

A Comprehensive Review of Artificial Intelligence Applications in Healthcare

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Abstract— This study uses a general literature review to examine how artificial intelligence (AI) has changed the healthcare industry. We explore the use of AI to medical imaging, developing drugs, digital patient care, research, administrative work, and interaction with patients. AI helps healthcare providers with their administrative workload, enhances patient participation and adherence to treatment plans, and manages electronic medical records more effectively. It is essential for virtual patient care and for using cutting-edge medical imaging for recognizing clinical problems. AI has an impact on a wide range of areas, such as vaccine research, medication mistake detection, drug discovery, and massive data storage and processing. However, there are issues with privacy, safety, decision rights, costs, consent for information sharing, access, and efficacy when incorporating AI into healthcare. Healthcare workers' faith in one another, patient safety, and responsibility all depend on the governance of AI applications. This study highlights how AI has the potential to completely change healthcare, but it also highlights the necessity for careful thought and efficient control.

Keywords— Artificial Intelligence, Drug Design, Remote Monitoring, Pharmacy, Diagnosis.

I. INTRODUCTION (HEADING 1)

Artificial Intelligence (AI) is a transformative privilege, algorithmic approach are used to diagnosis and treatment of diseases. AI epitomizes the convergence of human intelligence and machine learning. Coined by John McCarthy, a pioneer in AI, as "the science and engineering of making intelligent machines" [1], Intelligent machines have become prevalent in modern sectors such as finance, autonomous transportation, and smart home technologies within healthcare, machine learning finds widespread application in constructing automated clinical decision systems and addressing diverse diseases [2]. AI harnesses sophisticated algorithms to assimilate insights from healthcare data, aiding professionals in clinical practice with

self-correcting and learning capabilities to enhance precision based on analysis. Its capabilities extend to tracking the spread of epidemics by monitoring animal and plant diseases and accessing global airline ticketing data to pinpoint movements of infected residents, predicting potential pandemics. AI's progression in emulating human intelligence is moving towards surpassing the Turing test and is positioned to be a key player in the forthcoming industrial revolution.

Global health systems are facing difficulties to their sustainability due to rising healthcare expenses that are surpassing GDP growth rates [3]. This challenge has been further accentuated by global events such as the conflict in Ukraine. Financial constraints, an aging population, a rise in chronic diseases, and increased demand for services are placing strain on healthcare systems. Moreover, certain countries, including India, Brazil, and Indonesia, are experiencing health system failures [4]. As health systems pivot towards robust disease management and evidence based care, the concept of a "Highly Reliable Organization (HRO)" underscores the significance of managed services through entities like "accountable care organizations (ACOs)" or "health maintenance organizations (HMOs)" [5]. Despite these efforts, the prevalence of chronic diseases in the United States (USA) is steadily rising, resulting in annual healthcare costs of USD 3.3 trillion. This landscape has prompted a digital transformation in healthcare, reshaping fundamental aspects of medical care.

The pressure exerted on global healthcare systems, infrastructure, supply chains, and workforce has accelerated the adoption of digital technologies. Post-pandemic, the healthcare sector has witnessed foundational changes, with patients actively participating in healthcare decision-making, driven by the widespread acceptance of virtual healthcare systems and digital innovations. Challenges may arise, necessitating strategies to navigate the path towards the future of healthcare. Patient experiences and needs are driving innovations in the healthcare sector, emphasizing digitally empowered interactions between physicians and patients to ensure patient-centric services globally. The deployment of advanced digital devices has become imperative to enhance

customer satisfaction, enabling health status monitoring, and ensuring improved drug adherence. These aspects are particularly valuable during the post hospitalization period using digital health platforms. Simultaneously, healthcare consumers are cautious about sharing confidential data, prompting healthcare organizations (HCOs) to prioritize transparency, empathy, and reliability to maintain customer trust. The rise of biomedical science—which includes digital medicine, genomics, artificial intelligence (AI), and its branch, machine learning (ML) provides the foundation for the transformation of healthcare by bringing cutting-edge technology that call for a new workforce and set of professional standards. Technologies like genomics, biometrics, tissue engineering, and advancements in the vaccine industry having the capacity to transform diagnostics, therapeutics, care delivery, regenerative treatments, and precision medicine models.

II. BUILDING BLOCKS OF AI

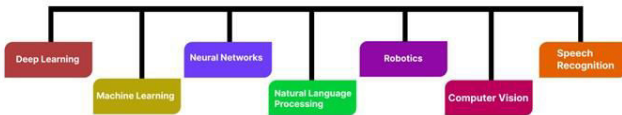


FIG-1: Building Blocks of AI

A. Deep Learning

Deep learning is a specialized field within machine learning that employs artificial neural networks to address intricate problems. At its core, artificial neural networks consist of interconnected nodes organized into layers, including an input layer, one or more hidden layers, and an output layer. The strength of connections between nodes, represented by weights, and the contribution of each node, known as biases, are crucial components.

