

# Investigating Explainable Methods for LLMs

## Case Study: Emotionally Aware Chatbot

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# What is the project about?

## Goal



Analyze explainable AI techniques to interpret the decisions made by the emotion detection model.

## Key Features:


Analysis and Implementation of explainable methods

Fine-tuned conversational models for emotion detection

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# Why this project?



Existing LLMs often lack transparency

Regulatory Compliance:  
Adhering with GDPR and European  
regulatory frameworks (AI Act)

Bridging the knowledge gap between  
human/computer interaction

# Project Structure

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1. Literature Review and Comparative Analysis of Explainable Methods for LLMs
2. Implementation of an Emotion Recognition Model
3. Model Explanation
4. Fine-Tuning and Evaluation

# System Architecture



## Input Handling:

Text input: Direct processing.  
Voice input: Speech-to-text conversion



## Emotion Analysis:

Emotion classifier pipeline (BERT)



## Response Generation:

Generate an empathetic response based on the detected emotion



## Output Delivery:

Response text returned to the user.

# Datasets

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## 1. GoEmotions:

1. Annotated with 27 emotions + neutral.
2. Source: Curated from Reddit comments.
3. Statistics:
  1. 58,000 annotated examples.
  2. Balanced across multiple emotion categories.

## 2. DailyDialog:

1. Multi-turn conversational dataset.
2. Annotated for emotions and dialogue acts.
3. Statistics:
  1. 13,000 multi-turn dialogues.
  2. Includes annotations for intent and sentiment.



# Model Choices

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## 1. Emotion Detection:

