare workers and minimizing documentation errors [40].

Another significant use case is predictive analytics of patient outcomes. Heart rate, blood pressure and glucose levels are time-series data that can be processed using recurrent neural networks (RNNs) and transformer based models. These models can forecast critical events like patient deterioration, onset of sepsis, and hospital readmission risk. This means that clinicians can intervene earlier and better inform their decisions about the care of their patients [41]. In the case of intensive care units (ICUs), where patient data is constantly collected, deep learning can be used for real-time decision making. Deep learning systems can combine various data sources, like vital signs and lab test results, to alert and give suggestions to clinicians, enhancing response times during emergencies [42].

Deep learning offers promising prospects for clinical decision support, it also comes with drawbacks, including the inability to explain its models, the need for a large number of labeled data, and high computational demands. But there are continuous improvements in explainable AI that are helping to overcome these drawbacks, making the decisions made by the model more transparent and trustworthy [43].

INTEGRATION OF AI INTO CLINICAL WORKFLOWS

The use of AI, especially machine learning systems, within clinical workflows is a pivotal move towards revolutionizing healthcare delivery, making it more efficient, data-driven, and patient-centric. AI models have proven themselves to be quite effective in research settings, but only once they are being used in the day-to-day clinical routine. This integration will enable access to predictive insights and decision-support tools at the point of care for healthcare professionals [44].

One of the primary areas of integration is decision support systems in hospitals. AI-powered clinical decision support systems (CDSS) support doctors in decision-making by analyzing patient information and giving evidence-based suggestions. These systems can flag potential diagnoses, suggest appropriate diagnostic tests, and recommend treatment options based on clinical guidelines and patient-specific information [45]. For instance, if a patient comes to the clinic with more complicated symptoms, an AI system can quickly process their EHR and point to any potential conditions that might not be suspected. This enhances the diagnostic accuracy and lightens the cognitive load of clinicians [46].

AI-driven diagnostics products are also a critical application. In radiology, pathology and cardiology, these are being used more and more to assist in interpreting images and diagnosing disease. For example, AI can automatically interpret medical scans and identify areas of concern that need to be further examined [47]. This not only accelerates the diagnostic process, but also improves consistency

of interpretation, thus minimizing variability from clinician to clinician. In medical writing, AI can help generate accurate and effective content, including reports and articles on medical research, product labels, and patient education materials, saving valuable time and resources [48]. In medical writing, AI can also be used to create accurate and effective medical content, such as reports and articles on medical research, product labels, and patient education materials, which can save medical writers a lot of time and resources.

AI is also important in optimizing workflows in healthcare systems. Some common issues in hospitals include patient overcrowding, limited resources, and poor patient scheduling. By leveraging the insights gained from analyzing hospital operations data, machine learning models can help optimize staffing resources and make better predictions about patient admission numbers, which can then be used to schedule appointments more efficiently. This results in shorter waiting times, better patient flow and better use of hospital resources [49].

