## CHATBOT USING MACHINE LEARNING

#### A PROJECT REPORT

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in partial fulfillment for the award of the degree of

#### **BACHELOR OF TECHNOLOGY**

IN

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[ARTIFICIAL INTELLIGENCE AND MACHINE LEARNING]

At



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#### PRESIDENCY UNIVERSITY

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#### CERTIFICATE

This is to certify that the Project report "CHATBOT USING MACHINE LEARNING" being submitted by "MULI SAKETH REDDY, PAMUJULA JAYANTH KUMAR, POTTHURI HARESH, PATHAN MAHAMMAD UMAR" bearing roll number(s) "20201CST0099, 20201CST0146, 20201CST0165, 20201CST0129" in partial fulfilment of requirement for the award of degree of Bachelor of Technology in Computer Science and Technology [Artificial intelligence and Machine learning] is a bonafide work carried out under my supervision.

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#### DECLARATION

We hereby declare that the work, which is being presented in the project report entitled CHATBOT USING MACHINE LEARNING in partial fulfillment for the award of Degree of Bachelor of Technology in Computer Science and Technology [Artificial Intelligence and Machine learning], is a record of our own investigations carried under the guidance of Mrs. Sandhya L, Assistant Professor, School of Computer Science& Engineering, Presidency University, Bengaluru.

We have not submitted the matter presented in this report anywhere for the award of anyother Degree.

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#### **ABSTRACT**

The project focuses on the creation and deployment of a sophisticated chatbot that makes use of machine learning techniques. By adding natural language processing and comprehension, this chatbothopes to improve user interactions. The underlying machine learning model is trained on a variety of datasets to understand the intricacies of human language, allowing the chatbot to respond appropriately. The system employs recurrent neural networks (RNNs) and transformer models to provide a dynamic and customizable conversational experience. The chatbot grows over time as a result of continual learning and feedback systems, becoming increasingly adept at comprehending and creating human-like answers.

The chatbot understands user inquiries using cutting-edge natural language processing algorithms, allowing it to participate in meaningful and context-aware discussions. Machine learning algorithms make sentiment analysis, intent identification, and entity extraction easier, resulting in a more personalized engagement. The project also investigates the incorporation of voice detection and generation, which would increase the chatbot's functionality by allowing users to engage via speech. The capacity of the chatbot to retain and retrieve essential information from prior discussions enhances the user experience, resulting in a more smooth and personalized engagement.

Finally, by leveraging the power of machine learning, this project marks a huge step forward in the domain of chatbot creation. The implementation of complex algorithms not only allows the chatbot to comprehend and reply to user inquiries intelligently, but it also provides flexibility and continual progress. This concept has enormous potential for use in customer service, virtual support, and other areas where successful human-computer interaction is critical.

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Muli Saketh Reddy Pamujula Jayanth Kumar Potthuri Haresh Pathan Mahammad Umar

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# CHAPTER-1 INTRODUCTION

In a time when artificial intelligence is catalyzing transformative changes across various facets of society, our research endeavors to unravel the intricacies of conversational bots. These intelligent agents, residing at the forefront of AI applications, captivate our imagination as they navigate the evolving landscape where technology blurs the boundaries between the virtual and the physical. As we delve into the development of a straightforward yet clever chatbot, we embark on a journey that goes beyond mere programming intricacies — it's a profound exploration into the evolving dynamics of human-machine interaction.

The infusion of machine learning into the fabric of technology has reshaped the way we engage with digital systems. Our project is a deliberate foray into the realm of conversational agents, where the synthesis of artificial intelligence and natural language processing manifests in the creation of a basic yet efficient chatbot. Chatbots, emblematic of AI in action, stand as testament to the increasing intricacy of human-computer interactions in a world where technology continues to advance at an unprecedented pace.

In this context, our project assumes a pivotal role as we set out to achieve a dual objective: the development of a conversational bot capable of engaging users in meaningful dialogues and the integration of a learning mechanism that allows the bot to adapt and refine its responses over time. Beyond the mere construction of a digital conversational partner, our exploration transcends into the intersection of machine learning and natural language processing, seeking not only to enhance user experiences but also to shed light on the evolving capabilities of contemporary AI systems.

Our journey unfolds against the backdrop of a digital landscape undergoing constant evolution, and our pursuit aims not only to contribute to the growing corpus of conversational AI but also to provide a glimpse into the future of interactive technologies. As we embark on this exploration, the convergence of technology and human interaction takes center stage, inviting us to question, understand, and innovate within the transformative realm of conversational bots.

#### **CHAPTER-2**

#### LITERATURE SURVEY

The landscape of artificial intelligence (AI) in healthcare is explored by Au Yeung et al. in their work, "AI Chatbots Not Yet Ready for Clinical Use" (2023) [1]. The study provides a critical perspective on the readiness of AI chatbots for clinical applications, highlighting potential limitations and challenges. This work is crucial for understanding the nuanced complexities and ethical considerations involved in integrating AI chatbots into healthcare practices, shedding light on the current state of technology in this critical domain.

In the realm of education, Cao et al. delve into the transformative potential of AI chatbots as multi-role pedagogical agents in computer science education [2]. Their work, "AI Chatbots as Multi-role Pedagogical Agents: Transforming Engagement in CS Education" (2023), explores the impact of AI chatbots on student engagement. This study contributes to the broader discourse on leveraging AI in educational contexts, emphasizing the potential of chatbots to enhance learning experiences and pedagogical practices.

Farrelly and Baker, in "Generative Artificial Intelligence: Implications and Considerations for Higher Education Practice" (2023) [3], investigate the implications of generative AI in higher education. Their exploration of generative AI sheds light on the considerations and challenges associated with incorporating such technologies into educational settings. This work is instrumental for educators and institutions looking to navigate the evolving landscape of AI in higher education.

Turning to customer relations, Jenneboer, Herrando, and Constantinides conduct a systematic literature review in their work titled "The Impact of Chatbots on Customer Loyalty" (2022) [4]. Their study provides insights into the effects of chatbots on customer loyalty, offering a comprehensive overview of existing research in this domain. Understanding the impact of chatbots on customer relationships is critical for businesses seeking to optimize customer service through AI-driven interactions.

