

# An Analysis of Artificial Intelligence in Healthcare

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**Abstract:** *This study aims to provide a summary of the use of artificial intelligence in healthcare. Artificial intelligence has had a significant impact on this subject. The fast advancement of analytics technology and the growing availability of healthcare data have led to a paradigm shift in the healthcare industry. Structured data is managed by machine learning methods such as natural language processing, deep learning neural networks, and support vector machines. Natural language processing is used to handle unstructured data.*

**Keywords:** AI applications in healthcare, Machine learning in medicine

## I. INTRODUCTION

Artificial intelligence is becoming significant in healthcare. AI has proven effective in medicine. Discussed is "Will AI replace doctors in the future eventually?" It seems unlikely soon. Sometimes it helps make better healthcare judgments. The growing availability of health care data and the rapid development of big data analysis technologies are helping build medical AI applications. Practical AI algorithms may find clinically useful information in massive data sets to aid clinical decision-making when prompted by relevant clinical questions. Changes in demography, logistical needs, faculty shortages, growing morbidity, and data innovation norms and interest are putting doctors and wellness administrations under unprecedented pressure.

AI's clinical research and healthcare applications are expanding. The study proves AI-enabled health solutions' potential. Governments and innovation clusters are collaborating on medical AI. The FDA wants more AI-enabled healthcare equipment. AI-powered medical services delivery may affect patient follow-up, clinical decision support, medical services organization, and medical care intercessions.

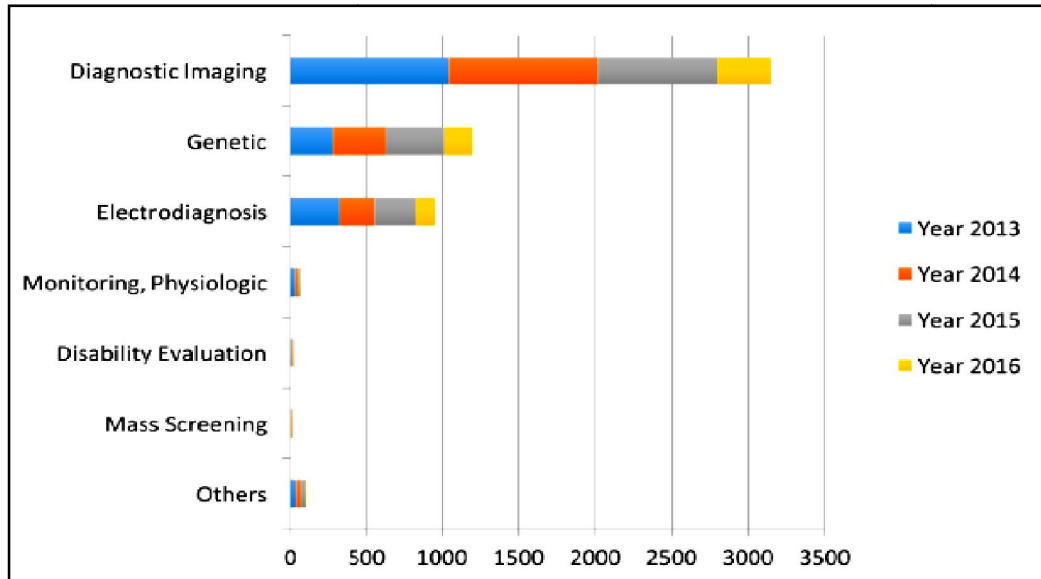
Innovative healthcare services integrate cutting-edge technology like cloud computing, IoT, and AI to provide a more efficient, useful, and tailored medical services framework. Continuous health monitoring using mobile phone or wearable technology applications empowers individuals to take charge of their health. AI-enabled patient-level health data may be given to clinicians for early illness detection, treatment plan assurance, and wellness screening.

## II. LITERATURE REVIEW

The clinical literature has extensively examined AI's advantages. Artificial intelligence (AI) may "learn" from massive patient data and improve healthcare using complex algorithms. It may also learn and self-correct to improve accuracy based on data. AI systems that deliver superior clinical knowledge from textbooks, journals, and medical practices may aid clinicians discuss patient care. Human doctors make errors in diagnosis and therapy, but AI can reduce them. In addition, an AI gadget gathers important data from a broad patient population to continually analyze and anticipate health risks.

