

A Review of Artificial Intelligence in Medical field

PRERAN E

Computer Science & Engineering
Alva's Institute of Engineering and Technology
Moodbidri, India
preranegowda@gmail.com

PRUTHVI B R

Computer Science & Engineering
Alva's Institute of Engineering and Technology
Moodbidri, India
Pruthvibr6013@gmail.com

PUNEETH C K

Computer Science & Engineering
Alva's Institute of Engineering and Technology
Moodbidri, India
Puneethck24@gmail.com

RAHUL GOWDA G V

Computer Science & Engineering
Alva's Institute of Engineering and Technology
Moodbidri, India
rahulgowdagv18@gmail.com

RAKESH

Computer Science & Engineering
Alva's Institute of Engineering and Technology
Moodbidri, India
rakeshnayaknayak235@gmail.com

Dr. Senthil Kumar

Computer Science & Engineering
Alva's Institute of Engineering and Technology
Moodbidri, India
senthil@aiet.org.in

Abstract – The concept of AI was first proposed in 1956, and since then, it has revolutionized human medicine through a plethora of technology advancements and fundamentally altered the conventional medical paradigm. The main method for perspectives to explain how artificial intelligence is being applied in the medical domain: machine learning, intelligent robots, image recognition, and expert systems. We also go over current issues and upcoming developments in these fields. Due to consequences of globalization, numerous research institutes worldwide having complete many studies on this topic in recent years. As such, the field of medical artificial intelligence has made significant progress and is anticipated to continue growing in the future.

1. Overview

John McCarthy first used the term artificial intelligence (AI) in 1956 at the Dartmouth Conference, and that is when the field of AI was born. The ability to process external data systematically and learn from it to achieve specific goals and tasks is how Kaplan and Haenlein described artificial intelligence.[2]. Artificial Intelligence (AI) is the process of utilizing machines to mimic intelligent behaviors and human thought processes, such as learning, reasoning, and thinking. Its goal is to deal with challenging issues that are best left to professionals.[3]. The study of AI a subfield of computer science, focuses mostly on the following topics: machine learning, intelligent robots, neural networks, natural language understanding, language

recognition, picture recognition, and expert systems.[4]

The concept of AI in medicine (AIM) first emerged in the early 1970s.[5] It sought to increase the effectiveness of medical diagnosis and treatment through the use of AI systems. Following the 1980s, the development of AIM might be broadly divided into four stages: [1] infancy (1980s): the “decision tree” algorithm was proposed, and artificial neural networks continued to develop; [2] adolescence (1990s): “expert systems” continued to mature because of the emergence of support vector machines; [3] coming-of age (2000s): the concept of “deep learning” was proposed, and machine learning became a prominent theme of AIM; and [4] currently, we are in the “maturation period” (2010s): as the technologies are relatively advanced. However, the ability to communicate with others still requires improvement. Therefore, we are still in the stage of “weak” AI.[6]

2. Application of AIM

2.1. Machine learning

The phrase "machine learning" was first used in 1959 by Arthur Samuel to refer to a class of algorithms and classifier development.[7]. To be able to effectively forecast incoming data, the algorithm automatically learns from the source data and creates a prototype based on it.[8]

The backpropagation technique was first developed in the early 1960s, and numerous more

significant advancements in machine learning algorithms followed.[9]. Paul implemented the automatic differentiation technique in neural networks in 1982.[10] A popular machine learning method recognised as the "decision tree" was created by Ross Quinlan in 1986. It entails classifying data based on predetermined rules. Tin Kam Ho developed a significant algorithm called as the random forest (double spatial feature extraction algorithm). According to the decision tree-based random subspace method.[11] Vladimir created the support vector machine model in 1995. The profound learning algorithm was given by prominent deep learning expert Geoffrey Hinton in 2006. Convolutional neural networks are among machine learning's representative algorithms, and deep learning is essentially built on them.

CNN-BP, or convolutional recurrent neural network-blood pressure, is a revolutionary blood pressure measuring model developed in 2008 to meet the criteria of the British and Irish Hypertension Society and the Association for the Advancement of Medical Instrumentation. Its goal was to increase the model's precision, address concerns with poor resilience, and extract pulse waveform feature points in conventional medicine.[12]

As assisted diagnosis technology advances, a sizable data bank is created during the illness screening, diagnosis, and treatment processes.[13] For the doctors, organizing and interpreting this data in a short amount of time can be difficult. Consequently, machine learning is being employed more and more in medicine to assist physicians in forecasting patient outcomes and disease states.

