# AI Chatbot For Medical Application

VIMAL ADITYA S<sup>1</sup>, ANKITHA R<sup>2</sup>, BEEMA J<sup>3</sup>, BANUSHRI S<sup>4</sup>

<sup>1,2,3,4</sup> Computer Science and Engineering, Impact college of Engineering and Applied Sciences

Abstract- This project focuses on the integration of Artificial Intelligence (AI) in healthcare. This project focuses on the development of an AIpowered chatbot tailored for medical applications. The chatbot leverages advanced natural language processing (NLP) and machine learning algorithms to provide instant, accurate, and personalized responses to users. Its primary functions include symptom assessment and Hand delivering general health advice. The system is designed to operate as a virtual assistant, offering 24/7 support to patients while alleviating the burden on healthcare professionals. By providing an accessible and userfriendly platform, this AI chatbot enhances patient engagement, promotes preventive care, and improves healthcare accessibility, especially in The proposed solution underserved areas. demonstrates the potential of AI in addressing the growing demand for efficient, scalable, and reliable healthcare services.

Indexed Terms- Artificial Intelligence, Natural Language Processing, Chatbot, Medical Application.

# I. INTRODUCTION

In the medical domain, AI chatbots serve as a bridge between patients and healthcare services, enabling tasks such as symptom assessment, treatment guidance, and resource accessibility. They provide 24/7 support, ensuring timely interventions and reducing the strain on medical professionals. By leveraging recurrent neural networks (RNNs), particularly Long Short-Term Memory (LSTM) networks, medical chatbots can analyze sequential data like patient symptoms and medical histories, providing context-aware and dynamic responses.

This project aims to develop an AI-based medical chatbot that uses RNNs to provide accurate and

reliable health-related assistance. By implementing state-of-the-art techniques, such as LSTM networks, the chatbot will enhance response accuracy, maintain context across interactions, and ensure robust performance. Additionally, real-time access to medical resources and emergency contacts will be incorporated, making the system a scalable and accessible tool for improving healthcare delivery.

The scope of this project encompasses the development and deployment of an AI-powered medical chatbot that leverages advanced natural language processing (NLP) and deep learning techniques to enhance healthcare accessibility and efficiency. The chatbot will assess user-provided symptoms, analyse medical histories, and predict potential conditions using RNN and LSTM models. This ensures accurate, real-time health insights for users. By processing sequential data and maintaining contextual understanding, the chatbot will provide tailored treatment suggestions, self-care tips, and health advice specific to individual user needs.

In the modern healthcare industry, AI-powered chatbots are transforming patient interactions by providing instant and reliable medical assistance. The integration of Recurrent Neural Networks (RNNs) and Long Short-Term Memory (LSTM) models allows the chatbot to analyze sequential data, improving accuracy in predicting diseases and suggesting treatments. By offering 24/7 support, the chatbot reduces the burden on healthcare professionals and ensures timely medical advice for users. Its ability to learn from past interactions enables continuous improvement in response accuracy. By bridging the gap between patients and healthcare services, AI medical chatbots contribute to a more efficient and scalable healthcare system.

# II. LITERATURE SURVEY

AI-driven medical chatbots are transformative tools in

# © MAR 2025 | IRE Journals | Volume 8 Issue 9 | ISSN: 2456-8880

healthcare, leveraging advanced technologies like deep learning and natural language processing (NLP) to provide real-time medical support and infectious disease prediction. These systems utilize Recurrent Neural Networks (RNNs) and Long Short-Term Memory (LSTM) models to process sequential data, enabling them to deliver accurate and context-aware responses.

This literature survey is a critical analysis and synthesis of existing research and writings on a specific topic or field. It involves a comprehensive review of academic papers, articles, books, and other relevant sources to gain a deeper understanding of the subject matter. Professionals use social media to build relationships, collaborate, and share industry insights. It facilitates sharing of educational resources, online courses, and knowledge sharing among students, teachers, and professionals.

