

MeetingGPT

Collaboration empowers teams through communication



NLP Solution for Extracting Key Insights from Online Meetings

MIDS Capstone

Introduction of Team



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At MeetingGPT, we harness the power of advanced NLP to transform the essence of meetings into actionable insights, fostering meaningful connections and efficiency in the modern digital landscape.



Problem Space

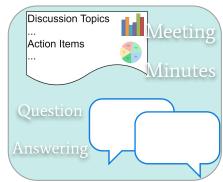




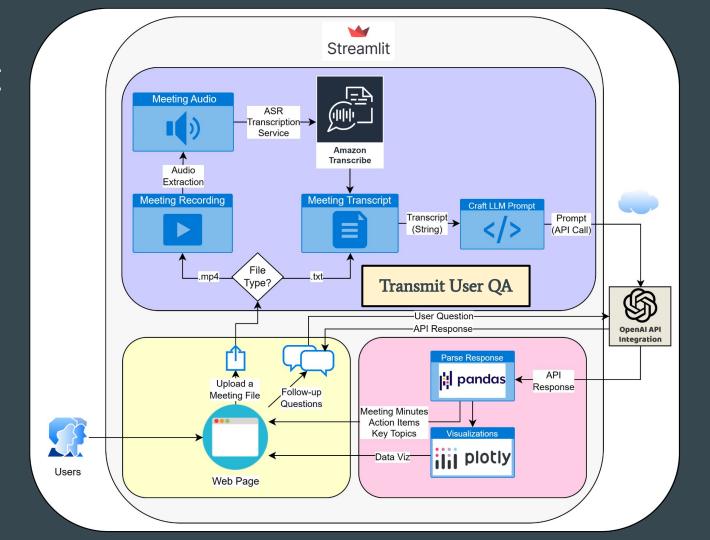
Problem ... Too Many Meetings Fix ... Comprehensive AI Tool Experts ... Summary & Review Users ... Business & Education Impact ... Save Time & Organize







How It Works

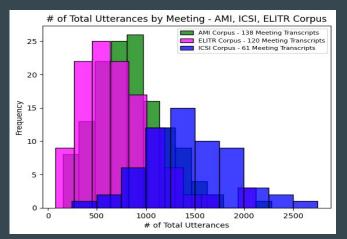


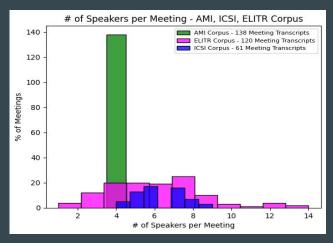
Demo meetingapt-final.streamlit.app



Datasets

Corpus	# of Meetings	Meeting Durations	# of Participants	Domain
AMI	138	10-50 mins	4	Industrial product design
ICSI	61	30-100 mins	4-9	Academic research
ELITR	120	10-50 mins	2-14	Computer science
Capstone Project Meetings	15	45-60 mins	4	Data Science





Four Technical Meeting Datasets used for model fine-tuning & evaluation throughout the project Datasets include wide range of Meeting Duration, # of Participants & Technical topics

Model Selection → Fine-Tuned Transformers vs. LLMs

Model Evaluation Results - Meeting Summary Generation for AMI Dataset							
		Max Input	Fine-Tuning Dataset -	ROUGE (vs. Ref Summ) - AMI Test (89 meetings)		· · · · · · · · · · · · · · · · · · ·	
Model Type	Model	Tokens	Training	ROUGE-1	ROUGE-2	ROUGE-L	
Pre-Trained Transformers	Longformer	16384	N/A	0.218	0.026	0.119	
(In-House)	Longformer w/ Fine-Tuning	16384	AMI (50 meetings, 5 epochs)	0.394	0.105	0.202	
API-accessed	OpenAl - GPT3.5 Turbo	16384	N/A	0.418	0.104	0.223	
LLMs (External)	Meta - CodeLlama	16384	N/A	0.419	0.109	0.231	

Fine-Tuned Transformer falls short in ROUGE and lacks the versatility of LLMs
Proceeding w/ high-performing Base LLMs will enable team to expand tool capability beyond just a summary

Model Outputs / Prompt Engineering

- The use of LLMs allowed us to expand from a simple meeting summary to five output sections:*
 - Abstract Summary

*Latest Prompt details documented in backup

- Key Topics
- Action Items
- Sentiment
- Speaking Time %
- Meeting Datasets used to iterate on optimal prompt & evaluate LLM performance (GPT3.5 Turbo & CodeLlama) in the following metrics to downselect optimal LLM for our product:
 - Response Time
 - Cost / Meeting
 - % of Meetings requiring Truncation (Exceeds Max Token Length)
 - Qualitative Team Evaluation → % Acceptability of Final Outputs for each sub-section

Final Solution Flow:

Transcript→ Multi-Question LLM Prompt → Reformat LLM Prompt → Post-Processing Functions → Final Dashboard Viz

LLMs solution enabled team to extract insights beyond a summary & add novelty to our product output Final Solution required two sequential LLM prompts & post-processing functions to ensure consistent output

Qualitative Team Evaluation - % Acceptable Output

% Acceptable Scoring Ground Rules:

