

Evaluation on Audio-LLMs and Beyond

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Event Space

LLM - Can do a lot of things in just One Model

- LLM solves problems by following human instructions
 - Why does the sky appear blue during the day? -> QA
 - Summarize the following article in one paragraph. -> Summarization
 - What is the sentiment of the above sentence? -> Sentiment Analysis
 - Write a Python function that returns the factorial of a number. -> Coding

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Can we expand the one model fits all task paradigm? And how?

An intelligent system takes all the surrounding information, E.g. as humans

- Information
 - Textual data
 - Audio and speech
 - Image and video
 - Sensor data
 - Numerical or structured data
 - Location and geospatial data
 - Knowledge, reasoning, and others

Towards AGI – Artificial General Intelligence



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AudioLLM: Audio-based Large Language Models

Agenda

- What is AudioLLM? What is new?
- What is missing in its performance measurement?
- AudioBench: A Universal Benchmark for Audio Large Language Models – NAACL 2025
- What is next?

Conventional Audio / Speech Solutions

"One task: one model" paradigm

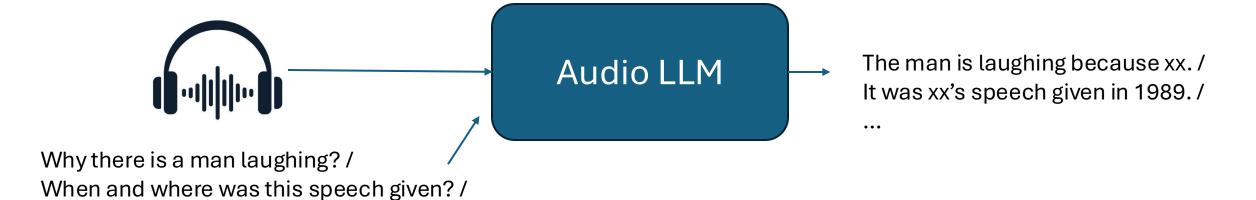
- Automatic Speech Recognition Whisper, Conformer, Wav2Vec 2.0
- Speech Translation Whisper, SeamlessM4T, Translatotron 2
- Audio Captioning CLAP, PANNs, AudioCLIP
- Sentiment Recognition EmoReact, emotion2vec

Characteristics:

- Usually supervised learning
- Predefined prediction space

AudioLLM – One Model for all audio/speech understanding tasks

Audio + Human Instruction = Answer



Characteristics:

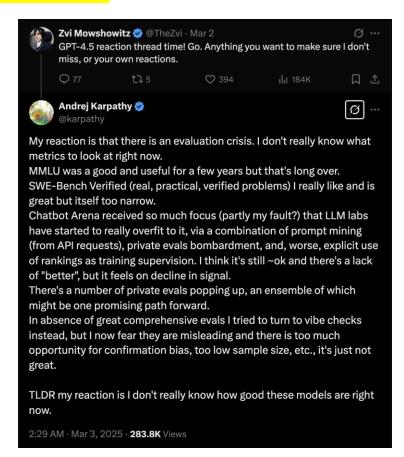
Can you help me transcribe the speech? /

Is it an indoor or outdoor environment? / ...

- Complete the task by following human instructions
- Can generalize to zero-shot / unsupervised scenarios
- Response in human language

How do researcher evaluate **LLMs**?

- Multiple choices questions
 - Unnatural for its common use cases
- Open-ended questions
 - How can we judge answer A is better than B?
- Human rating
 - Troublesome, no immediate feedback



Comments from X.com for GPT4.5 Release March 3, 2025

TLDR: It is hard.

How do researcher evaluate AudioLLMs?

- Conclusion: No unified paradigm yet.
- Major paradigm shift:
 - 1. One model that handles (any) audio/speech tasks
 - 2. Open answer generation
 - 3. Follow human instructions

How do researcher evaluate AudioLLMs?

- Conclusion: No unified paradigm yet.
- Major paradigm shift:
 - 1. One model that handles (any) audio/speech tasks → The more, the better
 - 2. Open answer generation Judgement by LLMs with/without references
 - 3. Follow human instructions Instruction following measurement
- Performance measure
 - Multiple Choice Questions (MCQ)
 - Model as Judge
 - Others?

