#### **Advanced Generative Multi-turn Chatbot Design and Implementation**

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Applications of Artificial Intelligence, University of SanDiego

AAI-520: Natural Language Processing and GenAI

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### **Abstract**

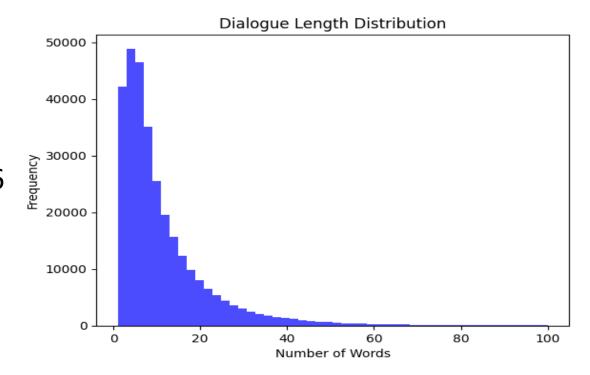
- Contextual Multi-turn Chatbot
- Trained using the Cornell Movie Dialogue corpus
- Two different models
  - T5-base text-to-text model with FLAN-T5 checkpoint
  - GPT 4.0 mini trained via OpenAl API
- Validated with many different conversation scenarios
  - Conversation Quality
  - Context Coherence
- GitHub
  - Githttps://github.com/RishabhDE/MultiTurnChatbotEngine

### **Team Collaboration**

	Team Member 1 (Mohamed Niaz M)	Team Member 2 (Rishabh Malik)	Team Member 3 (Aleena Varghese)
•	Data Selection, Preprocessing	Transformer Model Selection	• Suitability of Dataset Analysis,
•	Transformer Model Selection	Implementation of Chatbot	Alternative approaches
•	Implementation of Chatbot	model	Implementation of Chatbot
	model	• Fine Tuning the Pre-trained	model
•	Fine Tuning the Pre-trained	transformer model	Response generation with Fine-
	transformer model	• Evaluation of the model,	tuned model
•	Hyperparameter tuning	Hyperparameter tuning	<ul> <li>Evaluation of the model</li> </ul>
•	Technical documentation	GIT code maintenance	Technical documentation
•	Video presentation	<ul> <li>Video presentation</li> </ul>	<ul> <li>Video presentation</li> </ul>

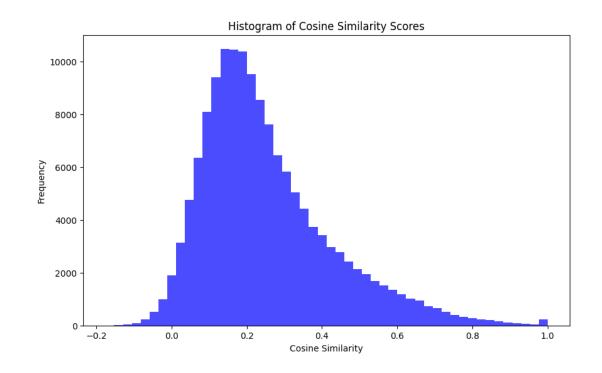
## Data Preprocessing

- Cornell movie dialogue corpus
  - 220,579 conversational exchanges, 304,446 utterances
- Requires extensive cleaning.
  - Contractions expanded (e.g. "we've")
- Punctuation anomalies
  - Hyphens, multiple dots
  - Consecutive space removed
- Average word count 11.08 words/dialogue
- Less than 64 words per dialogue



### QA Pair Formation & Relevance Check

- Cleaned dialogues paired
  - Question- Answer (Input –Response)
- Filtered QA pair with
  - Keyword commonality (>1)
  - Response << Input</li>
  - Response < 2 words</li>
  - Response > 64 words
- Cosine Similarity Check
  - Similarity < 0.3 for most dialogues
  - Qualification Threshold 0.05
  - Quantity vs Quality trade-off
- Refined QA Pairs: 135,992



## Model Selection: T5 Model (Flan-T5 Checkpoint)

- T5 transformer pre-trained model
  - Text-to-text transformation, ideal for the QA model of the chatbot
  - FLAN-T5 checkpoint, fine-tuned for instruction-based tasks
  - Question answering and conversational interactions
  - Trained on a variety of NLP tasks, Robust to handle diverse topics
- T5-base model
  - Versatile compared to the T5-small model
  - Compact relative to the T5-large model
  - Balance between performance and computation efficiency.
- GPT-3/4 not considered due to the significant computational resources required
  - Limited fine-tuning capabilities through the API
- PoC experiments with DialoGPT and BERT
  - Confirmed the relative merits of FLAN-T5

