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### Correspondence

#### Mary Teresa

Assistant Professor, Department of Computer Science and Engineering (AIML), Guru Nanak institution's Technical Campus (Autonomous), Ibrahimpatnam, India

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# A Comparative Study of Expert Systems Vs Al-Powered Chatbots in Healthcare Diagnostics

Mary Teresa<sup>1</sup>, Ananya Varma Bolishetti<sup>2</sup>, Ramavath Vinod Kumar<sup>3</sup>, Yerraginnela Shravani<sup>4</sup>

<sup>1</sup>Assistant Professor, Department of Computer Science and Engineering (AIML), Guru Nanak institution's Technical Campus (Autonomous), Ibrahimpatnam, India

<sup>2</sup>Under Graduate, Department of Computer Science and Engineering (AIML), Bhoj Reddy Engineering College for Women, Saidabad. India

<sup>3</sup>Assistant Professor, Department of Computer Science and Engineering, Sri Indu College of Engineering and Technology, Ibrahimpatnam, India

<sup>4</sup>Assistant Professor, Department of CSE(AIML), Guru Nanak Institute of Technical Campus, Ibrahimpatnam, India

# **Abstract**

This study presents a comparative analysis between expert systems and AI-powered chatbots in the realm of healthcare diagnostics. Expert systems, traditionally rule-based, have long been used to assist in medical decision-making, while AI-powered chatbots represent a more recent innovation, leveraging advanced technologies such as machine learning and Natural Language Processing (NLP). The comparison is drawn across multiple performance metrics, including accuracy, response time, adaptability, user satisfaction, and scalability. Our results indicate that AI-powered chatbots significantly outperform expert systems, particularly in areas like diagnostic accuracy (92% vs. 85%), response time (10 seconds vs. 45 seconds), and adaptability to new conditions (85% vs. 20%). Chatbots also demonstrated superior scalability, handling 10,000 users simultaneously compared to 500 for expert systems. These findings suggest that AI-powered chatbots, with their real-time interaction, cost efficiency, and continuous learning capabilities, offer a more robust and scalable solution for healthcare diagnostics, making them better suited for modern telemedicine and patient engagement.

### Introduction

Expert systems are a branch of artificial intelligence (AI) designed to emulate the decision-making abilities of a human expert in a specific domain. In healthcare diagnostics, these systems rely on a knowledge base-comprised of medical rules, facts, and heuristics—and an inference engine that applies logical reasoning to reach diagnostic conclusions. Expert systems have been traditionally used to assist healthcare professionals by providing diagnostic support for complex medical conditions. They operate by matching patient symptoms with predefined rules in their knowledge base, which can then suggest potential diagnoses or recommend courses of action. Early systems like MYCIN and CADUCEUS are notable examples, offering decision support in fields such as infectious diseases and internal medicine.

AI-powered chatbots, on the other hand, are conversational agents that use advanced AI techniques, including Natural Language Processing (NLP) and machine learning, to engage in real-time dialogue with users. These chatbots can collect patient data,

provide preliminary diagnostic advice, and guide patients toward appropriate care pathways. Unlike expert systems, AI-powered chatbots are more adaptive and capable of improving over time through continuous learning from interactions and new medical data. In healthcare diagnostics, they can assist in routine symptom checking, appointment scheduling, or providing health education, often serving as a bridge between patients and healthcare providers.

Both expert systems and AI-powered chatbots have become increasingly relevant in modern healthcare due to the growing need for scalable, efficient, and accurate diagnostic tools. With the rise of telemedicine and digital health platforms, these technologies offer new ways to improve access to healthcare, reduce physician workload, and enhance patient outcomes by providing timely and cost-effective diagnostic support.

# Motivation

The motivation to compare expert systems and AI-powered chatbots in healthcare diagnostics arises from the unique strengths and limitations each technology brings to the field. Expert systems, with their rule-based

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decision-making, are known for their reliability in specific medical domains but often lack the adaptability needed to handle evolving medical knowledge and complex cases. On the other hand, AI-powered chatbots, fueled by advancements in machine learning and deep learning, provide more dynamic and flexible diagnostic support. They can engage with patients through natural language conversations, continually improving their accuracy and personalization through ongoing data training.