One of the best machine learning algorithms is Random Forest. Nowadays, random forests are widely used in medicine, especially in disease prediction. People who have had idiopathic bleeding ulcers in the past may be more likely to develop an ulcer. Serious complications such as ulcers can make the patient unsafe. In 2018, an IPUML model was developed using machine learning to accurately predict rebleeding in idiopathic peptic ulcer. [14]

On the other hand, serious complications such as myocarditis and pulmonary edema may develop in a small number of children with serious enterovirus infection and foot and mouth disease. [15] The CatBoost model was developed in 2019 to predict the severity of hand, foot, and mouth disease and was discovered to have better specificity and sensitivity compared to other models such as decision tree and SVM. [16]

Machine learning can also tell the benefit of radiation therapy. For example, people with cancer, especially people with small brain cancers, often receive radiation therapy. However, long-term radiation therapy causes serious side effects, such as radiation pneumonia, which make way to respiratory failure and death. [17] Researchers have used neural networks to develop lung cancer prediction strategies. They also created a network heavily trained on hardware and memory to accurately predict various problems. [18]

The "black box" problem of machine learning needs to be solved. A "black box" neural network includes a recurrent neural network with short-term memory and a CNN for feature extraction. Generally speaking, a neural network consists of ideas, actions, and neural processes. "Black box" refers to the central processing of neural networks. Users cannot see the inner workings of the system. [19] By solving problems in the black box, machine learning can become more accurate, more powerful, and expand its application. This could lead to further advances in treatment.

2.2. Intelligent robots4

The American Robotics Association defined a robot in 1979 as "a reprogrammable, multifunctional manipulator designed to move materials, parts, tools, or other special objects by a variety of programs programmed to do many things." 20 smart robots participated in the surgery. [21] For example, PUMA 560 was used in prostate surgery in 1988, and PUMA 560 was used in neurosurgery in 1985. In 1992, the US Food and Drug Administration (FDA) approved the first intelligent robot called ROBODOC. It is used only in hip treatment in orthopedic surgery.

To date, the FDA has approved three types of robotic surgery: ZUES, da Vinci, and automated endoscopic systems for robotic approval. [22] Smart robots are widely used in fields such as urology, orthopedics, and dentistry due to their accuracy, intelligence, and minimal invasiveness.

Depending on the type of orthopedic surgery performed, robots can be divided into three groups: joint surgery robots, trauma orthopedic robots, and other robots. [24] Femoral neck fractures may develop in elderly people with disabilities, poor function, or hip deformities. The best treatment for these fractures is surgery because they can cause diseases such as nonunion and avascular necrosis of the femoral head. A study was conducted in 2018 investigating ways to reduce bleeding during surgery for femoral neck fractures. After comparing two surgical methods (robotic surgery and manual surgery), they concluded that surgeons could

ld define and shrink the surgical area, thus reducing the bleeding that occurs during surgery. [25]

Smart robots are also widely used in gynecological surgery. For example, patients with early-stage ovarian cancer may experience symptoms such as abdominal pain, torsion of the ovarian pedicle, and ovarian cancer. Treatment of this tumor, which is still in its early stages, is very important. A meta-analysis found that the da Vinci system provides many benefits, including removing more lymph nodes during surgery and reducing patients' need for additional blood. Therefore, this surgery is safer than laparoscopic surgery. [26]

Most robots that assist surgeons in clinical settings are discrete devices without any movement. However, continuum robots – specialized biomimetic robots with evolutionary patterns similar to those of “invertebrates” – have recently begun to emerge and will eventually replace Unique robots due to their flexibility and flexibility. In the future, robots will continue to be at the forefront of surgery. [27] Although smart robots are widely used in the field of orthopedics, they have limitations such as high cost, size and narrow application area. [21] Future surgeries will require smart robots to gradually adapt to the medical environment in a fast and intelligent manner. [twenty three]

2.3. Image recognition technology

The evolution of image recognition technology consists of three stages: object recognition, image recognition and text. [28] The analysis process consists of five steps: concept creation, image preproduction, image acquisition, design, and production. [29] This technology can be used to process image files quickly and efficiently. For example, one study found that a mechanical map was more effective than an electronic device at identifying bones most likely to break. [30] They found that the patient's age and gender affected the area of the broken line. [30] The application of image recognition technology in the diagnosis and treatment of intertrochanteric fractures is very important. It is also widely used in pain analysis, diagnosis and prognosis. [31]

Imaging technology is now widely used in medicine. One of the four leading causes of death in women is breast cancer (ranked 32nd), most commonly caused by human papillomavirus infection. In the early stages, patients do not have obvious symptoms.

