



GENERATIVE ARTIFICIAL INTELLIGENCE CHATBOT APPLICATION

D. Usha Rajeswari(Guide), K. Balaji Vamsi Krishna, K.V.R. Tejaswi, Ch. Raghu Ram, S. Santhoshini

Professor, 4thyear B.TechStudent , 4thyear B.TechStudent , 4thyear B.TechStudent , 4thyear B.TechStudent
Computer Science & Engineering Department, Sanketika Vidya Parishad Engineering College, Visakhapatnam, India

Abstract : Experience the power of conversational AI with our Generative Artificial Intelligence chatbot. Simply input your text prompt or question, and watch as our advanced technology generates accurate and informative responses in real-time. Introducing our cutting-edge Generative Artificial Intelligence chatbot - the ultimate solution for interactive and intelligent conversations. We use powerful tools such as Langchain, Google Colab, Hugging Face, OpenAI, Play HT, and more to ensure the highest quality chatbot experience. Introducing our state-of-the-art Generative Artificial Intelligence chatbot, designed to revolutionize the way you interact with technology. With its advanced capabilities, this chatbot is capable of providing text-based answers to any user prompt or question. But that's not all - our chatbot goes beyond text-based answers. With voice recognition capabilities, it can provide spoken responses to your prompts for a truly interactive experience. Discover a new level of conversational AI with our state-of-the-art technology. Google Colab is a free cloud-based platform provided by Google that allows users to run and execute Python code through a web browser. Offers a Jupyter Notebook-like environment. Widely used in ML & DS communities for prototyping, experimentation. In this platform we include all other tools like gradio, hugging face, playHT, langchain.

IndexTerms – Generative Artificial Intelligence, Langchain, Google Colab, Gradio, PlayHT, LLM(Large Language Model), Python

1. INTRODUCTION

A generative artificial intelligence (AI) chatbot application is a software program that utilizes generative models, such as recurrent neural networks (RNNs) or transformers, to generate human-like responses in a conversational context. These chatbots are designed to interact with users in a natural language format, simulating a conversation with a human being. Generative AI chatbots work by analyzing input from users, understanding the context and intent of the conversation, and generating appropriate responses. They can be trained on large datasets of human conversations or other sources of text to learn patterns, language structures, and appropriate responses. This allows them to generate coherent and contextually relevant replies. One popular approach to generative chatbot applications is using sequence-to-sequence models. These models take an input sequence, such as a user's message, and produce an output sequence, which is the chatbot's response. By training on a large corpus of conversational data, the chatbot can learn to generate meaningful and coherent responses based on the input it receives. Generative AI chatbots have various applications across industries. They can be used for customer service, providing information and support to users, virtual assistants, language translation, content generation, and more. With advancements in natural language processing and generative models, these chatbots can simulate human-like conversations and provide increasingly sophisticated interactions. It's worth noting that while generative AI chatbots can generate impressive responses, they still have limitations. They may occasionally produce incorrect or nonsensical replies, struggle with understanding ambiguous queries, or exhibit biases present in the training data. Ongoing research and development are focused on improving these aspects and creating more robust and reliable chatbot applications.

Applications:

1. Text Generation tools: It can generate text based content based on the user prompt like Rytr, Jasper, copy.ai etc
2. Image Generation tools: It can generate images based on the AI tools like Mid Journey, Dall E2 etc.
3. Video Generation tools: It can also generate videos based on the Runway ML, Pictory etc.
4. Speech Generation tools: It provides voice based text content by using tools like Murfai, Fliki etc.
5. Music Generation tools: It is used for generating music by using tools like Beatbot, Mubert etc.

Tools & Technologies used in Generative AI:

1. Langchain
2. Gradio
3. Google Colab
4. Large Language Model (LLM)

5. Hugging Face
6. Play HT

2. BACKGROUND

2.1 SPATIAL DATABASE

The spatial database refers to a specialized database that is designed to efficiently store and manage spatial data, such as geographic or location-based information. However, in the context of a generative AI chatbot application, the term "spatial database" may not directly relate to the chatbot itself. Instead, the focus is primarily on the language generation and conversation management aspects. In a generative AI chatbot application, the key components involve natural language processing (NLP), machine learning models, and conversational context management. These components are responsible for understanding user input, generating appropriate responses, and maintaining the flow of the conversation. The chatbot application may utilize various data structures and algorithms to facilitate these functionalities, but they are not typically referred to as a "spatial database." That being said, depending on the specific requirements of the chatbot application, there might be cases where spatial data integration is necessary. For instance, if the chatbot needs to provide location-based services or handle spatial queries, it may need access to a spatial database or external APIs that provide geospatial information. In such cases, the chatbot application might interact with a spatial database to retrieve relevant spatial data, but this usage would be specific to the integration requirements and not a core aspect of the chatbot itself.

2.2 SCOPE OF PROJECT

The scope of a generative artificial intelligence (AI) chatbot application is broad and encompasses various domains and industries. Here are some key areas where generative AI chatbots find application:

Customer Service and Support:

Generative AI chatbots are widely used in customer service to provide automated support and assistance. They can handle frequently asked questions, provide information about products or services, assist with troubleshooting, and offer basic customer support. Chatbots can handle a large volume of customer queries simultaneously, providing timely responses and reducing the need for human intervention.

Virtual Assistants:

Generative AI chatbots serve as virtual assistants, helping users with tasks such as scheduling appointments, setting reminders, answering general inquiries, and providing personalized recommendations. Virtual assistants can be integrated into various platforms, including websites, mobile apps, or smart home devices, to offer convenient and interactive user experiences.

E-commerce and Sales:

Chatbots are used in e-commerce platforms to assist customers with product selection, offer personalized recommendations, process orders, and provide post-purchase support. They can also handle sales inquiries, cross-selling, and upselling by analyzing customer preferences and purchase history.

Content Generation:

Generative AI chatbots can assist in content generation by providing suggestions, writing drafts, or generating specific types of content. They can help with tasks such as writing articles, creating product descriptions, generating social media posts, or composing personalized emails.

Language Translation:

Chatbots can facilitate language translation and communication across different languages. Users can input text or speech in one language, and the chatbot can generate responses in another language, enabling multilingual conversations and bridging language barriers.

Education and Training:

Chatbots are utilized in educational settings to provide personalized learning experiences, answer student queries, deliver instructional content, and offer interactive practice exercises. They can adapt to individual learning styles and provide targeted guidance to students.

