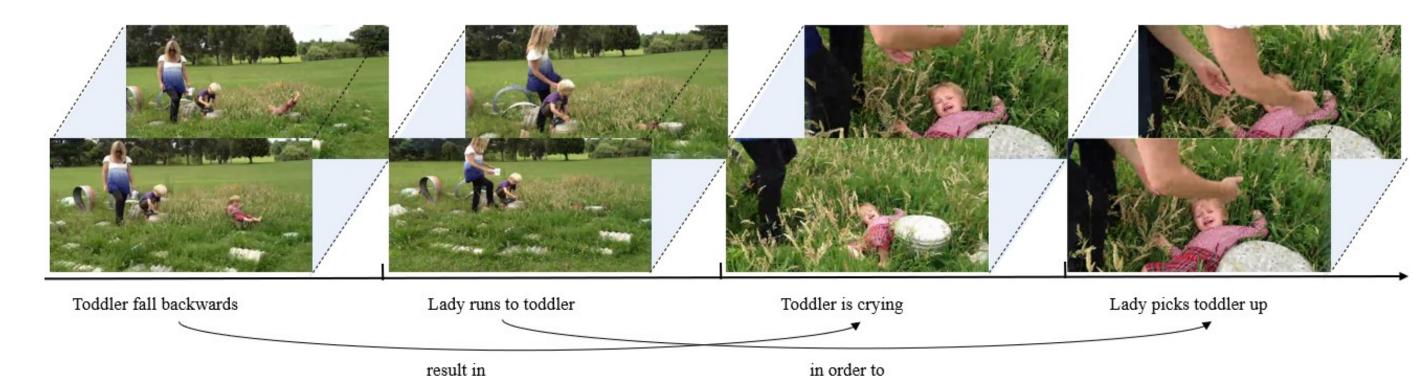


NExT-QA: Next Phase of Question-Answering to Explaining Temporal Actions Junbin Xiao, Xindi Shang, Angela Yao, Tat-Seng Chua Department of Computer Science, National University of Singapore



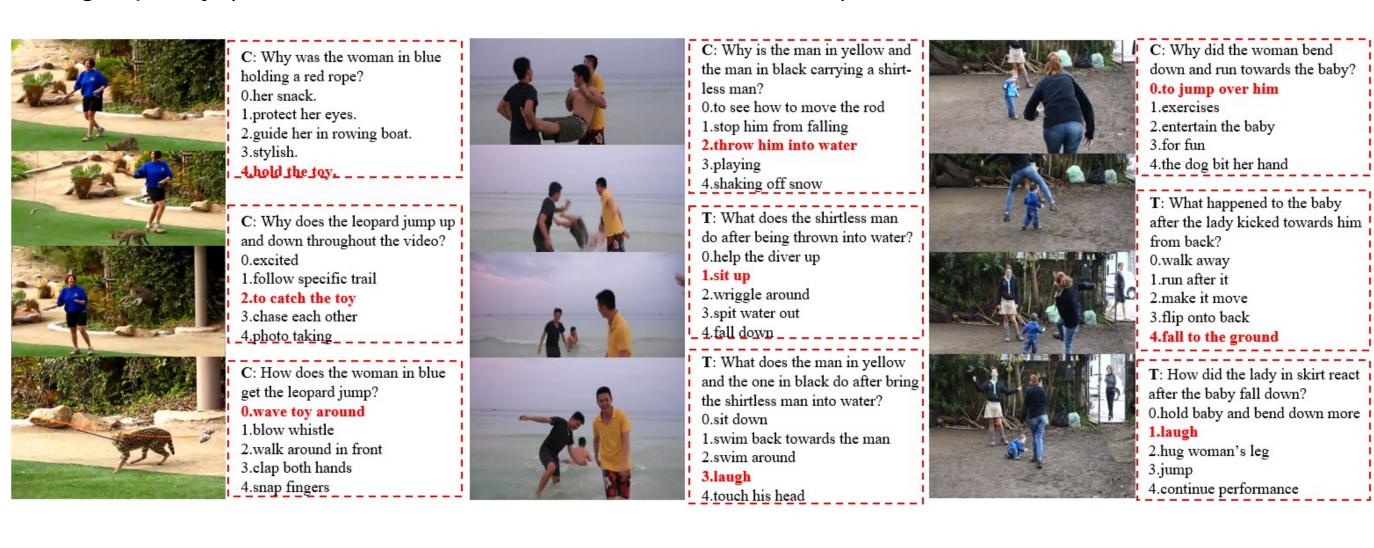
> Introduction

- Actions in videos are often not independent but rather related with causal and temporal relationships.
- Recognizing/describing the independent object/action in a video is now attainable with advanced neural network models.
- Understanding the causal and temporal interaction lies at the core of human intelligence,
 but remains a great challenge for AI and is also much less explored.
- **NExT-QA** is proposed to benchmark such a problem; it hosts causal and temporal action reasoning in VideoQA and is rich in object interaction in daily activities.