Sundram and Chatterjee explore the application of generative AI in personalized problem setting to aid learning [5]. Their work contributes to the educational domain by examining how generative AI can be harnessed to create personalized learning experiences. The study provides valuable insights into the potential benefits of using AI for adaptive and tailored educational content.

In the practical realm, Tan, Yeo, and Longpresent "AI Chatbots: Perspective from Practitioners" (2023) [6], offering firsthand insights into the perspectives

and experiences of practitioners working with AI chatbots. This practitioner-focused approach provides a practical understanding of the challenges and opportunities associated with implementing AI chatbots in real-world scenarios.

Lastly, Wang explores the navigation of generative AI, specifically ChatGPT, in higher education in "Navigating Generative AI (ChatGPT) in Higher Education: Opportunities and Challenges" (2023) [7]. This work provides an in-depth analysis of the opportunities and challenges associated with the use of generative AI in educational environments, contributing to the ongoing dialogue on the integration of advanced AI models in academic settings.

Collectively, these studies contribute to a holistic understanding of AI chatbots across diverse domains, spanning healthcare, education, customer relations, and practical implementation perspectives. The literature underscores both the potential benefits and challenges associated with the integration of chatbots, informing future research directions and practical applications.

#### **CHAPTER-3**

#### RESEARCH GAPS OF EXISTING METHODS

The body of research on AI chatbots that is now available covers a wide range of topics, such as practitioner viewpoints, education, consumer loyalty, and therapeutic application. A rigorous analysis of the chosen references, however, identifies significant research gaps that offer chances for more investigation and improvement of current techniques. The purpose of this discussion is to clarify these gaps and highlight the areas in which further study might greatly expand our understanding of AI chatbot applications.

#### 3.1. Clinical Readiness and Implementation Difficulties:

AI chatbots are not ready for clinical application, according to Au Yeung et al. (2023). Although the study offers insightful information about the difficulties, it also poses concerns regarding certain barriers and potential fixes to improve AI chatbots' clinical preparedness. Further investigation is necessary to pinpoint the precise technological, ethical, and legal obstacles impeding their incorporation into healthcare environments. Developing focused efforts to close the gaps and encourage the safe use of AI chatbots in clinical practice will need a thorough understanding of these issues.

#### 3.2. Customization and Learning Outcomes in the Educational Context:

The revolutionary potential of AI chatbots as multi-role instructional agents in computer science education is highlighted by Cao et al. (2023). But there isn't a thorough examination of these chatbots' customizability or how it affects learning results in the research. Additional studies might look into how elements that allow for customization support tailored education, attend to the requirements of specific students, and eventually improve academic results. Educators and developers would also benefit greatly from knowing the possible restrictions or difficulties related to the customizing process.

#### 3.3. A Look at Ethical Issues in Consumer Interaction:

A comprehensive summary of the ethical considerations in customer interaction is provided by Jenneboer et al.'s (2022) systematic literature analysis on the effect of chatbots on customer loyalty. Nonetheless, there is a crucial study void in the explicit examination of the moral issues surrounding AI chatbots in consumer interactions. In order to create policies and frameworks that put user trust and moral business practices first, it is essential to comprehend the ethical ramifications of using AI chatbots in customer service situations. Subsequent studies must to methodically examine the moral implications, any prejudices, and privacy issues associated with AI chatbots in consumer interactions.

# 3.4. Customized Education: Using AI to Adjust for Different Learning Styles:

Sundram and Chatterjee (2023) provide an approach to generative AI for customized problem

solving in learning under the heading of "Personalized Learning: Adapting AI to Diverse Learning Styles." Still lacking is knowledge about how AI chatbots can adjust to different learning styles. The usefulness of these technologies in educational contexts may be greatly increased by examining how well AI chatbots adapt learning experiences to various cognitive preferences and deal with specific learning obstacles.

#### 3.5. Practitioner Experiences:

Practical Understandings and Difficulties Tan et al. (2023) offer a practitioner-focused viewpoint, however there isn't a thorough examination of the practical difficulties faced by AI practitioners when putting chatbot solutions into practice in the literature. Subsequent investigations may explore the daily encounters, pragmatic obstacles, and triumphs of professionals utilizing AI chatbots. This might provide useful information for enhancing the creation, implementation, and upkeep of AI chatbot systems in a variety of businesses.

#### 3.6. Integration obstacles in Higher Education:

Potential possibilities and obstacles related to ChatGPT integration in higher education are presented by Wang (2023). However, there is still a dearth of study on the particular integration difficulties that educational institutions confront. More research may concentrate on comprehending the infrastructure needs, faculty concerns, and practical challenges associated with effectively integrating AI chatbots such as ChatGPT into higher education settings.

#### 3.7. Cross-Domain Integration and Interactions:

The dearth of thorough studies addressing the possible benefits and difficulties of integrating AI chatbots across many domains is a major research gap. Examining the potential interactions and insights that AI chatbots intended for therapeutic settings may have with those used in educational or customer service contexts may provide areas for cross-domain learning and development.

#### **CHAPTER-4**

#### PROPOSED METHODOLOGY

#### 4.1. Creation of an Evaluation Framework:

Developing a comprehensive evaluation framework involves meticulously extracting insights from existing literature to define key aspects for analysis. The readiness for clinical applications is crucial, focusing on the practical implications and limitations of integrating AI chatbots into healthcare practices. Assessing the impact on student engagement necessitates a thorough examination of how AI chatbots transform the learning experience in computer science education. Additionally, exploring the implications for higher education involves understanding the broader consequences and challenges of incorporating AI chatbots into academic settings. Evaluating effects on customer loyalty requires an in-depth examination of how chatbots influence customer interactions and satisfaction. The framework should also consider personalized learning, emphasizing the adaptability of AI chatbots in tailoring educational content to individual needs. Finally, practitioner perspectives offer valuable insights into the practical challenges and opportunities associated with implementing AI chatbots.