AudioBench: A Universal Benchmark for AudioLLMs

 Task categories: Speech Understanding, Audio-Scene Understanding, Voice Understanding (Paralinguistic), Music Understanding, Singlish Understanding, ...

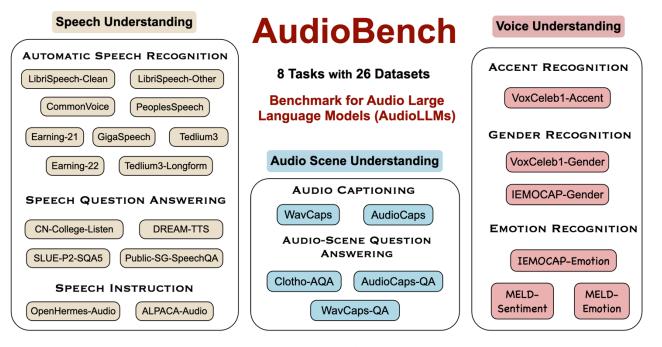
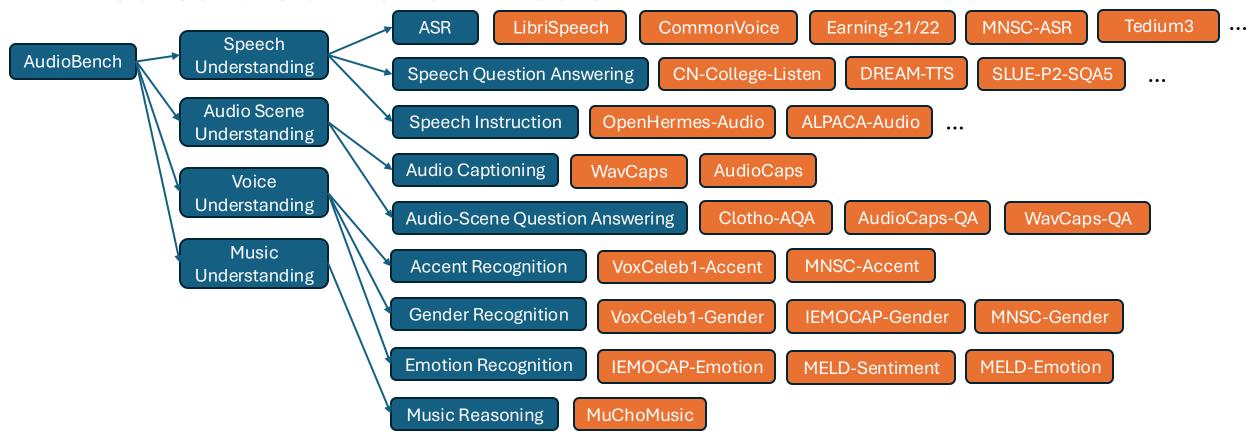


Figure 1: Overview of AudioBench datasets.

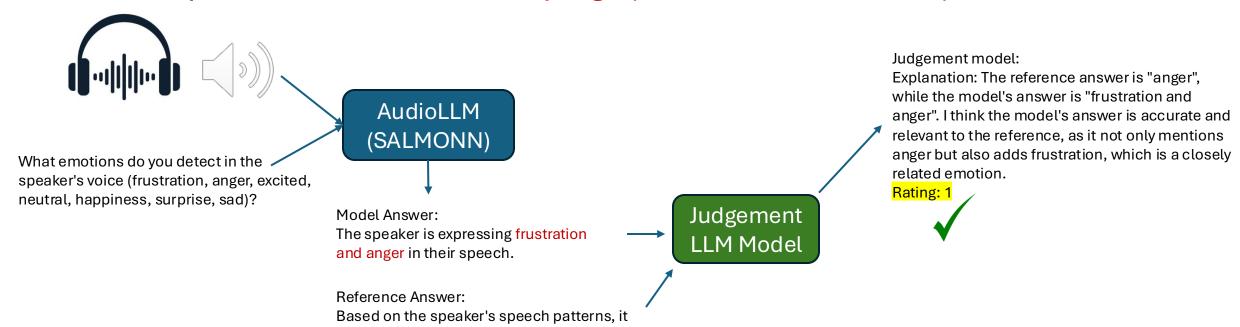
Collections of Evaluation Tasks



- Evaluation Measurement
 - ASR Word Error Rate
 - Speech Translation BLEU score

seems like they are feeling anger.

