

# Cross-Modal and Hierarchical Modeling of Video and Text # ECCV 2018

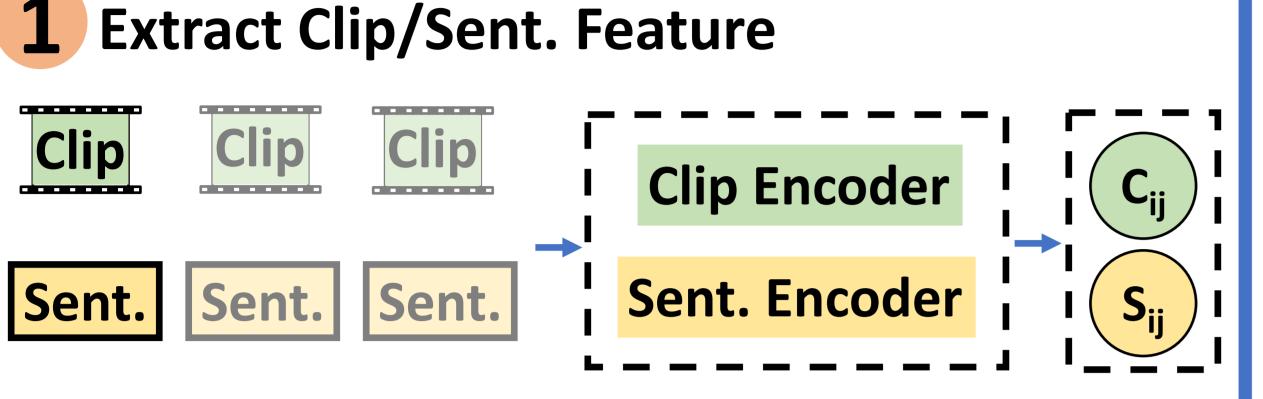
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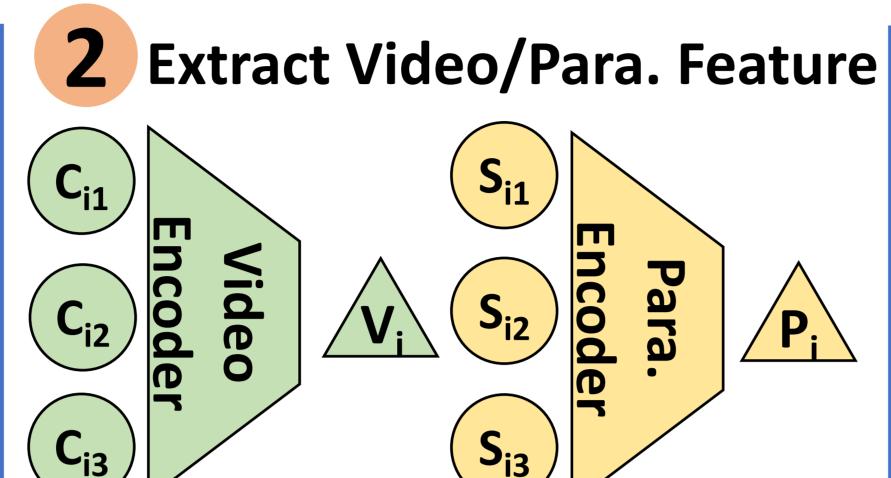
<sup>1</sup>University of Southern California, <sup>2</sup>Netflix

