





HEAR ME OUT (& THINK): MAESTRO, A MULTIMODAL AGENTIC MODEL WITH EFFICIENT, SYNERGISTIC TEXT-REASONING OPTIMISATION FRAMEWORK

"And then he was

sentenced to death

Yeah, I'm sure."

"And then the guy

who said it actually

Classification Layer Threshold = 0.65

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Specific Use-Case: Hateful Video Classification

500 hours of videos/min 1 Billion users

- 1. Demands Inference Reasoning
- 2. Multimodal Reasoning

Idea-frame pairs, $P = \{(c_1, F_{\Delta t_1}), ..., (c_k, F_{\Delta t_k})\}$

- 3. Temporal Understanding
- 4.Adaptability

Semantic Chunks

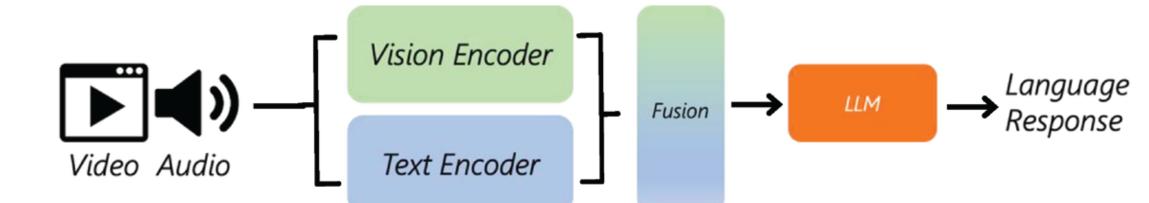
Boundary Labels

Transcript,

 $T = \{t_1, t_2, \dots, t_n\}$

00:00

CURRENT VLMs



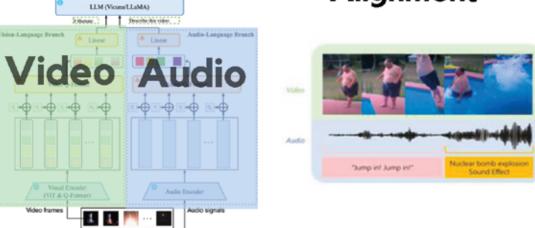
High Computational Overhead

Exhaustive frame-**Use of Q-formers** by-frame analysis / transformers

Achieving fine-grained understanding in VLMs often necessitates processing a vast number of video frames, leading to high computational costs that scale significantly with input length

Missing integration of audio modality

Separate Streams Neglects Time Alignment



Weak/Static Reasoning

Static Reasoning Pipeline

USER: Is this video hateful? Explain why or why not Video-LLaMA 🐂 : Based on the video descriptions, it is not clear whether the video is hateful or not. However, the video shows a large crowd of people at an airport, and some of them appear to be excited or happy. The video also shows a person with a phone, which could be a sign of annoyance or frustration

METHODOLOGY

Transcript Chunking

"I was at work

"and I said, um,

sir, actually it's a

C T1 T2 T[SEP] Tx TM

BERT

EICLS E1 E2 EISEPI Ex EM

CLS W1 -- CLS SEP Wx -- Wm

Speaker 1: "Now, let's hear from this ma'am with this stirring story, which I'm sure is true."

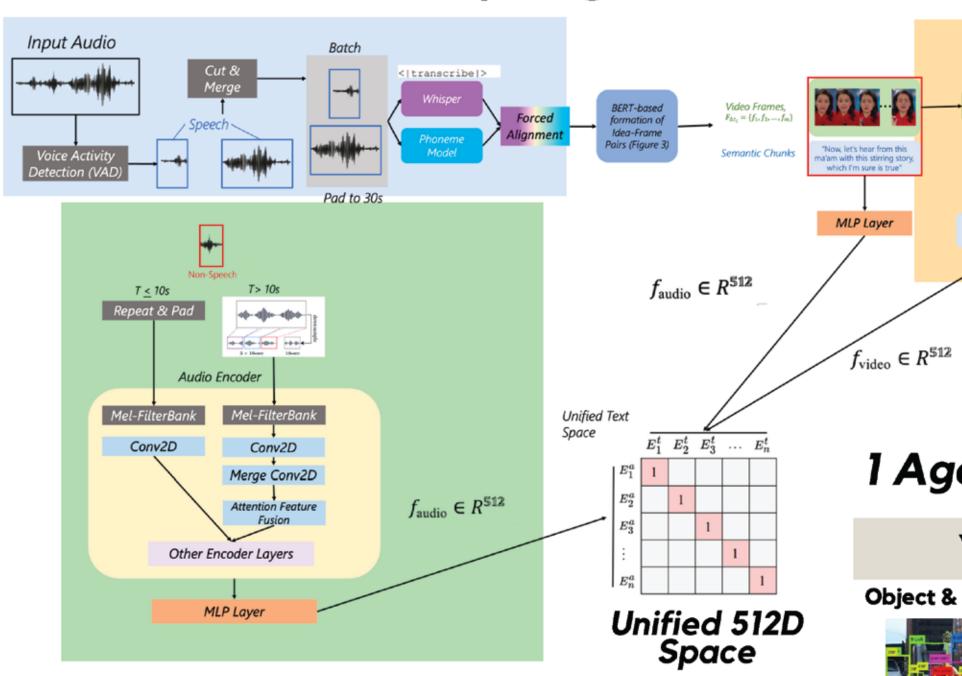
Speaker 2: "I was at work and a customer called me, sir, and I said, um, sir, actually it's a ma'am.