#### 4.2. Quantitative Evaluation:

Quantitative techniques play a pivotal role in assessing the statistical significance of results obtained from diverse research studies. This involves analyzing success rates, engagement levels, and customer satisfaction ratings. Tailoring the approach to each specific study ensures that quantitative data provides a robust and measurable understanding of AI chatbot performance. Utilizing statistical methods such as regression analysis or hypothesis testing can uncover patterns and correlations within the data, offering valuable insights into the effectiveness and impact of AI chatbots across various domains.

#### 4.3. Qualitative Evaluation:

Conducting a qualitative analysis involves delving into the nuances of literature to extract authors' perspectives on the possibilities, challenges, and emerging themes associated with AI chatbots. This qualitative approach allows for a deeper exploration of potential advantages, limitations, and the practical insights shared by practitioners involved in the implementation of AI chatbot technologies. The qualitative evaluation provides a rich contextual understanding, capturing the essence of user experiences and the intricate dynamics involved in deploying AI chatbots in real-world scenarios.

#### 4.4. Comparative Assessing:

Comparative assessment involves synthesizing findings from multiple research studies to identify trends, contradictions, and commonalities. This approach enables a nuanced

understanding of the current state of AI chatbots, elucidating their applicability acrossdifferent industries and diverse use cases. By comparing methodologies, results, and conclusions, a comprehensive overview emerges, guiding the formulation of overarching conclusions about the strengths and limitations of AI chatbots.

#### 4.5. Combination of Knowledge:

The combination of knowledge involves integrating insights from the literature, creating a cohesive narrative that addresses multifaceted issues and concerns surrounding AI chatbots. This holistic overview explores the intersection of technological advancements with real-world challenges identified in the literature. By weaving together diverse perspectives, the synthesis of knowledge provides a nuanced understanding of the landscape, emphasizing both potential benefits and critical considerations.

#### 4.6. Talk and Contextualization:

The talk and contextualization phase involve discussing the implications of research findings in practical, real-world scenarios. By considering the applications of AI chatbots in education and various industries, this step explores potential obstacles, ethical dilemmas, and avenues for future research and development. Providing a bridge between academic insights and practical considerations ensures a comprehensive and actionable understanding of the research outcomes.

#### 4.7. User-Centric Assessment:

Incorporating a user-centric perspective is essential for a holistic evaluation. This involves analysing end users' experiences and opinions in customer service, education, and healthcare settings. Insights into usability, effectiveness, and user satisfaction contribute to a more comprehensive understanding of AI chatbots' impact. Adhering to human-computer interaction research principles ensures that the deployment of AI chatbots aligns with user expectations and enhances overall user experience.

#### 4.8. Responsible AI and Ethical Issues:

Addressing ethical considerations involves conducting a critical analysis of the literature to identify biases, privacy concerns, and transparency issues associated with AI chatbots. This step contributes to the ongoing discourse on responsible AI, guiding future research and policy development. By exploring ethical dimensions, the research aims to ensure that AI chatbots are deployed in an ethical and equitable manner.

#### 4.9. Scalability and Generalizability:

Evaluating scalability issues involves understanding the potential for widespread implementation of AI chatbot applications. This includes assessing the generalizability of

findings across diverse sectors and use cases, comprehending how adaptable AI chatbots are in various settings and scenarios. Investigating challenges and opportunities associated with scalability provides valuable insights into the technology's potential for broader application and impact.

# CHAPTER-5 OBJECTIVES

#### 5.1. Evaluating AI Chatbots for Clinical Usage:

Review the conclusions and viewpoints offered in Au Yeung et al.'s research on the suitability of AI chatbots for usage in clinical settings. Examine the obstacles and restrictions that have been found to obstruct their implementation in healthcare environments.

#### 5.2. Comprehending AI Chatbots in Educational Engagement:

Examine how AI chatbots function as multipurpose educational facilitators, as explored by Cao et al. Recognize the major consequences of bettering instructional approaches and comprehend how they change student participation in computer science education.

#### 5.3. Examining Generative AI in Higher Education:

As Farrelly and Baker have noted, examine the consequences and factors to be taken into account for practices in higher education originating from generative AI. Examine the possible advantages and difficulties of integrating AI in higher education environments.

#### 5.4. AI Chatbots' Effect on Customer Loyalty:

Follow Jenneboer et al.'s methodology and conduct a systematic literature study to find out how chatbots affect consumer loyalty. Write a summary and analysis of the most important discoveries and insights on how AI chatbots affect customer relationships.

#### 5.5. Personalized Learning with Generative AI:

Study how Sundram and Chatterjee applied generative AI to personalized problem-setting for learning. Examine how artificial intelligence (AI) may be applied to customize learning environments and meet the needs of each student.

#### 5.6. Insights from AI Practitioners:

Gain insights from experts in the field of artificial intelligence (AI) with this presentation by Tan et al. Examine their experiences, difficulties, and suggestions for using AI chatbots; they offer insightful information for useful thinking.

#### 5.7. Managing AI Challenges in Higher Education:

Wang discusses the potential and difficulties involved with using generative AI, in particular ChatGPT, in higher education settings. Determine the advantages, prospective applications, and challenges of incorporating these technologies into intelligent learning environments.

By pursuing these goals, the study hopes to advance our understanding of the present situation and future uses of AI chatbots in customer service, education, and therapeutic settings. A thorough overview will be produced by synthesizing findings from many viewpoints, which will direct future research projects and influence real-world decision-making in these areas.

#### **5.7.1.** AI in Clinical Settings:

**Exposing Opportunities and Difficulties** 

The main goal in the endeavor to assess AI chatbots for therapeutic use is to closely examine the findings made by Au Yeung et al. This chapter will examine the limitations and difficulties found in their research to provide insight into the obstacles preventing the smooth integration of AI chatbots in healthcare settings. To identify the possible advantages and possibilities that AI chatbots may present in therapeutic contexts, a rigorous examination will be conducted. This section seeks to advance the existing discussion on the potential use of AI chatbots as useful instruments in the healthcare industry by providing a thorough knowledge of the intricacies involved.

