



# **SVLTA:** Benchmarking Vision-Language Temporal Alignment via Synthetic Video Situation

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Project: http://svlta.csail.mit.edu

# **Problem Formulation**

Vision



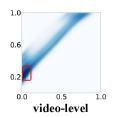
Temporal Alignment Timestamp [0.4s, 6.2s] Timestamp [22.5s, 30.8s]

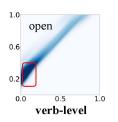
Language

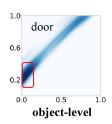
a person opens a door.

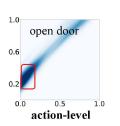
person sit in a chair at a desk.

# **Limitations of Existing Benchmarks**









#### Limitations

- Biased temporal distributions
- Imprecise annotations
- Insufficient compositionally

## Reasons

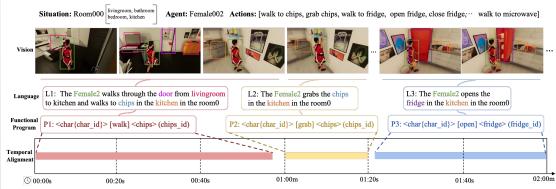
- Inherent properties of natural videos
- Human Annotation Challenges

TJSD: a diagnostic tool to analyze and quantify video temporal imbalance.

## Question

Can we design a temporally fair benchmark to evaluate MLLMs?

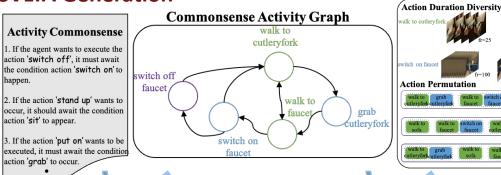
# **SVLTA Overview**



SVLTA includes synthetic videos, language, and high-quality temporal alignment.

		<b>Dataset Characteristics</b>						
Benchmark	# Videos / # Annotations	# Actions	Avg. Video / Moment Duration (s)	Scalable	Controllable	Synthetic	Compositional	Unbiased
TACoS	0.1K / 18.8K	60	287.1 / 27.9	X	Х	Х	X	×
ActivityNet Captions	14.9K / 54.9K	N/A	117.6 / 37.1	×	×	×	X	X
Charades-STA	6.7K / 16.1K	157	30.0 / 8.1	×	X	×	X	X
DiDeMo	10.5K / 40.5K	N/A	30.0 / 6.5	X	X	X	X	X
TVR	21.8K / 109K	N/A	76.1 / 9.1	×	X	×	X	X
MAD	0.7K / 384.6K	N/A	6646.2 / 4.1	~	×	×	X	X
Ego4D	1.6K / 19.2K	N/A	495.3 / 11.2	×	X	×	X	X
Ego4D Goal-Step	0.8K / 48K	N/A	1560.0 / 32.5	×	×	×	X	X
E.T.Bench	7K / 7.3K	N/A	129.0 / 11.0	X	×	×	×	×
SVLTA (ours)	25.3K / 77.1K	96	134.1 / 24.3	V	~	V	✓	~

## **SVLTA Generation**



# **Holistic Temporal Alignment Evaluations**





R@1

Temporal Bias Comparison

#### **Temporal Question Answering**

Method	# Frames	es Size Visual Encoder		LLM	IoU=0.1	IoU=0.3	IoU=0.5	IoU=0.7	mIoU
Gener	al Open-source	d Models:	All models use their de	fault setting, Ex	cept LLaVA-V	Video, due to	the GPU mem	ory limits.	
LLaVA-Video	16	7B	SIGLIP-SO400M	Qwen2	2.52	0.89	0.40	0.27	0.84
Videochat2	16	7B	UMT-L/16	Vicuna-0	2.93	0.87	0.32	0.13	0.87
Video-LLaVA	8	7B	LanguageBind-ViT-L/14	Vicuna-1.5	8.22	3.19	0.96	0.23	2.59
Video-ChatGPT	100	7B	CLIP-ViT-L/14	Vicuna-1.1	10.68	3.17	0.90	0.21	2.94
Video-LLaMA2	16	7B	CLIP-ViT-L/14	Mistral-7B	35.48	16.02	6.64	2.28	12.33
		Time-awai	re Open-sourced Models	s: All models uti	lize their defa	ult configurati	ion.		
E.T.Chat	1FPS	3.8B	EVA-ViT-G/14	Phi-3-Mini	17.86	8.07	3.48	1.36	6.29
TimeChat	96	7B	EVA-ViT-G/14	Llama-2	23.29	13.58	6.96	3.25	9.61
VTimeLLM	100	7B	CLIP-ViT-L/14	Vicuna-1.5	29.97	13.29	5.26	1.71	10.29
		Cl	ose-sourced Models: Ev	aluated on a sub	set with 2000	samples.			
GPT-4o-mini	32		_	_	24.79	6.49	1.57	0.42	6.70
Gemini 1.5 Pro	1FPS	_	_	_	32.30	17.45	7.45	3.15	12.48
GPT-40	32	_	<u> </u>	<u>-</u>	49.54	27.38	11.69	5.62	18.90