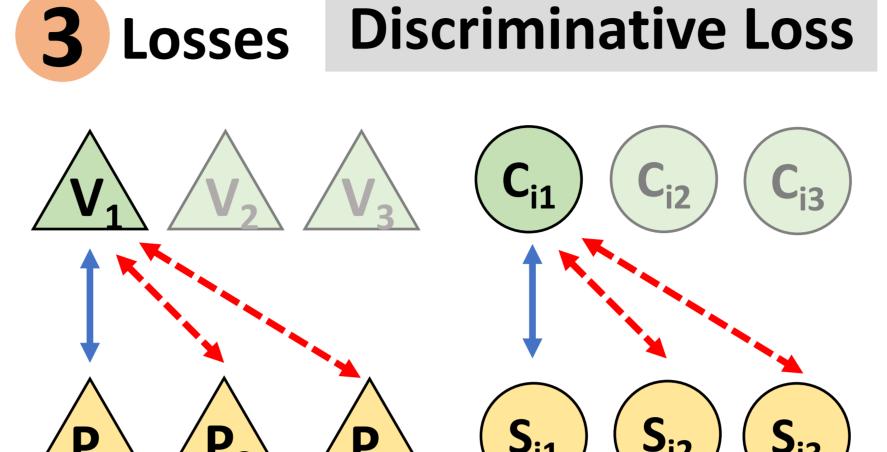


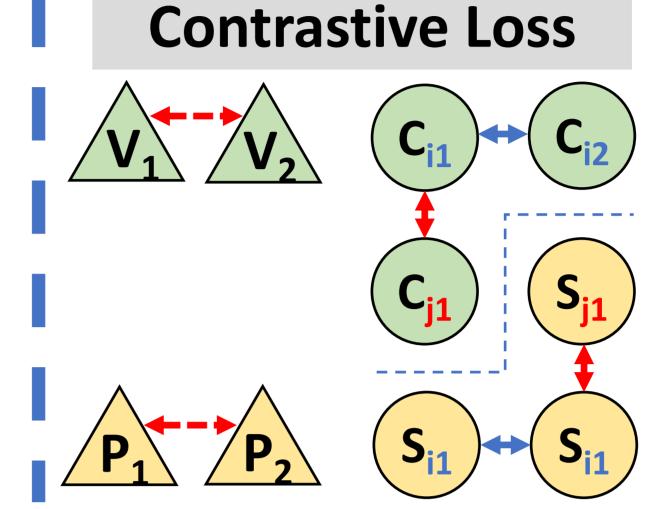


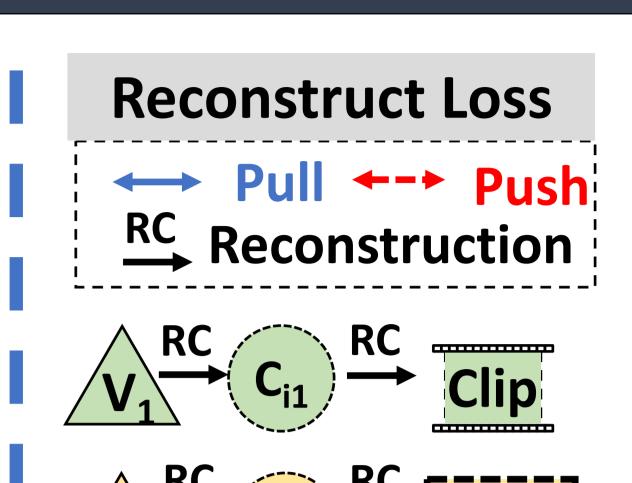
# Approach











## Highlights

- Propose to hierarchically model cross-modal sequential data.
- Preserve correspondence of complex structures across modalities through discriminative losses and contrastive losses.
- State-of-the-art performance on video and paragraph retrieval.
- Systematical study on several tasks involving video and language.

#### Goal

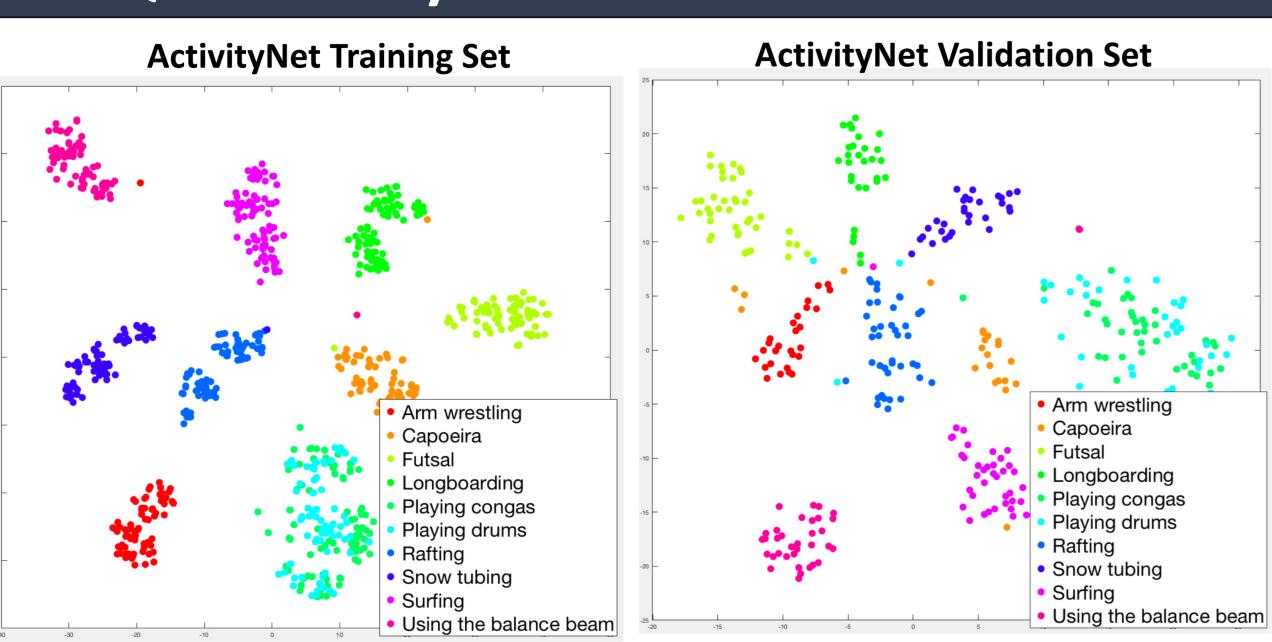
 Learn embeddings for hierarchical sequential data (video and) text) where they have correspondence across multiple modalities.