- For each of the meeting, the team scored each subsections w/ a 0, 0.5 or 1 score:
 - **0** → **Unacceptable** Minimal/No output generated
 - **0.5** → **Moderately Acceptable** Output generated, but the content has one primary issue
 - 1 → Fully Acceptable Output generated, formatted per intent w/ content that makes sense

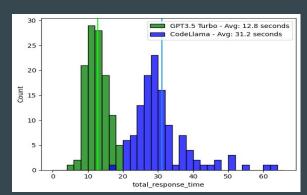
% Acceptable Scores Final Viz Output Evaluation (120 Test Set Meetings (AMI/ELITR/ISCI))						
LLM	Summary	Action Items	Topics	Speaking %	Sentiment	
GPT3.5 Turbo	97.9 %	80.8 %	95.4 %	84.2 %	99.2 %	
CodeLlama	97.1 %	75.0 %	87.9 %	55.8 %	98.3 %	

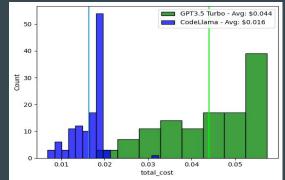
Test set outputs scored to collect quantitative data to determine relative model performance GPT3.5 Turbo is best in every sub-section, especially % speaking time, topics & Action Items

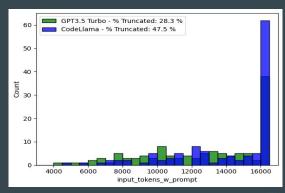
Final Model Evaluation - Quantitative Metrics

Quantitative Evaluation Metrics - GPT3.5 vs. CodeLlama (120 Meeting Test Set)							
	Avg Response Time (sec)		Avg Cost / Meeting (\$)		% of Meetings Truncated		
	Week 10	Week 14	Week 10	Week 14	Week 10	Week 14	
GPT3.5 Turbo	50	13	0.15	0.04	18	28	
CodeLlama	60	31	0.06	0.02	30	48	

- Cost & Response time of models driven down by consolidating into one multi-question prompt
- With more lengthy prompt this drove the input tokens up impacting % of meeting truncated







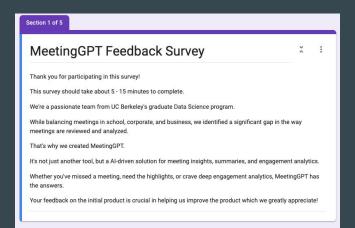
GPT3.5 Turbo superior to CodeLlama in every metric aside from cost (\$0.04 vs. \$0.02/meeting) However, with it's faster response time & more reliable output it was selected for our product

Tie Back

- Corporate Professionals
 - Efficient Review of Key Points
 - Actionable Insights and Follow-ups
 - O Time Management
 - Improved Communications
- Education Instructors & Students
 - Engagement Analytics
 - Meeting Recap for Learning
 - Collaborative Learning:
- General Users
 - Q&A Features
 - Visual Analytics
 - Summarize general audio/transcripts

Top 2-3 Roadmap items if more time

- Cost
 - Reduce API costs further (near \$0)
- UI/UX beautification
 - Customizable dashboards
 - o **Tableau**
- Business Enterprise functions
 - o login/password, archive meetings, group access
- Multi Language Support
 - Any language to English; Any language to any language
- Human Feedback Loop
 - Option to loop back based on user feedback
- Time based Q&A
- Continue User Based Survey



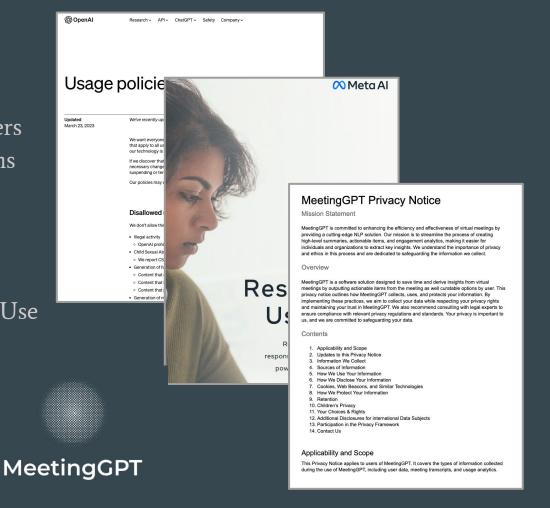
Conclusion

MeetingGPT harnesses the power of advanced NLP to transform the essence of meetings into actionable insights, fostering meaningful connections and efficiency in the modern digital landscape. Collaboration empowers teams through communication where no idea is left behind.

MeetingGPT

Privacy Notice

- Transparency is key to the Users
- Notice is outlined in 14 sections
- Contact Us
 - Privacy Inquiries
 - o General Inquiries
- Adhered to OpenAI's Usage
 Policies & Meta's Responsible Use
 Guide



Acknowledgement & Resources

University of California - Berkeley for their invaluable support, expertise, and guidance provided by the faculty and fellow students that have been integral to the success of our capstone. OpenAI for the ChatGPT model, which played a significant role in the development of our project by providing advanced natural language processing capabilities. Meta AI for the LLaMA-2 model, which enhanced our project's deep learning capabilities and another LLM for comparison.

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