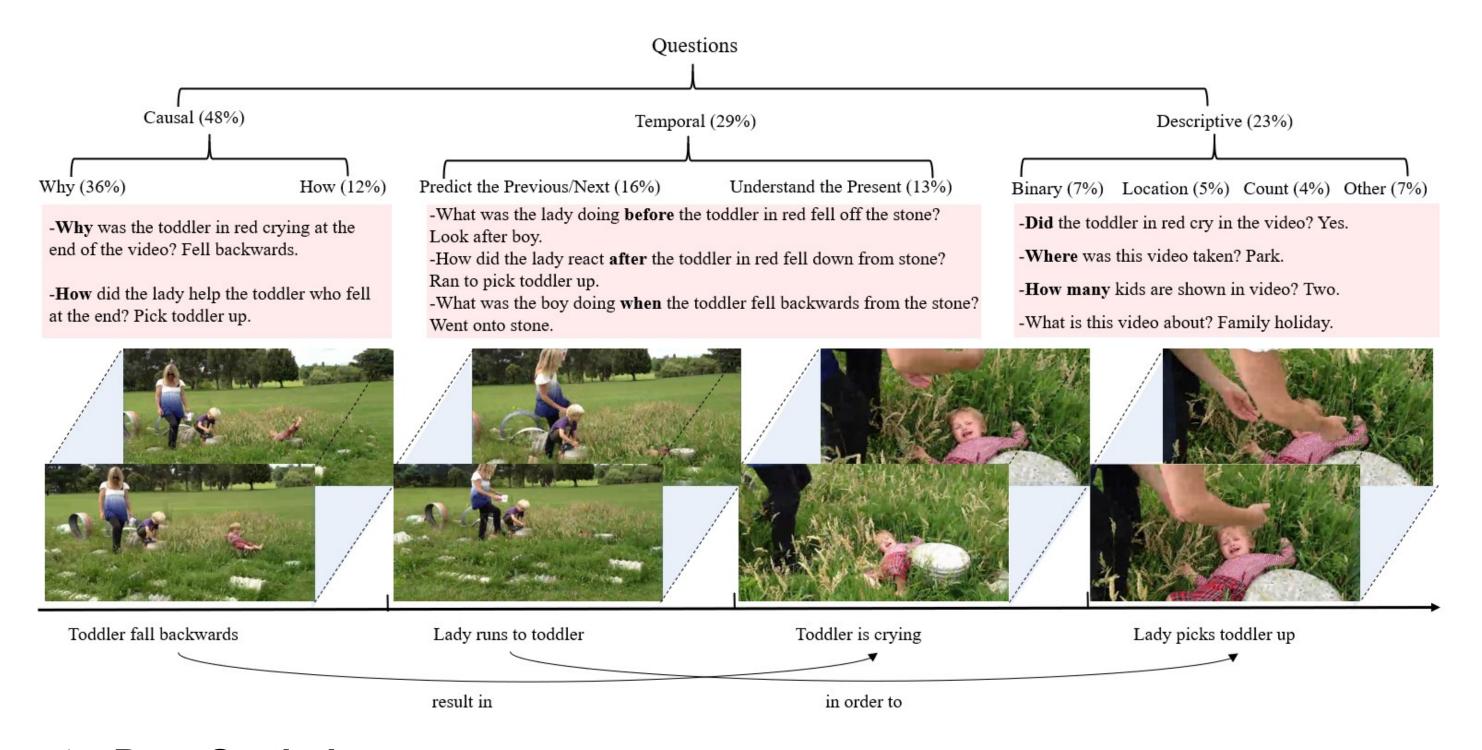
Attractive Property

- The 1st VideoQA benchmark challenges comprehending the causal & temporal action interaction, advancing video understanding from recognition to explanation.
- Comprehensive baselines and analysis for both multi-choice and open-ended QA, along with detailed question types to help analyze the VQA models.
- High quality (with multi-turn manual annotation & check).



