Active Scene Understanding for Video Labeling

Public Version

Team Moonshot:



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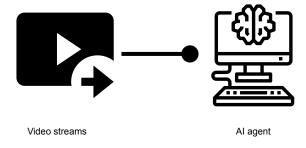
Co-supervisor:

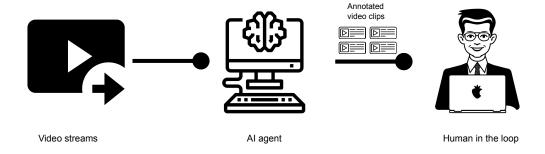


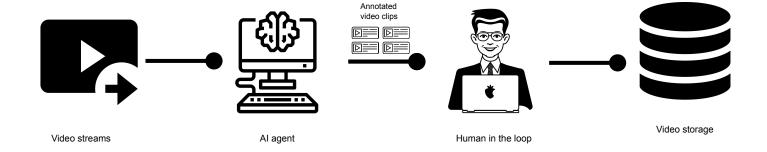


TUM Data Innovation Lab 24 February 2021









DESIGN Ni



AUDIO TRANSCRIPT

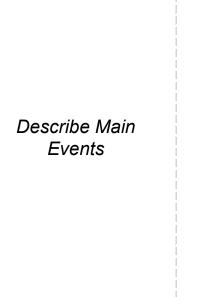






TIME





PLOT SUMMARY



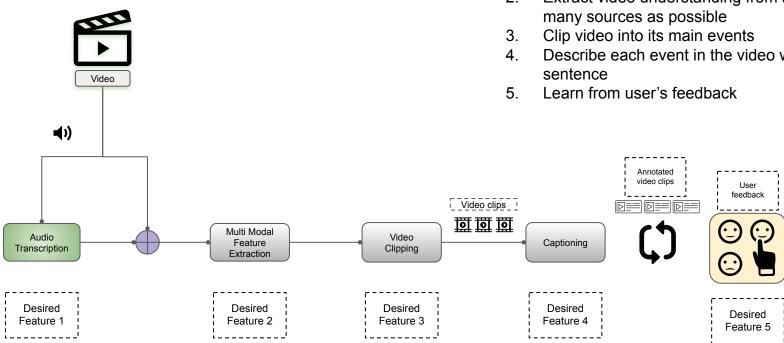




Scene 4#

Scene 3#

Task Overview

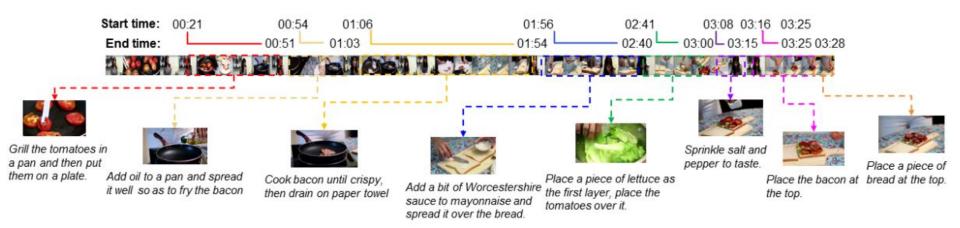


Why?

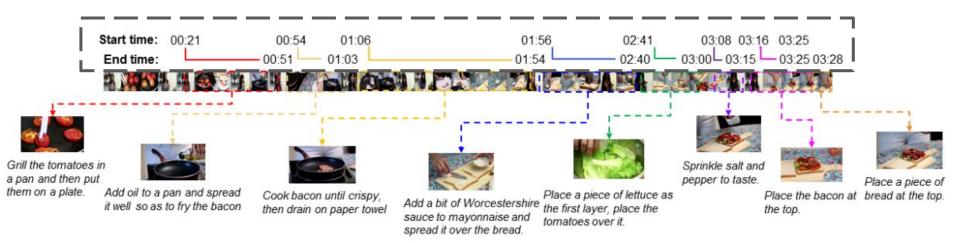
- Automatically show video conversations split by scene
- Extract video understanding from as
- Describe each event in the video with one