### **Tasks & Datasets**

- Tasks: Video/Text Retrieval, Video Captioning, Zero-shot Action Recognition
- Datasets: ActivityNet Dense Caption; ActivityNet V1.3; DiDeMo

## Qualitatively Results

T-SNE visualization of video embedding of **HSE** on ActivityNet V1.3



# **Experiments & Analysis**

Video and Text Retrieval: With Ground-truth clip proposal

Table 1. Performance on Activity Net Dense Caption

	Paragraph => Video			<b>V</b> ideo => <b>P</b> aragraph		
	R@1	R@5	R@50	R@1	R@5	R@50
C3D with Dimension Reduction						
DENSE	14.0	32.0	65.0	18.0	36.0	74.0
FSE	12.6	33.2	77.6	11.5	31.8	77.7
HSE	32.7	63.2	90.8	32.8	63.2	91.2
Inception-V3						
FSE	18.2	44.8	89.1	16.7	43.1	88.4
HSE	44.4	76.7	97.1	44.2	76.7	97.0

#### Table 2. Performance on DiDe Mo

	Paragraph => Video			Video => Paragraph		
	R@1	R@5	R@50	R@1	R@5	R@50
Inception-V3						
S2VT	11.9	33.6	76.5	13.2	33.6	76.5
FSE	13.9	36.0	78.9	13.1	33.9	78.0
HSE	29.7	60.3	92.4	30.1	59.2	92.1

Our approach **HSE** outperforms SotA by a large margin.

## Ablations: With heuristic clip proposal

Table3. Performance on ActivityNet Dense Caption w/o clip proposal

Proposal Method		<b>P</b> aragraph => Video		Video => Paragraph	
Inception-V3	#Seg.	R@1	R@5	R@1	R@5
FSE	-	18.2	44.8	16.7	43.1
HSE+GT	-	44.4	76.7	44.2	76.7
HSE + Uniform	3	20.0	48.6	18.2	47.9
HSE + Uniform	4	20.5	49.3	18.7	48.1
	·				

With a poor uniform proposal, **HSE** can already outperform **FSE** methods.

# Retrieval with incomplete video and paragraph Video to Paragraph Paragraph to Video Recall@5-HSE Recall@1-FSE Recall@5-HSE **Number of Segments/Sentences Observed**

## Video Captioning and Zero-shot Action Recognition:

Table 4. Results for video captioning on ActivityNet

		B@1	B@2	B@3	Meteor	CiDER
	DENSE	26.5	13.5	7.1	9.5	24.6
	DVC	19.6	9.9	4.6	10.3	25.2
•	FSE	17.9	8.2	3.6	8.7	32.1
	HSE	19.8	9.4	4.3	9.2	39.8

Table 5. Results for action recognition on ActivityNet

	Zero-shot	Transfer	Train Classifier		
	Top-1 Top-5		Top-1	Top-5	
FV-VAE	_	-	78.6	_	
TSN	-	-	88.1	-	
FSE	48.3	79.4	74.4	94.1	
HSE	51.4	83.8	75.3	94.3	

Check paper for more results and ablations studies!