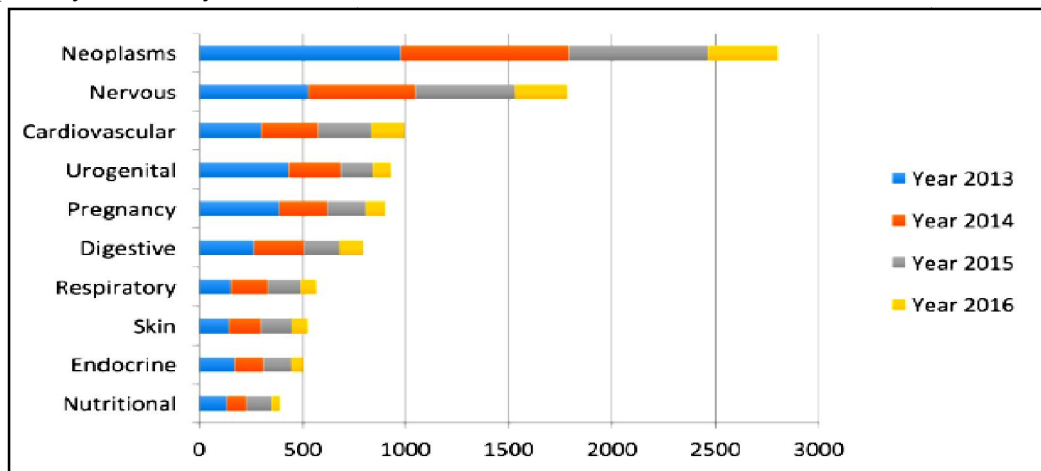
Medical information from patient discovery, screening, and treatment underpins artificial intelligence. Demographical data, medical records, digital medical equipment recordings, clinical exams, laboratory tests, and photos are clinical data.

We have two types of AI gadgets. Genomic, imaging, and EP data are analyzed by machine learning (ML) techniques first. Natural language processing (NLP) extracts features from clinical notes and medical literature to improve organized medical data.



### The types of data considered in AI Literature

Despite the fact that medical AI research is expanding, the bulk of it is focused on three disease categories: cancer, neurological, and cardiovascular. It is not surprising that the same place has high rates of all three illnesses. Since these three illnesses are leading causes of mortality, people should get a diagnosis as soon as possible to save their health from becoming worse. Furthermore, the AI system's capacity to improve imaging, genomics, EP, and EMR research techniques may lead to early detection.



### In the artificial intelligence (AI) literature, the top ten disease forms are considered.

Medical professionals curated the previous generation of artificial intelligence, which is presently overgrowing. Modern machine learning methods have been used in AI research to find data patterns that may explain complicated connections. AI helps clinical researchers comprehend patient conditions.

Intelligent robotics and AI enhance the lives of the elderly and disabled in assisted living. A man-machine interface (HMI) lets handicapped persons utilize wheelchairs and robot aids without controllers or sensors. People with impairments are independent. Such people may become self-sufficient with AI. Blind persons may work with able-bodied people in informatics and innovation on RUDO, a "ambient intelligent system." Fall detection may help seniors prevent falls and other issues. A skilled computational modeling assistant (CMA) may help biomedical examiners turn relevant models into "executable" replication models.

AI is revolutionizing healthcare by managing data, developing medicines, and doing manual jobs. All things have pros and cons. AI can examine data and enhance diagnosis, which may assist with administrative tasks, monitoring, and virtual consultations.