Information Retrieval:

Generative AI chatbots can be used as information retrieval systems, answering questions, and providing relevant information from vast knowledge bases or databases. They can assist with research, provide news updates, retrieve weather information, or offer general knowledge.

It's important to note that the scope of generative AI chatbot applications is continuously expanding as technology advances and new use cases emerge.

3. LITERATURE REVIEW

Generative Artificial Intelligence (AI) chatbot applications have gained significant attention in recent years due to their ability to simulate human-like conversations and provide interactive and intelligent responses. This literature review aims to explore the existing research and advancements in generative AI chatbot applications, including their methodologies, techniques, and potential applications across various domains. Generative AI chatbots utilize generative models, such as recurrent neural networks (RNNs) or transformers, to generate human-like responses in a conversational context. These chatbots analyze user input, understand the context and intent of the conversation, and generate appropriate responses. They can be trained on large datasets of human conversations or other sources of text to learn patterns, language structures, and suitable replies.

Several studies have focused on the development and evaluation of generative AI chatbot applications. For instance, Radford et al. (2019) introduced GPT-2, a large-scale language model capable of generating coherent and contextually relevant responses. The model was trained on a diverse range of internet text and demonstrated impressive conversational abilities.

3.1 Classification of algorithm

Generative artificial intelligence (AI) chatbot applications utilize various algorithms to enable conversational capabilities. Here are some commonly used algorithms in generative AI chatbot systems:

1. Recurrent Neural Networks (RNNs): RNNs, including Long Short-Term Memory (LSTM) and Gated Recurrent Units (GRUs), are frequently employed for language generation tasks. RNNs are capable of capturing sequential dependencies in text data and are often used in chatbot systems to generate coherent and contextually relevant responses.
2. Transformer Models: Transformer models, such as the ones used in the OpenAI GPT (Generative Pre-trained Transformer) series, have gained popularity in generative AI chatbots. Transformers leverage attention mechanisms to capture contextual information effectively. Models like GPT have been pre-trained on large amounts of text data and can be fine-tuned for specific chatbot tasks.
3. Sequence-to-Sequence Models: Sequence-to-sequence models, also known as encoder-decoder models, are widely used in chatbot systems. They consist of two components: an encoder that processes the input text and a decoder that generates the response. These models can be based on RNNs or transformer architectures and are effective in generating conversational responses.
4. Reinforcement Learning: Reinforcement learning techniques can be employed to improve the performance of chatbot systems. In reinforcement learning, the chatbot interacts with users and receives feedback in the form of rewards or penalties based on the quality of its responses. This feedback helps the chatbot optimize its behavior over time.
5. Rule-Based Approaches: Rule-based approaches involve defining a set of predefined rules or patterns to handle specific user queries or intents. These rules can be based on regular expressions or handcrafted logic. Rule-based approaches are often used to handle simple or domain-specific tasks in chatbot applications.
6. Statistical Models: Statistical models, such as Naive Bayes classifiers or Support Vector Machines (SVMs), can be employed for tasks like intent recognition or sentiment analysis. These models use statistical techniques to classify user intents or determine the sentiment of user queries, enabling the chatbot to understand and respond appropriately.
7. Information Retrieval: Information retrieval techniques can be used in chatbot systems that integrate a knowledge base or database. These techniques involve indexing and querying the knowledge base to retrieve relevant information or answers to user queries. Methods like TF-IDF (Term Frequency-Inverse Document Frequency) or BM25 (Best Matching 25) are commonly used for information retrieval.

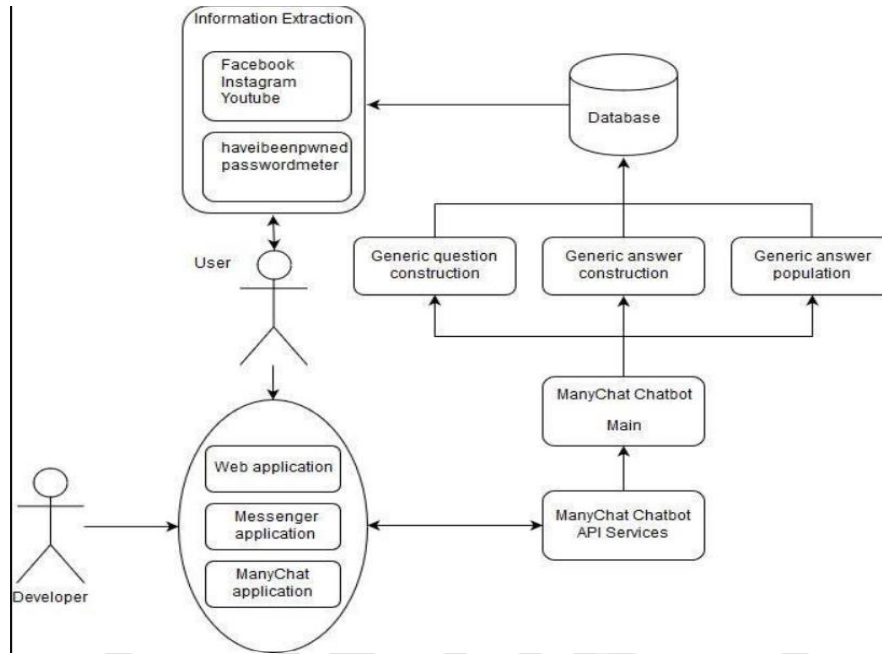
3.3 UML(Unified Modelling Language)

Unified Modeling Language (UML) is a general-purpose modeling language. The main aim of UML is to define a standard way to visualize the way a system has been designed. It is quite similar to blueprints used in other fields of engineering. UML is not a programming language, it is rather a visual language. We use UML diagrams to portray the behavior and structure of a system. UML helps software engineers, businessmen, and system architects with modeling, design, and analysis.