The term "deep" in deep learning indicates the presence of multiple hidden layers in neural networks, allowing for hierarchical learning. This depth facilitates the capture of progressively complex and abstract features as data passes through the layers [6].

The training process involves supervised learning, where the model adjusts its weights and biases using labelled data. Backpropagation, a common technique, helps minimize the error between predicted and actual outputs by updating the model parameters.

B. Machine Learning

Machine learning is a computational approach enabling systems to learn and improve performance without explicit programming. It encompasses supervised, unsupervised, and reinforcement learning. Supervised learning involves labeled data, while unsupervised learning detects patterns in unlabeled data. Reinforcement learning guides agents to make decisions to maximize rewards [7].

Reinforcement learning guides agents to make decisions to maximize rewards. Algorithms like decision trees, random forests, SVM, and clustering are used. Feature engineering enhances model performance by selecting and transforming features. Evaluation metrics include accuracy, precision, and recall. Machine learning finds applications in healthcare, finance, marketing, and natural language processing.

Challenges include overfitting, underfitting, and the need for large datasets.

C. Neural Networks

Neural networks are computational models inspired by the human brain's structure and functioning. Comprising interconnected nodes arranged in layers, neural networks process information through input, hidden, and output layers. Weights and biases define the strength and influence of connections between nodes. The depth of neural networks, often referred to as "deep learning," involves multiple hidden layers for hierarchical learning. During training, supervised learning adjusts weights and biases using labeled data. Backpropagation minimizes prediction errors by updating model parameters.

Activation functions, such as sigmoid or ReLU, introduce non-linearity, enabling the network to capture complex patterns. Neural networks have proven successful in diverse applications, including image and speech recognition, natural language processing, and generative tasks [8].

D. Natural language Processing(NLP)

Natural Language Processing (NLP) is a branch of artificial intelligence (AI) that focuses on enabling computers to understand, interpret, and generate human language. It involves the development of algorithms and models that allow machines to interact with and process natural language data. NLP encompasses a range of tasks, including language translation, sentiment analysis, text summarization, and chatbot development[9].

In practical terms, NLP enables machines to comprehend the nuances of human language, extracting meaningful information from text or speech data. This technology is widely used in applications such as virtual assistants, language translation services, and automated customer support systems. NLP algorithms analyze linguistic patterns, semantic relationships, and context to derive insights and perform tasks that traditionally required human intelligence. The continual advancement of NLP contributes to improved language understanding and communication between humans and machines.

E. Robotics

Robotics is a multidisciplinary field that involves the design, construction, programming, and operation of robots. Robots are autonomous or semi-autonomous machines that can perform tasks in the physical world. Robotics combines elements of mechanical engineering, electrical engineering, computer science, and artificial intelligence to create machines capable of interacting with their environment and completing specific actions.

Robots are equipped with sensors to perceive their surroundings and actuators to carry out movements or tasks. The field of robotics spans various applications, from industrial robots used in manufacturing processes to robots designed for healthcare, exploration, and even entertainment. The goal is to create machines that can operate in diverse environments, adapt to changing conditions, and perform tasks efficiently.

Advancements in robotics include developments in machine learning and computer vision, enabling robots to learn from experience and perceive the world more intelligently. As technology progresses, robotics continues to play a crucial role in automation, enhancing productivity,

and expanding the range of tasks that machines can perform autonomously[10].

F. Computer Vision

Computer vision is a field within computer science that focuses on enabling machines to interpret and make sense of visual information from the world. It involves the development of algorithms and models that allow computers to analyze and understand images and videos. The primary goal of computer vision is to replicate human vision capabilities, enabling machines to perceive, interpret, and react to visual data.

Key tasks in computer vision include image recognition, object detection, image segmentation, and facial recognition. Through the use of various techniques such as image processing, feature extraction, and machine learning, computer vision systems can identify patterns, objects, and relationships within visual data.

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G. Speech Recognition

Speech recognition is a technology that enables computers to convert spoken language into written text. It involves the development of algorithms and systems capable of accurately interpreting and transcribing spoken words. The primary objective of speech recognition is to allow machines to understand and respond to human verbal commands or input [12]. Speech recognition systems use a combination of signal processing, pattern recognition, and machine learning techniques to analyze audio signals and identify the spoken words. These systems are trained on large datasets to recognize diverse accents, languages, and speech patterns. The technology has applications in various domains, including voice-activated virtual assistants, transcription services, and hands-free control of devices.

Advancements in deep learning and neural network architectures have significantly improved the accuracy and performance of speech recognition systems. As a result, speech recognition has become an integral component of modern human-computer interaction, providing a natural and convenient way for users to communicate with technology [13].