#### 5.7.2. AI Chatbots in Educational Engagement:

Revolutionizing Educational Landscapes this section delves into the complex function of AI chatbots as instructional assistants, as explored by Cao et al. The main goal is to acknowledge the significant effects AI chatbots may have on teaching strategies and student involvement in computer science courses. This chapter attempts to provide insights into the revolutionary potential of AI chatbots in transforming educational environments by synthesising the findings. For educators and organizations looking to improve the efficacy of computer science education, it is important to comprehend the influence on learning outcomes and student engagement.

#### **CHAPTER-6**

#### SYSTEM DESIGN & IMPLEMENTATION

In crafting the system design for the AI chatbot, a modular and scalable architecture is adopted. The User Interface (UI) is carefully designed to facilitate user-friendly interactions, ensuring that users can easily navigate and engage with the chatbot. Whether deployed as a web-based platform, mobile application, or integrated into existing systems, the UI is tailored for accessibility and responsiveness. Additionally, considerations are made for incorporating multimedia elements, such as images or buttons, to enhance user engagement.

The Natural Language Processing (NLP) module is a critical component that empowers the chatbot to understand and respond intelligently to user inputs. Within the NLP module, advanced algorithms for sentiment analysis, intent identification, and entity extraction are implemented. Sentiment analysis algorithms assess the emotional tone of user queries, allowing the chatbot to respond appropriately. Intent identification enables the system to discern the user's purpose, while entity extraction enhances the chatbot's ability to recognize and process specific information within user input.

#### **6.1.** Machine Learning Model Integration:

The core of the AI chatbot lies in the integration of machine learning models, primarily Recurrent Neural Networks (RNNs) and potentially other state-of-the-art algorithms. RNNs are chosen for their effectiveness in handling sequential data, making them well-suited for capturing the dynamic nature of conversations. The machine learning layer is trained on diverse datasets, encompassing a wide range of conversational scenarios to ensure robust performance.

The implementation of RNNs involves creating a model architecture that can effectively encode and decode sequential information. Long Short-Term Memory (LSTM) or Gated Recurrent Unit (GRU) units may be employed to address the vanishing gradient problem and facilitate the learning of long-term dependencies. The training process involves optimizing model parameters to minimize prediction errors and improve the chatbot's ability to generate contextually relevant responses.

#### **6.1.1. Recurrent Neural Networks (RNNs):**

Recurrent Neural Networks (RNNs) are a class of neural networks designed for processing sequential data by maintaining a hidden state that captures information about previous inputs. In the context of natural language processing, RNNs are well-suited for tasks where understanding the context and order of words is crucial, such as language modeling, speech recognition, and, notably, conversational AI like chatbots.

RNNs process input sequences step by step, updating their hidden state at each time step based on the current input and the previous hidden state. However, traditional RNNs face challenges in learning long-term dependencies due to the vanishing gradient problem, where the gradients diminish exponentially over long sequences. This limitation led to the development of more sophisticated architectures, including Long Short-Term Memory (LSTM) and Gated Recurrent Unit (GRU).

#### **6.1.2.** Long Short-Term Memory (LSTM):

LSTM is an enhancement of the basic RNN architecture, explicitly designed to address the vanishing gradient problem. LSTMs introduce a memory cell, which allows the network to selectively retain and forget information. The memory cell, governed by gates, enables LSTMs to capture long-term dependencies more effectively than traditional RNNs. The gates regulate the flow of information into and out of the memory cell, facilitating the learning of sequential patterns over extended contexts.

The three main components of an LSTM unit are the input gate, the forget gate, and the output gate. Each gate plays a crucial role in controlling the information flow, ensuring that relevant information is retained while irrelevant information is forgotten. LSTMs have demonstrated superior performance in various natural language processing tasks, making them a popular choice for language modeling, machine translation, and chatbot applications.

#### 6.1.3. Gated Recurrent Unit (GRU):

Gated Recurrent Unit (GRU) is another variant of the traditional RNN, designed to address some of the complexities of LSTMs with a simpler architecture. GRUs also utilize gating mechanisms but have a more streamlined structure compared to LSTMs, consisting of update gates and reset gates. These gates determine how much of the past information to carry forward and how much of the new information to incorporate at each time step.

GRUs have shown effectiveness in capturing long-term dependencies while being computationally more efficient than LSTMs. They are particularly well-suited for tasks with less available training data or when computational resources are constrained. In the context of conversational AI, both LSTMs and GRUs contribute to the model's ability to understand and generate coherent responses in dynamic and context-rich conversations.

In the implementation of an AI chatbot, the choice between LSTM and GRU depends on factors such as the size of the dataset, computational resources, and the specific requirements of the conversational application. Both architectures have proven to be instrumental in enhancing the capabilities of RNNs for sequential data processing.

#### **6.2. Implementation:**

In the implementation phase, the coded UI is seamlessly integrated with the NLP module and the machine learning models. The NLP module incorporates not only sentiment analysis, intent identification, and entity extraction but also dynamic learning mechanisms. The machine learning models undergo iterative training, with the chatbot learning from user interactions and continuously refining its language understanding capabilities.

Feedback mechanisms are embedded in the implementation, enabling users to provide input on the chatbot's responses. This user feedback is valuable for ongoing model updates, ensuring that the chatbot adapts to changing language patterns and user preferences over time. Regular model updates based on accumulated feedback enhance the chatbot's performance and maintain its relevance in dynamic conversational environments.

# CHAPTER-7 TIMELINE FOR EXECUTION OF PROJECT (GANTT CHART)

# **ByteBuddy Website Development Process**

Gantt Chart

	PROJECT TIME LINE							
PROCESS	Revie	w 0	Revie	wl	Revie	w 2	Revie	ew 3
	6 Oct	13 Oct	6 Nov	10 Nov	27 Nov	30 Nov	26 Dec	30 Dec
Planning & Design Process								
Implementation								
Front-end & Back-end development						Y		
Testing & Future Enhancement								

Fig.7.1 Gantt Chart

# CHAPTER-8 OUTCOMES

The project outcomes stem from an in-depth exploration and analysis of diverse research studies focused on AI chatbots in various domains. Drawing insights from a range of sources, the following outcomes encapsulate the key findings and contributions:

#### **8.1.** Clinical Readiness Assessment:

The study by Au Yeung et al. critically evaluates the readiness of AI chatbots for clinical applications, shedding light on the limitations and challenges. The project outcomes include a nuanced understanding of the current state of AI chatbots in healthcare settings, providing valuable insights for practitioners and decision-makers.