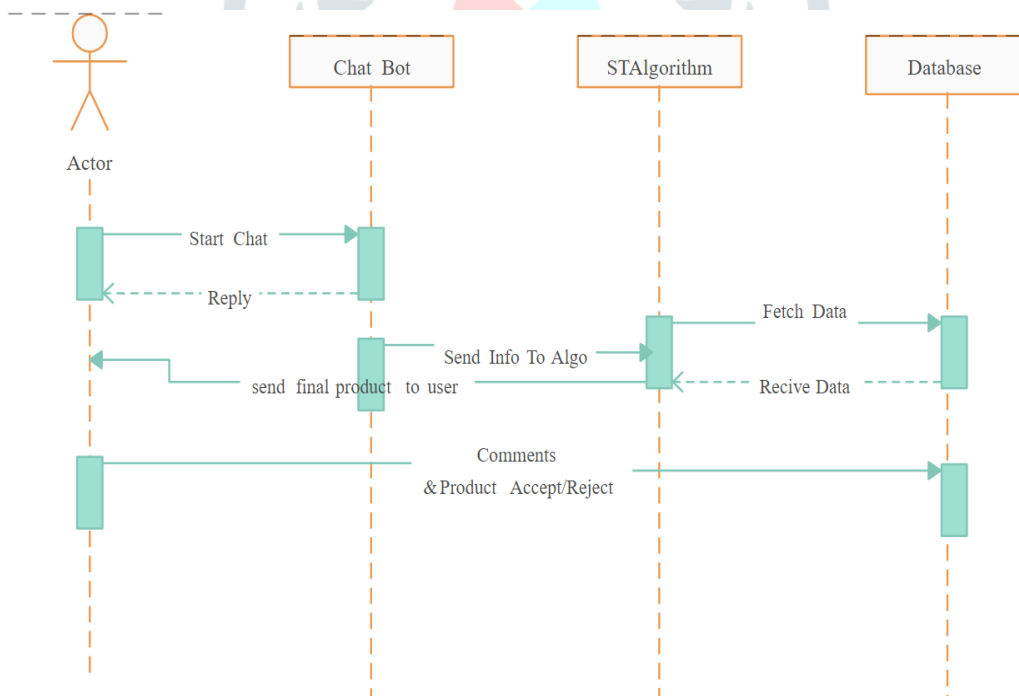
UML is linked with object-oriented design and analysis. UML makes use of elements and forms associations between them to form diagrams.