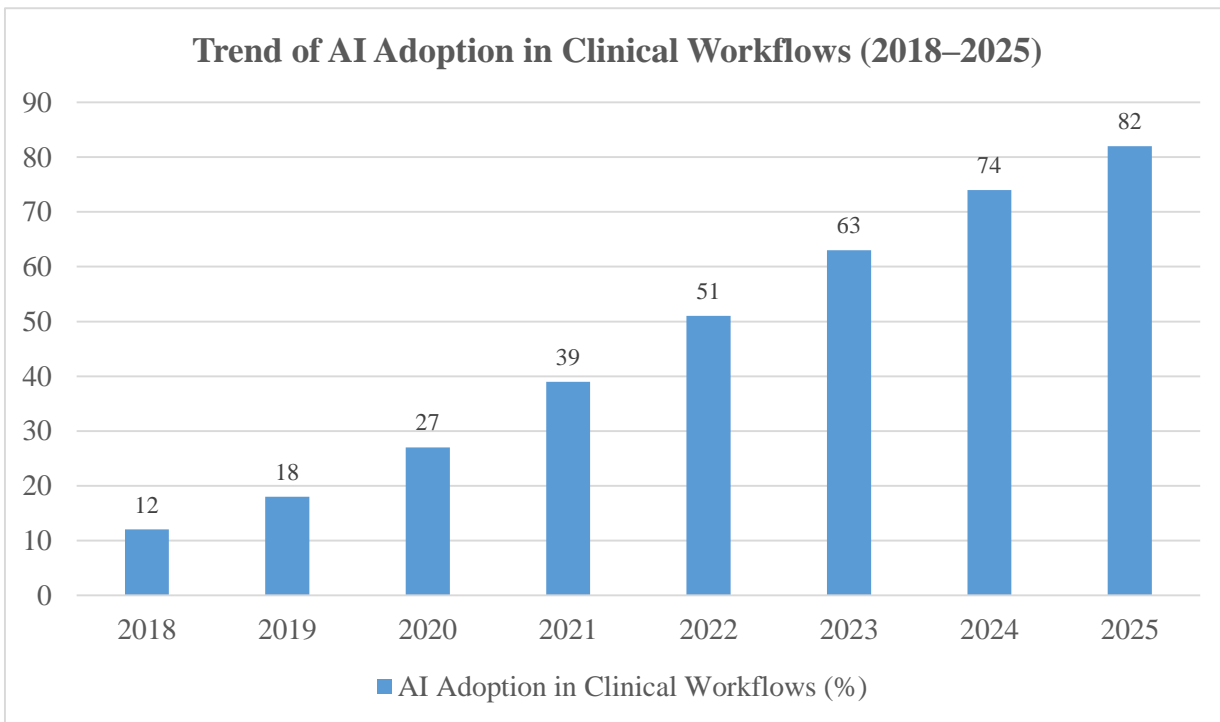


Figure 3. Trend of AI Adoption in Clinical Workflows (2018–2025)

The use of AI enables real-time patient monitoring, which is particularly crucial in intensive care units and emergency rooms. AI systems have the capacity to continuously monitor vital signs and clinical data, which can help detect early signs of deterioration, allowing healthcare providers to take proactive action. This proactive approach helps to ensure patient safety and better clinical results. Using AI in clinical practice, however, comes with its own set of obstacles [50]. There are a number of challenges to overcome, including interoperability with current hospital systems, security, regulatory compliance, and user trust. But it's not just AI tools that need to be mastered healthcare

professionals also need training to use AI tools and understand their results.

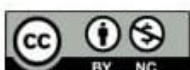
CHALLENGES AND LIMITATIONS

Although the field of personalized medicine is rapidly advancing with the potential for a wide range of applications, several challenges and limitations still remain in the widespread use of machine learning in clinical practice. They involve technical, ethical, operational, and regulatory challenges, and resolving these is crucial for the implementation of healthcare systems that are safe and effective. Data privacy and security is one of the most crucial issues that has to be addressed [51]. The healthcare industry has data which is very sensitive in nature, including personal, genetic and medical information. However, machine learning systems need considerable data to be trained, and there is a risk of data breaches or unauthorized access [52]. Compliance with data protection laws and maintaining patients' privacy is a significant issue. Data can still be re-identified, particularly by merging multiple data sets, even after anonymization.

Bias and fairness in AI models is another major drawback. Patterns learned from past data via machine learning algorithms can perpetuate existing bias in healthcare systems. For instance, if one ethnic group or population is under-represented in the training data, its predictions will be inaccurate. This can lead to inequity in treatment outcomes and pose significant ethical issues with personalized medicine. Interpretability and explain ability of the model is also a big challenge [53]. A significant number of advanced machine learning methods, and especially deep learning models, work as “black boxes” in that the human cannot easily fathom how they make a decision. Healthcare professionals need to have trust and understanding in the recommendations generated by AI before incorporating them into patient care in clinical settings [54]. If clinicians are not confident, or if it is not used in a real world setting, it can limit uptake. Lack of transparency can decrease clinician confidence and limit uptake in real world environments.

Integration into clinical workflows is another barrier. Many healthcare systems and hospitals are using antiquated systems that might not integrate well with today's AI solutions. Healthcare organizations and hospitals may be using legacy systems that are not compatible with modern AI tools. The transition to machine learning systems demands a lot of change in the workflow design, training of employees, and IT environment. Even a very accurate model can be of little use if it is not well integrated [55]. The deployment of AI in healthcare is further hindered by regulatory and legal barriers. There are strict regulations that medical AI systems must meet to ensure their safety and effectiveness. But, the regulation is still in a process of change and accountability is hard to come by if the AI system fails to predict or recommend correctly [56].