III. LITERATURE REVIEW

A. AI-Driven Drug Development

Artificial intelligence plays a vital component that drives tremendous advancements in many different sectors on a daily basis. A lot of work has been achieved in areas like robotics, image and speech recognition, and natural language processing, which points to significant advancements in technology and human-machine interaction [13].

Strict requirements, including as appropriate potency for the biological target, selectivity against undesirable targets, and advantageous physicochemical and ADMET characteristics, must be met by clinical candidate molecules in the drug development process. The multifaceted nature of compound optimization makes the use of many in silico prediction techniques necessary for effective design. In this setting, machine learning approach

like Bayesian learning, Random Forests (RF), and support vector machines (SVM) have to be effective.

Large datasets are essential for the successful application of machine learning for property prediction, especially in the pharmaceutical sector[14]. Large-scale datasets covering a variety of attributes are gathered during compound optimization. Machine learning models are trained on these datasets, which contain targets and anti-targets from various chemical series, in a methodical manner, guiding compound improvement.

B. AI-Based Disease Diagnostic

AI technologies, from deep learning to machine learning, are crucial to the development of new clinical systems and management of patient data and records, and the treatment of a wide range of diseases [15].

The most efficient techniques for diagnosing diseases make use of artificial intelligence. There are previously unheard-of opportunities to repeat patient and clinical group outcomes thanks to the application of artificial intelligence (AI) in healthcare. Researchers have employed a range of AI-based methods, including deep learning and machine learning models, to identify diseases that require early diagnosis, including those of the skin, liver, heart diseases so on [16].

The artificial intelligence (AI) tool IBM Watson for Oncology was created to help doctors identify and treat cancer. It examines a sizable number of patient information, clinical trial data, and medical literature using machine learning techniques. Breast cancer research is one prominent area of use. The Watson system allows oncologists to enter patient data, like as clinical and genetic data. In order to provide individualized therapy alternatives, the AI evaluates this data in addition to the most recent medical literature. It gives oncologists access to the most recent research and evidence-based guidance to assist treatment choices. Oncologists will benefit in the real world by having access to more data and insights to help them decide on treatment options. This artificial intelligence technology enhances human skills by providing a wealth of information quickly and efficiently [17].

Artificial intelligence is used extensively in several medical sectors. Technologies based on artificial intelligence (AI) have made it easier to diagnose acute as well as chronic diseases such as Alzheimer's disease and acute appendicitis. Integrative AI, which employs numerous algorithms rather than a single algorithm, dramatically improves the identification of cancerous cells and raises diagnostic accuracy. The development of various AI approaches also used in the prediction of recurrence of breast cancer. In home AI systems may manage patients with insulin abnormalities and swallowing difficulties instead of physicians [17].

C. AI-Enhanced Pharmacy Solutions

Artificial Intelligence is utilized in hospital pharmacy-based health care systems in a number of ways, including treatment recommendations, dose form organization for individual patients, and the selection of viable or acceptable administration routes. Modernizing health records: Keeping track of patient medical records may be challenging. Data gathering, storing, standardization, and tracing are made easy by the AI system's implementation. During critical situations, an AI system becomes crucial for taking control, especially when a patient faces a severe

condition and determining the right treatment plan becomes challenging [18].

In the healthcare domain, artificial intelligence (AI) has gained recognition as a valuable tool, particularly in offering assistance with medications and providing support services. Its main goal is to assist individuals dealing with chronic conditions between their visits to the doctor, giving them the ability to actively manage their own treatment. Within the healthcare system, people greatly benefit from AI's capacity to collect and compare data using social awareness algorithms. This involves gathering a wealth of information, such as medical history, treatment records, profiles from birth, as well as details about habits and lifestyle choices. The reliability of robots for physicians has grown, thanks to the integration of big data and AI. Currently, numerous organizations use robots under human supervision to undertake tasks that were traditionally performed by people [19].

Pharmaceutical companies, despite having access to a diverse range of compounds, often lack the necessary resources for self-identification. AI plays a notable role in streamlining the drug development process, resulting in shorter development times, reduced costs, increased returns on investment, and potentially lower expenses for end users. One of the primary advantages of artificial intelligence is its ability to analyze extensive datasets, a task that would typically surpass the capacity of conventional computers. This capability positions AI as an exceptionally efficient tool, especially in managing complex healthcare data [20].

D. Remote Patient Monitoring and Telemedicine with AI

One of the most popular applications of telemedicine was remote patient monitoring. It presents a more efficient and economical way to do routine doctor-patient consultations, allowing clinical outcomes and the patient's current status to be evaluated remotely.

