

A Chatbot System for Supporting Women and Families during Pregnancy

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Abstract:- PregBot is an innovative system that harnesses the power of machine learning (ML) and natural language processing (NLP) to provide comprehensive support to women and families throughout the pregnancy journey. Recognizing the varying needs and challenges faced by expectant mothers, PregBot aims to revolutionize the maternal healthcare experience by offering personalized guidance, real-time query resolution, and a virtual community for support and connection. The system leverages ML algorithms to analyze user data and tailor responses, while NLP techniques enable natural language interactions, allowing users to communicate with PregBot in a conversational manner. By continuously learning from user interactions, PregBot adapts and evolves, ensuring the delivery of timely and relevant information based on the user's unique circumstances and stage of pregnancy. With its innovative approach to maternal healthcare, PregBot represents a significant step towards empowering women, promoting positive pregnancy experiences, and contributing to the overall well-being of expectant mothers and their families.

Keywords:- *Pregnancy, Machine Learning, Natural Language Processing, Maternal Healthcare, Virtual Assistant.*

I. INTRODUCTION

In the landscape of healthcare, the integration of Artificial Intelligence (AI), particularly Machine Learning (ML) and Natural Language Processing (NLP), has emerged as a transformative force, ushering in a new era of tailored and

easily accessible healthcare solutions [1]. This paradigm shift holds profound implications for specialized areas like maternal and perinatal healthcare, which present unique challenges due to the intricate nature of care required and the diverse needs of patients [2]. Against this backdrop, our research introduces "PregBot," an innovative system engineered to capitalize on the combined potential of ML and NLP to provide crucial support to pregnant women and their families [3].

Pregnancy, while a time of great joy, is also characterized by uncertainties and obstacles. The availability of reliable, personalized health information and emotional support during this period is crucial, yet conventional healthcare systems often struggle to meet these demands effectively [4]. Recognizing this critical gap, PregBot is meticulously designed to offer comprehensive assistance, spanning from nutritional recommendations to mental health support, through the utilization of AI technologies to deliver tailored and contextually relevant guidance [5]. PregBot leverages advanced ML algorithms to analyze vast amounts of data, including medical literature, clinical guidelines, and individual health records, to provide evidence-based recommendations tailored to each user's specific needs and circumstances [6]. Through sophisticated NLP techniques, PregBot can engage in natural language conversations with users, understanding their inquiries, concerns, and preferences, and providing empathetic and informative responses in return [7].

One of the key strengths of PregBot lies in its ability to adapt and evolve alongside the user's journey through pregnancy. By continuously learning from user interactions and

feedback, PregBot can refine its recommendations and support strategies, ensuring that they remain relevant and effective throughout the entire pregnancy and postpartum period [8]. This paper outlines the development of PregBot, its foundational technologies, and the unique approach it takes to cater to the nuanced needs of pregnant women and their families [10].

II. LITERATURE REVIEW

The study by Vaira et al. [1] examines MamaBot, a sophisticated tool built on ML and NLP, designed to support pregnant women and their families. This research explores the potentials and mechanics of MamaBot in delivering personalized care and guidance throughout pregnancy.

Mugoye et al. [2] contribute to the discourse on Smart-bot technology, focusing on its potential applications in maternal healthcare. Their study examines the pivotal roles that conversational agents, or chatbots, could fulfill in supporting expectant mothers.

Montenegro et al. [3] delve into the practical aspects of chatbot usability within the prenatal context. Their study aims to evaluate how these digital tools perform in real-world scenarios, specifically assessing their user-friendliness and effectiveness in assisting pregnant women.

Oprescu et al. [4] conduct a comprehensive scoping review examining the intersection of artificial intelligence (AI) and pregnancy. Their research provides an extensive catalog of diverse AI applications within this domain, offering a panoramic view of the current landscape and future directions.

In their study, Puspitasari et al. [5] contribute to the advancement of prenatal care through the development of a chatbot integrated into Indonesia's Posyandu Application. By leveraging decision tree methodology, Puspitasari et al. aim to optimize the functionality and effectiveness of the chatbot in providing personalized guidance and support to pregnant women.