The lack of quality and large-scale datasets is a constraint. Healthcare systems often have limited,



disorganized, and inaccurately labeled data, making it difficult to train machine learning models. machine learning has shown promising potential in personalized medicine, but it is crucial to address important issues of privacy, bias, interpretability, integration, regulation, and data quality to ensure that machine learning is applied in a safe, fair, and effective way in clinical practice [57].

FUTURE PERSPECTIVES

The future of personalized medicine using machine learning looks promising, as new developments are likely to continue transforming healthcare into a more predictive, preventive, and patient-centered system. With greater computational power and more integrated healthcare data comes greater potential for how machine learning will be used in the diagnosis, treatment and management of disease. The emphasis is no longer on reactive care but on proactive and adaptive healthcare systems [58].

The development of fully personalized treatment system is one of the most important future directions. For these systems, machine learning algorithms will process a patient's genes, medical records, dietary habits and other physiological data continuously to provide highly personalized treatment suggestions. This will enable practitioners to tailor therapies over time to be more effective and safe, depending on the patient's response [59]. One new trend is the application of digital twins in the healthcare sector. A digital twin is an abstract model of a patient that mimics the patient's biology and physiology. Digital twins can be combined with machine learning algorithms and real-time data to anticipate patient responses to various therapies or disease progression [60]. This enables physicians to trial treatment options virtually without risk, and with a high level of accuracy, in real-life scenarios.

It is also anticipated that developments in federated learning will be vital to the future of personalized medicine. In federated learning, sensitive patient data is not sent to a central server to train machine learning models. This method will enable training of large models over a variety of datasets while maintaining data privacy [61]. It may be especially helpful to enhance model generalizability across populations and healthcare systems. The other exciting news is Edge AI. Edge AI brings the power of AI directly to the end-user's device, like wearable sensors or a mobile health application, and can help lower latency and enable real-time health monitoring. This can be particularly useful in acute situations where there's a need for rapid response [62].

Integration with genomics and precision oncology will also continue to grow. The use of machine learning to identify rare mutations in genes and develop targeted therapies for complex diseases like Cancer will become more common and increase the chances of survival and decrease side effects. But progress toward this improvement will necessitate the solution of problems of data

standardization, regulatory approval, ethical issues, and system interoperability [63]. It will require a partnership between clinicians, data scientists, policymakers, and those working in technology to implement it successfully. The future of personalized medicine with machine learning promises highly adaptable, intelligent and patient-centric healthcare systems. These advancements will have a profound impact on the landscape of medical care, providing more accurate insights into disease, more precise treatment, and greater efficiency in health care [64].

CONCLUSION

Healthcare systems are being revolutionized by machine learning, shifting from a one-size-fits-all treatment strategy to highly personalized and data-driven care. In each of the areas covered in this review, it is clear that the use of artificial intelligence (AI) in the healthcare sector has opened up new possibilities for enhancing diagnosis, treatment planning, disease prediction, and managing patients. As the fusion of sophisticated algorithms and a wide array of medical data sources unfolds, it heralds a fresh era in healthcare, one defined by precision.

The shift towards personalized medicine represents a gradual shift from population-based treatment strategies towards genomics-based and increasingly AI-assisted clinical decision making. The Human Genome Project and early advances in pharmacogenomics were the beginnings of understanding of individual variability in disease and in response to treatment. In the age of machine learning, this base has been greatly reinforced, and can now interpret very large and complex data sets that could not be interpreted well before now.

Healthcare applications rely primarily on machine learning principles, supervised, unsupervised and reinforcement learning. These techniques allow systems to identify patterns in electronic health records, genomic information, imaging and wearable devices. The seamless combination of these data sources promotes a more comprehensive view of patient health, enabling early detection and tailored intervention strategies.

This shows the diverse use cases of machine learning in healthcare. AI systems are now playing a crucial role in various medical applications, including disease risk prediction, treatment response modeling, drug discovery, and precision oncology. One of the key areas where deep learning can make a significant difference is in medical image analysis, extracting information from clinical notes and real-time patient monitoring, which all improve clinical decision support.