Dataset Overview

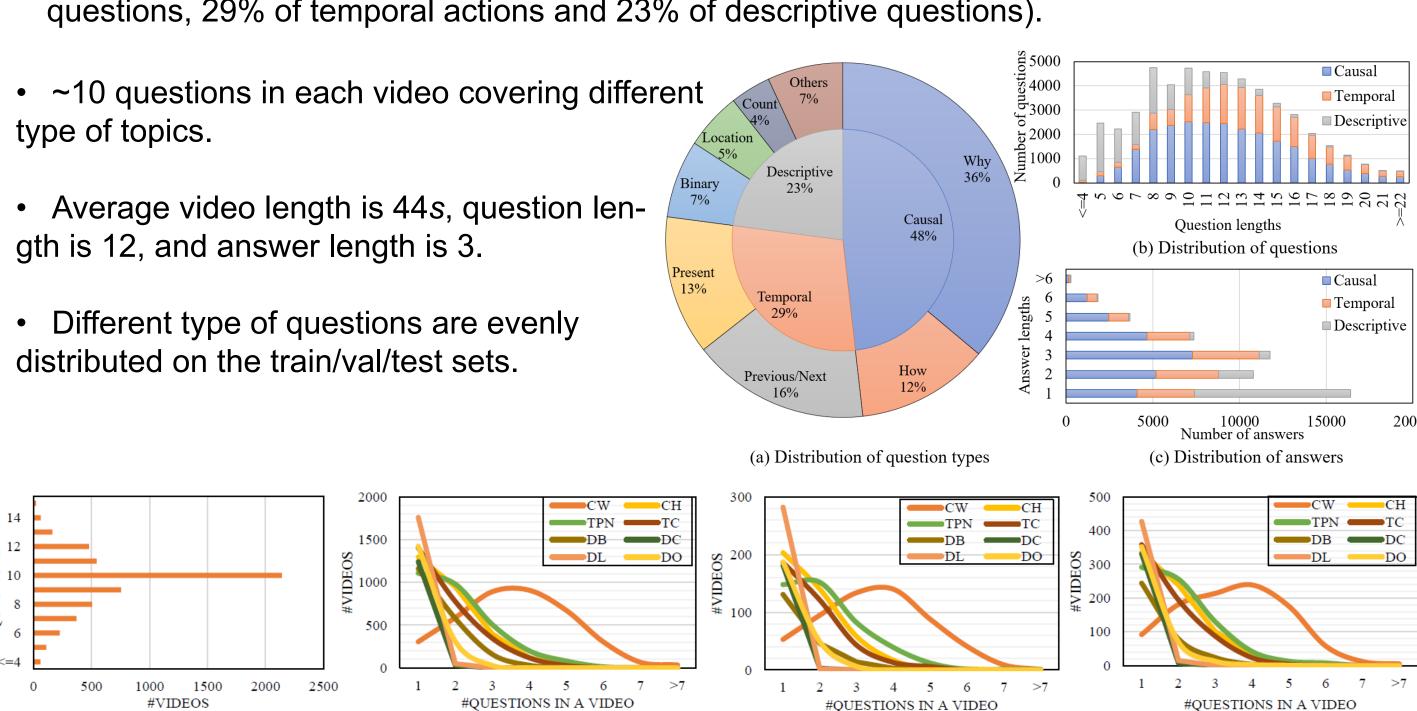
- Causal questions explain why something happen or how to bring a visual effect.
- Temporal questions challenge understanding the order of the actions.
- Descriptive Questions aim at a scene-level recognition of the visual facts in videos.



Data Statistic

(a) Distribution of questions per video.

• 5440 Videos (train/val/test: 3870/570/1000) with 52K question-answer pairs (48% of causal questions, 29% of temporal actions and 23% of descriptive questions).