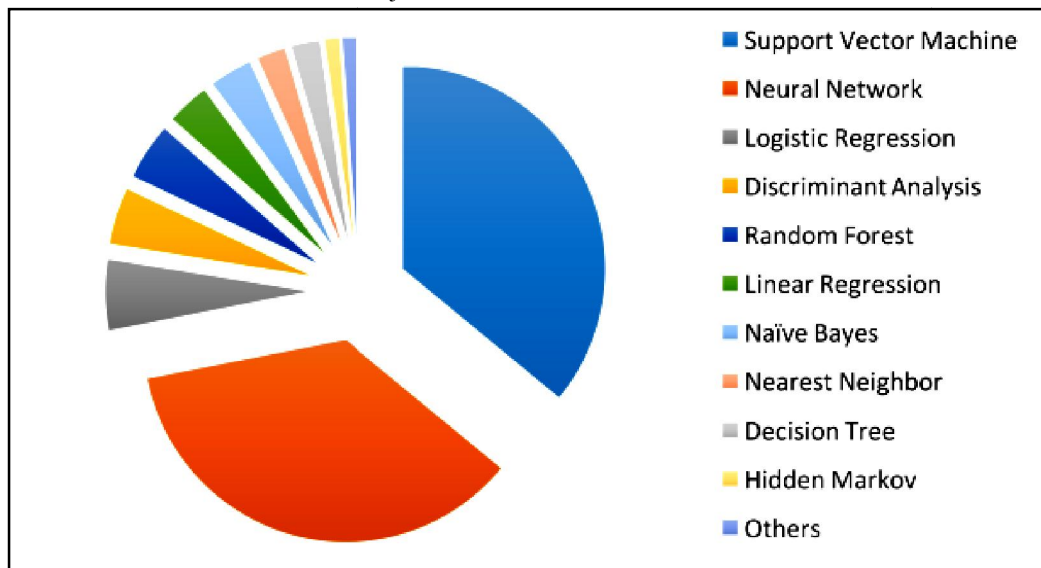
Training algorithms may be difficult due to exponential data growth. Privacy concerns may make it impossible to get some of the data needed for the model to predict the desired outcome. Since the patient's life is at danger in the digital era, switching to AI may be difficult. According to the approach, we need proof that AI will work. Given that investors and the healthcare industry are spending heavily, is it cost-effective?

**Challenges in AI development**

Artificial intelligence in healthcare has several obstacles. Training neural networks or machine learning algorithms requires plenty of data. However, objective facts are uncommon. Prejudice, noise, uneven medical data, missing facts, etc. may corrupt diverse healthcare data. The model constructed using one institution's data may not work for another. Thus, scientists must ensure that their data appropriately represents the desired patient group. Data's exponential growth and, most critically, its faultless decision-making knowledge provide the fewest challenges. The system's accountability is especially important since inaccurate AI knowledge might kill a patient. Who will pay for his death? AI or medical staff? Medical staff exploited the system to improve patient care, making this harder to reply to. Trust underpins doctor-patient relationships. When a patient trusts a doctor and thinks their sickness will be treated, medical consequences may result. When the user is unfamiliar with AI, how would the system build trust? Doctor-patient trust is essential for treatment.