But, the adoption of AI in clinical workflows comes with its challenges. Data privacy, algorithmic bias, interpretability, and regulatory challenges are just some of the issues that need to be resolved for the safe and ethical use of AI tools. Likewise, issues of ethics and legality around patient consent, accountability and data ownership underscore the need for strong governance structures.

Next steps include further innovation in personalized medicine including digital twins, federated learning and edge AI. These innovations will create real-time, secure and highly accurate healthcare systems that would continuously adjust to meet patient needs. Machine learning is an important part of the transformation of healthcare into a more accurate, efficient, and patient-focused system. Despite the obstacles, research and technological advancements continue to address these challenges, promising a future where individual disease treatment is the norm.

REFERENCES

- [1]. Boppiniti ST. Machine learning for predictive analytics: Enhancing data-driven decision-making across industries. *International Journal of Sustainable Development in Computing Science*. 2019;1(3):13.
- [2]. Ghanem M, Ghaith AK, Bydon M. Artificial intelligence and personalized medicine: transforming patient care. In *The new era of precision medicine 2024* Jan 1 (pp. 131-142). Academic Press.
- [3]. Bandi M, Masimukku AK, Vemula R, Vallu S. Predictive analytics in healthcare: enhancing patient outcomes through data-driven forecasting and decision-making. *International Numeric Journal of Machine Learning and Robots*. 2024;8(8):1-20.
- [4]. Mijwil MM, Al-Mistarehi AH, Abotaleb M, El-Kenawy ES, Ibrahim A, Abdelhamid AA, Eid MM. From pixels to diagnoses: deep learning's impact on medical image processing-a survey. *Wasit Journal of Computer and Mathematics Science*. 2023 Sep 30;2(3):9-15.
- [5]. Alharbe N, Almalki M. IoT-enabled healthcare transformation leveraging deep learning for advanced patient monitoring and diagnosis. *Multimedia Tools and Applications*. 2025 Jun;84(19):21331-44.
- [6]. Kolossvary M, De Cecco CN, Feuchtner G, Maurovich-Horvat P. Advanced atherosclerosis imaging by CT: radiomics, machine learning and deep learning. *Journal of cardiovascular computed tomography*. 2019 Sep 1;13(5):274-80.
- [7]. Kolossvary M, De Cecco CN, Feuchtner G, Maurovich-Horvat P. Advanced atherosclerosis imaging by CT: radiomics, machine learning and deep learning. *Journal of cardiovascular computed tomography*. 2019 Sep 1;13(5):274-80.
- [8]. Chianumba EC, Ikhalea N, Mustapha AY, Forkuo AY, Osamika D. Exploring the role of AI and machine learning in improving healthcare diagnostics and personalized medicine. *Journal of Frontiers in Multidisciplinary Research*. 2023 Jan;4(1):177-82.

- [9]. Fatima S. Transforming healthcare with AI and machine learning: revolutionizing patient care through advanced analytics. *International Journal of Education and Science Research Review*. 2024;11(6):58-75.
- [10]. Alum EU, Ugwu OP. Artificial intelligence in personalized medicine: transforming diagnosis and treatment. *Discover Applied Sciences*. 2025 Mar 1;7(3):193.
- [11]. Kasula BY. Harnessing machine learning for personalized patient care. *Transactions on Latest Trends in Artificial Intelligence*. 2023;4(4):1-9.
- [12]. Mandala G, Reddy R, Nishanth A, Yasmeeen Z, Maguluri KK. AI and ml in healthcare: redefining diagnostics, treatment, and personalized medicine. *International Journal of Applied Engineering & Technology*. 2023 Oct 19;5(S6).
- [13]. Mulani AO, Deshmukh M, Jadhav V, Chaudhari K, Mathew AA, Salunkhe S. Transforming Drug Therapy with Deep Learning: The Future of Personalized Medicine. *Drug Research*. 2025 Oct;75(08):326-33.
- [14]. Saeed A, Husnain A, Rasool S, Gill AY. Healthcare Revolution: How AI and Machine Learning Are Changing Medicine. *Journal Research of Social Science, Economics & Management*. 2023 Oct 1;3(3).
- [15]. Kothinti RR. Deep learning in healthcare: Transforming disease diagnosis, personalized treatment, and clinical decision-making through AI-driven innovations. *World J Adv Res Rev*. 2024 Nov 30;24(2):2841-56.
- [16]. Rahman MH, Islam T, Hossen ME, Chowdhury ME, Hayat R, Shovon MS, Alamgir M, Akter S, Chowdhury R, Sunny AR. Machine learning in healthcare: From diagnostics to personalized medicine and predictive analytics. *Journal of Angiotherapy*. 2024 Dec 12;8(12):1-8.
- [17]. Kolluri VE. Revolutionizing Healthcare With AI: Personalized Medicine: Predictive. *JETIR-Int. J. Emerg. Technol. Innov. Res*. 2016;3(11):2349-5162.
- [18]. Khan O, Badhiwala JH, Grasso G, Fehlings MG. Use of machine learning and artificial intelligence to drive personalized medicine approaches for spine care. *World neurosurgery*. 2020 Aug 1;140:512-8.
- [19]. Sharma A, Pesarlanka DB, Lankadasu NV, Sharma S. Revolutionizing Patient Care: A Personalized Medicine Approach Using Artificial Intelligence for Better Healthcare. In *AI-Driven Breakthroughs in Antimicrobial Resistance 2025* (pp. 169-194). IGI Global Scientific Publishing.

