

Chatbot For Medical Diagnosis

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Abstract

Healthcare costs could be decreased and access to medical information could be enhanced with the help of medical chatbots, conversational agents created with medical applications in mind. We created a text-to-text diagnosis bot that interacts with patients about their health problems and offers a customised diagnosis based depending on their profile and symptoms. From user inputs, our chatbot technology was able to recognise ailments with an recall and precision on average are 64% and 66%, respectively. Correct the problem with these retrieved symptoms.

A recall of 64% and a precision of 72% were used to identify codes. In the end, the chatbot provided the anticipated diagnosis is made in 34% of cases. This illustrates that a medical chatbot can make a diagnosis with some degree of precision.

This essay examines the methodology, lingo, and various platforms employed in the creation of medical chatbots. Additionally, it offers some genuine, relatable, and common examples and applications of medical chatbots. The lack of funding for this initiative serves as the impetus for its work. As this initiative primarily focuses on medical institutions in cities and metropolitan areas, contemporary chatbot facilities focuses on all healthcare facilities in urban and suburban locations. As a result, it entails responding to common and answers to frequently asked questions are given whenever the person requests them.

Key words: Chatbot, Rasa, Infermedica API

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I. INTRODUCTION

This paper has the potential to be useful in conditions in which conventional methods cannot be adopted. Using natural language, our text-to-text conversational agent diagnoses patients and describes their condition. The bot requests pertinent details like age and sex as well as a list of symptoms. In order to get an accurate diagnosis, the system keeps track of previous responses and asks increasingly more detailed questions. Our system's three main parts are: (1) identifying and extracting symptoms from the user's conversation; (2) accurately mapping extracted (and potentially ambiguous) symptoms to documented symptoms and their corresponding codes in our database; and (3) creating a personalised diagnosis and, if necessary, referring the patient to the right specialist. Our system, which is different from other medical diagnosis chatbots in that it only analyses natural language to extract symptoms, may make it simpler for elderly and less tech-savvy users to communicate their symptoms and make it relatively simple to add ASR and NLG components to support spoken language. In its current state, the optimum usage for our bot would be as a tool for preliminary diagnosis that patients could use to evaluate their symptoms before to seeing a doctor, possibly using the bot's capability for specialist referral to select the best healthcare professional.

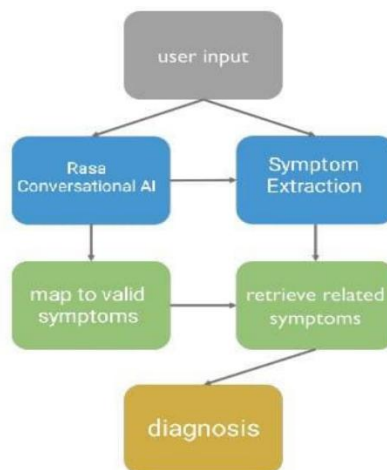
II. LITERATURE SURVEY

The article provides information on products that help consumers get precisely what they desire. Answering questions (QA) Systems that access information can be categorised as systems that aim to provide natural language responses answering questions with appropriate responses and creating a utilising a natural language property technique. In [1] A chatbot for consulting that uses techniques for emotion identification and a chat assistant platform to offer conversational services for health treatment. In [2] The suggested method is a

chatbot-based mobile healthcare service that can react right away to accidents that occur in daily life and to changes in a patient's condition if they have a chronic illness. Additionally, it suggests a structure for interactions between humans and robots that will hold up under a successful implementation of the chatbot service. Despite its extensive functionality, it is a text-based bot. In [3] To respond to users' questions, the system employs a chatbot that follows a question-and-answer protocol. An expert views and responds to the intricate inquiries and answers that are contained in the database. This chatbot takes a fair amount of time to use. In [4] The Bot Transition programme offers a structure and materials based on the guidelines of the AAP, AFP, and ACP to encourage the development of self-care skills. It is practical to use a scripted text messaging platform, and patients and caregivers seem to like it. It is only intended for those transitioning into adolescence who have particular healthcare needs. In [5]. Bigram, which divides the input sentence into two halves, completes the examination of sentence equivalence. The chatbot's data are put into the database. The database is chosen to serve as information is stored using a predictor and information storage. The process and carry out pattern recognition. This Programming can be used to create applications.

III. METHODOLOGY

The software development process has a component called implementation. The step of the paper known as implementation is where a theoretical design is transformed into a functional system. Placing a fully functional, tested software system into a real-world working context is known as implementation. The goal of implementation is to translate design specifications into source code. Making the source code as simple and clear as possible will help you achieve the main objective of implementation, which is to write the code according to its specifications. This paper uses various technologies like Scikit learn, Rasa Framework, Infermedica API. Python's **Scikit-learn** (Sklearn) library for machine learning is the most effective and reliable. Through a Python consistency interface, it offers a variety of effective tools for statistical modelling and machine learning, including classification, regression, clustering, and dimensionality reduction. This library is based on NumPy, SciPy, and Matplotlib and was written primarily in Python. **Rasa** is a tool to build custom AI chatbots using Python and natural language understanding (NLU). Rasa provides a framework for developing AI chatbots that use natural language understanding (NLU). It also allows the user to train the model and add custom actions. **The Infermedica API**, a potent NLP endpoint, handles user messages and analyses them before returning a list of symptoms and risk factors that were found in the message.



IV. EXPERIMENTS AND RESULTS

The research undertaken in this section, which describes the ubiquity of chatbots among businesses, particularly within the medical industries, is supported by the data acquired during the testing phase. The results demonstrate that using a chatbot to distribute services is a very suitable application of technology. The created chatbot gives users more accessible and effective access to about their own health. Users can communicate with their medical institution through natural language interactions.

The key goals and needs of the project were all met as it continuously evolved and advanced throughout. With this chatbot, users can communicate using only one channel and natural language. The project's concept was itself highly intriguing.

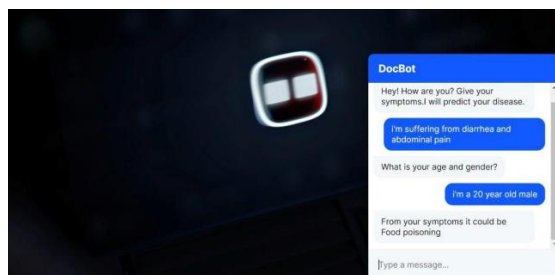


Fig 1: User Interface

Through a sequence of experiments, the result is consistent with the empirical evidence and shows that the detection rate of the proposed system exceeds previous studies.

V. CONCLUSIONANDFUTURESCOPE

Reviewing the project enables us to determine if this system will produce correct results. Due to the enormous dataset we are employing and the newest technologies, the results will be better. performance. As a result, we create a functional system. for a medical centre or hospitals to assist users in openly texting medical-related questions. system obtainsAPI output and display of the outcome. We need a computer becauseuse their language while communicating with users. Consequently, utilising Natural Language Processing a disease's system of symptoms can forecast it. Users may view related answeron their screen and use this response as a reference for more analysis.

Future iterations of our chatbots may, after diagnosing a condition, recommend doctors from their respective organisations, which would ultimately grow business and improve the reputation of the company.

In the upcoming years, chatbots will significantly improve customer experience while also helping businesses acquire market share. In the near future, we will be able to broaden the scope of queries that the chatbot can answer. By utilising the system's extensibility, it will eventually be used for voice and facial recognition to simulate a human counsellor while also engaging the patient on a deeper level.

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