#### Distributional Shift Sensitiveness

		R@1					
Method	Test set	IoU=0.3	IoU=0.5	IoU=0.7	IoU=0.9	mIoU	$\mathbf{RC}\downarrow$
2D-TAN	high bias low bias	93.82 84.40 <sup>(-9.42)</sup>	87.08 76.10 <sup>(-10.98)</sup>	72.55 60.75 <sup>(-11.8)</sup>	35.06 22.75 <sup>(-12.31)</sup>	76.41 66.66 <sup>(-9.75)</sup>	10.85
VSLNet	high bias low bias	98.14 85.59 <sup>(-12.55)</sup>	97.03 83.22 <sup>(-13.81)</sup>	95.26	83.40 67.34 <sup>(-16.06)</sup>	92.63 79.16 <sup>(-13.47)</sup>	14.31
LGI	high bias low bias	97.02 89.70 <sup>(-7.32)</sup>	94.26 82.98 <sup>(-11.28)</sup>	87.38 68.74 <sup>(-18.64)</sup>	56.36 31.49 <sup>(-24.87)</sup>	85.25 72.67 <sup>(-12.58)</sup>	14.94
QD-DETR	high bias low bias	98.96 95.59 <sup>(-3.37)</sup>	98.35 93.93 <sup>(-4.42)</sup>	96.46 90.17 <sup>(-6.29)</sup>	82.61 72.43 <sup>(-10.18)</sup>	93.05 87.72 <sup>(-5.33)</sup>	5.92
DCM	high bias low bias	92.89 79.55 <sup>(-13.34)</sup>	85.72 68.11 <sup>(-17.61)</sup>	69.75 46.15 <sup>(-23.6)</sup>	32.29 13.49 <sup>(-18.8)</sup>	74.85 58.88 <sup>(-15.97)</sup>	17.86
Shuffling	high bias low bias	93.78 93.26 <sup>(-0.52)</sup>	89.43 88.61 <sup>(-0.82)</sup>	82.25 80.23 <sup>(-2.02)</sup>	49.63	81.62 80.36 <sup>(-1.26)</sup>	1.04
	2D-TAN VSLNet LGI QD-DETR DCM	2D-TAN high bias low bias high bias low bias high bias low bias Shuffling high bias low bias	2D-TAN high bias 93.82 low bias 84.40(-9.42) high bias 98.14 VSLNet low bias 95.59(-12.55) low bias 97.02 low bias 89.70(-7.32) low bias 95.59(-3.37) DCM high bias low bias 95.59(-3.37) DCM high bias 92.89 low bias 79.55(-13.34) low bias 93.78	Method         Test set         IoU=0.3         IoU=0.5           2D-TAN         high bias low bias low bias high bias low b	Method         Test set         IoU=0.3         IoU=0.5         IoU=0.7           2D-TAN         high bias low bias low bias low bias low bias low bias low bias bias low	Method         Test set         IoU=0.3         IoU=0.5         IoU=0.7         IoU=0.9           2D-TAN         high bias low bias low bias low bias high bias of the pias with pias low bias low b	2D-TAN

### Temporal Alignment Adaptation

	R@1				2		Entity			
ethod	IoU=0.3	IoU=0.5	IoU=0.7	mIoU	Benchmark	Process	Verb	Object	Compo	
	100 0.0	100 0.0	100 0.7		TACoS	0.243	0.786	0.787	0.8	
-TAN	15.81	5.03	1.94	11.8	ActivityNet Captions	0.107	0.764	0.827	0.9	
LNet	28.33	8.52	3.87	19.66	Charades-STA	0.287	0.739	0.877	0.8	
.GI	33.96	12.52	3.30	22.24	TVR	0.229	0.779	0.84	0.9	
State of the same					MAD	0.628	0.842	0.860	0.9	

## References

Controllable Activity Manuscript

[1] Puig, Xavier, et al. "Virtualhome: Simulating household activities via programs." CVPR. 2018.
 [2] Otani, Mayu, et al. "Uncovering Hidden Challenges in Query-Based Video Moment Retrieval."
 BMVC. 2020.