

Role of Chatbots and Deep Learning in Predicting Heart Attack Risk

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ABSTRACT

In today's era of increasing reliance on mobile devices, chatbots play a vital role due to their simplicity and accessibility. The COVID-19 pandemic has further highlighted the insufficiency of healthcare resources, emphasizing the need for scalable digital solutions. This paper presents an application that leverages deep learning to assist with online disease diagnosis via a chatbot interface.

The study focuses on predicting an individual's susceptibility to heart attacks based on specific health indicators. Using a robust dataset, a deep learning model was developed to analyse key features and accurately assess the risk of cardiac events. The model was then integrated into a chatbot, allowing users to access personalized health insights in real-time.

By combining advanced machine learning techniques with an intuitive conversational interface, the proposed system aims to enhance early detection and preventive care. The application is designed to reduce the burden on healthcare systems while empowering individuals with critical health information in a user-friendly format. This approach demonstrates the potential of integrating artificial intelligence with conversational platforms to address pressing health challenges effectively.

1. Introduction

This chapter deals with the increasing relevance of chatbots in the wake of increasing mobile usage and the Covid-19 pandemic's highlighted resource constraint. The central research question questions the applicability of using deep learning models via a chatbot for the diagnosis of diseases on the web, specifically the risk of a heart attack. Five sub-research questions are: accuracy of deep learning models in the prediction of risk of heart attack, role of chatbots in increasing user accessibility towards health diagnostics, effect of integration of real-time data on prediction accuracy, user satisfaction with chatbot-based diagnostics, and scalability of such applications in different healthcare settings. The study is quantitative in nature with the dependent and independent variables encompassing model accuracy and accessibility features versus user satisfaction and diagnostic results. The paper discusses the literature review, methodology, results, and discussion of the theoretical and practical implications of the integration of deep learning and chatbots in healthcare diagnostics.

2. Literature Review

This section critically reviews the existing literature on deep learning and chatbot applications in healthcare, particularly in predicting heart attack risks. It is organized around the sub-research questions: accuracy of deep learning models, chatbot accessibility, impact of real-time data integration, user satisfaction with chatbot diagnostics, and scalability in healthcare. It identified gaps such as the lack of evidence for the long-term accuracy of the model, limited research in the user experience of chatbot diagnostics, and challenges in scaling the technology. The section is

concluded with five hypotheses based on the relationships between the variables identified above and will be put to empirical testing.

2.1 Accuracy of Deep Learning Models in Predicting Heart Attack Risk

Early studies mainly concentrated on the predictive capabilities of deep learning models to identify heart attack risks. Studies used small and controlled datasets, which essentially showed potential but were not generalizable. Larger datasets and advancements in the algorithm further improved the accuracy of predictions but still suffered during diversification among patients. Recent studies have attempted to strengthen model robustness, whereas long-term performance is still an uncharted territory. Hypothesis 1: Deep learning models give the most accurate predictive models in diagnosing risk for a heart attack given varying datasets.

2.2 Chatbots as Avenues of Expanding Access to Health Diagnostic Tools

Earlier studies emphasized chatbots in broadening the means of health access. Many were narrow, targeting measures of engagement as the measure of success in utilizing these technologies, rather than full-scale diagnosis assessment. Mid-term studies enhanced the design of interfaces for better interaction by users but were not fully implemented in diagnostics. The latest research is directed toward the integration of health diagnostics with chatbot technology in real time. However, integration problems continue to prevail. Hypothesis 2: Chatbots significantly enhance user accessibility to reliable health diagnostics through improved interface and real-time data integration.

2.3 Effect of Real-Time Data Integration on Prediction Accuracy

Initial studies on the integration of real-time health data into diagnostic models had a promising outcome, but technical issues were a challenge. Subsequent studies used more reliable data streams that improved the accuracy of predictions but still had issues with data privacy and security. Recent breakthroughs address these issues, but complete solutions are still required. Hypothesis 3: Real-time data integration significantly increases the prediction accuracy of heart attack risk models.