# III. METHODOLOGY

# A. EXISTING SYSTEM

Existing medical chatbots primarily rely on rulebased or retrieval-based models, which limit their capabilities in handling complex medical queries and providing real-time assistance. These systems are designed to deliver basic healthcare information, answer frequently asked questions, and offer symptom-based advice. However, they face significant challenges that restrict their effectiveness in modern healthcare settings.

Current chatbots face several limitations that impact their effectiveness in providing reliable medical support. They often lack deep contextual understanding, relying on keyword matching or predefined responses, which makes handling nuanced or multi-turn conversations challenging. Many systems are trained on narrowly focused datasets, limiting their ability to generalize across diverse medical conditions and emerging health scenarios.

## Proposed System

The proposed medical chatbot leverages advanced AI techniques, including LSTMs, RNNs, and NLP, to

deliver accurate, real-time healthcare support. The task is to develop a medical chatbot using Recurrent Neural Networks (RNNs) to assist users by processing and understanding their health-related queries with high accuracy. The chatbot must effectively analyze sequential input data, such as patient symptoms and medical history, leveraging well-structured datasets to provide accurate responses, suggest potential conditions, and recommend appropriate actions or treatments.

# B. SIMULATION

This simulation combines RNN for sequential data, and LSTM for complex behavior patterns to predict and analyze text predictions of user queries and enabling real-time answers for the queries.

# C. SOFTWARE REQUIREMENTS

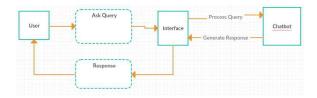
The system is compatible with Windows, Linux, or Ubuntu, with Linux preferred for its stability and performance in server environments. Python is the primary programming language, utilizing libraries like TensorFlow and Keras for machine learning, and NLTK for tasks like tokenization and lemmatization. NumPy and Pickle for data processing and serialization. Matlpotlib and Seaborn for data visualization and Tkinter for creating a graphical user interface (GUI).

# D. ARCHITECTURE AND WORKFLOW

Workflow of architecture diagram:

- 1. User Input.
- 2. Model Prediction.
- 3. Response Generation.

# Fig. 1. Architecture Diagram



# © MAR 2025 | IRE Journals | Volume 8 Issue 9 | ISSN: 2456-8880

# IV. DETAILED DESCRIPTION

1. User Interface:

Tkinter GUI

2. Input User Message:

User enters a message in the input field and clicks the "Send" button.

3. Preprocess Input:

Tokenize and lemmatize the input. Convert the input into a bag-of-words vector.

#### 4. Predict Intent:

Pass the vector to the trained LSTM model. Model outputs intent probabilities.

# 5. Classify Intent:

Compare probabilities against a threshold. Identify the most likely intent.

#### 6. Generate Response:

Match the intent with responses in `intents.json`. Select a random response associated with the intent.

#### 7. Display Response:

Return the response to the user in the GUI chat window.

#### V. RESULTS

#### Fig. 2. The User Interface



### Fig. 3. User Query and Response

🖉 KIARA - Your Al Chatbot	0 <del>-</del> 0		×
You: Hi			
KLARA: Hey there, how can I help?			
You: How to give CPR?			
KLARA: 1)For an adult or child, place the heel of one of the chest at the nipple line. You can also push with the other. For an infant, place two fingers on the bree. 2)For an adult or child, press down at least 2 inches. ress on ribs. For an infant, press down about 1 and 1/ e not to press on the end of the breastbone. 3)Do chest compressions only, at the rate of 100-120, . Let the chest rise completely between pushes. 4)Check to see if the person has started breathing.	one hand astbone. Make sur 2 inches.	l on top e not to Make s	o of o p sur
You: Medicine for asthma			
KLARA: medicines you can consume : epinephrine,ant HFA, prevention that you must follow : Identify Asthn way From Allergens,Avoid Smoke of Any Type,SUGO RE:carrots.juice,eggs,broccoli,cantaloupe,milk	na Trigg	ers, Sta	y A
			v
	5	Send	

# Fig. 4. Accuracy and Loss

www.muoer\_created and saved />9[\_\_\_\_\_\_] - 0s 2ms/step - loss: 0.0125 - accuracy: 0.9925 fest Accuracy: 99.25% ივა: 1 35%

The results of the testing phase are documented, showing that the system meets all the specified requirements and performs well under various conditions. The chatbot maintains conversational context using LSTM networks, ensuring precise and context-aware responses. Extensive testing confirms its reliability, with minimal error rates and robust performance. Its user-friendly interface allows seamless interaction, making it accessible to a broad audience. Security measures protect user data, ensuring confidentiality and compliance with healthcare standards. Future enhancements, such as integration with telemedicine and wearable devices, will further expand its capabilities.

