

# MentalChat16K: A Survey on Conversational Mental Health AI

## CMPE 255 - Data Mining Assignment

**Student:** Bala Anbalagan  
**Course:** CMPE 255 - Data Mining Sec 47  
**Semester:** Fall 2025  
**Topic:** MentalChat16K: A Benchmark Dataset for Conversational Mental Health Assistance

## Paper Information

Attribute	Details
Title	MentalChat16K: A Benchmark Dataset for Conversational Mental Health Assistance
Authors	Jing Xu, Tianming Wei, Bohan Hou, Patryk Orzechowski, et al.
Institution	University of Pennsylvania
arXiv Link	<a href="https://arxiv.org/abs/2503.13509">https://arxiv.org/abs/2503.13509</a>
Dataset	<a href="#">HuggingFace - MentalChat16K</a>
Official Code	<a href="#">GitHub - MentalChat16K</a>

## Overview

This repository contains my comprehensive analysis and presentation of the MentalChat16K paper, which introduces a benchmark dataset for developing AI-powered mental health conversational agents. The paper addresses a critical gap in mental health AI research by combining:

- **Real clinical data:** 6,338 QA pairs from 378 anonymized behavioral health intervention transcripts
- **Synthetic data:** 9,775 QA pairs generated using GPT-3.5 Turbo covering 33 mental health topics
- **Total:** 16,000+ question-answer pairs for training empathetic AI assistants

## Why This Paper Matters for Data Mining

Data Mining Aspect	Application in Paper
Data Collection & Curation	Multi-source dataset creation (real + synthetic)
Data Preprocessing	Privacy-preserving paraphrasing using local LLMs
Data Quality	Manual filtering and de-identification pipelines
Feature Engineering	7 therapeutic evaluation metrics

Data Mining Aspect	Application in Paper
Model Evaluation	Multi-evaluator framework (GPT-4, Gemini, Human)
Benchmark Creation	Standardized evaluation for mental health AI

## Repository Structure

```
MentalChat16K/
├── README.md                # This file
├── slides/
│   └── MentalChat16K_Presentation.pdf    # Slide deck
├── images/
│   ├── architecture.png      # System architecture diagram
│   ├── dataset_composition.png
│   ├── evaluation_metrics.png
│   └── results_comparison.png
├── notebooks/
│   └── data_exploration.ipynb  # Optional: Dataset exploration
├── code/
│   └── evaluation_demo.py      # Optional: Evaluation demo
├── ARTICLE.md               # Medium article draft
└── VIDEO_LINK.md            # Link to video presentation
```

## Key Contributions of the Paper

### 1. Dataset Creation

- First large-scale mental health dialogue dataset combining real and synthetic data
- 16,000+ QA pairs covering depression, anxiety, grief, relationships, and more
- Doubles the size of previous comparable datasets (Psych8K)

### 2. Privacy-Preserving Pipeline

- Uses local Mistral-7B for paraphrasing sensitive clinical transcripts
- Avoids uploading patient data to commercial APIs
- Manual de-identification of PII (names, addresses, financial info)

### 3. Novel Evaluation Framework

Seven mental health-specific metrics:

Metric	Description
Active Listening	Reflects and validates user concerns
Empathy & Validation	Shows understanding of emotional states