1. Class Diagram
2. Sequence diagram
3. Activity diagram
4. Use case diagram

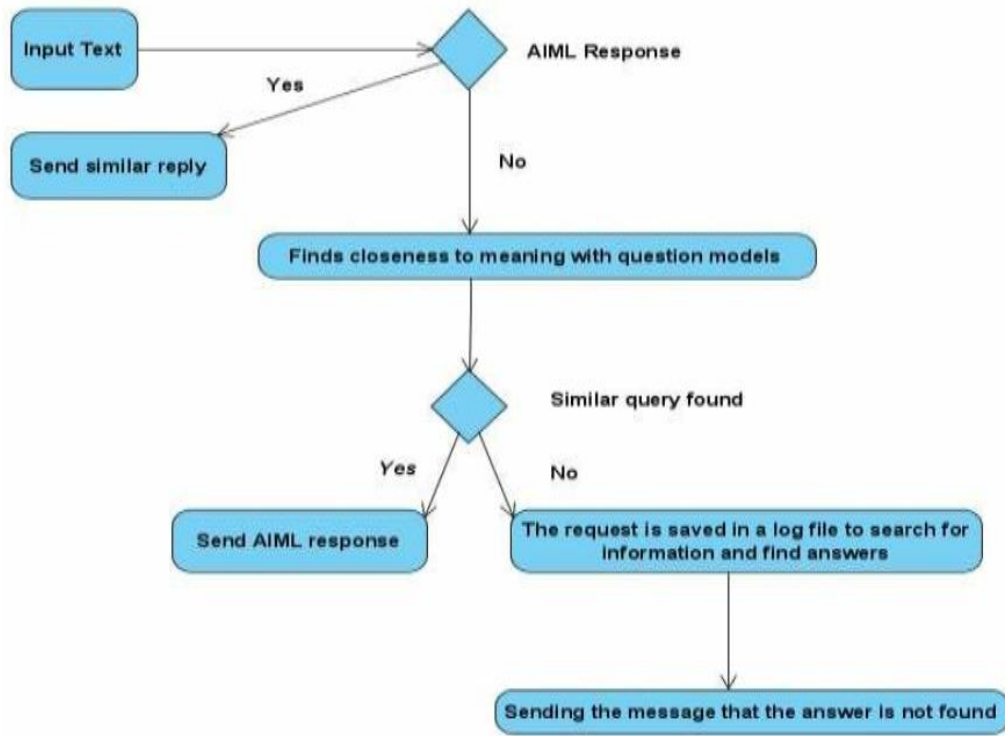
Class Diagram



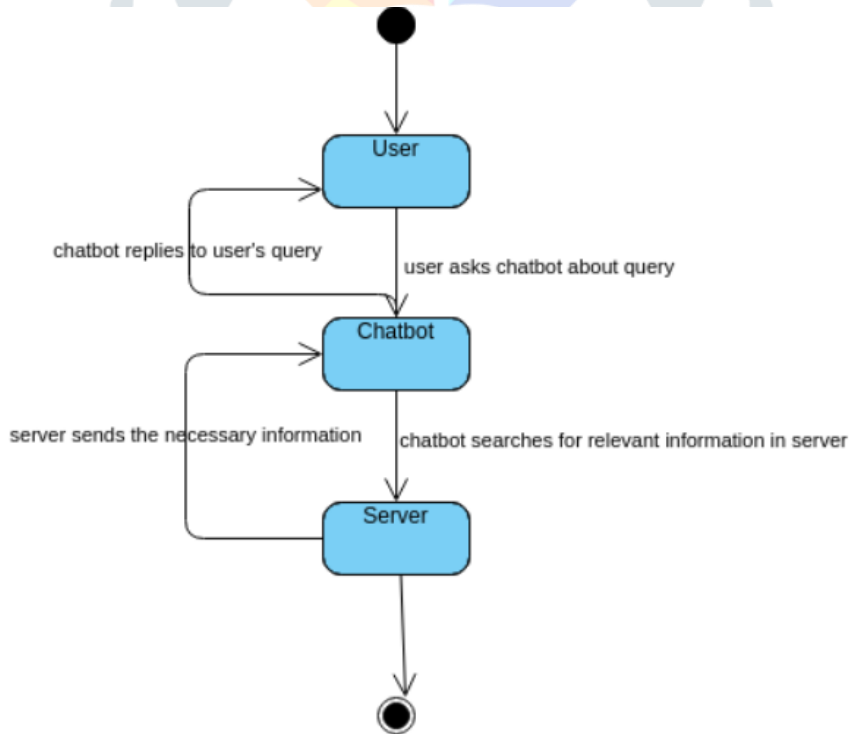
Sequence diagram



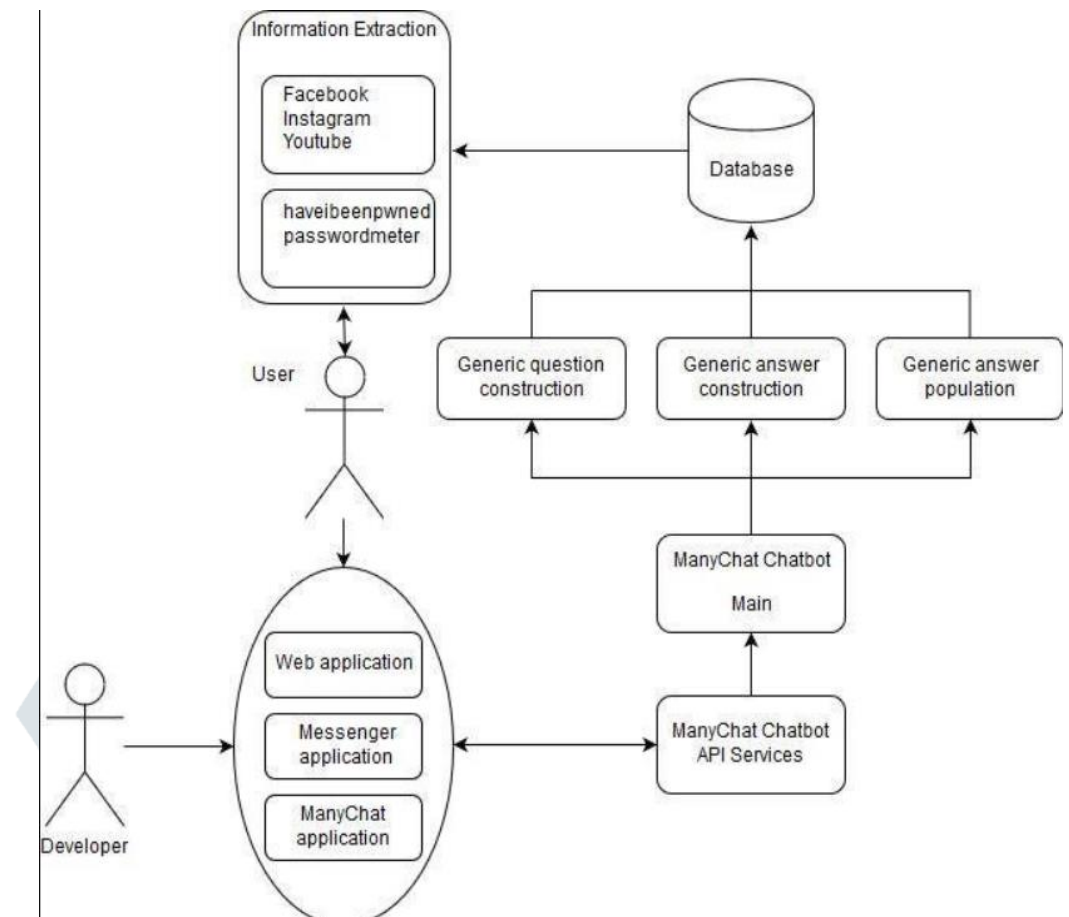
Activity Diagram



State Chart Diagram



Use Case Diagram



4. Data Implementation

Data implementation in generative artificial intelligence (AI) chatbot applications involves several key aspects. Here are some considerations for data implementation in such systems:

- 1. Data Collection:** Gather a diverse and representative dataset of human conversations or other relevant textual sources. This dataset will be used to train the generative AI chatbot model. Ensure that the data covers a wide range of topics, conversational styles, and language patterns to enable the model to learn effectively.
- 2. Preprocessing:** Clean and preprocess the collected data before training the chatbot model. This step involves removing noise, such as irrelevant characters or special symbols, and standardizing the text format. Common preprocessing techniques include tokenization, lowercasing, stemming, and removing stop words.
- 3. Training Data Format:** Convert the pre-processed data into a suitable format for training the generative AI model. This format typically involves representing the input-output pairs, where the input is the user's message, and the output is the corresponding chatbot response. The data can be organized as a sequence of input-output pairs or in a more complex structure, depending on the model architecture and requirements.