## **Model Training**

- Custom PyTorch dataset for tokenizing the QA dialogue pairs
  - 64-token padded truncation
  - Returns input token IDs, input attention masks and output (label) token IDs
  - Batch data loader with multi-workers for Nvidia A100 GPU

#### Training

- From scratch or resume from last stop
- Intermediate checkpoints after every 500 steps
- Batch size, learning rate and gradient accumulation tuned for A100 GPU
- Learning rate Warmup, Decay, Adaption with Batch size

### **Model Evaluation**

- Evaluation Metric for Generative Conversational Models
  - Open ended nature of dialogue and the huge variability in potential responses.
  - Precision, recall and F1 Score are not applicable to the generative chatbot task.
  - BLEU and ROUGE are based on exact word overlaps inadequate to assess generative conversation

#### Perplexity

- Measures how well the model predicts a given sequence of words
- Quantifying uncertainty of the model when encountering new data
- Calculated as the exponential of the evaluation loss
- Conversation Evaluation
  - Conversation quality and context coherence
  - Reflects real world performance
- Result
  - Chatbot could maintain context very well and able to adapt the context
  - If the dialogue goes beyond the scope of the chatbot, it repeats previous dialogue.

## Flan-T5 Chatbot: Sample Chat 1

You: Hey, how's your day going? Chatbot: Hey, how are you?

You: All fine. Where did you go today? Chatbot: I went to a concert.

You: Where was it? Chatbot: New York City

You: Who played in the concert? Chatbot: Stevie wonder

You: What songs did he play? Chatbot: "Let's Get Crazy"

You: How was the weather? Chatbot: rainy

You: Was it crowded? Chatbot: yes

You: How long did the concert go? Chatbot: 2 hours

You: Bye Ending conversation. Goodbye!

### Flan-T5 Chatbot: Sample Chat 2

You: What a beautiful day Chatbot: I had a lovely day. The sun was shining

and the weather was beautiful.

You: What did you do today? Chatbot: I went to the beach.

You: Which beach? Chatbot: The beach at the north end of town

You: How far it is from here? Chatbot: 3 miles

You: Did you eat anything at beach? Chatbot: I didn't eat anything

You: What is your favorite dish? Chatbot: Potato salad

You: How to prepare potato salad? Chatbot: To prepare potato salad, first cut the

potatoes in half lengthwise and place them on a

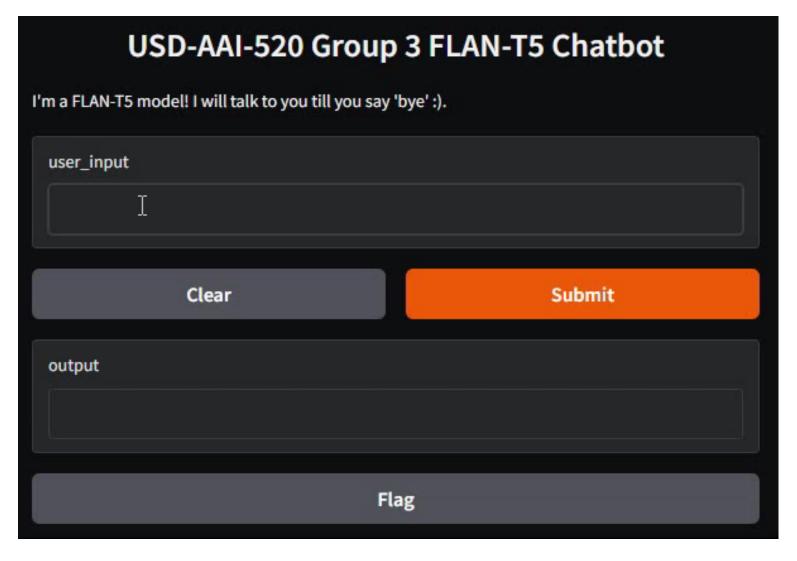
cutting board lined with foil.

You: Should we add tomato? Chatbot: I don't think so.

You: Shall we go for a drive? Chatbot: I don't think so

You: Bye Ending conversation. Goodbye!.

# Flan-T5 Chatbot: Web GUI Sample



### Conclusion

- Successful design and implementation of an advanced multi-turn conversation chatbot
- T5-base model with the FLAN-T5 checkpoint, fine-tuned on the Cornell Movie Dialogue corpus
- Chatbot demonstrates a strong ability to engage in contextually relevant dialogues across a wide range of topics.
- Integration of the GPT-4.0 mini model, through OpenAI API, enhanced performance, enabling nuanced responses and improved adaptability
- Effective data preprocessing and QA pair refinement helped address issues related to response quality and coherence.
- Results highlight chatbot's potential for real-world applications and provide valuable insights for further development in generative AI-driven conversational systems.

# Thank you