



Using Artificial Intelligence (AI) Technology in the Health Sector has Several Goals

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Abstract: In a broad sense, artificial intelligence (AI) refers to any computer or system behavior that resembles human behavior. A subfield of artificial intelligence called "machine learning" enables computers to learn from data without explicit human programming. One of the most significant contemporary trends in global healthcare is the use of artificial intelligence (AI) technologies in medicine. Technologies based on artificial intelligence are profoundly transforming the world's healthcare system, enabling a dramatic reconstruction of the medical diagnostics system while simultaneously lowering healthcare expenditures. Identifying the class of diseases to which a disease belongs is crucial before treating it. It is feasible to categorize the type of disease based on the feature space of the condition. Algorithms for machine learning can address this issue.	Review Paper
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I. INTRODUCTION

Traditional programming techniques are finding it harder and harder to structure, analyze, and respond to the ever-growing amount of data being produced every day because they provide less flexibility. Systems that can learn from data by identifying patterns and connections across data sets to improve predictions are becoming more and more in demand today. Artificial intelligence's machine learning subfield enables computers to learn from data without explicit human programming [32]. In a broad sense, artificial intelligence (AI) refers to any computer or system behavior that resembles human behavior. The most basic kind of artificial intelligence is the "imitation" of human behavior by computers, which is based on considerable data on previous instances of the same behavior. The same task of utilizing computers to understand human intelligence is connected to artificial intelligence (AI), which is not always restricted to biologically logical approaches. In the twenty-first century, the field of AI has grown steadily. With dramatic revolutions influenced by both ideas and tactics, the evolution of AI has improved the development of human society in our own time [1]. Deep learning, machine learning, and artificial intelligence are all areas of active research. And it appears that it may soon be able to fully replace human intelligence. The study of artificial intelligence, or AI, is a branch of computer science that focuses on creating intelligent computer systems, or systems that possess the

skills that are typically associated with the human mind, such as language comprehension, learning, the capacity for reasoning, problem-solving, etc. Later, a variety of software programs and algorithms started to be referred to as AI; its defining characteristic is that they have the ability to answer some issues in the same way that a human would. For instance, AI is just starting to permeate medicine through speech processing [2], natural language text processing [3], object identification [4], voiceprint recognition [5], robotics [6], handwritten character recognition [7], expert systems [8], and medical diagnostics [9]. There have already been a lot of intriguing computer algorithms and inventions in this field, but they are still a long way from being widely used because they lack clinical evidence of their efficacy. However, it should be acknowledged that narrowly focused artificial intelligence will firmly take its position given how swiftly this subject has progressed over the past few years and the fact that computers are now outperforming people in solving specific medical problems, which will rise significantly. All areas of human activity—including medicine and healthcare—have been affected by AI technologies. The medical professional must stay current with the most recent developments in medical science. A doctor cannot treat patients, rest, update knowledge, and maintain it in his thoughts at the same time, hence they cannot perform this task as quickly as AI [25–26, 31]. AI can keep all the information gathered and regularly update research data.