- 4. Model Training:** Train the generative AI chatbot model using the prepared dataset. Popular architectures for training generative chatbot models include recurrent neural networks (RNNs), such as long short-term memory (LSTM), and transformer models. During training, the model learns to generate appropriate responses given the input sequence by optimizing certain objective functions, such as maximum likelihood estimation or reinforcement learning.
- 5. Data Augmentation:** To enhance the performance and diversity of the chatbot responses, data augmentation techniques can be applied. These techniques involve generating additional training examples by adding variations to the original dataset. For example, introducing paraphrases, synonyms, or perturbations to the input messages can help the chatbot handle different user inputs effectively.
- 6. Fine-tuning and Transfer Learning:** After the initial training, the chatbot model can be fine-tuned on domain-specific or task-specific data to improve its performance in specific contexts. Fine-tuning allows the model to adapt to the target domain or application by leveraging a smaller, more specific dataset while preserving the knowledge learned from the initial training phase.

Transfer learning techniques can also be employed, where pre-trained models on large-scale datasets are fine-tuned for chatbot applications.

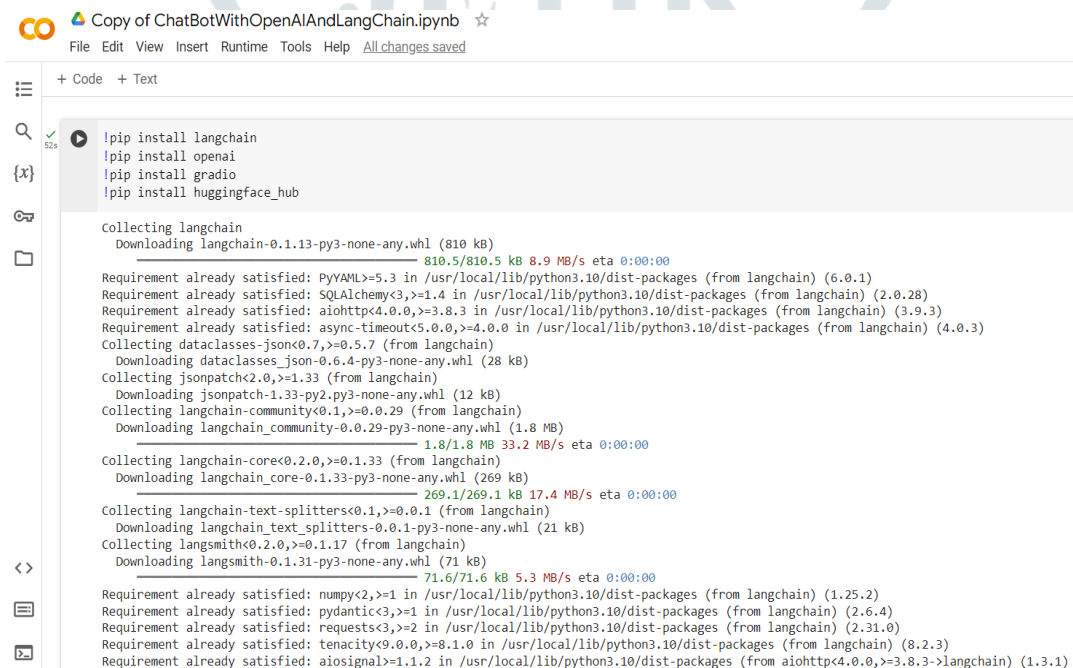
7. Evaluation Data: Prepare a separate evaluation dataset to assess the performance and quality of the generative AI chatbot. This dataset should consist of input-output pairs that were not used during the training process. Evaluation metrics such as perplexity, BLEU score, or human evaluation can be employed to measure the chatbot's conversational quality, coherence, and relevance.