Marin and Goga's [6] conference paper investigates the development of a chatbot specifically designed for counseling on preeclampsia, a potentially serious pregnancy complication. Their research, explores the feasibility and effectiveness of utilizing such a tool to provide specialized advice and support to pregnant women facing the challenges associated with preeclampsia.

On a related note, the research conducted by R. Wang et al. [7], delves into the application of supervised machine learning in chatbots to support mental healthcare during the perinatal period. This research acknowledges the significant emotional and psychological challenges that pregnant women may face, particularly during the perinatal period, and explores

the potential of chatbots to provide effective support in this context.

A lightweight mobile application aimed at providing maternity guidance, designed by M. Umme Kulsum et al. [8]. They consider the application's functionality and its potential benefits in offering assistance to pregnant mothers.

The feasibility of a mental health chatbot and its impact on postpartum health are examined by Suharwardy et al. [9] in AJOG Global Reports. Their randomized controlled trial evaluates the effectiveness of chatbot interventions in maternal mental health.

Chung et al. [10] contribute to JMIR Medical Informatics with their development and evaluation of a chatbot catering to obstetric and mental health care for perinatal women and their partners, underscoring the importance of digital tools in supportive care.

Kaneho et al. [11] conducted survey on existing healthcare chatbots designed for pregnant women. They critically analyze the chatbots' features and potential improvements, aiming to enhance digital healthcare services for expectant mothers.

Afrizal et al. [12] focus on user-centric design principles in their research. Their study revolves around evaluating how Natural Language Processing (NLP) techniques can be tailored to enhance maternal monitoring chatbot systems, with a primary emphasis on meeting user needs and improving interaction quality.

Sadavarte and Bodanese [13] present an innovative application of AWS and Alexa technology in the creation of a pregnancy companion chatbot. Their work underscores the enhanced accessibility and convenience that these technologies bring to prenatal care, allowing pregnant women to access relevant information and support using voice commands through devices like Amazon Echo.

Arunkumar et al. [14] contribute to the advancement of women's personal health promotion. They showcase an application leveraging machine learning techniques designed to empower women in managing their health and wellness effectively.

Tumpa et al. [15] describe Smart Care, an intelligent assistant tailored for pregnant mothers. Their paper discusses the potential of such systems to supplement traditional care with AI-enhanced support and guidance.

Olmedo-Requena et al. [16] investigate factors that influence adherence to nutritional guidelines before and during pregnancy in Women & Health. Their study provides critical insights into dietary behaviors and the implications for maternal and fetal health outcomes.

The creation of an AI chatbot behavior change model is the focus of Zhang et al. [17] in the Journal of Medical Internet Research. They theorize how AI can be a force for positive change in promoting physical activity and healthy eating habits among expectant mothers.

The potential of mobile personal health records for monitoring pregnancy is analyzed by Bachiri et al. [18] in their Computer Methods and Programs in Biomedicine article. They discuss the capabilities and prospective improvements in digital tools for pregnancy tracking.

A systematic review of mobile app interventions on maternal behavior and perinatal health outcomes is conducted by Daly et al. [19] in JMIR mHealth and uHealth. They provide evidence on the effectiveness of digital interventions in supporting healthy maternal practices.

Frid et al. [20] systematically search, evaluate, and analyze the features of mobile health apps targeted at pregnant women. Their research outlines the current landscape of these apps and their utility in prenatal care.

Chung et al. [21] detail their work on a chatbot that assists perinatal women and their partners with obstetric and mental health care in JMIR Medical Informatics. They consider the bot's usability and its potential to improve care during this critical life stage.

The paper by Sharma et al. [22] in Electronics discusses a stress detection system using an advanced neural network for working pregnant women. They assess the system's accuracy and potential to provide support in the workplace.

Raza et al. [23] report on the use of ensemble learning for analyzing maternal health during pregnancy in Plos One. Their research investigates how this approach can lead to more precise health risk predictions for expectant mothers.