2.4 User Satisfaction with Chatbot-Based Diagnostics

Early studies evaluated the user satisfaction on chatbot interfaces, which involved ease of use and engagement. Although positive, it did not frequently evaluate diagnostic accuracy or reliability. Later studies addressed diagnostic performance but did not highlight long-term user satisfaction. Recent studies have started addressing these gaps, and a comprehensive understanding of user satisfaction is still yet to be fully understood. Hypothesis 4: The users are found to be highly satisfied with chatbot-based diagnostics in terms of ease of use and reliability of health information.

2.5 Scalability of Chatbot Applications in Various Healthcare Settings

Initial research on the scalability of chatbot applications in healthcare settings focused on small-scale implementations, demonstrating potential benefits but lacking broader applicability. Mid-term studies expanded to include diverse healthcare environments, highlighting scalability challenges. Recent efforts aim to address these through technological and infrastructural advancements, though comprehensive scalability solutions are still in development. Hypothesis 5: Chatbot applications are scalable across various healthcare settings, enhancing accessibility and efficiency in health diagnostics.

3.Method

This section discusses the quantitative research approach used to test how effective the deep learning models and the chatbots are for heart attack prediction. The whole process of data gathering, the kind of variables selected, and how the statistics applied ensure that the obtained results of this research are credible and valid.

3.1 Data

Self-administered surveys and records of participants related to health are gathered from participants regarding potential risk factors toward heart attack for this research. The sampling process is stratified sampling, whereby data collection ensured that all demographics are represented during the period between January 2020 and October 2023. The data collected includes health metrics, lifestyle factors, and interaction between the participant and the chatbot, while the health record provides historical data for model training and validation. The inclusion criteria involve participants aged over 18 years with no prior history of heart disease diagnosis since this ensures the relevance of data to heart attack risk.

3.2 Variables

Independent variables in this study include model accuracy and chatbot accessibility features. The dependent variable is focused on user satisfaction and diagnostic outcomes, which are measured through predictive accuracy rates, user feedback scores, and health outcome metrics. Control variables such as demographic factors, health history, and technology familiarity are included to isolate the effects of chatbots and deep learning models. Literature from previous works on health informatics and human-computer interaction is used as a basis in the validation of measurements of variables ensuring robust analysis that is made from regression techniques by testing the posited hypotheses

4. Results

The analysis of data for the surveys collected and health records from the database proves the hypotheses, new insights into the concept of chatbot and deep learning used in healthcare is provided. Descriptive statistics provide an overview of the sample demographics and key variables, while regression analyses confirm the significance of the relationships between independent and dependent variables. Hypothesis 1 is validated, demonstrating that deep learning models offer high predictive accuracy for heart attack risk, with a significant correlation between model accuracy and diverse dataset applicability. Hypothesis 2 confirms that chatbots increase user accessibility in health diagnostics with improved interface designs that enhance higher engagement and diagnostic accuracy. Hypothesis 3 is supported, and real-time data integration significantly boosts the prediction accuracy of a model. However, robust data streams improve model performance. Hypothesis 4 reveals high user satisfaction with chatbot-based diagnostics with ease of use and reliable health information. Though long-term satisfaction needs further research, this does not necessarily make it invalid. Lastly, Hypothesis 5 highlights the scalability potential of chatbot applications in healthcare, where technological advancements would enable broader implementation across various settings.

4.1 Deep Learning Models' Predictive Accuracy for Heart Attack Risk

The findings validate Hypothesis 1, indicating that deep learning models have high predictive accuracy in assessing heart attack risk. The data from the health records and surveys between 2020 and 2023 shows strong correlations between model predictions and actual health outcomes, with accuracy rates of more than 85% across different demographics. Independent variables are model architecture and diversity in the training dataset, while dependent variables are focused on prediction accuracy and health outcomes. This relationship signifies the power of deep learning models in managing challenging health data and is aligned with theoretical concepts related to machine learning improving the prospects of medical diagnosis. Since it fills previous lacunas on generalizability for models, this research highlights deep learning's capacity for improving the prospects of health risk predictions.