- [20]. Obiany CM, Nzeyimana P, Alehegn HZ, Drammeh A, Lawal A, Gadzama LR, Yelwa JM. The Integration of Artificial Intelligence (AI) and Machine Learning (ML) in Diagnostics and Personalized Medicine. *Biomed J Sci & Tech Res*. 2025 Apr 25;61(5).
- [21]. Ali SA, Ali S, Shamim S. Role of Artificial Intelligence in Transforming Diagnosis, Treatment, and Patient Care: A Review. *Current Pharmacogenomics and Personalized Medicine*. 2026 Jan;23(1):E18756921404728.
- [22]. Matcha A. Innovations in healthcare: Transforming patient care through technology, personalized medicine, and global health crises. *International Journal of Science and Research (IJSR)*. 2023 Dec;12(12):1668.
- [23]. Blobel B, Ruotsalainen P. Healthcare Transformation Towards Personalized Medicine- Chances and Challenges. *InpHealth 2019 May 15* (pp. 3-21).
- [24]. Johnson KB, Wei WQ, Weeraratne D, Frisse ME, Misulis K, Rhee K, Zhao J, Snowdon JL. Precision medicine, AI, and the future of personalized health care. *Clinical and translational science*. 2021 Jan;14(1):86-93.
- [25]. Rani S, Kataria A, Bhambri P, Pareek PK, Puri V. Artificial intelligence in personalized health services for better patient care. In *Revolutionizing healthcare: AI integration with IoT for enhanced patient outcomes 2024 Sep 24* (pp. 89-108). Cham: Springer Nature Switzerland.
- [26]. Hossain MD, Rahman MH, Hossain KM. Artificial Intelligence in healthcare: Transformative applications, ethical challenges, and future directions in medical diagnostics and personalized medicine. DOI: <https://doi.org/10.30574/ijjsra>. 2025;1.
- [27]. Bilakanti G. AI-Powered Precision Medicine Transforming Personalized Healthcare. *IJLRP- International Journal of Leading Research Publication*.;4(10).
- [28]. Recharla M, Chakilam C, Kannan S, Nuka ST, Suura SR. Harnessing AI and machine learning for precision medicine: Advancements in genomic research, disease detection, and personalized healthcare. *American Journal of Psychiatric Rehabilitation*. 2025;28(1):112-23.
- [29]. Susilo YK, Rahman SA, Amgain K. Machine learning in precision medicine: transforming personalized diagnostics and treatment. *Journal of Mathematics and Computing Science*. 2025 Jun 12;11(1):1-6.
- [30]. Lin B, Wu S. Digital transformation in personalized medicine with artificial intelligence and the internet of medical things. *Omics: a journal of integrative biology*. 2022 Feb 1;26(2):77-81.