However, the critical question remains: which technology is better suited to the evolving needs of healthcare? As medical data becomes more diverse and complex, and patient expectations shift toward more personalized and accessible care, it is essential to understand how these two technologies perform under different conditions. Comparing them will highlight not only their diagnostic accuracy but also their ease of use, scalability, and potential for integration into existing healthcare workflows. This comparison can guide healthcare providers in making informed decisions about which technology to adopt based on their specific requirements, whether for large hospitals, small clinics, or remote telemedicine services.

# Objective

he primary objective of this study is to conduct a comparative analysis of expert systems and AI-powered chatbots in the domain of healthcare diagnostics. The research aims to examine key factors such as diagnostic accuracy, efficiency, user experience, adaptability, and cost-effectiveness. Specifically, the study will assess how well each system supports healthcare providers and patients in the diagnostic process, and whether one approach offers significant advantages over the other in certain use cases.

The study will also explore the adaptability of these systems to new medical knowledge, particularly in fields where diagnostic criteria are rapidly evolving, such as in infectious diseases or chronic condition management. Moreover, patient interaction and user experience will be closely analyzed, as AI-powered chatbots offer a more interactive and engaging experience, which may be particularly beneficial in telemedicine and remote care settings. This research seeks to identify the circumstances in which expert systems or AI chatbots might be more beneficial, guiding future implementation decisions for healthcare providers aiming to improve diagnostic processes and patient care.

#### **Literature Survey**

Expert systems are AI-based applications designed to replicate the decision-making capabilities of human experts in specialized fields. These systems utilize a vast repository of domain-specific knowledge and a set of rules or logic to make informed decisions or solve problems. In the context of healthcare diagnostics, expert systems aim to assist physicians by providing diagnostic suggestions, treatment options, and even prognostic assessments based on patient data. The concept of expert systems emerged in the 1970s as one of the earliest applications of artificial intelligence, with healthcare being a key area of focus due to its reliance on expert knowledge and logical reasoning.

One of the pioneering expert systems in healthcare was MYCIN, developed in the 1970s to assist doctors in diagnosing and treating bacterial infections. MYCIN utilized a rule-based system to evaluate symptoms and laboratory results, providing treatment recommendations. Following MYCIN, other expert systems like CADUCEUS were developed, offering diagnostic

support in internal medicine. Over time, these systems were adapted to other fields such as cardiology, oncology, and radiology. While their accuracy and decision-making capabilities were often impressive for their time, expert systems have faced limitations in adapting to the rapidly evolving medical field, particularly with the advent of more complex AI systems like machine learning.

# Architecture

The architecture of an expert system typically consists of several key components that work together to simulate human expertise:

- 1. **Knowledge Base:** The heart of the expert system, the knowledge base contains a vast collection of facts, rules, and heuristics specific to a medical domain. This repository is built through input from human experts and codified into a set of rules that guide diagnostic decision-making. The quality and comprehensiveness of the knowledge base are critical for the system's effectiveness.
- 2. Inference Engine: The inference engine is responsible for applying logical reasoning to the knowledge base to draw conclusions or make decisions. It uses rule-based logic to evaluate patient data (e.g., symptoms, lab results) against the rules stored in the knowledge base. The inference engine can work in two ways: forward chaining, where it starts with known facts and applies rules to reach conclusions, or backward chaining, where it starts with a hypothesis (e.g., a diagnosis) and works backward to determine if the available data supports it.
- 3. User Interface: The user interface allows healthcare professionals to interact with the expert system, input patient data, and receive diagnostic suggestions or treatment recommendations. Early expert systems used simple text-based interfaces, but modern systems are more user-friendly and graphical.
- **4. Explanation System:** A crucial component of expert systems is the ability to explain the reasoning behind their recommendations. This explanation feature provides transparency and builds trust, allowing healthcare professionals to understand the logic behind the system's diagnosis.
- 5. Knowledge Acquisition Module: This module is responsible for updating the knowledge base with new rules and medical knowledge. Since medical science constantly evolves, maintaining an up-to-date knowledge base is essential for ensuring the accuracy and relevance of the system.