Speech QA? -> Model-as-a-judge (LLAMA-3-70B-Instruct)



• Evaluation of AudioLLMs and Cascade Models

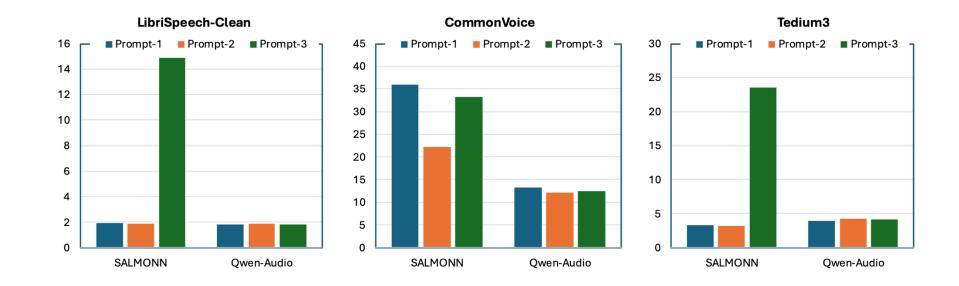
Dataset Name	AudioLLMs				Whisper+Llama3
	SALMONN	Qwen-Audio-Chat	WavLLM	Qwen2-Audio-Instruct	wnisper+Liama5
Speech Understanding	5				
$LibriSpeech-Clean_{(\downarrow)}$	55.58	2.25	2.10	3.20	1.83
$LibriSpeech-Other_{(\downarrow)}$	41.80	4.16	4.80	6.07	3.71
CommonVoice-15(1)	33.75	11.65	14.53	11.44	9.89
$PeoplesSpeech_{(\bot)}$	34.33	30.72	37.92	22.32	14.54
$GigaSpeech_{(\downarrow)}$	14.22	13.32	15.49	11.89	9.51
$Tedlium3_{(\downarrow)}$	8.56	4.00	6.62	6.39	3.81
Tedlium3-Longform(,)	18.39	45.29	45.37	95.35	4.75
Earning- $21_{(\downarrow)}$	26.87	38.46	64.47	98.65	11.77
Earning- $22_{(\downarrow)}$	36.38	51.18	66.72	98.84	15.61
CN-College-Listen	50.51	60.85	65.43	74.50	85.25
SLUE-P2-SQA5	78.24	76.12	83.92	80.05	82.99
DREAM-TTS	55.93	57.76	64.56	66.70	86.09
Public-SG-SpeechQA	56.77	57.47	58.55	58.31	64.94
OpenHermes-Audio	19.20	11.00	22.40	44.80	63.0
ALPACA-Audio	12.40	9.60	21.60	52.60	70.8
Audio Scene Understa	nding				
Clotho-AQA	51.18	58.20	43.01	50.92	29.47
WavCaps-QA	46.25	38.68	26.25	44.47	17.38
AudioCaps-QA	47.03	47.99	29.84	45.75	16.71
WavCaps _(M.J.)	21.16	29.25	6.40	33.78	3.45
$AudioCaps_{(M.J.)}$	34.37	47.99	4.17	40.78	2.47
WavCaps _(METEOR)	17.72	24.02	9.78	21.34	13.89
AudioCaps _(METEOR)	21.20	27.70	6.70	19.89	7.95
Voice Understanding					
IEMOCAP-Emotion	21.56	27.34	45.91	49.30	34.43
MELD-Emotion	33.06	50.57	41.07	40.54	33.36
MELD-Sentiment	41.87	43.87	50.08	53.49	43.87
VoxCeleb1-Accent	28.06	45.70	37.65	29.19	39.33
VoxCeleb1-Gender	88.90	70.56	70.51	99.12	53.41
IEMOCAP-Gender	51.60	51.13	45.29	49.30	51.50

What about diverse prompts?