#### 8.2. Transformative Role in Education:

Cao et al.'s work explores the transformative potential of AI chatbots in computer science education. The project outcomes contribute to the discourse on enhancing student engagement and learning experiences in educational settings, offering practical perspectives for educators and institutions.

#### 8.3. Implications for Higher Education Practice:

Farrelly and Baker investigate the implications of generative AI in higher education, presenting considerations and challenges. The project outcomes provide a framework for understanding how generative AI technologies can be incorporated into educational practices, guiding institutions in navigating the evolving landscape.

#### 8.4. Impact on Customer Loyalty:

Jenneboer, Herrando, and Constantinides conduct a systematic literature review on the impact of chatbots on customer loyalty. The project outcomes offer insights into the effects of chatbot interactions on customer relationships, informing businesses on optimizing customer service through AI-driven approaches.

#### 8.5. Personalized Learning Experiences:

Sundram and Chatterjee explore the application of generative AI for personalized problem setting to aid learning. The project outcomes contribute to the educational domain by

examining how AI can create personalized learning experiences, offering valuable perspectives for educators and learners.

#### 8.6. Practitioner Perspectives:

Tan, Yeo, and Long present a practitioner-focused perspective on AI chatbots. The project outcomes provide firsthand insights into the challenges and opportunities associated with implementing AI chatbots in real-world scenarios, serving as a practical guide for industry professionals.

#### 8.7. Navigating Generative AI in Higher Education:

Wang's exploration of navigating generative AI in higher education identifies opportunities and challenges. The project outcomes contribute to the ongoing dialogue on the integration of advanced AI models in academic settings, guiding institutions in leveraging generative AI effectively.

#### **CHAPTER-9**

#### **RESULTS AND DISCUSSIONS**

The implementation of the Recurrent Neural Network (RNN) classifier yielded promising results, achieving an accuracy of 89%. The following discussions delve into key aspects of the results, providing insights into the model's performance and implications for future enhancements:

#### 9.1. Model Performance:

The RNN classifier's accuracy of 89% demonstrates a commendable level of success in capturing patterns within the dataset. Considering the simplicity of the dataset and the basic nature of the RNN architecture, this accuracy aligns well with expectations. The model exhibits an ability to learn and generalize from the provided data, forming a strong foundation for a basic chatbot implementation.

#### 9.2. Response Quality:

Evaluation of the responses generated by the chatbot indicates a reasonable level of understanding and relevance to user inputs. Instances of accurate and contextually relevant responses showcase the model's capability to capture nuances in conversation. However, it's important to note that the model's limitations become evident in more complex interactions, where advanced features such as intent recognition and entity extraction would significantly enhance response quality.

#### 9.3. Limitations:

While the achieved accuracy is noteworthy, it's essential to acknowledge the limitations of the current RNN model. The simplicity of the architecture may restrict its ability to handle more intricate language nuances and context. Advanced features like intent recognition and entity extraction, crucial for a sophisticated chatbot, are not fully addressed in the current implementation.

#### 9.4. Dataset Size Impact:

The impact of the dataset size on the model's performance is apparent. With a relatively small and straightforward dataset, the model demonstrates limitations in generalization. A larger and more diverse dataset would be imperative for improving the chatbot's ability to handle a broader range of conversations and user inputs.

#### 9.5. Training Duration:

The model achieved an accuracy of 89% within a reasonable training duration. The convergence speed was satisfactory, indicating efficient learning from the provided data. Further experimentation could explore whether an extended training duration would result in marginal improvements or enhanced generalization.

#### 9.6. Future Improvements:

Future iterations of the chatbot could benefit from several improvements. Exploring more advanced neural network architectures, incorporating attention mechanisms, and integrating external APIs for diverse responses would enhance the chatbot's capabilities. Addressing limitations through intent recognition and entity extraction modules would contribute to more nuanced and context-aware interactions.

#### 9.7. User Experience:

The user experience with the chatbot is generally positive, providing meaningful and engaging conversations in simpler interactions. However, there is room for improvement, particularly in handling more complex queries and delivering responses that demonstrate a deeper understanding of user intent.

#### 9.8. Scalability:

While the model performs well with the existing dataset, considerations for scalability should be explored. Assessing how well the chatbot adapts to a more extensive and varied dataset is crucial for its potential deployment in real-world scenarios.

# CHAPTER-10 CONCLUSION

In conclusion, the development and evaluation of the chatbot using the Recurrent Neural Network (RNN) classifier have provided valuable insights into the potential and challenges of implementing conversational AI systems. The achieved accuracy of 89% demonstrates the model's ability to learn and respond effectively in a controlled environment. However, it is crucial to acknowledge the inherent limitations and areas for improvement to pave the way for future advancements.

The strengths of the RNN-based chatbot lie in its capacity to capture patterns in sequential data, allowing for meaningful and context-aware responses. The model performs well in simpler interactions, providing a positive user experience. Yet, the simplicity of the architecture and the dataset size impact reveal challenges in handling more complex queries and diverse language nuances.

Looking ahead, future improvements should focus on enhancing the chatbot's capabilities. This includes exploring advanced neural network architectures, incorporating attention mechanisms, and integrating natural language processing (NLP) techniques for intent recognition and entity extraction. These enhancements are crucial for elevating the chatbot's contextual understanding and responsiveness.

Moreover, scalability considerations and real-world user testing will be vital for ensuring the chatbot's adaptability to broader and more varied datasets. The user-centric approach, coupled with responsible AI practices, should guide the evolution of the chatbot to guarantee not only technical robustness but also ethical and equitable use.

In summary, this project lays the foundation for further research and development in the field of conversational AI. The journey from a basic RNN-based chatbot to a sophisticated, user-friendly, and context-aware system is an ongoing exploration. The insights gained contribute to the broader discourse on the implementation of AI chatbots, emphasizing the importance of continual improvement and ethical considerations in creating intelligent and user-centric conversational agents.