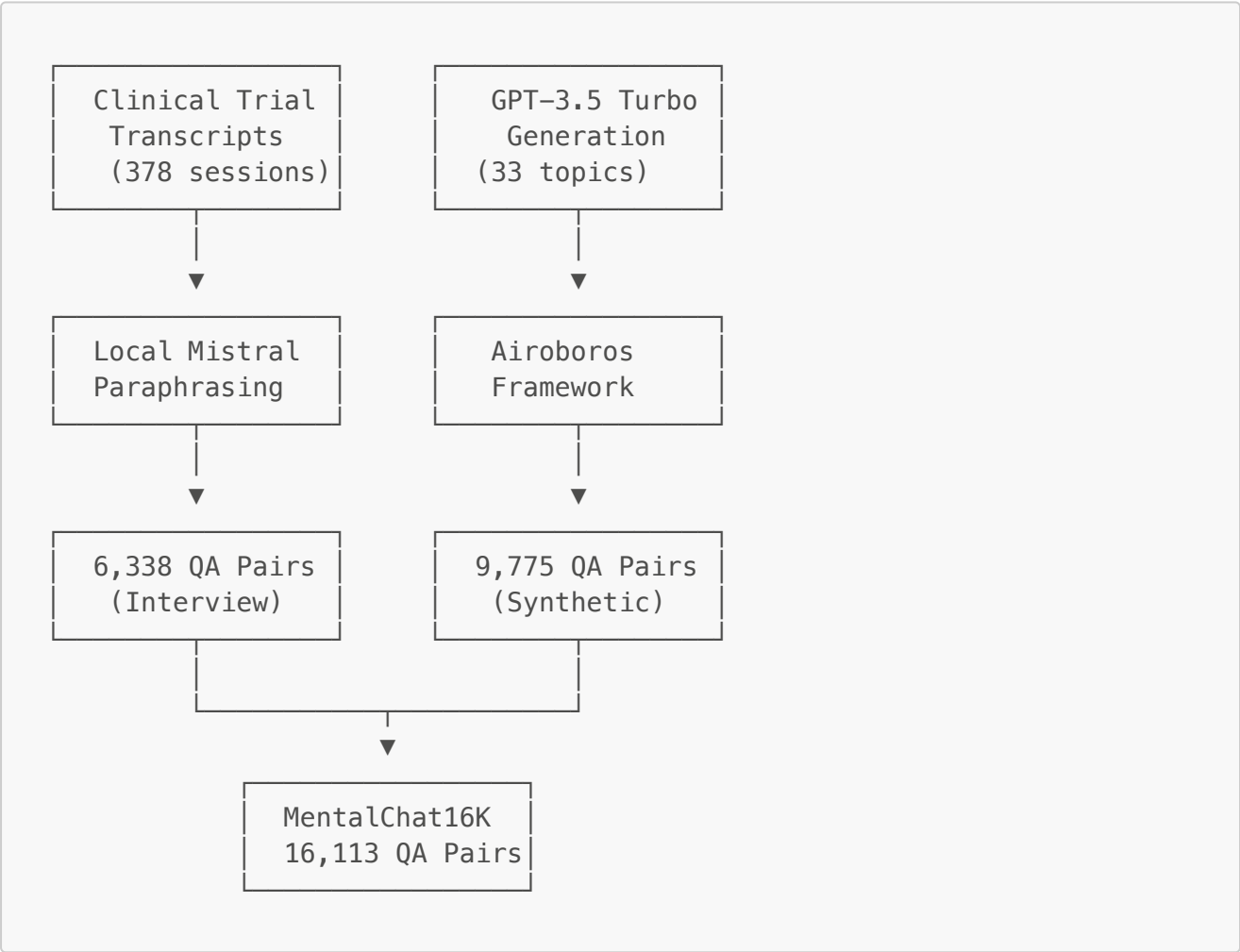
Metric	Description
Safety & Trustworthiness	Prioritizes user safety, suggests professional help
Open-mindedness	Non-judgmental, accepting of diverse perspectives
Clarity & Encouragement	Clear communication, positive reinforcement
Boundaries & Ethics	Maintains appropriate professional boundaries
Holistic Approach	Considers overall well-being

4. Comprehensive Benchmarking

- Fine-tuned 7 different 7B-parameter LLMs using QLoRA
- Compared performance with GPT-4, Gemini Pro, and human evaluators
- Statistical significance testing across all metrics

Methodology

Data Pipeline



Fine-tuning Configuration

Parameter	Value
Method	QLoRA (Quantized Low-Rank Adaptation)
Models	LLaMA-2-7B, Mistral-7B, Vicuna-7B, Zephyr-7B
Hardware	NVIDIA A100 (80GB)
Training Configs	Synthetic only, Interview only, Combined

## Results Summary

### Key Findings

1. **Fine-tuned models significantly outperform base models** across all 7 metrics
2. **GPT-4 evaluator** favored synthetic data fine-tuning (alignment with GPT-3.5 patterns)
3. **Gemini Pro evaluator** valued real interview data, especially for safety metrics
4. **Human evaluators** consistently preferred fine-tuned models
5. **Combined training** did not always outperform individual approaches

### Performance Comparison

Model Type	Avg Score (GPT-4)	Avg Score (Gemini)	Avg Score (Human)
Base Models	~6.5	~6.2	~5.8
Synthetic Fine-tuned	~8.2	~7.5	~7.8
Interview Fine-tuned	~7.8	~8.0	~7.6
Combined Fine-tuned	~8.0	~7.8	~7.7

## Deliverables

### 1. Medium Article

**Link:** [Add your Medium article link here]

A comprehensive article covering:

- Introduction to mental health AI challenges
- Dataset creation methodology
- Evaluation framework explanation
- Key results and implications
- Personal analysis and future directions

### 2. Slide Presentation

**SlideShare Link:** [Add your SlideShare link here]

**PDF Location:** [slides/MentalChat16K\\_Presentation.pdf](#)

### 3. Video Presentation

**Duration:** 10-15 minutes

**Link:** [Add your video link here]

---

## How to Use This Repository

### Explore the Dataset

```
from datasets import load_dataset

# Load MentalChat16K from HuggingFace
dataset = load_dataset("ShenLab/MentalChat16K")

# View sample
print(dataset['train'][0])
```

### Run Evaluation Demo (Optional)

```
cd code
python evaluation_demo.py
```

---

## References

1. Xu, J., Wei, T., Hou, B., et al. (2025). MentalChat16K: A Benchmark Dataset for Conversational Mental Health Assistance. arXiv:2503.13509
  2. Dettmers, T., et al. (2023). QLoRA: Efficient Finetuning of Quantized LLMs
  3. Touvron, H., et al. (2023). LLaMA 2: Open Foundation and Fine-Tuned Chat Models
  4. Chen, J., et al. (2023). ChatPsychiatrist: Evaluating LLMs for Mental Health Applications
- 

## License

This project is for educational purposes as part of CMPE 255 coursework.

---

## Contact

**Student:** [Your Name]

**Email:** [Your SJSU Email]

**LinkedIn:** [Your LinkedIn Profile]

---

*Last Updated: November 2024*