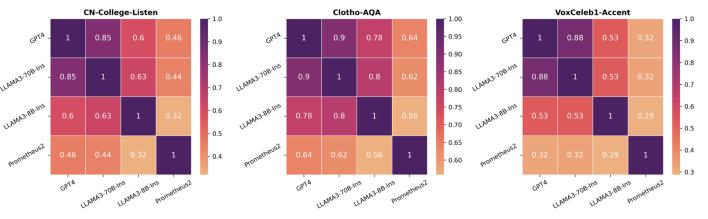
E.g. ASR Instructions

- "Please help me transcribe the speech into text.",
- "Transcribe the spoken words into written form.",
- "Listen to the speech and provide the text version.",
- "Transform the speech into a text document as transcriptions.",

- ..



Most recent models are becoming robust towards diverse prompts.

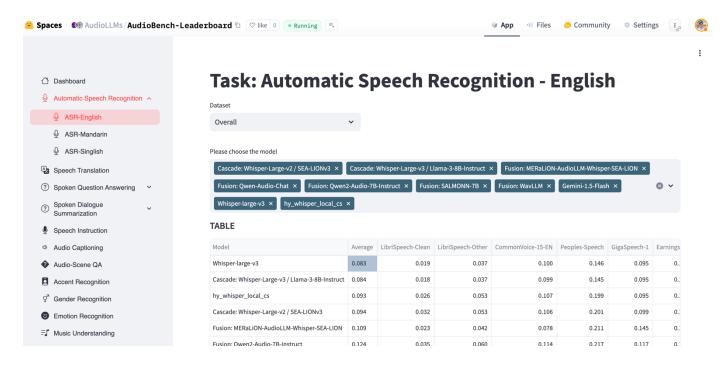


Correlation between judgement models

- What judgement model should we use?
 - LLAMA-3-8B-Instruct
 - Not have enough capability
 - GPT4o
 - Costly and will have version updates
 - LLAMA-3-70B-Instruct
 - Can fit in 1 H100 GPU after int4 quantization
 - Good instruction following and judgement quality
 - High correlation with GPT4o (especially for reference-based judgements)

AudioBench - Resources

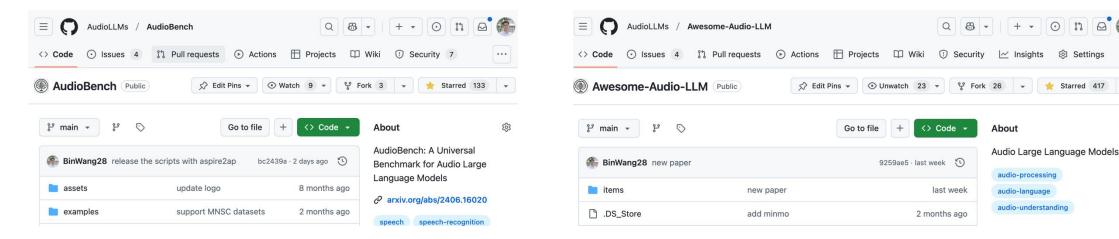
- Paper: https://arxiv.org/pdf/2406.16020 (NAACL 2025)
- Leaderboard: https://huggingface.co/spaces/AudioLLMs/AudioBench-Leaderboard



Wang, Bin, Xunlong Zou, Geyu Lin, Shuo Sun, Zhuohan Liu, Wenyu Zhang, Zhengyuan Liu, AiTi Aw, and Nancy F. Chen. "Audiobench: A universal benchmark for audio large language models." NAACL, 2025

AudioBench - Resources

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- Leaderboard: https://huggingface.co/spaces/AudioLLMs/AudioBench-Leaderboard
- Toolkit: https://github.com/AudioLLMs/AudioBench



AudioBench Toolkit

Collection of AudioLLM Publications

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What should we do next? AudioBench v2

- Instruction following of AudioLLMs
- Multilingual Support
- Multi-round Evaluation
- Evaluation on Speech Generation (Can hear and speak!)

Next Talks

- Our Models!
- MOWE-Audio (ICASSP 2025)
- MERaLiON-AudioLLM (Technical Report, Open-Sourced)