1. Model: BERT fine-tuned on GoEmotions.
2. Output: Probabilities for each emotion.

## 2. Response Generation:

1. Model: using DialoGPT or BlenderBot Fine-Tuning on DailyDialog.

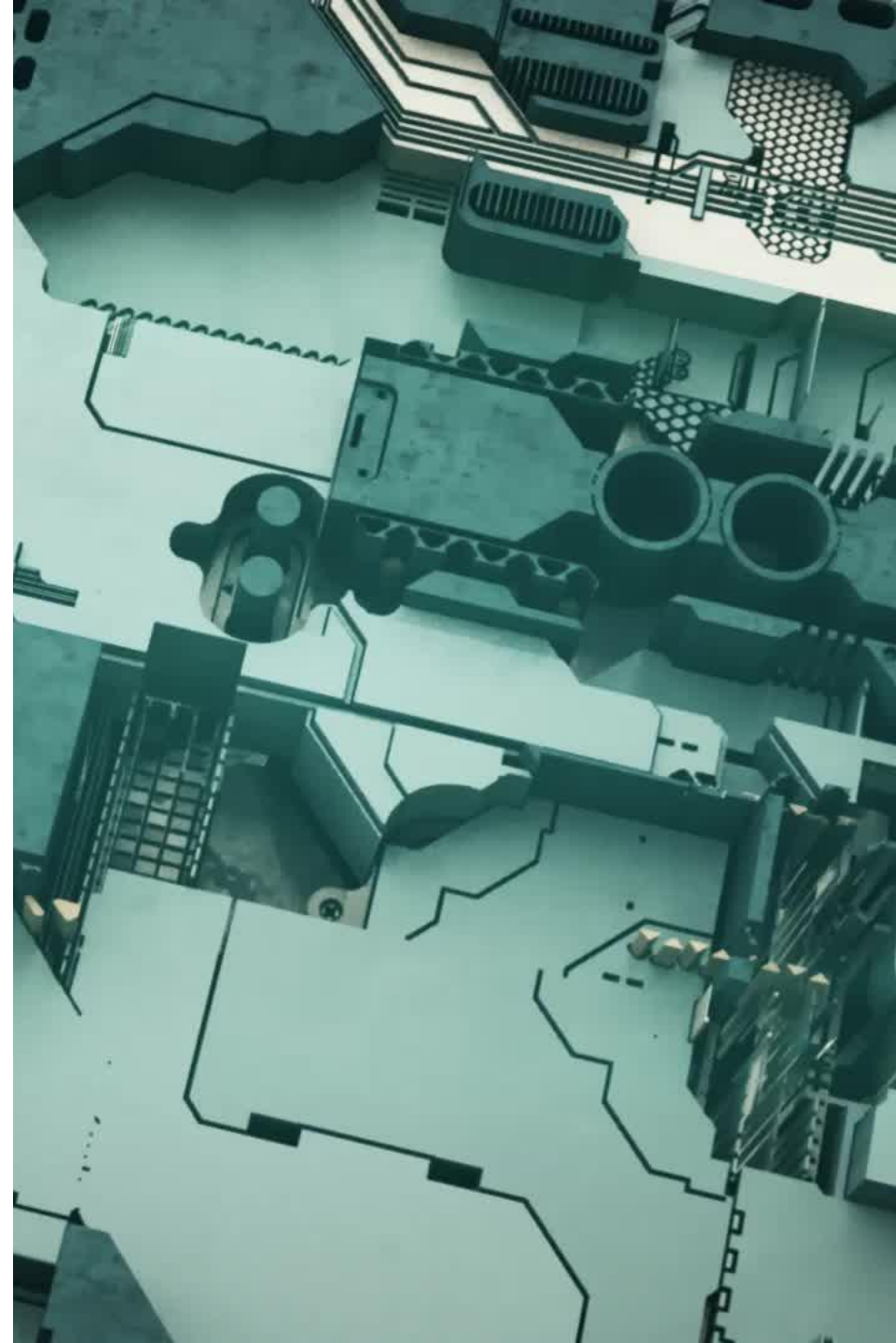




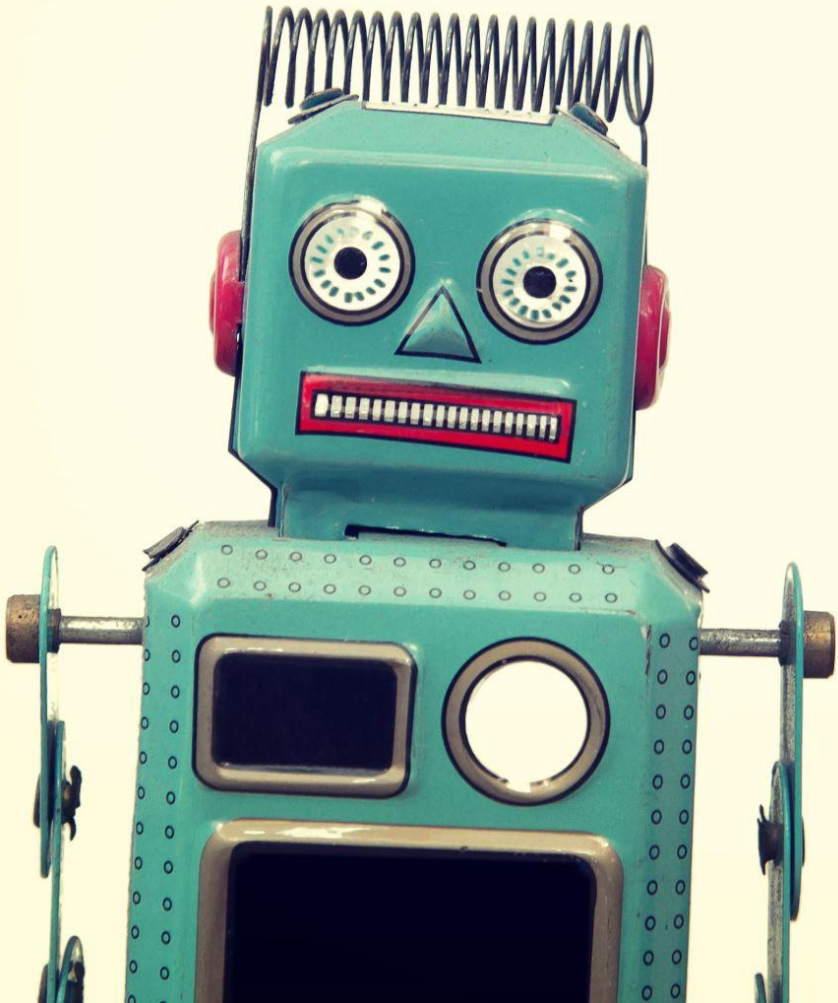
# Fine-Tuning Process

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- Dataset: DailyDialog (context-response pairs).
- Steps:
  - 1.Tokenize and preprocess conversations.
  - 2.Train using transformers library.
  - 3.Evaluate and save the fine-tuned model.