However symptoms such as cachexia and anemia may appear later. Cervical cancer patients have access to a variety of treatments, including surgery, radiation therapy and chemotherapy; however, the prognosis for these individuals

often depends on early diagnosis of the disease. The accuracy of cervical face recognition based on deep learning is almost 90%, which can help doctors detect cancer early. [33]

A 2017 study explored how imaging technology could improve the accuracy of fungal keratitis. [34] After analyzing experimental data using light microscopy, researchers concluded that images based on image recognition are more accurate and specific than bone smears when diagnosing fungal keratitis. Additionally, knowledge of imaging technology can help uninformed doctors detect diseases accurately.

Although knowledge of imaging technology can help doctors diagnose the condition, it cannot replace doctors' technical knowledge. The final diagnosis will affect some resolution of the images obtained, and this resolution will vary depending on the equipment used in different hospitals [31]. The use of image recognition technology brings many challenges. For illustration, in multilayer neural convolution, further effort must be done to improve the efficiency of the calculation as the training model requires more data. Additionally, the use of high-performance supercomputers is still rare. Therefore, more research is needed in the coming days to overcome the problems related to hardware, optimization and technology integration.

2.4. Expert system

An expert system of experts is a computer program designed to mimic the decision-making process of human experts. [36]. As one of the most effective of intellectual skills, it uses existing knowledge to think through and solve many complex problems. [37]

The professional development phase can be divided into the initial phase (1965-1971), the initial phase (1972-1977) and the developmental phase (1978-1978). [38] The first expert, the Deckard system, was developed in the early 1960s. The AAPHelp system was developed by the University of Leeds in 1972 to help diagnose abdominal pain. [39] The INTERNIST-I system was developed by the University of Pittsburgh in 1974 and is mainly used in the identification of clinical diseases. MYCIN, an intelligent diagnostic machine, was developed by Stanford University in 1976 and has the ability to help treat infectious diseases. It is not used in medicine for many reasons, including ethical issues. [40]

Research shows that first aid effectiveness can be greatly increased by using personal digital assistants (PDAs) to provide training from professionals to non-

professionals. After examining the data, it was determined that the use of professional technology could strengthen the weak link in the chain of survival while also improving the quality of first aid provided by agents. [41]

Another study used expert criteria to diagnose different types of headaches, including tension-type, migraine, and drug-induced migraine. The Computerized Headache Assessment Tool (CHAT) correctly diagnosed 93.4% of daily symptoms and 94.4% of migraines. The average accuracy rate is 98%. Therefore, guidance from CHAT experts can help doctors diagnose the correct origin of the headache, which is very important. [42] Fuzzy experts were used to analyze the MIT-BIH arrhythmia database to distinguish between ischemic heartbeats and arrhythmias. The outcome of the study showed an average specificity of 99% and sensitivity of 96%. [43]

Professionals have demonstrated to be useful in detecting and diagnosing disease and can make strong medical decisions. However, according to the doctor's experts, the precision of the system needs to be increased and the patient's medical history should be integrated with the system. Additionally, medical knowledge and research must be constantly updated in order to provide doctors with the most advanced diagnoses and treatments. [44]

3. Conclusion

Future advances in artificial intelligence should present more challenges: in data mining and machine learning, researchers must develop predictive models to solve "black box" problems and create continuous integration between 5G and the IoT. Robots; Training models in image recognition technology must become more effective, and professionals must continually increase their knowledge base so that people seeking medical services can receive more information.

The concept of AI has made significant progress in the last decade. International collaboration between research institutions is exemplary in achieving these findings. Knowledge about intelligence is growing rapidly as researchers from many countries make progress in this field.

With the help of the high speed transmission of 5G network realtime remote guidance has been implemented to ensure longdistance c

ooperation to ensure the safety, trust and safety of surgery. It reduces the risk of surgery by allowing professionals to provide immediate and surgical intervention to the patient..

Artificial Intelligence (AI) has changed the traditional medical model, improving the quality of treatment and providing many guarantees for human health. The future development of medical intelligence is expected to be greater.

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