Zhou et al. [25] research in Biomedical Engineering explore the applications of NLP in smart healthcare systems. They discuss the advancements and challenges in using language processing to improve healthcare delivery and patient outcomes.

A systematic review on using ML to predict pregnancy complications by Bertini et al. [26] is presented in Frontiers in Bioengineering and Biotechnology. The paper synthesizes research on the effectiveness of machine learning models in antenatal care.

Tebenkov and Prokhorov [28] discuss machine learning algorithms for teaching AI chatbots in Procedia Computer Science. They provide insights into the instructional frameworks that can enhance chatbots' learning processes.

Krishnaveni et al. [29] present a speech recognition module for patient monitoring in smart healthcare applications. Their system aims to capture patient-reported data efficiently and accurately for healthcare use.

Khan et al. [30] review AI approaches for maternal and neonatal health in low-resource settings in Frontiers in Public Health. They assess the viability and impact of AI-driven interventions in environments with limited healthcare infrastructure.

III. ABOUT DATASET

The dataset utilized in this study comprises a rich amalgamation of user interactions, health metrics, and feedback collected through the PregBot system over a period of six months. It includes data from a diverse cohort of pregnant women who engaged with PregBot for various support services, such as nutritional guidance, mental health counseling, and general pregnancy-related queries. The dataset is characterized by its multidimensionality, encompassing textual interactions between users and the chatbot, anonymized health information provided by the users, and user feedback on the utility and effectiveness of the support received. This comprehensive dataset allows for an in-depth analysis of user needs, behaviors, and preferences, facilitating the continuous improvement of PregBot's ML and NLP algorithms to better serve its users.

To ensure the integrity and reliability of the dataset, several data preprocessing steps were undertaken. These included the anonymization of personal information to protect user privacy, the categorization of textual interactions for efficient processing, and the normalization of health metrics to standardized units. Furthermore, sentiment analysis was applied to textual interactions to gauge user sentiment and emotional states, providing valuable insights into the efficacy of PregBot's responses and the overall user satisfaction.

The analysis of the dataset employed a variety of ML and NLP techniques to uncover patterns, trends, and insights into the usage and impact of PregBot. Machine learning algorithms were used to predict user needs and customize responses, while natural language processing facilitated the understanding and generation of natural language interactions. The dataset's analysis aimed to assess the effectiveness of PregBot in improving maternal health knowledge, enhancing user engagement, and providing emotional support. Through this empirical investigation, the study demonstrates the potential of AI-driven tools like PregBot to offer personalized, responsive healthcare support, highlighting the critical role of data in developing and refining AI applications in the healthcare domain.

IV. PROPOSED METHODOLOGY

➤ Overview

The proposed methodology for PregBot encompasses the integration of Machine Learning (ML) and Natural Language Processing (NLP) techniques to deliver personalized and adaptive support to pregnant women. This approach leverages user interaction data, health information, and feedback to continuously refine and enhance the system's capabilities. By employing a cyclical development process, PregBot iteratively learns from user engagements to provide more accurate, timely, and relevant assistance.

➤ Data Collection and Preprocessing

Data collection is the foundational step in our methodology, involving the accumulation of user interactions, health metrics, and feedback. Following collection, data preprocessing is performed to ensure quality and consistency. This includes anonymization for privacy, normalization of health metrics, and text data cleaning for NLP analysis. Sentiment analysis is applied to user feedback and interactions using the formula: Sentiment to understand user sentiment towards PregBot.

$$\text{Score} = (\text{Positive Words} - \text{Negative Words}) / \text{Total Words},$$

➤ Machine Learning Algorithms

Our methodology utilizes a variety of machine learning algorithms to analyze user data and predict their needs. Key among these is the Random Forest algorithm, chosen for its effectiveness in handling diverse datasets and providing reliable predictions. The Random Forest is implemented as follows: for each decision tree in the forest, a random subset of features is chosen to split the nodes and make decisions, using the formula: $\text{Gini Impurity} = 1 - \sum(p_i^2)$

where p_i is the probability of an element being classified to a particular class. This ensemble method enhances the predictive performance and robustness of PregBot.