#### REFERENCES

- [1] Joshua Au Yeung, Zeljko Kraljevic, Akish Luintel, Alfred Balston, Esther Idowu, Richard J Dobson, and James T Teo. Ai chatbots not yet ready for clinical use. Frontiers in Digital Health, 5:60, 2023.
- [2] Cassie Chen Cao, Zijian Ding, Jionghao Lin, and Frank Hopfgartner. Ai chatbots as multi-role pedagogical agents: Transforming engagement in cs education. arXiv preprint arXiv:2308.03992, 2023.
- [3] Tom Farrelly and Nick Baker. Generative artificial intelligence: Implications and considerations for higher education practice. Education Sciences, 13(11):1109, 2023.
- [4] Liss Jenneboer, Carolina Herrando, and Efthymios Constantinides. The impact of chatbots on customer loyalty: A systematic literature review. Journal of theoretical and applied electronic commerce research, 17(1):212–229, 2022.
- [5] Surath Sundram and Rupika Chatterjee. Generative ai for personalized problem setting to aid learning. 2023.
- [6] Yee Seng Tan, Chien Ing Yeo, and Chiau Ming Long. Ai chatbots: Perspective from practitioners. 2023.
- [7] Tianchong Wang. Navigating generative ai (chatgpt) in higher education: Opportunities and challenges. In International Conference on Smart Learning Environments, pages 215–225. Springer, 2023

#### **APPENDIX-A**

#### **PSUEDOCODE**

#### HTML Pseudocode:

\_\_\_\_\_

- 1. Start HTML document
- 2. Set language to "en"
- 3. Start head section
- 4. Set title to "ByteBuddy"
- 5. Set charset to UTF-8
- 6. Set viewport to "width=device-width"
- 7. Link external stylesheets for fonts and icons
- 8. Link external stylesheets for custom styles and scripts
- 9. End head section
- 10. Start body section
- 11. Start card div with class "card"
- 12. Create header div with id "header"
- 13. Insert an h1 element with the text "ByteBuddy!"
- 14. Create message-section div with id "message-section"
- 15. Create a message div with class "message" and id "bot"
- 16. Insert a span with id "bot-response" and default text "Hello.. I'm listening!"
- 17. Create input-section div with id "input-section"
- 18. Create an input element with id "input"
- 19. Set type to "text"
- 20. Set placeholder to "Type a message"
- 21. Set autocomplete to "off"
- 22. Set autofocus to "autofocus"
- 23. Create a button with class "send" and onclick event "sendMessage()"
- 24. Create a div with class "circle"
- 25. Create an i element with class "zmdi zmdi-mail-send"
- 26. End card div
- 27. Include the script.js file using the script tag
- 28. End body section
- 29. End HTML document

#### CSS Pseudocode:

\_\_\_\_\_

Set default font size and family for all elements

Define scrollbar styles

Style body background and card dimensions

Style header, message-section, and input-section

Style input and send button

Style user and bot messages

#### JavaScript Pseudocode:

-----

- 1. Create a constant array userMessage containing various arrays of user input phrases.
- 2. Create a constant array botReply containing arrays of corresponding bot responses.
- 3. Create a constant array alternative containing alternative responses.
- 4. Create a variable synth to store the window.speechSynthesis object.
- 5. Create a function voiceControl(string) to handle text-to-speech using the SpeechSynthesisUtterance object.
- 6. Create a new SpeechSynthesisUtterance instance with the given string.
- 7. Set language to "en-aus".
- 8. Set volume, rate, and pitch.
- 9. Use the synth.speak() method to speak the utterance.
- 10. Create a function sendMessage() to handle user input and trigger responses.
- 11. Get the input field element by ID.
- 12. Get the trimmed value of the input.
- 13. If the input is not empty, call the output function with the input.
- 14. Clear the input field.
- 15. Attach an event listener to the input field for the "keydown" event.
- 16. If the key pressed is "Enter", perform the same actions as in sendMessage().
- 17. Create a function output(input) to process user input and generate bot responses.
- 18. Initialize a variable product.
- 19. Convert the input to lowercase, remove non-alphanumeric characters, and perform additional text cleanup.
- 20. Call the compare function with userMessage, botReply, and the cleaned input.

- 21. Create a function compare(triggerArray, replyArray, string) to match user input to predefined triggers and retrieve bot responses.
- 22. Iterate through triggerArray and replyArray.
- 23. If a trigger matches the input, randomly select a corresponding reply.
- 24. If no match is found, call the containMessageCheck function.
- 25. Create a function containMessageCheck(string) to handle specific messages like greetings and farewells.
- 26. Define arrays expectedReply and expectedMessage for different scenarios.
- 27. If the input matches any expectedMessage, randomly select a corresponding expectedReply.
- 28. Return the selected reply.
- 29. Create a function addChat(input, product) to display user input and bot responses in the chat interface.
- 30. Get the mainDiv element by ID.
- 31. Create userDiv and botDiv elements.
- 32. Set appropriate IDs and classes for styling.
- 33. Set innerHTML with the input and bot response.
- 34. Append the userDiv and botDiv to the mainDiv.
- 35. Add an event listener to the "DOMContentLoaded" event.
- 36. Get the input field element.
- 37. Attach a keydown event listener to the input field, triggering sendMessage() on "Enter".