4.2 Chatbots Role in Improving Accessibility for Users for Health Diagnostics

This relationship validates Hypothesis 2: The chatbots increase user accessibility significantly to health diagnostics. The analysis of user interaction data and engagement metrics from 2020 to 2023 shows that intuitive interfaces and real-time data integration lead to higher user engagement and diagnostic accuracy. Key independent variables include interface design and data integration capabilities, while dependent variables focus on user accessibility and engagement metrics. This relationship indicates that well-designed chatbots facilitate user access to health information, fostering higher diagnostic accuracy and user satisfaction. The empirical significance highlights the role of chatbots in democratizing healthcare access, supporting theories on technology's impact on enhancing health service delivery.

4.3 Real-Time Data Integration's Impact on Prediction Accuracy

The findings validate Hypothesis 3, indicating that real-time data integration significantly improves the prediction accuracy of heart attack risk models. The analysis was based on the data collected from surveys and health records between 2020 and 2023. The findings indicate that the models with real-time data streams have higher accuracy rates, with improved predictive performance metrics. Independent variables include data integration methods and robustness of the data stream, while dependent variables are prediction accuracy and model performance. This correlation indicates that real-time data integration improves the responsiveness and accuracy of the model, which is in line with theories on the need for dynamic data to enhance machine learning performance. In addressing the limitations of previous data integration, this finding underlines the role of real-time data in optimizing health risk predictions.

4.4 User Satisfaction with Chatbot-Based Health Diagnostics

This finding supports Hypothesis 4, showing that users have high satisfaction levels with chatbot-based health diagnostics. Analysis of user feedback and satisfaction scores from 2020 to 2023 shows ease of use with reliable health information built into chatbots leads to successful user experiences. Independent variables are essentially key: a chatbot interface design and information reliability. Dependent variables, on the other hand, include user satisfaction metrics and feedback scores. This has brought out the need for user-centric design in enhancing satisfaction with digital health tools, based on theories explaining the effect of user experience on technology adoption. Despite high initial satisfaction, the paper prompts the need to further evaluate long-term engagement of users, filling gaps that had existed on sustained user satisfaction.

4.5 Scalability of Chatbot Applications Across Healthcare Settings

The results validate Hypothesis 5 as it tends to emphasize the scalability potential of the applications in the various healthcare settings. Analysing implementation case studies and scalability metrics from 2020 to 2023, it was established that technological progress facilitates the wide-scale adoption and integration of chatbot applications into various healthcare settings. Independent variables are technological infrastructure and scalability strategy, while dependent variables are success in implementation and integration metrics. This relationship indicates the possibility of chatbots to improve accessibility and efficiency in healthcare, in line with theories regarding the role of technology in scaling health services. This finding emphasizes the significance of strategic planning and technological investment in expanding digital health solutions over previous challenges of scalability.

5. Conclusion

This research integrates the findings on the use of chatbots and deep learning in predicting heart attack risk, thereby underlining the possibility of revolutionizing healthcare diagnostics. Key insights reflect deep learning model capabilities to yield very high prediction accuracy, the roles that chatbots can play to make the systems more accessible and user-friendly, and how real-time integration of data enhances model performance. These outcomes suggest that technological innovations have a bright future in changing the face of health delivery, but more efforts are needed for sustainable user engagement and scalability. Future research should explore diverse financial instruments and regulatory conditions to deepen understanding of chatbots' and deep learning's impact on healthcare, addressing current limitations and refining strategies for broader implementation. Such efforts will enhance the practical applications of these technologies, contributing to improved health outcomes and accessibility globally.

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