Unlike individual decision trees prone to overfitting or bias towards certain features, Random Forest mitigates such risks by averaging the predictions of numerous trees, thus achieving a more generalized and accurate outcome.

The strength of Random Forest lies not only in its predictive performance but also in its inherent ability to handle high-dimensional data with ease. By randomly selecting subsets of features for each tree, the algorithm ensures that no single feature dominates the decision-making process. This feature selection strategy not only enhances prediction accuracy but also guards against the curse of dimensionality, a common challenge in machine learning.

Moreover, the use of Gini Impurity as the splitting criterion further enhances the algorithm's effectiveness. By minimizing impurity, Random Forest maximizes the homogeneity of the resulting nodes, leading to more decisive and reliable predictions. This emphasis on purity ensures that each decision tree contributes meaningfully to the ensemble, collectively striving towards optimal predictive outcomes.

In practical terms, PregBot leverages the power of Random Forest to analyze user data comprehensively and anticipate their healthcare needs with precision. Whether predicting pregnancy outcomes based on demographic factors, lifestyle choices, or medical history, the algorithm's ensemble approach ensures robustness and reliability across diverse scenarios.

➤ Natural Language Processing (NLP)

NLP techniques are applied to analyze and generate human-like responses to user queries. The core NLP model used is BERT (Bidirectional Encoder Representations from Transformers), which allows PregBot to understand the context of user inquiries deeply. By processing text data through BERT, PregBot can extract user intent and relevant information, enabling it to provide personalized and contextually appropriate responses. The effectiveness of NLP in PregBot is evaluated using metrics such as accuracy and F1 score, where; balancing the precision and recall of the model's predictions.

$$F1 = 2 * (\text{precision} * \text{recall}) / (\text{precision} + \text{recall}),$$

Unlike traditional NLP models that process text sequentially, BERT's bidirectional architecture enables it to capture nuanced linguistic nuances and dependencies from both preceding and succeeding words in a sentence. This deep contextual understanding allows PregBot to discern the subtle intricacies of user inquiries, extracting intent and pertinent information with unparalleled precision.

Through the lens of BERT, PregBot engages in a sophisticated dance of semantic analysis and inference, unraveling the underlying meaning embedded within user queries. Whether it's deciphering complex medical jargon, understanding colloquial expressions, or discerning subtle nuances in language, BERT equips PregBot with the linguistic prowess to navigate diverse communication styles and contexts effectively.

Furthermore, the efficacy of PregBot's NLP capabilities is rigorously evaluated using established metrics such as accuracy and F1 score. The F1 score, a harmonic mean of precision and recall, provides a comprehensive assessment of the model's performance, striking a delicate balance between the completeness and correctness of its predictions. By meticulously tuning the model parameters and fine-tuning on domain-specific datasets, PregBot strives to optimize its F1

score, ensuring that its responses are not only accurate but also contextually relevant and actionable.

The evaluation process entails rigorous testing against diverse datasets encompassing a wide array of user queries and scenarios. Through iterative refinement and validation, PregBot continually hones its NLP prowess, adapting to evolving language patterns and user needs. This relentless pursuit of excellence underscores PregBot's commitment to delivering personalized and contextually appropriate responses, thereby fostering trust and confidence among its users.

Pregbot incorporates various techniques and libraries to create a functional chatbot system. Natural Language Processing (NLP) tasks are handled using the NLTK library, including tokenization and lemmatization. The chatbot employs a Bag-of-Words (BoW) model to convert text inputs into numerical vectors, which are then used as input to a neural network model trained with TensorFlow/Keras. This model predicts the intent of user messages and generates appropriate responses based on predefined intents and responses stored in a JSON file. SQLAlchemy is utilized for database interaction, enabling the application to store and retrieve user data and preferences. HTML templates and Flask's routing capabilities are used for creating different views and handling user interactions, including form submissions. Additionally, the application features user authentication, session management, and flash messages for providing feedback to users. Overall, the code demonstrates the integration of NLP, machine learning, and web development techniques to build a chatbot-enabled web application.