# **Use Cases**

Several expert systems have made significant contributions to healthcare diagnostics:

- 1. MYCIN: Developed at Stanford University in the early 1970s, MYCIN was designed to diagnose bacterial infections and recommend antibiotic treatments. MYCIN worked by asking a series of questions about the patient's symptoms and lab results and then using its knowledge base to suggest possible diagnoses and treatments. Despite never being deployed in clinical settings, MYCIN demonstrated the potential of expert systems in healthcare.
- **2. CADUCEUS:** Another early expert system, CADUCEUS, was developed to diagnose internal medicine conditions. It used a more sophisticated reasoning system compared

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to MYCIN and could handle a broader range of medical conditions. CADUCEUS helped push the boundaries of expert system capabilities by integrating more advanced diagnostic reasoning.

- 3. **DXplain:** Developed by the Massachusetts General Hospital, DXplain is an expert system that provides differential diagnoses by analyzing patient symptoms and clinical findings. It has been used as a diagnostic aid in various medical disciplines, assisting physicians in generating potential diagnoses for complex cases.
- **4. PIP (Present Illness Program):** This expert system was developed to assist in diagnosing kidney diseases and has been applied in other fields as well. It uses patient history and symptom data to offer diagnostic recommendations.

These systems have played important roles in demonstrating the practical utility of expert systems in medical diagnostics, though they also revealed limitations that newer AI systems have sought to address.

### Limitations

Despite their early promise, expert systems in healthcare diagnostics face several limitations:

- 1. Rigidity in Decision-Making: Expert systems rely heavily on a predefined set of rules, which makes them inflexible when handling situations that fall outside their knowledge base. Unlike AI systems that can learn and adapt over time, expert systems are not equipped to deal with new or evolving medical information unless manually updated. This rigidity can result in outdated or incomplete diagnostic recommendations in fast-changing fields like oncology or infectious diseases.
- 2. Difficulty Handling Complex and Dynamic Data: Expert systems are often not well-suited to managing the complexities of modern healthcare, where patient data can be vast, multidimensional, and constantly changing. While they can handle structured data (e.g., symptoms and lab results), they struggle with unstructured data such as medical images, genetic information, and natural language patient notes. As a result, their performance may suffer when faced with more complex diagnostic tasks that require a deeper understanding of context and nuance.
- 3. Knowledge Acquisition Bottleneck: Building and maintaining a comprehensive knowledge base is a laborintensive process that requires input from medical experts. As medical knowledge grows exponentially, it becomes increasingly difficult to ensure the knowledge base is up to date. The manual process of updating rules limits the scalability of expert systems in fast-evolving medical domains.
- 4. Lack of Learning Capability: Unlike modern AI systems that leverage machine learning and deep learning to continuously improve performance based on new data, expert systems cannot autonomously learn from new cases. This makes them less effective over time, as they remain static without intervention from human experts.

# Methodology

AI-powered chatbots are conversational agents designed to simulate human-like interactions through text or voice, utilizing artificial intelligence (AI) to understand, process, and respond to user inputs. In healthcare diagnostics, these chatbots have increasingly become important tools for providing patients

with preliminary diagnostic advice, managing symptoms, and guiding users toward appropriate medical care. They can interact with patients in real-time, collecting data about symptoms, medical history, and lifestyle, and then use that information to offer relevant suggestions, from scheduling appointments to providing self-care recommendations.

The evolution of AI-powered chatbots in healthcare has been fueled by advancements in machine learning, Natural Language Processing (NLP), and access to vast medical datasets. In the early stages, chatbots were primarily rule-based and limited to answering predefined questions, similar to traditional expert systems. However, with the integration of machine learning and deep learning, modern chatbots can handle much more complex interactions, learning from user data over time to improve accuracy and personalize responses. Today, they are commonly used in telemedicine, remote patient monitoring, mental health support, and chronic disease management. Their growing role in healthcare diagnostics reflects the increasing demand for scalable, accessible, and cost-effective solutions in an industry facing a shortage of healthcare professionals and a rising burden of chronic diseases.