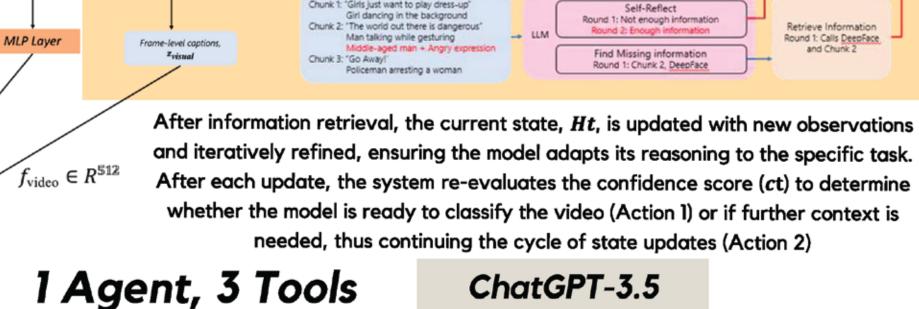
And then the guy who said it actually got arrested."

Speaker 1: "And then he was sentenced to death. Yeah, I'm sure."

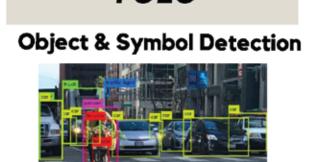
Unified Modality Alignment

Global-Local Reasoning Loop

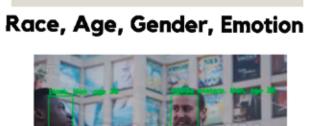




YOLO **LLaVAOneVision**



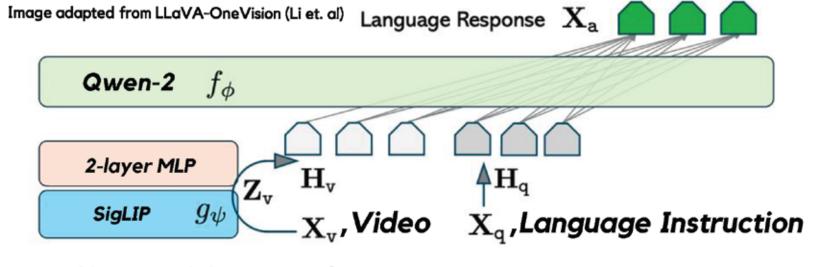
Action Recognition Accurate Descriptions • Weak Reasoning



videoDeepFace

Yes the video is hateful. In Chunk 2, the man makes

Video

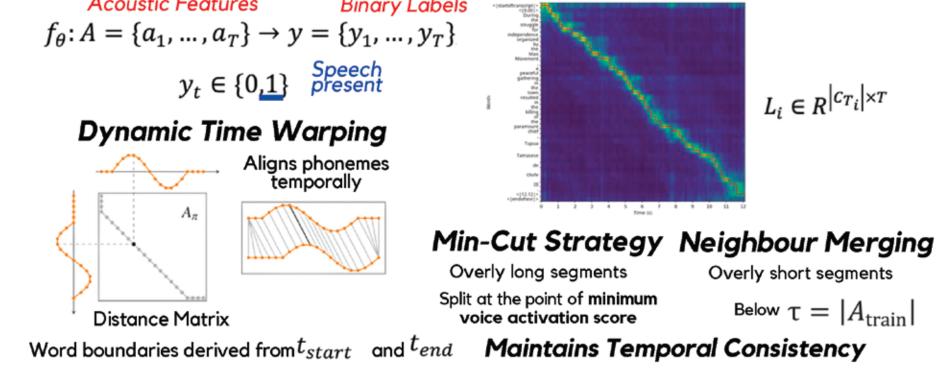


SigLIP Vision Encoder Pepper the aussie pup T₁ T₂ T₃ ... T_N $I_1 \cdot T_1 = I_1 \cdot T_2 = I_1 \cdot T_3 = \dots = I_1 \cdot T_N$

I_N I_N·T₁ I_N·T₂ I_N·T₃ ...