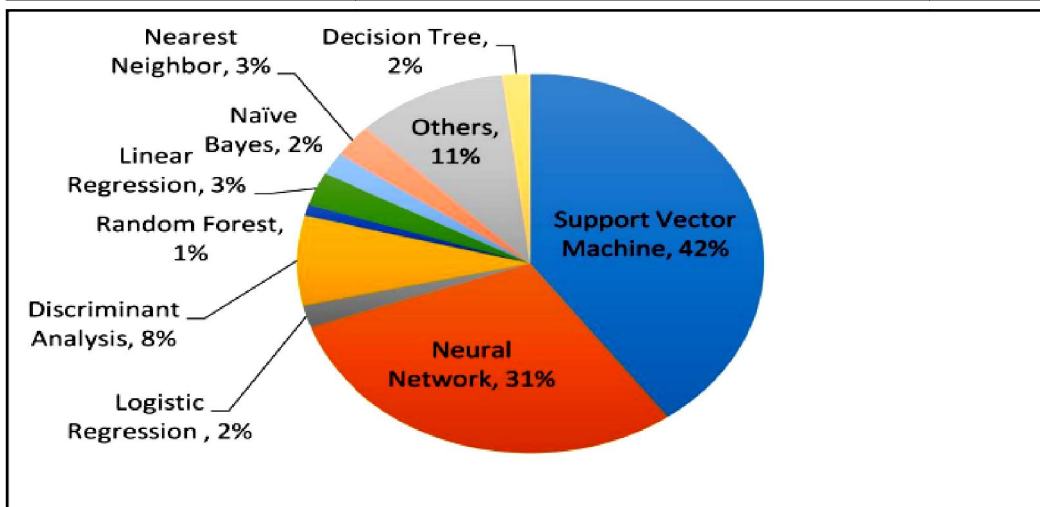
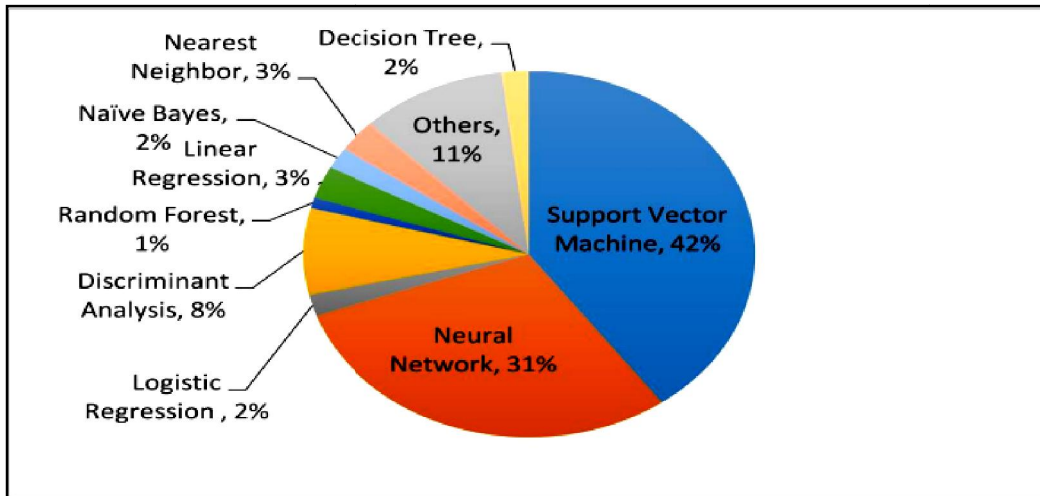
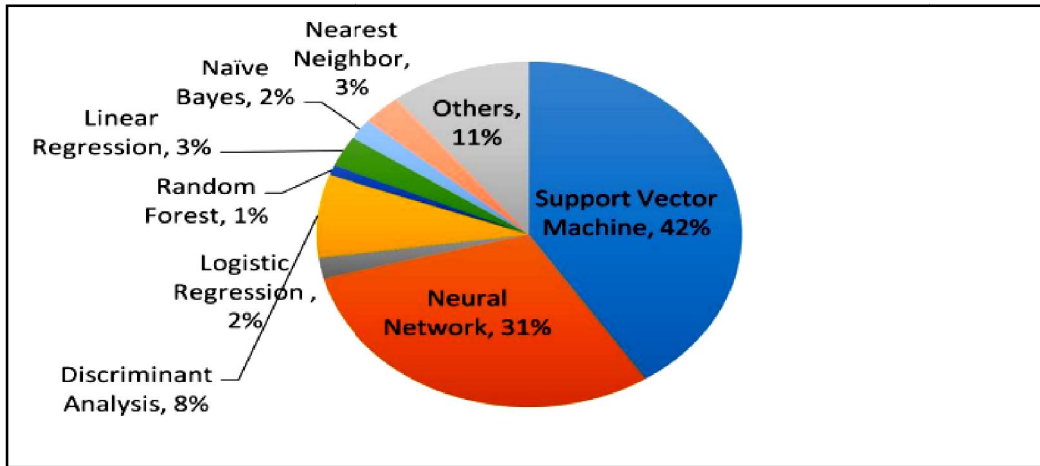
**Machine learning**

Machine learning algorithms analyze data and extract information. The machine learning system uses the patient's "characteristics" and, rarely, medical outcomes. Unsupervised and supervised machine learning algorithms exist. Unsupervised learning is good for feature extraction, while supervised learning is better for predictive modeling since it links patient data to the intended result. Partial supervised learning, a blend of unsupervised and supervised learning, has been offered as a solution to uncertain subject outcomes.



**Machine learning algorithm used in healthcare**

Healthcare AI solutions employ supervised learning more because unsupervised learning yields better clinical results. In medical applications, SVM and neural networks are the most frequent supervised learning methods.