8. Iterative Improvement: Continuously refine and update the chatbot's performance by incorporating user feedback and iteratively retraining the model. Collect user interactions with the chatbot in real-world scenarios to gather more data for ongoing improvement. This can involve a feedback loop where user interactions are used to improve the chatbot's responses and enhance its overall conversational capabilities.

Remember, data implementation in generative AI chatbot applications is an iterative process that requires careful consideration of data quality, preprocessing techniques, model training, and evaluation. It is crucial to regularly monitor and update the chatbot's performance based on real-world usage and user feedback to ensure its effectiveness and user satisfaction.

5. Execution

In this project we have used different tools like Langchain, Googlecolab, Gradio, Large language model, Huggingface, PlayHT. In addition to it we have added one more feature to this application i.e the user gives a prompt to the chatbot such as query or any other information whereas, the chatbot replies to the user in via voice. It can be possible by using tool called "PLAYHT". The PlayHT tool takes the user voice clone of one minute audio. Then it gives the output or solution via voice.

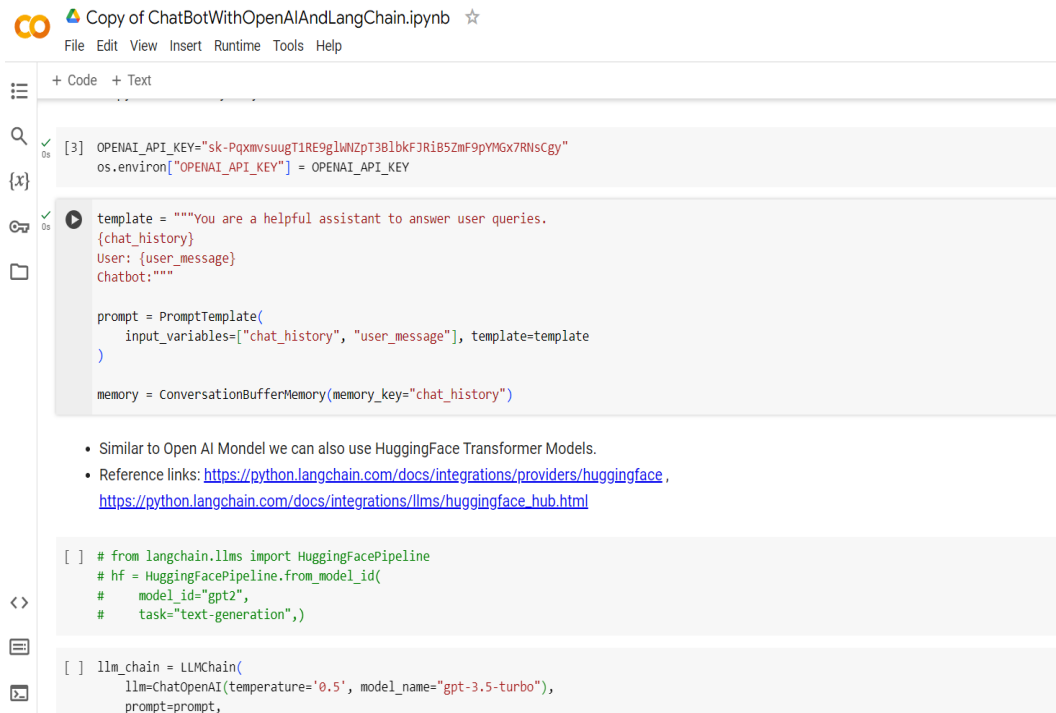


```
Copy of ChatBotWithOpenAIAndLangChain.ipynb
File Edit View Insert Runtime Tools Help All changes saved

+ Code + Text

!pip install langchain
!pip install openai
!pip install gradio
!pip install huggingface_hub

Collecting langchain
  Downloading langchain-0.1.13-py3-none-any.whl (810 kB)
    810.5/810.5 kB 8.9 MB/s eta 0:00:00
Requirement already satisfied: PyYAML>=5.3 in /usr/local/lib/python3.10/dist-packages (from langchain) (6.0.1)
Requirement already satisfied: SQLAlchemy<3,>=1.4 in /usr/local/lib/python3.10/dist-packages (from langchain) (2.0.28)
Requirement already satisfied: aiohttp<4.0.0,>=3.8.3 in /usr/local/lib/python3.10/dist-packages (from langchain) (3.9.3)
Requirement already satisfied: async-timeout<5.0.0,>=4.0.0 in /usr/local/lib/python3.10/dist-packages (from langchain) (4.0.3)
Collecting dataclasses-json<0.7,>=0.5.7 (from langchain)
  Downloading dataclasses_json-0.6.4-py3-none-any.whl (28 kB)
Collecting jsonpatch<2.0,>=1.33 (from langchain)
  Downloading jsonpatch-1.33-py2.py3-none-any.whl (12 kB)
Collecting langchain-community<0.1,>=0.0.29 (from langchain)
  Downloading langchain_community-0.0.29-py3-none-any.whl (1.8 MB)
    1.8/1.8 MB 33.2 MB/s eta 0:00:00
Collecting langchain-core<0.2.0,>=0.1.33 (from langchain)
  Downloading langchain_core-0.1.33-py3-none-any.whl (269 kB)
    269.1/269.1 kB 17.4 MB/s eta 0:00:00
Collecting langchain-text-splitters<0.1,>=0.0.1 (from langchain)
  Downloading langchain_text_splitters-0.0.1-py3-none-any.whl (21 kB)
Collecting langsmith<0.2.0,>=0.1.17 (from langchain)
  Downloading langsmith-0.1.31-py3-none-any.whl (71 kB)
    71.6/71.6 kB 5.3 MB/s eta 0:00:00
Requirement already satisfied: numpy<2,>=1 in /usr/local/lib/python3.10/dist-packages (from langchain) (1.25.2)
Requirement already satisfied: pydantic<3,>=1 in /usr/local/lib/python3.10/dist-packages (from langchain) (2.6.4)
Requirement already satisfied: requests<3,>=2 in /usr/local/lib/python3.10/dist-packages (from langchain) (2.31.0)
Requirement already satisfied: tenacity<9.0.0,>=8.1.0 in /usr/local/lib/python3.10/dist-packages (from langchain) (8.2.3)
Requirement already satisfied: aiosignal>=1.1.2 in /usr/local/lib/python3.10/dist-packages (from aiohttp<4.0.0,>=3.8.3->langchain) (1.3.1)
```



```

Copy of ChatBotWithOpenAIAndLangChain.ipynb
File Edit View Insert Runtime Tools Help

+ Code + Text

[3] OPENAI_API_KEY="sk-PqxmvsuugT1RE9g1wNzP381bkfJRI85Zmf9pYMGx7RNsCgy"
os.environ["OPENAI_API_KEY"] = OPENAI_API_KEY

template = """You are a helpful assistant to answer user queries.
{chat_history}
User: {user_message}
Chatbot:"""

prompt = PromptTemplate(
    input_variables=["chat_history", "user_message"], template=template
)

memory = ConversationBufferMemory(memory_key="chat_history")

• Similar to Open AI Model we can also use HuggingFace Transformer Models.
• Reference links: https://python.langchain.com/docs/integrations/providers/huggingface,
https://python.langchain.com/docs/integrations/lms/huggingface\_hub.html

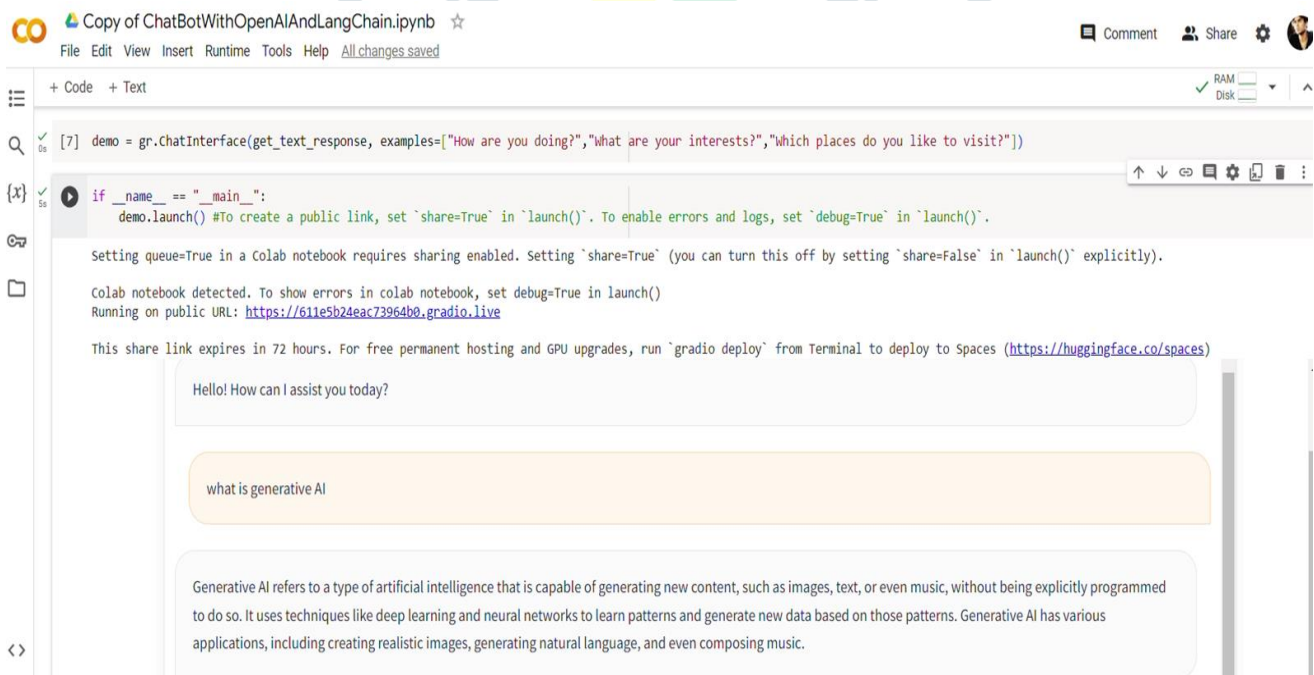
[ ] # from langchain.llms import HuggingFacePipeline
# hf = HuggingFacePipeline.from_model_id(
#     model_id="gpt2",
#     task="text-generation",)

[ ] llm_chain = LLMChain(
    llm=ChatOpenAI(temperature='0.5', model_name="gpt-3.5-turbo"),
    prompt=prompt,
  
```

PlayHT (Voice Cloning)

Uses artificial intelligence (AI) to generate a synthetic copy. Potential applications include creating audiobooks, dubbing movies, and generating personalized messages. Would create a Global Economic Impact of \$13.9 trillion by 2030. The PlayHT tool is used to intake the user voice prompt of one minute audio file and then it converts to the individual voice clone. Where the user asks query to the chatbot and then the chatbot can reply to the user via voice.

6. Result



```

Copy of ChatBotWithOpenAIAndLangChain.ipynb
File Edit View Insert Runtime Tools Help All changes saved

+ Code + Text

[7] demo = gr.ChatInterface(get_text_response, examples=["How are you doing?", "What are your interests?", "Which places do you like to visit?"])

if __name__ == "__main__":
    demo.launch() #To create a public link, set `share=True` in `launch()`. To enable errors and logs, set `debug=True` in `launch()`.

Setting queue=True in a Colab notebook requires sharing enabled. Setting `share=True` (you can turn this off by setting `share=False` in `launch()` explicitly).

Colab notebook detected. To show errors in colab notebook, set debug=True in launch()
Running on public URL: https://611e5b24eac73964b0.gradio.live

This share link expires in 72 hours. For free permanent hosting and GPU upgrades, run `gradio deploy` from Terminal to deploy to Spaces (https://huggingface.co/spaces)

Hello! How can I assist you today?

what is generative AI

Generative AI refers to a type of artificial intelligence that is capable of generating new content, such as images, text, or even music, without being explicitly programmed to do so. It uses techniques like deep learning and neural networks to learn patterns and generate new data based on those patterns. Generative AI has various applications, including creating realistic images, generating natural language, and even composing music.
  