Desired Feature 1 - Audio Transcription

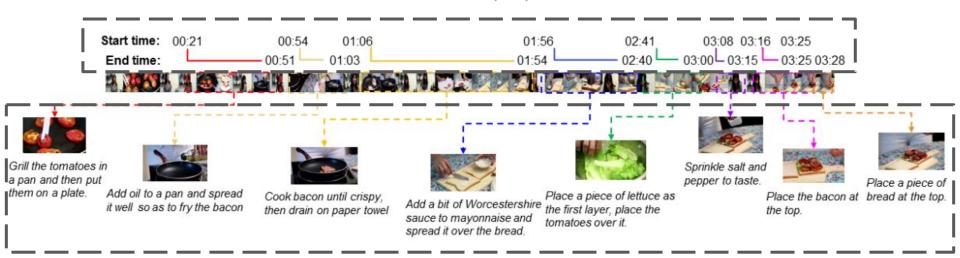
Requirements Accurate Speech-to-Text Alignment Offline Start (ms), end (ms), transcript 55000.56000.three were Language model: Acoustic model: 56000,57000, given to to the elves predict most performs 57000,58000,immortal likely emissions one-pass beam 58000,59000,wisest and transitions search decoding 59000,60000,and fairest of 60000,6100,all beings



Dense event proposals

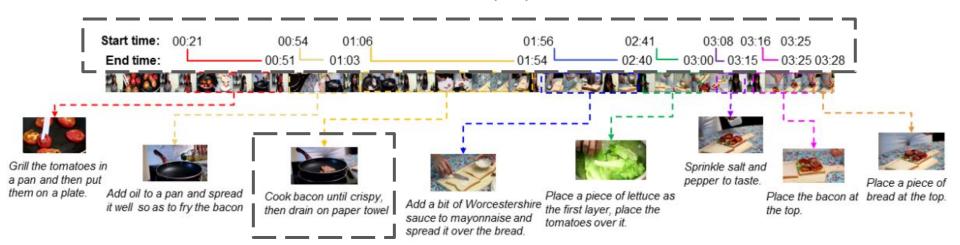


Dense event proposals



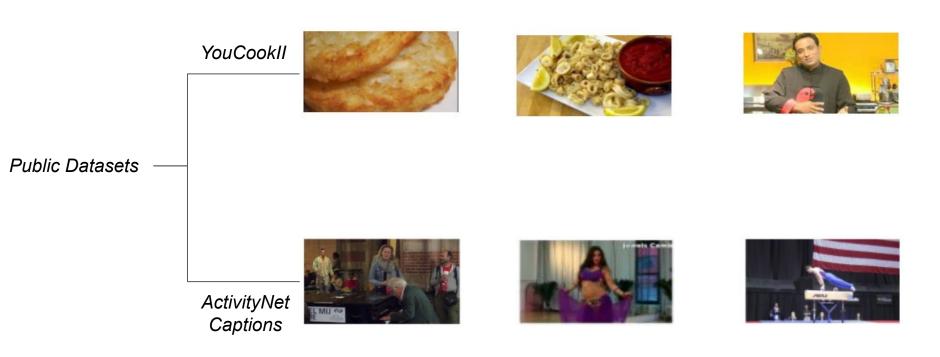
One-sentence captioning of each proposal

Dense event proposals



Each proposal represents one main event

Datasets



Datasets

Too application specific!

YouCookII Public Datasets ActivityNet Captions

Datasets

Too application specific!