The adoption of such technology will simplify life for medical professionals. In fact, one of the most significant aspects of healthcare that AI technologies may aid with is the treatment of chronic diseases. Broadly speaking, chronic diseases are problems that last for a year or longer and necessitate continuing medical care, restrict daily activities, or both. The main causes of death and disability worldwide are chronic illnesses such as renal disease, heart disease, cancer, and diabetes. The annual investment in AI had a modest decline in 2018, however it was only momentary. The majority of total corporate investments in AI are private. The amount invested in artificial intelligence initiatives for the healthcare industry worldwide in 2021 increased to \$11.2 billion from \$8 billion in 2020. The Stanford Institute for Human-Centered Artificial Intelligence released such statistics in March 2022. The study found that from 2017 to 2021, the "attractive" businesses for private investment in the artificial intelligence market were those related to medicine and healthcare. During this time, core projects received a total investment of more than \$28.9 billion [10]. Automation and increasing the precision of diagnostics are two crucial topics. The classification of diseases is one method for increasing the precision of diagnosis. AI in the form of machine learning (ML) [11] enables the classification of illness kinds that are similar to one another in terms of a parametric factor. And one of the fundamental machine learning algorithms used for classification is K-Nearest Neighbor (KNN) [12]. A neural network can also be used to tackle the categorization problem [13]. High-performance hardware is needed for deep learning algorithms that use huge datasets, such as heterogeneous computing systems [14] or parallel computing techniques. At the moment, parallel and distributed computing technologies [15, 16] can also be used to overcome this issue. The major goal of this study is to choose significant parametric variables from the gathered disease data that produce more F1-score outcomes. For classification, two forms of coronary heart disease were chosen. It is suggested to use the KNN algorithm for categorizing coronary heart disease. It can be viewed as an algorithm that, when used with the training dataset, generates predictions based on the characteristics of other data points that are present adjacent to it [22]. In medical data mining, hidden patterns in datasets are discovered. For the early diagnosis of cardiac disease, a supervised algorithm like KNN is employed. The most well-known, successful, and efficient algorithm for pattern recognition is KNN, a frequently used lazy classification algorithm. The distance measure and K value both affect how accurate KNN is. Cosine and Euclidean distance are two other methods for calculating the separation between two instances. KNN determines its closest neighbors and determines a class by majority vote in order to evaluate a fresh unknown sample [17]. When the training sample is large, lazy learning techniques like the KNN classifier can be expensive to use because they need to store the whole training sample. In order to reduce storage and

processing needs, the compressed closest neighbor classifier incrementally caches a portion of the sample [18]. Due to its ease of use and relatively quick convergence speed, KNN is growing in popularity [19]. Medical information technology has advanced toward intelligence as a result of the quick growth of information technology. For the intellectualization of medical information, the classification of large data in health care is extremely important. The KNN classification technique is straightforward, which has led to its widespread application in numerous disciplines [20]. One area of healthcare that might be categorized is coronary heart disease (CHD). The Center for Specialized Cardiology's medical personnel and the CHD statistics were both discussed. The CHD dataset was collected from the National Center for Health Statistics (NCHS) [21]. The main developments in machine learning will be discussed in this paper, including automated data analysis for patient health records and data-driven prediction. The advancements in computer-aided diagnosis, medication discovery, and personalized medicine will also be contrasted [23]. It is impossible to stress the importance of using big data analytics and machine learning to improve patient outcomes and healthcare performance. With the use of these technologies, healthcare professionals are now able to gain useful insights from big datasets that were previously unexplored, opening up a whole new world of opportunities. By utilizing these information, medical professionals can decide more intelligently about tailored medicine, treatment plans, and resource allocation, ultimately improving patient outcomes and making the healthcare system more effective. Healthcare professionals may now more easily spot trends, correlations, and risk factors thanks to the ability to analyze enormous amounts of healthcare data. This information enables early disease detection, disease prevention, and patient-specific treatment approaches [24]. In this opinion piece, we will examine AI's enormous influence on medicine while noting both its possible advantages and impending difficulties [27]. The application of AI and ML in healthcare has grown in importance, creating new opportunities for innovation, precision medicine, and better decision-making. It is essential to investigate the potential, difficulties, and ethical ramifications of integrating AI and ML into healthcare as we set out on this transformative journey. The field of diagnostics is one of the primary areas where AI and ML have demonstrated tremendous promise. These technologies can swiftly and precisely find trends, spot anomalies, and help with disease diagnosis by analyzing enormous amounts of medical data. The early diagnosis of diseases like cancer and better patient outcomes are made possible by AI-powered algorithms' outstanding accuracy in analyzing medical images like X-rays and MRIs [28]. A general taxonomy of machine learning algorithms is presented in this overview, which is followed by a more in-depth explanation of each algorithm class, its function and capabilities, and examples of applications, particularly in geriatric