```


7. Conclusion

In conclusion, the creation of a generative artificial chatbot application marks a significant milestone in the field of artificial intelligence. Through advanced natural language processing techniques and machine learning algorithms, these chatbots have the ability to generate human-like responses and engage in meaningful conversations with users. The development of a generative artificial chatbot application brings numerous benefits and opportunities. Firstly, it enables businesses to enhance customer service by providing round-the-clock support and personalized interactions. Chatbots can handle a wide range of customer queries efficiently, resolving issues and providing information in real-time. Moreover, generative chatbots have the potential to revolutionize various industries. They can be utilized in healthcare, assisting medical professionals with patient consultations and providing accurate information. In education, chatbots can serve as virtual tutors, offering personalized learning experiences to students. Additionally, they can aid in language learning, providing interactive conversations and instant feedback. The creation of generative chatbot applications also poses challenges and considerations. Ethical concerns regarding data privacy, security, and the potential for misuse must be carefully addressed. Striking a balance between automation and human involvement is crucial to ensure the chatbot's responses remain accurate, unbiased, and aligned with user expectations. Looking ahead, the continuous advancement of natural language processing and machine learning algorithms will likely result in even more sophisticated and intelligent chatbots. These future iterations may possess enhanced contextual understanding, emotional intelligence, and the ability to handle complex tasks.

In summary, the creation of generative artificial chatbot applications represents a significant leap forward in artificial intelligence. With their ability to engage in human-like conversations and provide valuable assistance, these chatbots have the potential to reshape various aspects of our lives, from customer service to education and beyond. However, careful attention must be given to ethical considerations to ensure their responsible and beneficial integration into society.

8. Future Scope

The future scope of generative AI chatbot applications is vast and promising, with several potential areas of advancement and impact:

- **Multimodal Interaction:** Future chatbots will likely incorporate multimodal capabilities, allowing for seamless interaction through voice, visuals, and other modalities, enabling more natural and intuitive communication.
- **Personalization and Adaptability:** Chatbots will become more personalized and adaptive, tailoring their responses and behavior to individual users' preferences, contexts, and interaction styles, leading to more engaging and effective conversations.
- **Emotional Intelligence:** With advancements in natural language processing and emotional AI, chatbots will become better at recognizing and responding to human emotions, enabling more empathetic and emotionally intelligent interactions.
- **Continuous Learning and Self-Improvement:** Chatbots will have the ability to continuously learn and improve from their interactions, expanding their knowledge and capabilities over time, potentially leading to more advanced and specialized applications.
- **Domain Expertise:** Generative AI chatbots will be developed for specific domains, such as healthcare, finance, or education, possessing in-depth knowledge and expertise to provide highly specialized assistance and advice.
- **Multilingual Support:** With the globalization of technology, chatbots will need to support multiple languages and cultural contexts, enabling seamless communication across diverse populations.
- **Integration with IoT and Smart Environments:** Chatbots will be integrated into Internet of Things (IoT) devices and smart environments, allowing for natural language control and interaction with various devices and systems.
- **Ethical and Responsible AI:** As chatbots become more advanced and influential, there will be a growing emphasis on developing ethical and responsible AI systems, ensuring fairness, transparency, and accountability in their decision-making processes.
- **Deployment at Scale:** With improved scalability and efficiency, generative AI chatbots will be deployed at a massive scale, enabling organizations to provide personalized assistance and support to large customer bases.
- **New Application Areas:** Generative AI chatbots will find applications in various domains, such as creative writing, content generation, virtual assistants, and even as companions or tutors, expanding the scope of their impact.

Overall, the future scope of generative AI chatbot applications is promising, with the potential to revolutionize how we interact with technology and access information and services, ultimately enhancing productivity, efficiency, and overall user experiences.

9. References

- Radford, A., Wu, J., Child, R., Luan, D., Amodei, D., & Sutskever, I. (2019). Language models are unsupervised multitask learners. OpenAI Blog, 1(8), 9. [Link: <https://openai.com/blog/better-language-models/>]
- Vaswani, A., Shazeer, N., Parmar, N., Uszkoreit, J., Jones, L., Gomez, A. N., ... & Polosukhin, I. (2017). Attention is all you need. In Advances in Neural Information Processing Systems (pp. 5998-6008). [Link: <https://proceedings.neurips.cc/paper/2017/file/3f5ee243547dee91fbd053c1c4a845aa-Paper.pdf>]
- Serban, I. V., Sordoni, A., Lowe, R., Charlin, L., Pineau, J., Courville, A., & Bengio, Y. (2017). A hierarchical latent variable encoder-decoder model for generating dialogues. In Proceedings of the 31st AAAI Conference on Artificial Intelligence (AAAI-17) (pp. 3295-3301). [Link: <https://www.aaai.org/ocs/index.php/AAAI/AAAI17/paper/view/14856>]
- Xu, J., Liu, C., & Zhao, K. (2020). A survey on dialogue systems: Recent advances and new frontiers. ACM Transactions on Information Systems (TOIS), 38(4), 1-40. [Link: <https://dl.acm.org/doi/10.1145/3380981>]
- Li, J., Monroe, W., Ritter, A., Galley, M., Gao, J., & Jurafsky, D. (2016). Deep reinforcement learning for dialogue generation. In Proceedings of the 2016 Conference on Empirical Methods in Natural Language Processing (pp. 1192-1202). [Link: <https://www.aclweb.org/anthology/D16-1127>]
- Wang, Y., Wu, C., Wu, S., & Zhou, X. (2020). A survey on chatbot design techniques in speech-based systems. ACM Computing Surveys (CSUR), 53(1), 1-33. [Link: <https://dl.acm.org/doi/10.1145/3390625>]
- Liu, C., Lowe, R., Serban, I. V., Noseworthy, M., Charlin, L., & Pineau, J. (2016). How NOT to evaluate your dialogue system: An empirical study of unsupervised evaluation metrics for dialogue response generation. In Proceedings of the 2016 Conference on Empirical Methods in Natural Language Processing (pp. 2122-2132). [Link: <https://www.aclweb.org/anthology/D16-1230>]
- Wolf, T., Sanh, V., Chaumond, J., & Delangue, C. (2019). TransferTransfo: A transfer learning approach for neural network based conversational agents. In Proceedings of the 2019 Conference on Empirical Methods in Natural Language Processing and the 9th International Joint Conference on Natural Language Processing (pp. 4374-4383). [Link: <https://www.aclweb.org/anthology/D19-1441>]