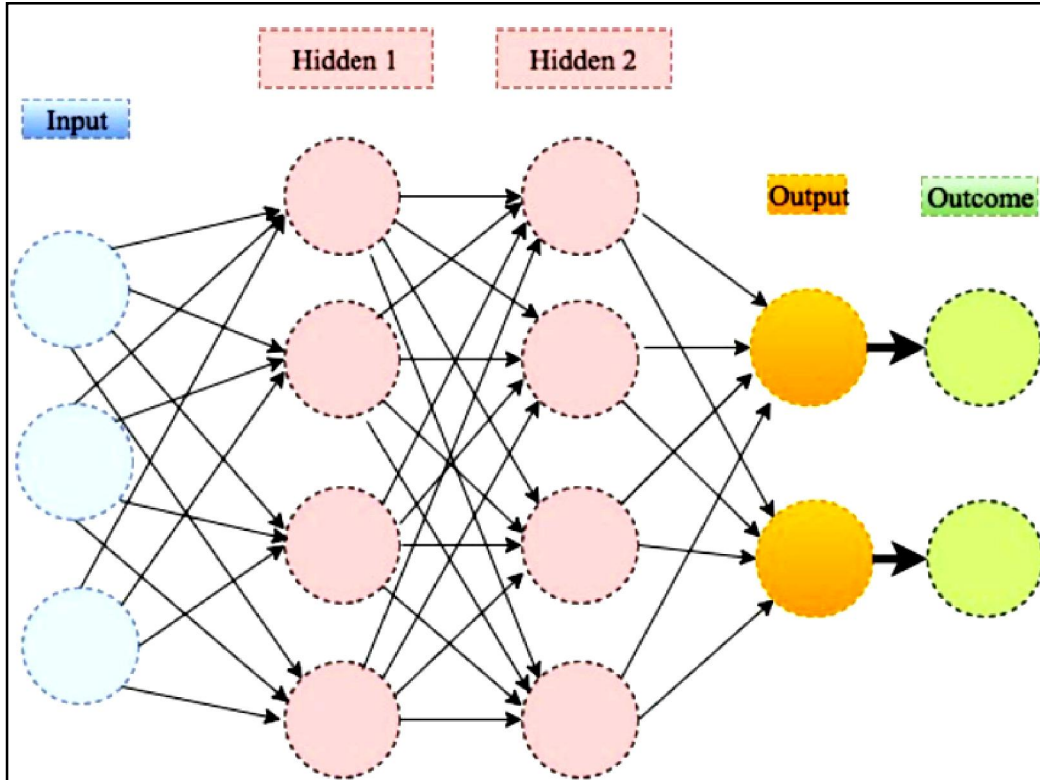


**Machine learning algorithm used for diagnosis analysis.**

**Neural networks**

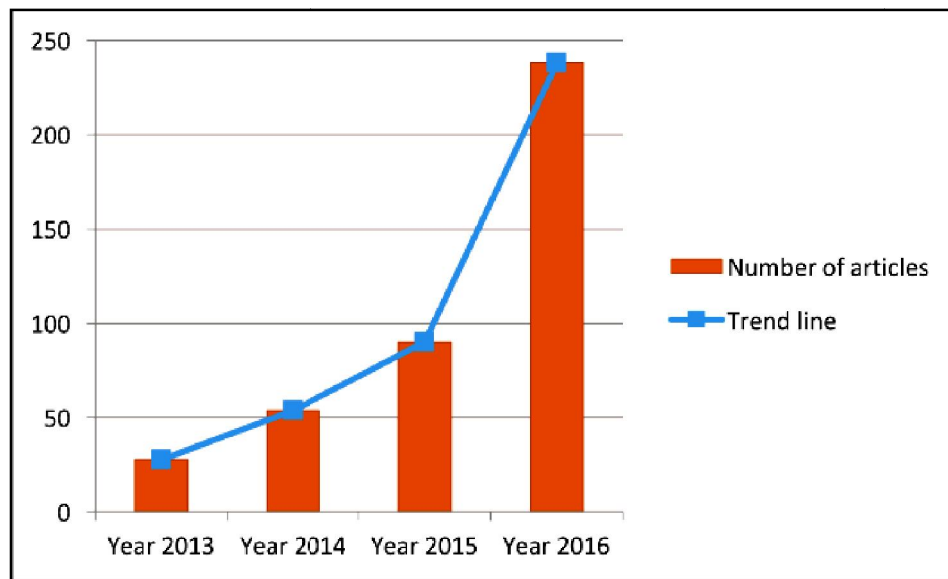
AI uses artificial neural networks. ANNs can solve issues humans and mathematicians cannot. ANNs simulate human data analysis and processing by mimicking the brain. ANNs self-learn without a curriculum. Artificial neural networks

include input, hidden, and output layers. The first layer's input layer neurons provide data to the second layer for processing. After passing through the second layer's hidden layer, activated neurons output the result via activation. For more complicated problems, the concealed layer may have several layers.