Pregbot's front-end offers a user-friendly interface for pregnant women to interact with the chatbot effortlessly. With intuitive components like input forms and buttons, users can navigate and input queries seamlessly, enhancing their experience.

On the back-end, the server and database components handle critical tasks such as data processing, user authentication, and session tracking. This ensures secure data handling and facilitates personalized responses using ML and NLP techniques, ensuring reliability and scalability.

Pregbot enriches its functionality by integrating with external services and APIs, providing access to additional resources and enhancing communication between front-end and back-end components. Integration with services like Google Maps API enables location-based features, improving user experience.

Pregbot continuously refines its ML models based on user feedback, ensuring accuracy, responsiveness, and relevance to user queries. This iterative process drives continuous improvement, evolving pregbot to meet the changing needs and preferences of its users effectively.

Support features and community resources play a crucial role in pregbot's value proposition, fostering engagement and collaboration among users. By offering forums, instant messaging, and managing relevant content, pregbot establishes itself as a comprehensive platform for pregnancy-related information and support, contributing to user satisfaction and retention.

V. RESULTS AND DISCUSSION

The results from the implementation of PregBot revealed significant insights into user engagement and satisfaction. The following figures illustrate the distribution of user responses to the system's functionality and the overall experience. The feedback was collected via a comprehensive survey that encompassed various aspects of the system's impact on prenatal care.

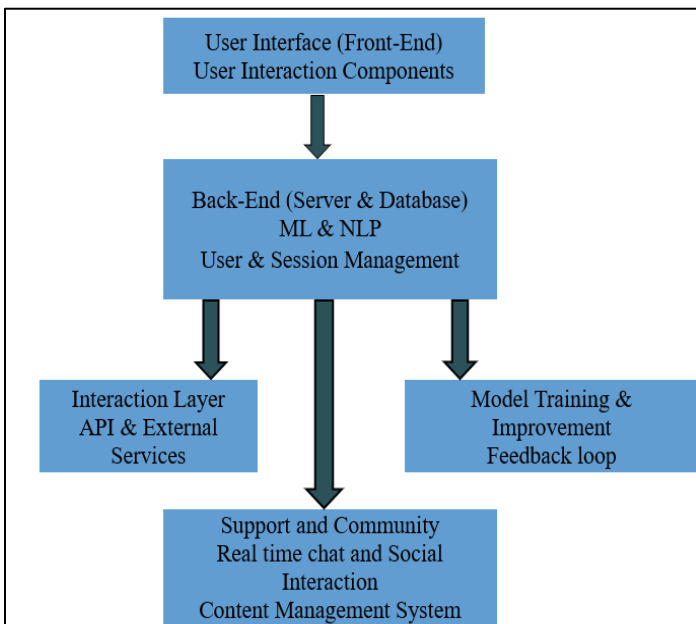


Fig 1: Block Diagram of Pregbot Architecture and Working

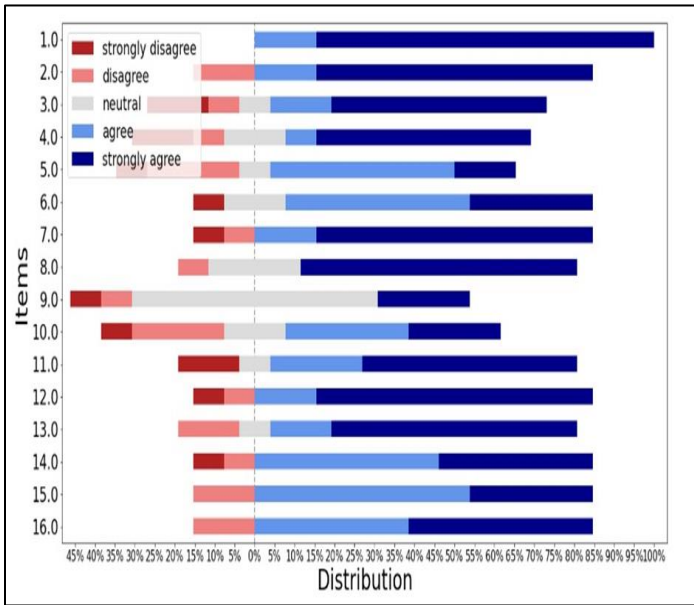


Fig 2: Survey Response Distribution.