- [31]. Adeyinka A, Lamina Y, Tawo O, Adeyeye Y, Minkah A. Machine learning efforts that enhance personalized patient care and chronic disease management. *International Journal of Scientific Research in Science and Technology*. 2022 Mar;647-71.
- [32]. Poalelungi DG, Musat CL, Fulga A, Neagu M, Neagu AI, Piraianu AI, Fulga I. Advancing patient care: how artificial intelligence is transforming healthcare. *Journal of personalized medicine*. 2023 Jul 31;13(8):1214.
- [33]. Mangrolia JR. *Revolutionizing Healthcare through Artificial Intelligence: The role of Machine Learning, Deep Learning and NLP in transforming Patient Care*. Deep Science Publishing; 2025 May 2.
- [34]. Sharma P, Kumra R, Bhaggi E. Transforming Healthcare: The Impact of AI and Machine Learning on Personalization. In *The Ethical Landscape of AI: Global Issues and Solutions 2026* (pp. 413-430). IGI Global Scientific Publishing.
- [35]. Babu CS, Bhargavi Y, Krishna PR. AI-Driven Personalized Healthcare Solutions: Transforming Healthcare Delivery Through Precision, Efficiency, and Innovation. *AI-Driven Personalized Healthcare Solutions*. 2025:241-76.
- [36]. Harry A. Revolutionizing healthcare: how machine learning is transforming patient diagnoses-a comprehensive review of ai's impact on medical diagnosis. *BULLET: Jurnal Multidisiplin Ilmu*. 2023;2(4):1259-66.
- [37]. Chianumba EC, Ikhalea N, Mustapha AY, Forkuo AY. Developing a framework for using AI in personalized medicine to optimize treatment plans. *Journal of Frontiers in Multidisciplinary Research*. 2022 Jan;3(1):57-71.
- [38]. Bisht N, Ali ZS, Amir NA, Krishna EP, Mudhafar M, Shakir KH, Shnain AH. Transforming Healthcare in the New AI Era through Integrated Personalized Healthcare. In *2024 4th International Conference on Advance Computing and Innovative Technologies in Engineering (ICACITE) 2024 May 14* (pp. 790-794). IEEE.
- [39]. Gill AY, Saeed A, Rasool S, Husnain A, Hussain HK. Revolutionizing healthcare: how machine learning is transforming patient diagnoses-a comprehensive review of AI's impact on medical diagnosis. *Journal of World Science*. 2023 Oct 27;2(10):1638-52.
- [40]. Neoaz N, Shah HH, Zainab H. AI in Personalized Medicine: Transforming Treatment Plans through Precision Health. *Global Journal of Emerging AI and Computing*. 2025 Jan 23;1(1):34-50.

- [41]. Ponnambalath Mohanadas H, Manikandan A, Ismail AF, Tucker N, Jaganathan SK. Artificial intelligence applications in medical devices for personalized health care solutions: systematic review. *Journal of Medical Internet Research*. 2026 Mar 4;28:e72410.
- [42]. Alzamily JY, Bakeer H, Almadhoun H, Abunasser BS, Abu-Naser SS. Artificial Intelligence in Healthcare: Transforming Patient Care and Medical Practices.
- [43]. Singh AP, Saxena R, Saxena S, Maurya NK. Artificial intelligence revolution in healthcare: transforming diagnosis, treatment, and patient care. *Asian Journal of Advances in Research*. 2024 Jun 3;7(1):241-63.
- [44]. Aluru KS. Precision Medicine: Leveraging AI for Personalized Patient Care. *International Journal of Advanced Engineering Technologies and Innovations*. 2023 Mar 25;1(02):491-516.
- [45]. Miftah NS. Transforming Healthcare Through Artificial Intelligence: From Predictive Diagnostics to Personalized Therapeutics. *Libyan Open University Journal of Medical Sciences and Sustainability*. 2025 Aug 8:37-44.
- [46]. Rane N, Choudhary S, Rane J. Towards autonomous healthcare: integrating artificial intelligence (AI) for personalized medicine and disease prediction. Available at SSRN 4637894. 2023 Nov 9.
- [47]. Khan MA, Kumaraguru S, Praveen R, Chinthamu N, Sarkar R, Deka N, Shrivastava A. Exploring the Role of Artificial Intelligence in Personalized Healthcare: From Predictive Diagnostics to Tailored Treatment Plans. *Frontiers in Health Informatics*. 2024 Apr 1;13(3).
- [48]. Paranjape K, Schinkel M, Nanayakkara P. Short keynote paper: Mainstreaming personalized healthcare—transforming healthcare through new era of artificial intelligence. *IEEE journal of biomedical and health informatics*. 2020 Feb 10;24(7):1860-3.
- [49]. Chen K, Wu L, Chang H. Artificial Intelligence in Clinical Medicine: Transforming Diagnosis, Treatment, and Patient Care. *AI in Clinical Medicine*. 2025 Jun 14:e3-.
- [50]. Amirineni S. The role of artificial intelligence in revolutionizing personalized medicine: A comprehensive review of techniques and applications. *World Journal of Advanced Research and Reviews*. 2024;23(2):2026-30.
- [51]. Rahmah L, Wianti S, Herdalisah W, Purwoko RY, Sari FE. The impact of AI-powered diagnostics, personalized medicine, and digital health records on patient care quality. *The Journal of Academic Science*. 2024 Jul 29;1(2):118-30.