Encompasses many domains

Datasets: ActivityNet Captions



20k videos



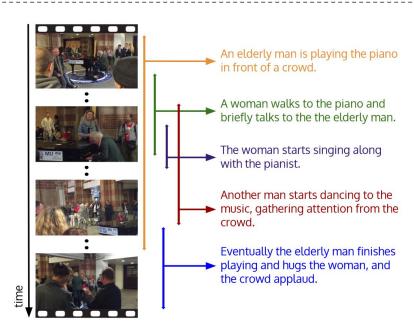
180s on average



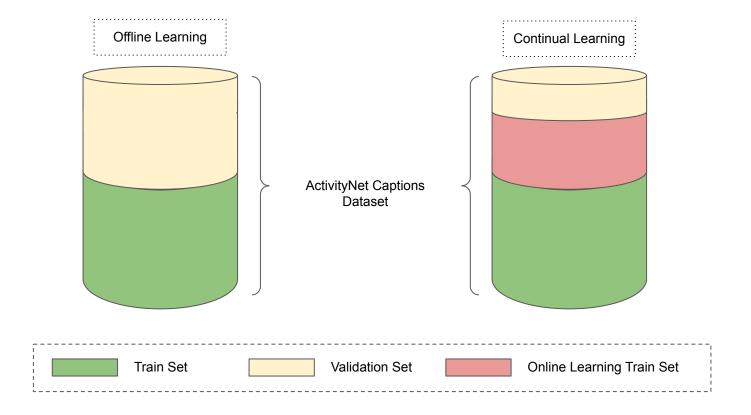
100k sentences



13.48 WPS



Results - Evaluation Procedure



Results - Proposal Generation & Captioning

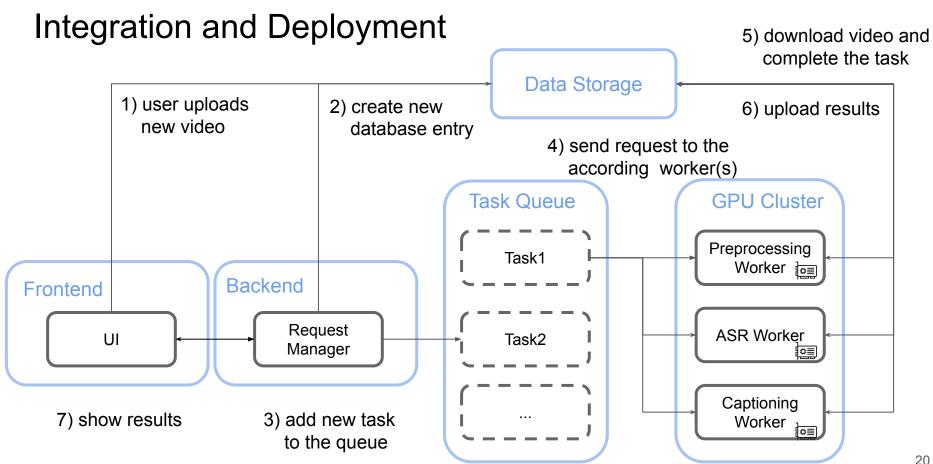
	Full Dataset Available	GT Proposals			Learned Proposals		
		B@3	B@4	METEOR	B@3	B@4	METEOR
Mun et al.	Yes	-	-	-	-	-	6.92
Krishna et al.	Yes	4.09	1.60	8.88	1.90	0.71	5.69
Li et al.	Yes	4.51	1.71	9.31	2.05	0.74	6.14
Zhou et al.	Yes	5.78	2.71	11.16	2.91	1.44	6.91
Wang et al.	Yes	-	-	10.89	2.27	1.13	6.10
	'						
Teng et al.	No	-	1.98	11.49	2.53	-	7.65
lashin et al.	No	4.52	1.46	11.07	1.85	1.01	7.46
Rahman et al.	No	3.04	1.99	7.23	3.84	0.90	4.93
ВМТ	No	4.63	1.99	10.90	3.84	1.88	8.44
MMT [Ours]	No	5.83	2.86	11.72	4.00	2.01	9.43

Multi-Modal Transformer (MMT) surpasses the state of the art in both GT and Learned Proposals on all metrics

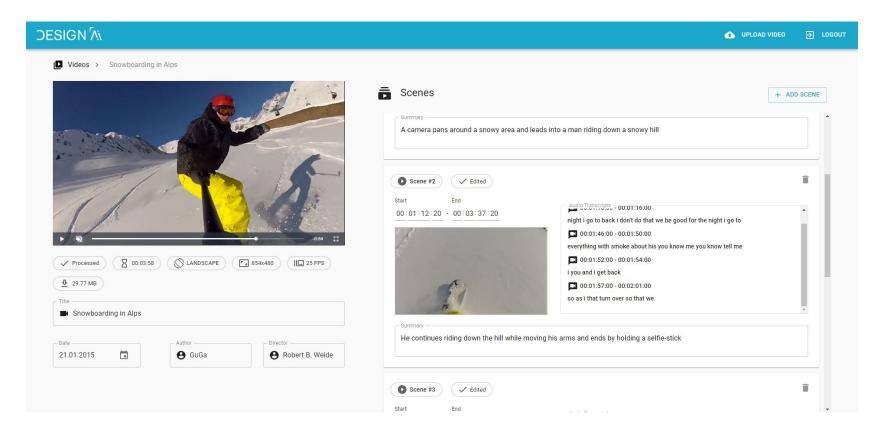
Results - Continual Learning

Method	B@3	B@4	METEOR
BMT(offline + Train Set)	4.9	2.3	10.51
BMT(offline + Train Set + 80% Val Set)	5.66	2.71	11.12
ALTC (ours)	5.7	2.8	11.23

- Bi-Modal Transformer (BMT) used as lower and upper bound offline baseline
- ALTC applied on BMT surpasses the offline upper bound



Demo - Application



Conclusion



Leveraged Multiple Modalities



Beat the State-of-the-Art in DVC



Adapted Continual Learning to DVC

Our Moonshot Goal