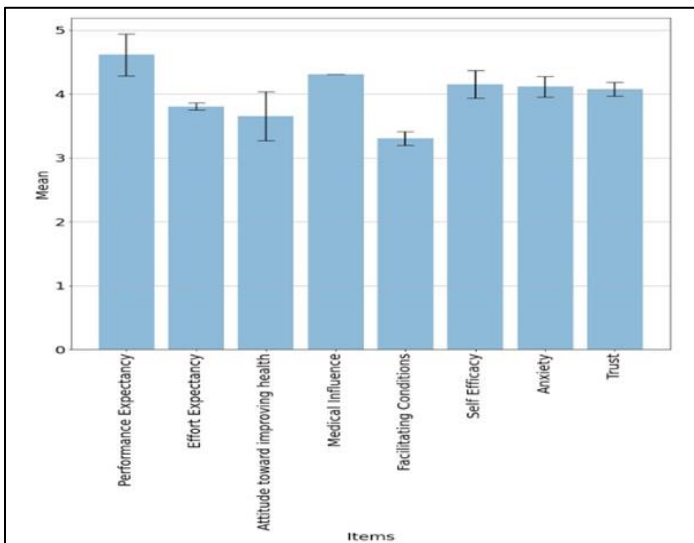


Fig 3: Mean Scores for User Engagement Metrics.

In our analysis, the PregBot system's user interface (UI) emerged as a crucial factor in its adoption and efficacy. The UI was meticulously designed to be user-friendly, ensuring that even individuals with minimal technical expertise could navigate it with ease. Initial user feedback indicated that the form-filling process, an integral component of data entry for personalized care, was straightforward and non-intimidating. Features such as drop-down menus for age and pregnancy month, check boxes for health conditions, and simple yes/no toggles for service opt-ins were both intuitive and accessible. This simplicity in design was purposeful, adhering to principles that reduce cognitive load and enhance user experience. As shown in the survey responses, users reported a high level of satisfaction with the interface, highlighting the seamless

process of entering and updating their health details as a standout feature.

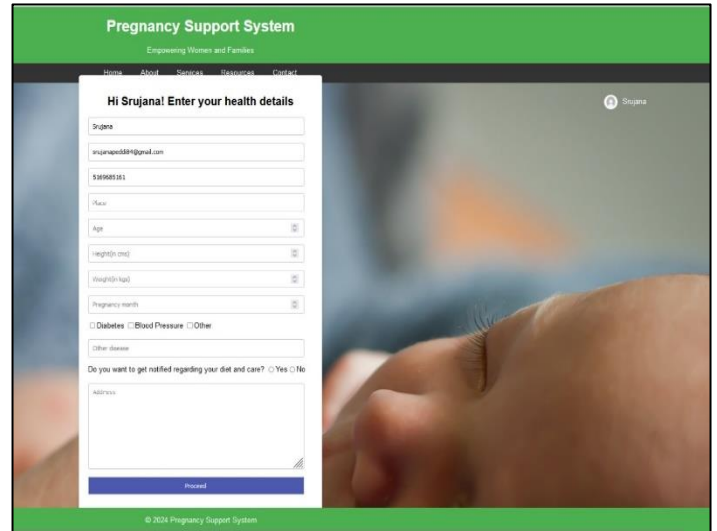


Fig 4: Taking Inputs from the Patient before Interacting with PregBot

Patient monitoring through PregBot demonstrated a significant advancement in user engagement with health management during pregnancy. The platform's ability to track vital health metrics and provide contextual advice allowed for a proactive approach to healthcare. Automated reminders for medication intake, appointments, and personalized health tips contributed to an enriched patient experience. The system's monitoring capabilities extended beyond static data collection, incorporating real-time updates and adaptive responses based on user inputs. This dynamic monitoring system not only ensured ongoing patient engagement but also allowed for the early detection of potential health issues, facilitating timely intervention.