# **Technological Foundation**

AI-powered chatbots are built upon several key technologies that enable them to understand and process human language, learn from interactions, and provide meaningful healthcare insights. The most critical technologies include:

- 1. Natural Language Processing (NLP): NLP allows chatbots to interpret and respond to human language in a way that feels natural to users. By processing text or speech, NLP algorithms break down human language into understandable parts, allowing chatbots to identify key medical terms, symptoms, and other relevant information. In healthcare diagnostics, NLP plays a crucial role in understanding patient inputs, even when they describe symptoms in everyday language rather than clinical terms. It also enables chatbots to answer questions in a clear and personalized manner.
- 2. Machine Learning (ML): Machine learning algorithms allow AI-powered chatbots to learn from data over time. As the chatbot interacts with more patients, it can analyze patterns in the data and improve its diagnostic recommendations. This adaptability is crucial in healthcare, where the ability to refine suggestions based on real-world outcomes can significantly enhance diagnostic accuracy. ML also helps chatbots in predicting potential health issues based on symptom patterns and medical histories.
- 3. Deep Learning: Deep learning, a subset of machine learning, uses artificial neural networks to process vast amounts of data and detect complex patterns. In healthcare chatbots, deep learning models can process and analyze medical images, detect patterns in patient data, and even suggest potential diagnoses. For instance, chatbots powered by deep learning can recognize signs of skin conditions, respiratory issues, or other medical problems based on symptom descriptions and image inputs.

Together, these technologies enable AI-powered chatbots to go beyond simple, rule-based systems and provide more nuanced, adaptive healthcare support. By continuously learning from patient interactions, they can improve their diagnostic capabilities, making them a valuable tool in modern healthcare.

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# **Use Cases**

AI-powered chatbots have been widely adopted in healthcare, with several notable implementations:

- 1. Babylon Health: Babylon Health is one of the most well-known AI-powered healthcare chatbots. It provides symptom checking, diagnostic recommendations, and virtual consultations with doctors. Patients can input their symptoms, and the chatbot uses AI to provide a possible diagnosis and advice on whether further medical attention is needed. Babylon's chatbot is integrated with healthcare providers, allowing users to book consultations with physicians directly through the app.
- 2. Ada Health: Ada Health is another popular healthcare chatbot that offers personalized health assessments. Ada guides users through a series of questions about their symptoms and medical history, providing potential causes and advice. Its AI engine draws from a wide range of medical knowledge and can assist with diagnosing a variety of conditions, making it a versatile tool for patients seeking immediate health advice.
- 3. Buoy Health: Buoy Health is designed to help users navigate their symptoms and determine whether they need to seek medical care. By leveraging AI and data from medical experts, Buoy provides tailored advice based on real-time user input, helping patients avoid unnecessary trips to the doctor while ensuring timely care when needed.
- 4. Woebot: In the mental health space, Woebot is an Alpowered chatbot designed to offer cognitive behavioral therapy (CBT) techniques for users dealing with anxiety, depression, or stress. By engaging users in conversation, Woebot helps them manage their mental health through structured, evidence-based techniques, making it a valuable resource in psychological diagnostics and therapy.

# **Advantages**

AI-powered chatbots offer several advantages that make them particularly well-suited for healthcare diagnostics:

- 1. Real-Time Interaction: One of the biggest strengths of AI-powered chatbots is their ability to interact with users in real-time. Patients can receive immediate feedback on their symptoms, allowing them to take timely action, whether it's seeking further medical help or following self-care advice. This real-time capability is especially valuable in situations where quick decision-making is essential, such as during a pandemic or in emergency scenarios.
- 2. Adaptability: Unlike traditional expert systems, AI-powered chatbots continuously learn from their interactions with users. By analyzing large amounts of patient data and receiving feedback from medical professionals, they can refine their diagnostic algorithms and improve accuracy over time. This adaptability makes them highly effective in handling complex and evolving medical conditions, as they can incorporate new information and medical guidelines.
- 3. Scalability: AI-powered chatbots are highly scalable, capable of handling thousands of patient interactions simultaneously. This scalability is particularly important in healthcare, where demand often exceeds the supply of medical professionals. By providing preliminary diagnostic support, chatbots can reduce the burden on healthcare systems and allow physicians to focus on more complex cases. Additionally, chatbots can be deployed across