Through the use of video conferencing and the connection of digital medical equipment to record and gather clinical data about the patient, this method has developed to mimic in-person consultations. When opposed to traditional physical patient monitoring, the main objective of this movement is to reduce expenses and increase accessibility, convenience, and efficiency [21].

In reality, Ada Health and the UK's National Health Service (NHS) collaborated to develop an AI-enhanced telemedicine service. Through a chat interface, users of this site, called Ask Ada, may discuss their medical history and symptoms. Ada's AI then compares this data with a vast medical knowledge library after processing it using machine learning and natural language processing. The AI creates individualized health suggestions, recommending whether to seek emergency medical assistance or pointing out possible causes of symptoms. Ada can occasionally link users to telemedicine services smoothly, facilitating online discussions with medical specialists [22].

By offering rapid and effective early assessments, this AI and telemedicine combination improves healthcare accessibility. It assists in relieving the strain on conventional healthcare services so they may concentrate on more urgent situations by providing AI-driven insights. This illustration shows how AI may be a useful tool for improving telemedicine's efficiency and accessibility to healthcare [23].

E. AI-Assisted Surgery

Artificial Intelligence (AI) has recently made significant strides in various fields such as object discovery, speech recognition, and natural language processing. In the medical sector, several success stories have emerged, leveraging Deep Learning (DL) for tasks like predicting cardiovascular threats based on retinal images and categorizing skin lesions. A nascent field, Surgical Data Wisdom, aims to enhance the quality of interventional healthcare and its value through the capture, association, analysis, and modeling of data, particularly using AI-based methodologies [24].

Present-day operating theaters are characterized by a multitude of information sources. Machine learning algorithms can operate in either a supervised or an unsupervised manner. While both approaches rely on data for learning, supervised learning necessitates annotated data. Similar techniques can be employed for workflow analysis by automatically segmenting procedures into phases or surgical activities.

Additional information is provided through annotation techniques such as numerical regression, semantic segmentation, and classification, enabling AI to effectively learn from raw data. However, this annotation process is often laborious, especially for large datasets, requiring expensive specialist knowledge, leading to a bottleneck.

Challenges in this domain include the absence of defined processes for data gathering and annotation, restricted digital availability of data, and the need for representative data from multiple centers accessible through open data registries. These challenges raise concerns related to privacy and confidentiality.[25]

F. Complexities of Integrating AI in Healthcare

Researchers and corporate executives are grappling with the challenges posed by the increasing integration of artificial intelligence (AI) into daily life and workplaces. A fundamental hurdle lies in the disparate architectures of AI applications compared to traditional enterprise applications. The transfer of data between servers and storage involves multiple protocols, creating complexity. Efforts to enhance data locality - enabling one server to process more data without waiting for others - can alleviate these issues. Enhanced storage and GPU integration have resulted in noticeable cost savings. In the realm of healthcare, AI faces obstacles related to information security and privacy. AI relies on data from medical histories, necessitating precautions to prevent unauthorized access. Robust defense against cyberattacks is crucial, particularly in healthcare, where the potential consequences on people's lives are severe. The pharmaceutical industry encounters challenges in developing new medications with heightened efficacy and fewer side effects. Creating an environment conducive to increased development, higher success rates, lower discovery costs, and direct patient access adds complexity to these efforts.[26, 27].

Customer demands in the pharmaceutical sector are evolving. Consumers seek new treatments that surpass current options clinically and financially. Concrete evidence from actual patient outcomes is increasingly important to validate claims of medication superiority. Scientific productivity in the pharmaceutical industry experienced a consistent level of stability over the past decade. However, concerns persist regarding the efficacy of existing processes for drug discovery and development.

The need for innovation and efficiency in this field remains a focal point for sustained progress [28].

IV. CONCLUSION

The artificial intelligence (AI) in biomedical domains is becoming more and more evident due to its capacity to address the complex problems that biomedicine presents and its ongoing development. This is especially clear when looking at explainable AI's place in clinical decisions involves taking into account perspectives from the patient, legal, medical, and technical standpoints. Despite the initial high costs involved with installations, the medical sector has demonstrated notable benefits from the incorporation of artificial intelligence. Long term projections indicate that the advantages will probably outweigh the disadvantages, allowing a wider audience to ultimately have obtain to this technology. AI is primarily utilized in healthcare to support two main goals: managing healthcare organizations more easily and assisting with diagnosis. The technology's importance in the medical field is highlighted by its ability to improve diagnostics and simplify medical procedures.

AI applications in pharmacy have garnered significant attention in recent times, spanning across domains like drug discovery, dosage form design, and hospital pharmacy administration. These applications make use of the idea that people are capable of understanding information, solving problems, and coming to decisions. Automated databases and processes are prime examples of how artificial intelligence may be used to undertake efficient assessments.

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