medicine. Additional emphasis is placed on the implications for clinical practice, the difficulties associated with depending on devices with limited interpretability, and the advancements made in overcoming the latter through the creation of explainable machine learning [29]. Examining how machine learning technologies might enhance healthcare operations management is the goal of this study. To accomplish this research goal, a machine-learning-based model to address a specific medical issue is created. This study specifically uses the CNN (convolutional neural network) technique to propose an AI solution for diagnosing malaria infection. A total of 24,958 photos were used for deep learning training using malaria microscopy image data from the NIH National Library of Medicine, and 2600 images were chosen for final testing of the suggested diagnostic architecture. The empirical findings show that, with minimal misclassification and performance metrics of precision (0.97), recall (0.99), and f1-score (0.98) for parasite cells and precision (0.99), recall (0.97), and f1-score (0.98) for uninfected cells, the CNN diagnostic model correctly identified the majority of malaria-infected and non-infected cases. The CNN diagnostic solution processed a large number of cases quickly and with a 97.81% accuracy that could be relied upon. The k-fold cross-validation test was used to further validate the performance of this CNN model. These findings imply that machine learning-based diagnostic techniques have an edge over traditional manual diagnostic techniques when it comes to enhancing operational capacities in the healthcare sector in terms of diagnostic quality, processing expenses, lead times, and productivity. In addition, by lowering the probability of unneeded medical disputes connected to diagnostic errors, a machine-learning diagnosis system is more likely to improve the financial viability of healthcare operations. Propositions with a research framework are offered to examine the effects of machine learning on healthcare operations management for safety and quality of life in international communities as an extension for future research [30, 33].

II. BENEFITS OF AI IN THE HEALTH SECTOR

AI is currently being extensively tested in hospitals for drug discovery as well as diagnosis and symptom prediction. Here are some of its most promising prospects:

1. Diagnostic Assessment

Electronic health records (EHRs), radiography, CT scans, and magnetic resonance images all produce large volumes of data that AI can analyze. AI systems can assist with early symptom forecasts by analyzing data from patients, identifying trends, and identifying relationships.

2. Virtual Health Assistants

Virtual health assistants are in charge of carrying out a range of duties, including returning normal

patient calls and emails, monitoring medical records, safeguarding private patient information, setting up doctor appointments, and reminding patients to schedule follow-up appointments. Because it offers patients a personalized experience in managing their health and responding to their questions, it is one of the most beneficial AI applications in healthcare.

3. Treatment of Rare Diseases

In order to speed up the discovery and development of cutting-edge breakthrough medications and vaccines and revolutionize the delivery of healthcare, BERG, an AI-based clinical-stage biotech platform, aims to map diseases. It combines interrogative biology with research and development (R&D) to enable medical professionals to create durable products for people battling rare diseases.

4. Targeted Treatment

Benevolent AI, a prominent clinical-stage AI-enabled drug development business, was able to offer the proper treatment to the required patients at the correct time with the use of technologies like deep learning and AI, leading to tailored treatment of patients with helpful insights. The company is currently focused on developing portable remedies for uncommon diseases and securing licensing for its medications.

5. Drug Discovery

Artificial intelligence uses neural networks to evaluate drug candidates' characteristics and bioactivity. With the aid of AI systems, researchers can determine the optimal therapeutic targets to investigate for specific diseases. The healthcare sector has seen an increase in speed and a decrease in investment in drug discovery as a result. It has proven important in clinical trials in the selection of the correct candidates [34].

CONCLUSION

This article describes the application of machine learning techniques in the healthcare industry to manage a large amount of patient data and decrease the time, money, and resources needed for its analysis. Since they have been demonstrated to be more accurate than professional doctors' diagnoses and to enhance the hospital experience for patients, these kinds of technologies are currently in high demand. It was also emphasized that because these tools are intended to work in conjunction with doctors, training is required before they can be used for their primary role. The advantages of AI in the medical field are highlighted.

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