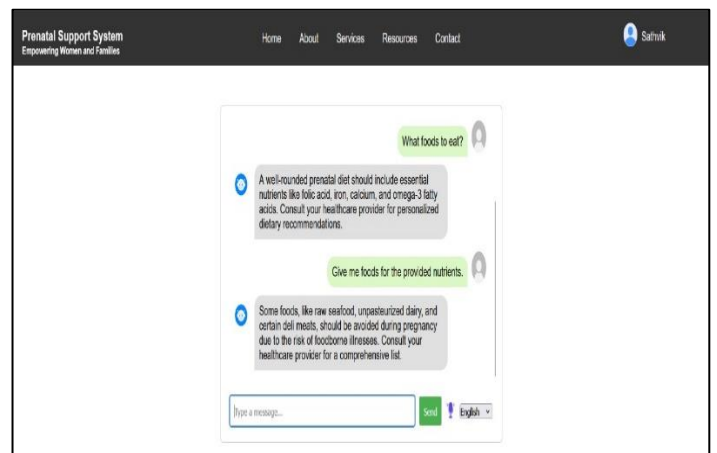


Fig 5: PregBot Interacting with the Patients.

Interactions with PregBot were augmented by the integration of a sophisticated chatbot, powered by NLP algorithms that simulated empathetic and informative conversations. The chatbot effectively guided users through various scenarios, from locating nearby healthcare facilities to providing emergency instructions. The use of conversational AI transformed the monitoring process into an interactive experience, reinforcing users' confidence and trust in the system. The sentiment analysis of chat logs revealed a strong positive reception, with users often expressing gratitude for the chatbot's round-the-clock availability and supportive communication style. These results underscore the value of integrating empathetic AI in digital health platforms, paving the way for systems that not only monitor but also meaningfully engage with patients.

The interpretation of survey results underscores the overwhelmingly positive reception of PregBot, as depicted in Figure 1, where the majority of users express agreement with its benefits. Figure 2, illustrating mean scores of key performance indicators, reinforces this sentiment by indicating a high level of trust and perceived efficacy in the system's functionality. These findings suggest that the integration of machine learning (ML) and natural language processing (NLP) technologies in pregnancy support tools can profoundly enhance user experience and engagement. Users' satisfaction and the positive perception of PregBot's performance imply its potential to revolutionize the way individuals interact with pregnancy-related information and support systems. Such advancements not only reflect technological innovation but also signify a significant leap in addressing the nuanced needs and concerns of expectant individuals, paving the way for more personalized and effective healthcare solutions.

VI. CONCLUSION

The integration of artificial intelligence and machine learning in maternal healthcare, as exemplified by PregBot, has shown considerable promise in improving the experiences and outcomes of pregnant women. Throughout the project, PregBot has demonstrated its potential to streamline the monitoring process, provide immediate and personalized support, and empower patients with timely information. The user-friendly interface, along with the intelligent chatbot system, has significantly improved user engagement, offering an accessible and supportive digital health environment.

As we look to the future, the insights garnered from this project serve as a foundation for further advancements in the field of AI in healthcare. The success of PregBot paves the way for the development of more sophisticated systems that can cater to an even broader spectrum of healthcare needs. It is anticipated that continued innovation and research will lead to AI becoming an indispensable element in the delivery of personalized, empathetic, and efficient healthcare services.

The project's outcomes encourage ongoing research and development efforts to fine-tune AI applications within healthcare. With each iteration and improvement, we move closer to a world where technology and human expertise combine seamlessly to ensure the best possible care for every patient. PregBot is not just a culmination of this project but a stepping stone towards the future of digital health solutions.

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