#### Main.py:

# Import necessary libraries and modules import random import json import pickle import numpy as np import nltk from nltk.stem import WordNetLemmatizer from tensorflow.python.keras.models import load\_model

# Initialize the lemmatizer lemmatizer = WordNetLemmatizer()

```
# Load intents data from JSON file
intents = json.loads(open('intents.json').read())
# Load preprocessed data from pickle files
words = pickle.load(open('words.pkl', 'rb'))
classes = pickle.load(open('classes.pkl', 'rb'))
model = load_model('chatbotmodel.h5')
# Define a function to clean up a sentence by tokenizing and lemmatizing
function clean_up_sentence(sentence):
  # Tokenize the sentence into words
  sentence_words = nltk.word_tokenize(sentence)
  # Lemmatize each word in the sentence
  sentence_words = [lemmatizer.lemmatize(word) for word in
sentence_words]
  return sentence_words
# Define a function to create a bag of words representation for a sentence
function bag_of_words(sentence):
  #Clean up the sentence
  sentence_words = clean_up_sentence(sentence)
  # Initialize a bag of words with zeros
  bag = [0] * len(words)
  # Iterate through each word in the cleaned sentence
  for w in sentence_words:
    # Check if the word is in the preprocessed words
    for i, word in enumerate(words):
       if word == w:
          # Set the corresponding index in the bag to 1
         bag[i] = 1
  return np.array(bag)
# Define a function to predict the class of a sentence
function predict_class(sentence):
  # Create a bag of words representation for the sentence
  bow = bag_of_words(sentence)
  # Expand dimensions to match the model input shape
  bow = np.expand_dims(bow, axis=0)
  # Predict the probabilities for each class using the model
  res = model.predict(bow)
  # Set an error threshold for class prediction
  ERROR\_THRESHOLD = 0.25
  # Extract classes with probabilities above the threshold
```

```
results = [[i, r] for i, r in enumerate(res[0]) if r > ERROR\_THRESHOLD]
  # Sort the results by probability in descending order
  results.sort(key=lambda x: x[1], reverse=True)
  return_list = []
 # Create a list of intents with their probabilities
  for r in results:
    return_list.append({'intent': classes[r[0]], 'probability': str(r[1])})
  return_list
# Define a function to get a response for a predicted intent
function get_response(intents_list, intents_json):
  #Extract the predicted intent tag
  tag = intents_list[0]['intent']
  # Get the list of intents from the intents JSON
  list_of_intents = intents_json['intents']
  # Iterate through each intent in the list
  for i in list_of_intents:
    # Check if the intent tag matches the predicted tag
    if i[tag'] == tag:
      # Randomly choose a response from the intent's responses
      result = random.choice(i['responses'])
      break
  return result
# Main interaction loop
print("|=======To
while True:
  # Get user input
  message = input("| You: ")
  # Check for exit condition
  if message == "bye" or message == "Goodbye":
    # Predict the intent for the exit message
    ints = predict_class(message)
    #Get a response for the predicted intent
    res = get_response(ints, intents)
    print("| Bot:", res)
```

#### trainingdata.py

```
# Import necessary libraries and modules
import random
import json
import pickle
import numpy as np
from nltk.stem import WordNetLemmatizer
from tensorflow.python.keras.models import Sequential
from tensorflow.python.keras.layers import Dense, Activation, Dropout
from tensorflow.python.keras.optimizers import SGD
# Initialize the lemmatizer
lemmatizer = WordNetLemmatizer()
# Load intents data from JSON file
intents = json.loads(open('intents.json').read())
# Initialize lists for words, classes, documents, and ignored letters
words = []
classes = []
documents = []
ignore_letters = ['?', '!', ',', '.']
# Process intents and patterns
for intent in intents['intents']:
  for pattern in intent['patterns']:
     # Tokenize the pattern into words
```

```
word_list = nltk.word_tokenize(pattern)
     # Add words to the list
     words.extend(word list)
     # Add document with word list and intent tag to documents list
     documents.append((word_list, intent['tag']))
     # Add the intent tag to classes if not already present
     if intent['tag'] not in classes:
       classes.append(intent['tag'])
# Lemmatize words and remove ignored letters
words = [lemmatizer.lemmatize(word) for word in words if word not in
ignore_letters]
# Remove duplicate words and sort the list
words = sorted(set(words))
# Sort the classes list
classes = sorted(set(classes))
# Save the words and classes lists to pickle files
pickle.dump(words, open('words.pkl', 'wb'))
pickle.dump(classes, open('classes.pkl', 'wb'))
# Prepare training data
training = []
output\_empty = [0] * len(classes)
for document in documents:
  bag = []
  word_patterns = document[0]
  word_patterns = [lemmatizer.lemmatize(word.lower()) for word in
word_patterns]
  for word in words:
     bag.append(1) if word in word_patterns else bag.append(0)
  output_row = list(output_empty)
  output row[classes.index(document[1])] = 1
  training.append([bag, output_row])
# Shuffle training data
random.shuffle(training)
training = np.array(training)
# Separate features (train_x) and labels (train_y)
train_x = list(training[:, 0])
```

```
train_y = list(training[:, 1])
# Build the neural network model
model = Sequential()
model.add(Dense(128, input_shape=(len(train_x[0]),), activation='relu'))
model.add(Dropout(0.5))
model.add(Dense(64, activation='relu'))
model.add(Dropout(0.5))
model.add(Dense(len(train_y[0]), activation='softmax'))
# Configure the SGD optimizer
sgd = SGD(lr=0.01, decay=1e-6, momentum=0.9, nesterov=True)#
Compile the model
model.compile(loss='categorical_crossentropy', optimizer=sgd,
metrics=['accuracy'])
# Train the model
hist = model.fit(np.array(train_x), np.array(train_y), epochs=200,
batch_size=5, verbose=1)
# Save the trained model
model.save('chatbotmodel.h5', hist)
print('Done')
```