- [52]. Trezza A, Visibelli A, Roncaglia B, Spiga O, Santucci A. Unsupervised learning in precision medicine: Unlocking personalized healthcare through AI. *Applied Sciences*. 2024 Oct 12;14(20):9305.
- [53]. Kumar D, Pawar PP, Gonaygunta H, Nadella GS, Meduri K, Singh S. Machine learning's role in personalized medicine & treatment optimization. *World Journal of Advanced Research and Reviews*. 2024 Feb 28;21(2):1675-86.
- [54]. Mahabub S, Das BC, Hossain MR. Advancing healthcare transformation: AI-driven precision medicine and scalable innovations through data analytics. *Edelweiss Applied Science and Technology*. 2024;8(6):8322-32.
- [55]. Panahi O. Farrokh.(2025). The Use of Machine Learning for Personalized Dental-Medicine Treatment. *Glob J Med Biomed Case Rep*. 2025;1(001):2.
- [56]. Thoppil IJ, Ashtalakshmi K, Chundi R. Transforming healthcare: Leveraging machine learning algorithms for diagnosis, treatment, and management. In *Bioinformatics and Beyond 2025* (pp. 92-114). CRC Press.
- [57]. Serrano DR, Luciano FC, Anaya BJ, Ongoren B, Kara A, Molina G, Ramirez BI, Sánchez-Guirales SA, Simon JA, Tomietto G, Rapti C. Artificial intelligence (AI) applications in drug discovery and drug delivery: revolutionizing personalized medicine. *Pharmaceutics*. 2024 Oct 14;16(10):1328.
- [58]. Jankauskas SS, Varzideh F, Kansakar U, Santulli G. Artificial intelligence in cardiovascular medicine: A giant step in personalized medicine?. *Journal of Personalized Medicine*. 2026 Apr 1;16(4):192.
- [59]. Mahabub S, Das BC, Hossain MR. Advancing healthcare transformation: AI-driven precision medicine and scalable innovations through data analytics. *Edelweiss Applied Science and Technology*. 2024;8(6):8322-32.
- [60]. Ahmadi A, RabieNezhad Ganji N. AI-driven medical innovations: transforming healthcare through data intelligence. *International Journal of BioLife Sciences (IJBLs)*. 2023 Oct 1;2(2):132-42.
- [61]. Jakka AL, Chacko RM, Vasam M, Alagarsamy S, Chandragiri SS, Gavini SD, Mathew MJ. From genomics to clinic: the transformative impact of AI in pharmacogenomics and personalized medicine. *Pharmacogenomics*. 2025 Sep 22;26(13-14):573-85.
- [62]. Khude H, Shende P. AI-driven clinical decision support systems: Revolutionizing medication selection and personalized drug therapy. *Advances in Integrative Medicine*. 2025 Dec 1;12(4):100529.



- [63]. Zeb S, Nizamullah FN, Abbasi N, Qayyum MU. Transforming Healthcare: Artificial Intelligence's Place in Contemporary Medicine. BULLET: Jurnal Multidisiplin Ilmu. 2024;3(4):592385.
- [64]. Danda RR, Maguluri KK, Yasmeeen Z, Mandala G, Dileep V. Intelligent Healthcare Systems: Harnessing Ai and MI To Revolutionize Patient Care And Clinical Decision-Making. International Journal of Applied Engineering and Technology <https://papers.ssrn.com/sol3/papers.cfm>. 2023.