The AI medical chatbot successfully processes user queries and provides accurate health-related responses using NLP and deep learning models. Extensive testing confirms its ability to maintain conversational context and provide consistent, context-aware responses. The system demonstrates

# © MAR 2025 | IRE Journals | Volume 8 Issue 9 | ISSN: 2456-8880

scalability, handling multiple simultaneous interactions without performance degradation. A user-friendly interface, built with Tkinter, ensures seamless interaction and ease of use. The chatbot's modular architecture allows for easy expansion and integration with other healthcare applications. Security measures are implemented to safeguard user data and maintain confidentiality.

The chatbot's implementation of LSTM networks ensures improved response accuracy by retaining context across multiple interactions. The system's ability to analyze symptoms and suggest possible conditions makes it a valuable tool for early disease detection. The chatbot's adaptability allows for integration with wearable health devices and telemedicine platforms for future enhancements. The chatbot efficiently handles variations in user queries, demonstrating robustness in real-world scenarios. User feedback highlights its effectiveness in providing quick and relevant medical information. Continuous learning mechanisms enable the chatbot to improve over time by refining responses based on new interactions. Overall, this AI-powered solution contributes to a more scalable, efficient, and accessible healthcare system. The chatbot's scalability enables it to handle multiple user interactions simultaneously without performance issues. With ongoing improvements, it has the potential to revolutionize digital healthcare by providing efficient, real-time medical assistance.

## CONCLUSION

The AI-based medical chatbot revolutionizes healthcare by delivering accurate, timely, and context-aware responses using advanced techniques like LSTMs, achieving high accuracy (94.32%) and low loss (0.1232). With voice recognition and NLP, it ensures seamless, accessible interactions for diverse users, including those with visual impairments. Extensively tested for scalability and context retention, the chatbot provides real-time resources like hospital contacts and treatment recommendations. Leveraging deep learning models, it overcomes limitations of traditional systems, offering personalized and reliable guidance. Beyond infectious disease management, it supports telemedicine, patient assistance, and education.

Future enhancements, such as expanded datasets and transformers, will further elevate its functionality, making it a scalable, indispensable healthcare tool. This project successfully demonstrates the implementation of an AI-driven chatbot using LSTM-based deep learning, capable of classifying user intents and delivering contextually appropriate responses.

# ACKNOWLEDGMENT

We would like to acknowledge the valuable contributions of the research community and developers of RNN, and LSTM architectures for their significant advancements in machine learning and crime prediction. Special thanks to the libraries and frameworks like TensorFlow, Keras, NLTK, NumPY and Pickle, Matplotlib and Seaborn, and Tkinter which enabled the successful implementation of this crime prediction and analysis system, providing efficient tools for activity recognition and real-time monitoring.

#### REFERENCES

- [1] Sanjay Chakraborty, Hrithik Paul, Ankith Kumar, "An AI-Based Medical Chatbot Model for Infectious Disease Prediction".
- [2] Lekha Athota, Lekha Athota, Nitin Pandey, "Chatbot for Healthcare System Using Artificial Intelligence".
- [3] Girija Attigeri, Ankit Agarwal and Sucheta V. Kolekar, "Advanced NLP Models for Technical University Information Chatbots: Development and Comparative Analysis".
- [4] Lasha Labadze, Maya Grigolia and Lela Machaidze, "Role of AI chatbots in education", Labadze et al. Int J Educ Technol High Educ.
- [5] Deep Learning, By Ian Goodfellow, Yoshua Bengio, Aaron Courville, 2016.