Speech Audio

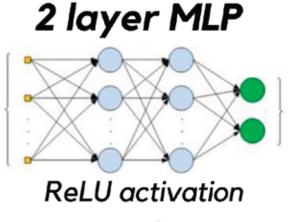
Voice Activity Detection (VAD)



MAESTRO-CLAP (Non-Speech Audio)

2 Audio Encoders 7 downsampling and 7 upsampling blocks • Penultimate layer : $Z_{\mathrm{PANN}} \in R^{2048}$ SOTA Transformer model featuring 4

HTSAT groups of Swin Transformer blocks, • Output: $\mathbf{Z}_{\text{HTSAT}} \in R^{768}$



Contrastive Loss

 $-\frac{1}{N} \sum_{i=1}^{N} log \frac{exp(f_{\text{audio}}^{i} \cdot f_{\text{text}}^{i} / \tau)}{\sum_{j=1}^{N} exp(f_{\text{audio}}^{i} \cdot f_{\text{text}}^{i} / \tau)}$

 $MLP_{non-speech}(z_{non-speech}), f_{non-speech} \in R^{512}$

Projected into unified space

Output, $f_{\text{output}} \in R^{512}$

Forced Phoneme Alignment

RESULTS

			М	ultiHat	eClip	o (Hateful Videos)
		Binary				Video
Model	Acc	M-F1	F1(O)	R(O)	P(O)	
mBERT	0.57	0.57	0.52	0.42	0.68	
GPT-4	0.81	0.79	0.73	0.69	0.78	
Qwen	0.72	0.71	0.65	0.57	0.75	2000 AND
MFCC	0.54	0.50	0.36	0.33	0.40	Audio
Wav2Vec	0.53	0.48	0.64	0.50	0.90	Addio
ViViT	0.73	0.73	0.68	0.57	0.86	Male Speaker: "Pay close attention. Notice a trend here?"
Vit	0.63	0.58	0.44	0.46	0.45	Female Speaker: "You're just mad because trans girls don't want to put trash like you in our designer bag. And we're
VLM	0.70	0.64	0.48	0.59	0.41	probably, most likely a better candidate of, uh, dating than like, you know, your average bish."
GPT-4V	0.81	0.79	0.73	0.72	0.73	Male Speaker: "Notice we never see trans men, biological women who identify as men, doing shenanigans like this? I
Qwen-VL	0.62	0.61	0.56	0.46	0.72	guess that male misogyny transcends even after you now identify as a female."
T1 ⊙ A1 ⊙ V1	0.75	0.74	0.67	0.61	0.77	
MAESTRO (Ours)	0.96	0.95	0.93	0.87	1.0	MAESTRO (Ours) Video LLaMA 7B LLaVA-OneVision (without MAESTRO)
Achieves SOTA by ≥ 15%				5%		HATEFUL. In Segment 7, the <u>male speaker</u> targets non-trans girls and discriminates against them by calling them "trash." NOT HATEFUL. It shows various objects and people in different situations. The video shows a variety of people and objects in different situations, such as a
Correc	Correct Reasoning Localisation Ability			In Segment 12, the <u>female speaker</u> targets woman standing on the street, a man standing in a store. In Segment 12, the <u>female speaker</u> targets woman standing on the street, a man standing in a store. In Segment 12, the <u>female speaker</u> targets woman standing on the street, a man standing in a store.		
Accurat	e Desci	ription	Corre	ct Label		Both instances meet the criteria for the video to be classified as 'hateful'. Hallucinations Incorrect Reasoning

Qwen-2

offers various model size and exhibits

strong language capabilities to date among

publicly available checkpoints

Industry Benchmarks (General VQA)

Model	Modality	MSRVTT-QA	MSVD-QA	ActivityNet-QA	
QueST	V	34.6	34.6	-	
ClipBERT	V	37.4	1-	-	
JustAsk	V	41.5	46.3	38.9	
GIT	V	42.7	55.1	-	
MERLOT	V	43.1	16.	41.4	
Singularity	V	43.5	14	43.1	
Clover	V	43.9	51.9	12	
VideoChat	V	45.0	56.3	26.5	
Video-ChatGPT	V	49.3	64.9	35.2	
VALOR	V,A	46.7	56.4	44.8	
FrozenBiLM	V	47.0	54.8	43.2	
Valley	V	45.7	65.4	42.9	
Video-LLaMA	V,A	29.6	51.6	12.4	
PandaGPT	V,A	25.5	42.1	14.5	
LLaVA-OneVision-7B	V	49.8	51.7	56.6	
MacawLLM	V,A	25.5	42.1	14.5	
MAESTRO (Ours)	V,A	82.0	86.9	87.2	

Achieves SOTA by 32.2%, 21.5%, 30.6%
Advances general multimodal understanding of VLMs

MAESTRO-CLAP

Multiclass (Zero-shot)								
Model	Acc	M-F1	F1(O)	R(O)	P(O)			
ZAC	0.21	0.23	0.20	0.26	0.33			
LAION-CLAP	0.32	0.28	0.30	0.34	0.31			
MAESTRO-CLAP (Ours)	0.82	0.89	0.84	0.92	0.90			