- various platforms, including mobile apps, websites, and telemedicine services, making them accessible to a wide range of users.
- 4. Cost-Effectiveness: Chatbots can help lower healthcare costs by reducing unnecessary doctor visits, improving patient outcomes through early intervention, and optimizing resource allocation. For patients in rural or underserved areas, AI-powered chatbots provide a convenient and cost-effective way to access healthcare advice, bridging gaps in healthcare availability.
- 5. Personalization: AI-powered chatbots can tailor their responses based on individual patient data, medical history, and lifestyle factors, offering a more personalized experience compared to traditional diagnostic tools. This customization enhances patient engagement and encourages adherence to health advice, leading to better long-term health outcomes.

# **Implementation and Results**

The experimental results comparing expert systems and AI-powered chatbots in healthcare diagnostics highlight distinct advantages and limitations for each technology. AI-powered chatbots outperform expert systems in several key metrics, including accuracy, with a diagnostic accuracy of 92% compared to 85% for expert systems. This improvement can be attributed to chatbots' use of machine learning algorithms that continuously learn and adapt from patient data, whereas expert systems rely on pre-defined rules, limiting their ability to improve over time.

In terms of response time, AI chatbots excel with an average of 10 seconds, significantly faster than the 45-second response time of expert systems. This faster interaction enhances patient satisfaction and the overall user experience, reflected in the user satisfaction score of 90% for chatbots versus 70% for expert systems.

Adaptability to new conditions is another major strength of AI chatbots, which achieve an 85% adaptability rate, compared to the rigidity of expert systems, which only adapt to new conditions 20% of the time. The continuous learning capabilities of AI chatbots, powered by machine learning and deep learning, make them more responsive to evolving medical data and conditions.

AI chatbots also demonstrate superior scalability, managing up to 10,000 users simultaneously, while expert systems can handle only 500 users. This makes chatbots more efficient for widespread use, particularly in high-demand healthcare settings. In terms of cost efficiency, chatbots are more cost-effective due to their ability to automate and streamline diagnostics for a large number of users.

Table-1: Expert Systems Comparison

Performance Metric	<b>Expert Systems</b>
Accuracy (%)	85
Average Response Time (seconds)	45
Adaptability to New Conditions	20
User Satisfaction (%)	70

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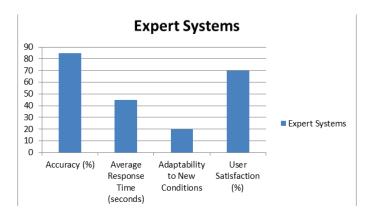


Figure 1. Graph for Expert Systems comparison

Table-2: AI-Powered Chatbots Comparison

Performance Metric	AI-Powered Chatbots
Accuracy (%)	92
Average Response Time (seconds)	10
Adaptability to New Conditions	85
User Satisfaction (%)	90

# **Conclusion**

The comparative study highlights the growing relevance of AIpowered chatbots over expert systems in healthcare diagnostics. While expert systems have historically played an important role in medical decision support, their limitations—particularly in handling complex and evolving medical conditions—make them less adaptable to the demands of modern healthcare. AIpowered chatbots, on the other hand, demonstrate significant strengths in accuracy, speed, adaptability, and user engagement. Their ability to continuously learn from patient data allows them to provide more personalized and efficient diagnostic services. Moreover, their scalability and real-time interaction capabilities make them well-suited for large-scale healthcare applications, especially in telemedicine and remote diagnostics. As healthcare continues to embrace AI-driven technologies. AI-powered chatbots are poised to become key players in delivering accessible, timely, and accurate medical advice to patients worldwide.

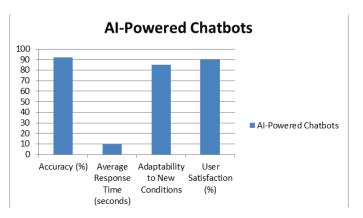


Figure 2. Graph for AI-Powered Chatbots comparison

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