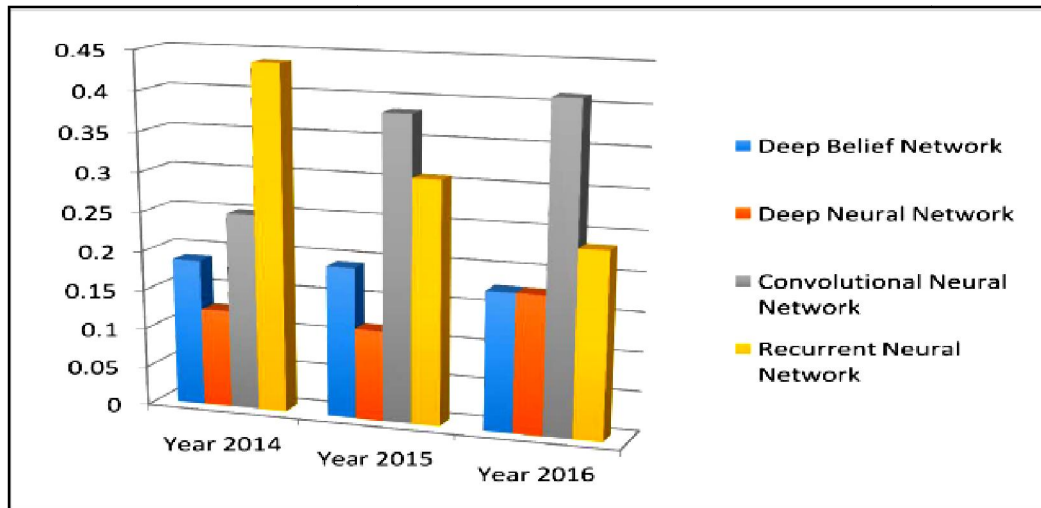


**The architecture of Artificial Neural Network**

Deep learning is a more complicated neural network. Deep learning involves multi-layered neural networks. Deep learning will find more complex non-linear patterns after digging deeper. The imaging researcher used deep learning to show that pictures are varied and vast.



The information was gathered by searching pubmed for deep learning in the healthcare and disease categories.



**The four most common profound deep algorithms, as well as their popularity**

#### Natural language processing

A computer programmer cannot understand textual clinical data like physical exams, lab results, operation papers, and discharge summaries. NLP extracts meaning from written texts to aid therapeutic decision-making. NLP pipelines focus on text processing and classification. The NLP pipeline helps physicians choose treatments, notify patients, and monitor adverse effects.

#### Applications of artificial intelligence in stroke

Over the world, millions of individuals suffer from stroke. It ranks sixth in North America but is the leading cause of mortality in China. Therefore, research on stroke therapy and prevention is crucial.

#### Early detection and diagnosis

The failure to identify early stroke symptoms meant that only a small number of patients could get timely treatment. The detection procedure comprised stages for identifying human activity and detecting the start of a stroke. A stroke warning is activated when a patient's movement significantly deviates from their typical pattern, and the patient is evaluated for treatment as quickly as feasible.

#### Prediction of outcome and analysis of prognosis

Clinical mortality and stroke diagnosis are influenced by several variables. Machine learning approaches are superior to traditional methods when it comes to improving prediction accuracy. To assess the data and increase prediction accuracy, the authors used SVM and ANN. 107 individuals who had intra-arterial therapy for a severe anterior or posterior circulation stroke were examined by Asadi et al. Applying an enhanced algorithm outperformed previous techniques for forecasting the 30-day death rate, according to Birkner et al.

### III. CONCLUSION

The justification for AI's usage in healthcare, the types of medical data it has evaluated, and the most frequent illness categories were discussed. Machine learning and natural language processing are the core AI devices investigated. We employed a novel deep learning technique together with SVM and neural networks, two popular classical methods. Any effective AI system needs NLP for unstructured text mining and mlcomponents for structured data. The IBM Watson computer pioneered this. The NLP-ML system has showed promising cancer results. AI is projected to analyze increasingly complex and practical clinical issues by leveraging massive amounts of data and expertise, improving stroke treatment decision-making. AI systems need ongoing clinical trial data to work properly. The data source must be maintained to construct and improve an AI system trained on past data.

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