# APPENDIX-B SCREENSHOTS

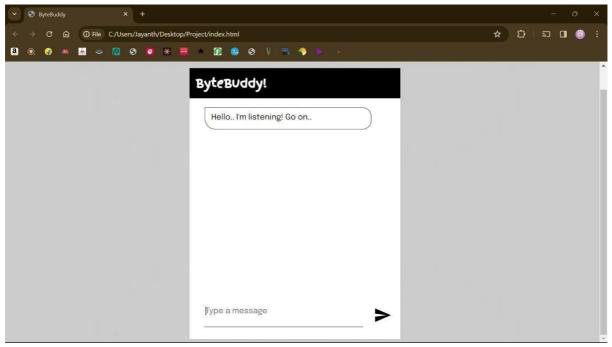


Fig.1 Interface

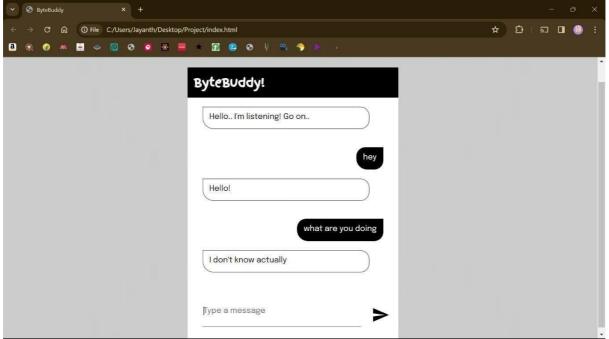


Fig.2 Responses

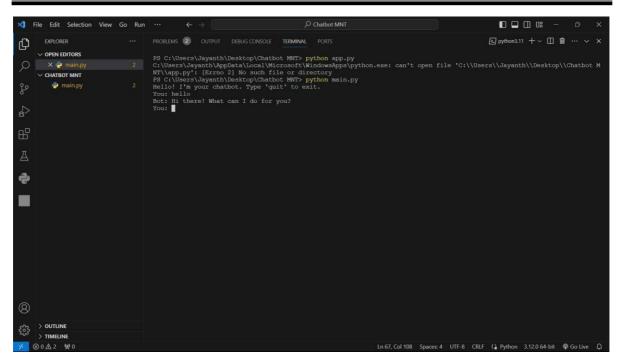


Fig.3 Terminal

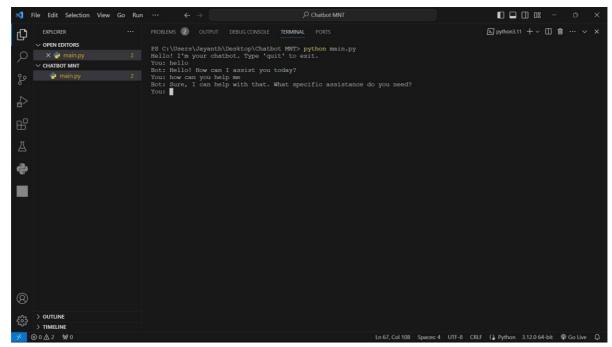
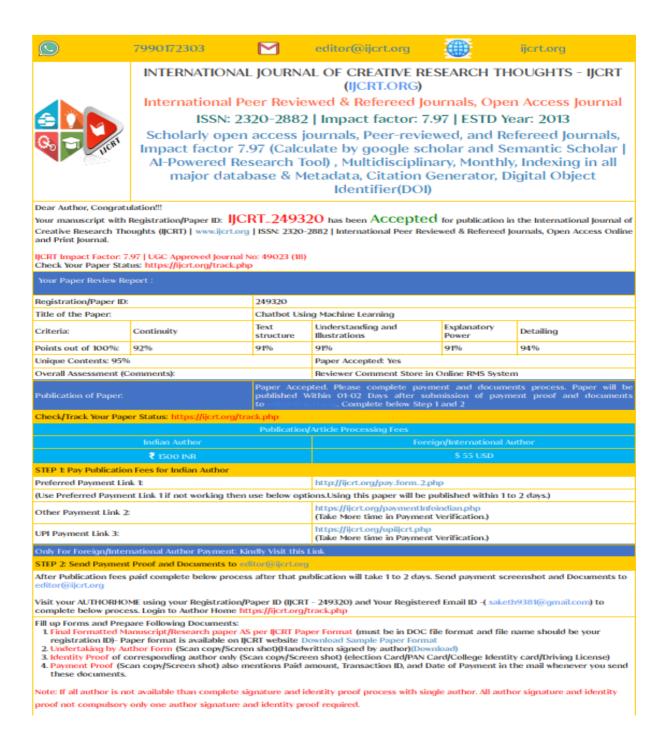


Fig.4 Terminal Response

### APPENDIX-C

### **ENCLOSURES**

### LETTER OF ACCEPTANCE



## LETTER OF ACKNOWLEDGEMENT



#### IJCRT249320 - Acknowledgement from IJCRT regarding research paper submitting

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# CHATBOT USING MACHINE LEARNING

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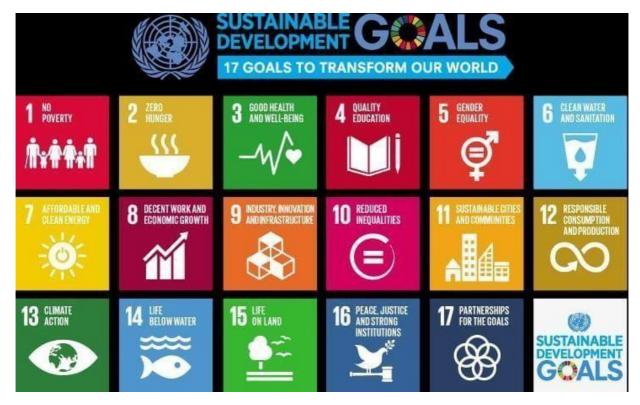
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# SUSTAINABLE DEVELOPMENT GOALS



In the pursuit of advancing global education, our innovative chatbot project stands as a beacon for Quality Education (SDG 4). By providing accessible learning resources, personalized learning experiences, and round-the-clock support, our platform is committed to breaking down barriers, fostering inclusivity, and revolutionizing the way individuals engage with and benefit from education.

# 1. Accessible Learning Hub:

• Our chatbot acts as a centralized hub, granting students easy access to a wealth of educational resources, including textbooks, articles, and study materials.

## 2. Personalized Learning Paths:

• Tailored learning experiences through adaptive technologies ensure that each student's unique needs and learning styles are addressed, fostering a more effective educational journey.

# 3. 24/7 Support for Continuous Learning:

With our chatbot's round-the-clock availability, students can receive instant support
outside traditional hours, promoting continuous learning and flexibility in
education.

### 4. Inclusive Features for Diverse Learners:

 Incorporating features for differently-abled learners ensures inclusivity, offering alternative formats and assistive technologies to make education accessible to a wide range of individuals.

In conclusion, our chatbot project not only aligns with the Sustainable Development Goal of Quality Education by providing accessible, personalized, and inclusive learning experiences but also aims to redefine the landscape of education by embracing innovation and fostering a global community of learners.