BIBLIOGRAPHY

DAVULURY USHARAJESWARI IS CURRENTLY WORKING AS ASSISTANT PROFESSOR FROM THE DEPARTMENT OF COMPUTER SCIENCE AND ENGINEERING AT SANKETIKA VIDYA PARISHAD ENGINEERING COLLEGE IN VISAKHAPATNAM, AFFILIATED TO ANDHRA UNIVERSITY, ACCREDITED BY NAAC WITH A GRADE. MAM HAS TEN YEARS OF EXPERIENCE AS TEACHER IN COLLEGES AFFILIATED WITH ANDHRA UNIVERSITY AND JNTUK. SHE OBTAINED HER B.TECH IN CSE FROM REGENCY INSTITUTE OF TECHNOLOGY, WHICH WAS AFFILIATED TO PONDICHERRY UNIVERSITY AND THE M.TECH FROM ACHARYA NAGARJUNA UNIVERSITY. HER PRIMARY AREAS OF INTEREST ARE ARTIFICIAL INTELLIGENCE, MACHINE LEARNING, THE INTERNET OF THINGS, AND DATA MINING.



K. BALAJI VAMSI KRISHNA IS PURSUING B.TECH IN COMPUTER SCIENCE & ENGINEERING FROM SANKETIKA VIDYA PARISHAD ENGINEERING COLLEGE, AFFILIATED TO THE NAAC A GRADE ACCREDITED AND ISO CERTIFIED ANDHRA UNIVERSITY. WITH A KEEN INTEREST IN GENERATIVE ARTIFICIAL INTELLIGENCE, HE HAS TAKEN UP AN ACADEMIC PROJECT FOCUSED ON DEVELOPING A CONVERSATIONAL CHATBOT APPLICATION. THE PROJECT AIMS TO LEVERAGE STATE-OF-THE-ART NATURAL LANGUAGE PROCESSING (NLP) AND GENERATIVE AI TECHNIQUES TO CREATE AN INTELLIGENT CHATBOT CAPABLE OF ENGAGING IN HUMAN-LIKE DIALOGUES. THE CHATBOT WILL BE TRAINED ON LARGE LANGUAGE MODELS AND VAST DATASETS TO UNDERSTAND AND GENERATE CONTEXTUALLY RELEVANT RESPONSES, MAINTAINING COHERENCE AND FOLLOWING THE FLOW OF CONVERSATIONS.



K.V.R. TEJSWI IS PURSUING B.TECH IN COMPUTER SCIENCE & ENGINEERING FROM SANKETIKA VIDYA PARISHAD ENGINEERING COLLEGE, AFFILIATED TO THE NAAC A GRADE ACCREDITED AND ISO CERTIFIED ANDHRA UNIVERSITY. WITH A KEEN INTEREST IN GENERATIVE ARTIFICIAL INTELLIGENCE, HE HAS TAKEN UP AN ACADEMIC PROJECT FOCUSED ON DEVELOPING A CONVERSATIONAL CHATBOT APPLICATION. THE PROJECT AIMS TO LEVERAGE STATE-OF-THE-ART NATURAL LANGUAGE PROCESSING (NLP) AND GENERATIVE AI TECHNIQUES TO CREATE AN INTELLIGENT CHATBOT CAPABLE OF ENGAGING IN HUMAN-LIKE DIALOGUES. THE CHATBOT WILL BE TRAINED ON LARGE LANGUAGE MODELS AND VAST DATASETS TO UNDERSTAND AND GENERATE CONTEXTUALLY RELEVANT RESPONSES, MAINTAINING COHERENCE AND FOLLOWING THE FLOW OF CONVERSATIONS.



CH. RAGHU RAM IS PURSUING B.TECH IN COMPUTER SCIENCE & ENGINEERING FROM SANKETIKA VIDYA PARISHAD ENGINEERING COLLEGE, AFFILIATED TO THE NAAC A GRADE ACCREDITED AND ISO CERTIFIED ANDHRA UNIVERSITY. WITH A KEEN INTEREST IN GENERATIVE ARTIFICIAL INTELLIGENCE, HE HAS TAKEN UP AN ACADEMIC PROJECT FOCUSED ON DEVELOPING A CONVERSATIONAL CHATBOT APPLICATION. THE PROJECT AIMS TO LEVERAGE STATE-OF-THE-ART NATURAL LANGUAGE PROCESSING (NLP) AND GENERATIVE AI TECHNIQUES TO CREATE AN INTELLIGENT CHATBOT CAPABLE OF ENGAGING IN HUMAN-LIKE DIALOGUES. THE CHATBOT WILL BE TRAINED ON LARGE LANGUAGE MODELS AND VAST DATASETS TO UNDERSTAND AND GENERATE CONTEXTUALLY RELEVANT RESPONSES, MAINTAINING COHERENCE AND FOLLOWING THE FLOW OF CONVERSATIONS.



S. SANTHOSHINI IS PURSUING B.TECH IN COMPUTER SCIENCE & ENGINEERING FROM SANKETIKA VIDYA PARISHAD ENGINEERING COLLEGE, AFFILIATED TO THE NAAC A GRADE ACCREDITED AND ISO CERTIFIED ANDHRA UNIVERSITY. WITH A KEEN INTEREST IN GENERATIVE ARTIFICIAL INTELLIGENCE, HE HAS TAKEN UP AN ACADEMIC PROJECT FOCUSED ON DEVELOPING A CONVERSATIONAL CHATBOT APPLICATION. THE PROJECT AIMS TO LEVERAGE STATE-OF-THE-ART NATURAL LANGUAGE PROCESSING (NLP) AND GENERATIVE AI TECHNIQUES TO CREATE AN INTELLIGENT CHATBOT CAPABLE OF ENGAGING IN HUMAN-LIKE DIALOGUES. THE CHATBOT WILL BE TRAINED ON LARGE LANGUAGE MODELS AND VAST DATASETS TO UNDERSTAND AND GENERATE CONTEXTUALLY RELEVANT RESPONSES, MAINTAINING COHERENCE AND FOLLOWING THE FLOW OF CONVERSATIONS.