(c) Val

(d) Test

(b) Train

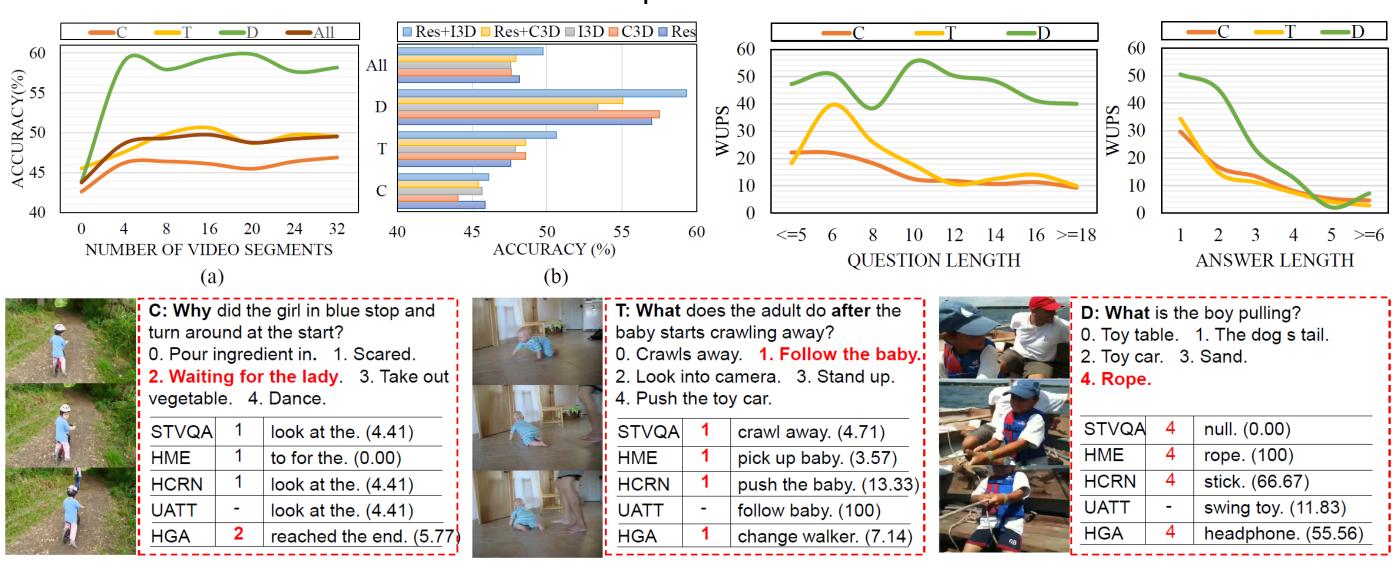
> Experiment

Methods	Text Rep.	Acc_C			Acc_T			Acc_D			Acc	
		Why	How	All	Prev&Next	Present	All	Count	Location	Other	All	7100
EVQA [2]	GloVe	28.38	29.58	28.69	29.82	33.33	31.27	43.50	43.39	38.36	41.44	31.51
PSAC [28]	GloVe	35.81	29.58	34.18	28.56	35.75	31.51	39.55	67.90	35.41	48.65	35.57
PSAC+ [28]	GloVe	35.03	29.87	33.68	30.77	35.44	32.69	38.42	71.53	38.03	50.84	36.03
CoMem [12]	GloVe	36.12	32.21	35.10	34.04	41.93	37.28	39.55	67.12	40.66	50.45	38.19
STVQA [20]	GloVe	37.58	32.50	36.25	33.09	40.87	36.29	45.76	71.53	44.92	55.21	39.21
HGA [21]	GloVe	36.38	33.82	35.71	35.83	42.08	38.40	46.33	70.51	46.56	55.60	39.67
HME [9]	GloVe	39.14	34.70	37.97	34.35	40.57	36.91	41.81	71.86	38.36	51.87	39.79
HCRN [25]	GloVe	39.86	36.90	39.09	37.30	43.89	40.01	42.37	62.03	40.66	49.16	40.95
EVQA [2]	BERT-FT	42.31	42.90	42.46	46.68	45.85	46.34	44.07	46.44	46.23	45.82	44.24
STVQA [20]	BERT-FT	45.37	43.05	44.76	47.52	51.73	49.26	43.50	65.42	53.77	55.86	47.94
CoMem [12]	BERT-FT	46.15	42.61	45.22	48.16	50.38	49.07	41.81	67.12	51.80	55.34	48.04
HCRN* [25]	BERT-FT	46.99	42.90	45.91	48.16	50.83	49.26	40.68	65.42	49.84	53.67	48.20
HME [9]	BERT-FT	46.52	45.24	46.18	47.52	49.17	48.20	45.20	73.56	51.15	58.30	48.72
HGA [21]	BERT-FT	46.99	44.22	46.26	49.53	52.49	50.74	44.07	72.54	55.41	59.33	49.74

Results of multi-choice QA on val set

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	Methods	$WUPS_C$	$WUPS_T$	$WUPS_D$	WUPS
-	Popular	9.73	8.95	28.39	13.40
	BlindQA	12.14	14.85	40.41	18.88
-	STVQA [19]	12.52	14.57	<u>45.64</u>	20.08
	HME [9]	12.83	14.76	45.13	20.18
	HCRN [25]	12.53	<u>15.37</u>	45.29	20.25
	UATT [56]	13.62	16.23	43.41	20.65
	HGA [21]	14.76	14.90	46.60	21.48

Results of open-ended QA on val set.



Discussion & Conclusion

- SOTA methods perform well on descriptive questions but are weak in causal and temporal action reasoning.
- Methods that are effective in multi-choice QA struggle in generating the answers in open-ended QA scenario.
- Future efforts can be made in understanding the rich object interactions and capturing the causal and temporal relationship.
- Data and code is available at: https://github.com/doc-doc/NExT-QA.