# Live Demonstration

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- Input examples:
  - "I passed my exam!"
    - Detected Emotion: Joy (95%).
    - Bot Response: "Congratulations! You worked hard for this!"
  - "I feel so sad today."
    - Detected Emotion: Sadness (85%).
    - Bot Response: "I'm sorry to hear that. Want to talk about it?"

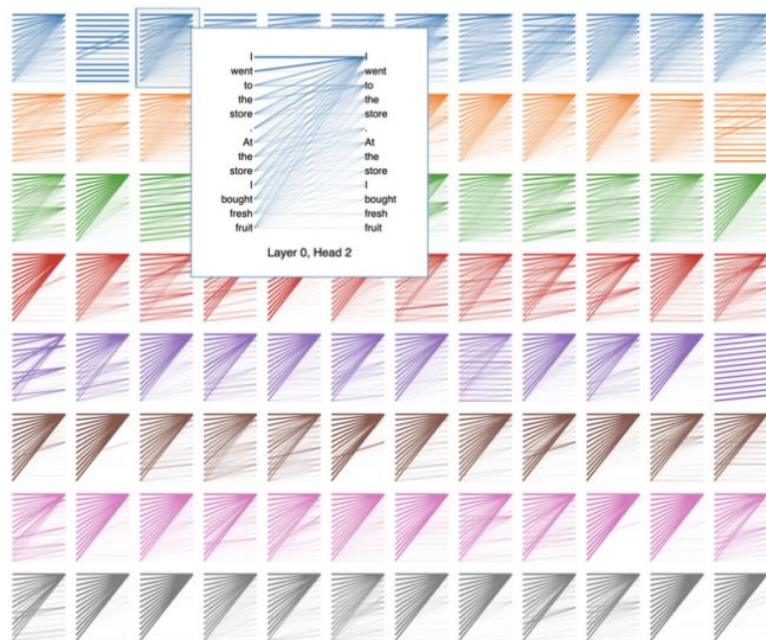
# Explanations

Feature-Attribution Based Explanations

Attention-Based Explanations

Example-Based Explanations

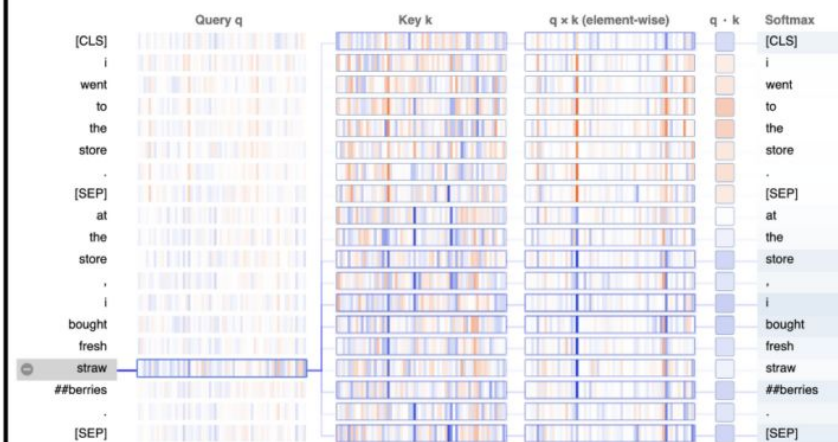
# BERTViz



model view



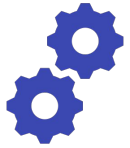
attention head view



neuron view

# Key Challenges

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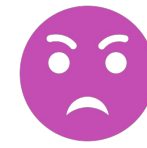
Balancing response diversity and coherence.



Selecting the right method



Managing computational requirements for fine-tuning.



Interpreting ambiguous results

# Thank you for your attention

## Questions?

### References:

BertViz: <https://www.comet.com/site/blog/explainable-ai-for-transformers/>

Zhao, H., Chen, H., Yang, F., Liu, N., Deng, H., Cai, H., ... & Du, M. (2024). Explainability for large language models: A survey. *ACM Transactions on Intelligent Systems and Technology*, 15(2), 1-38.