

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

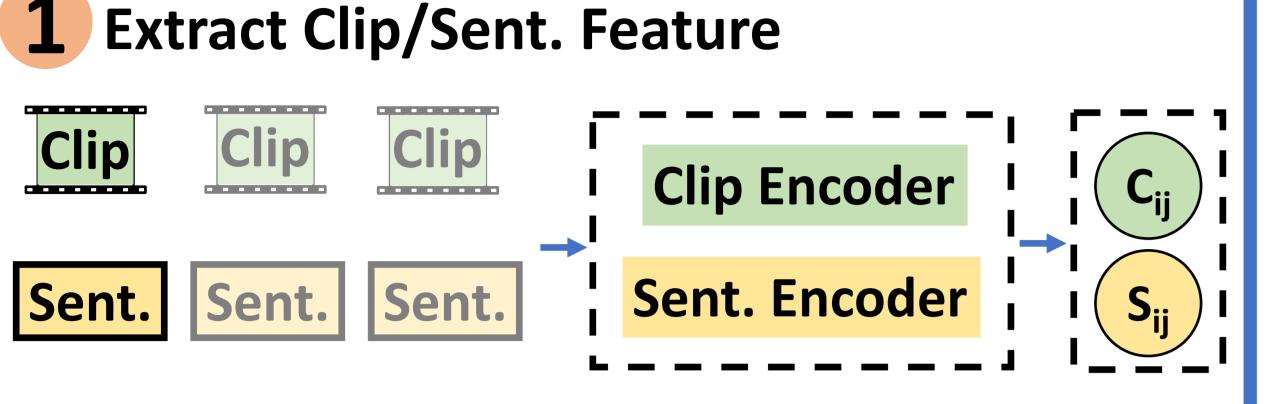
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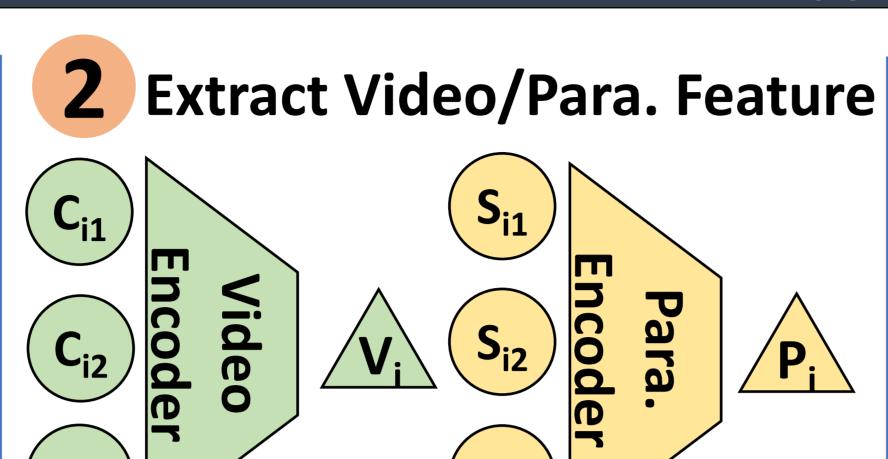
<sup>1</sup>University of Southern California, <sup>2</sup>Netflix

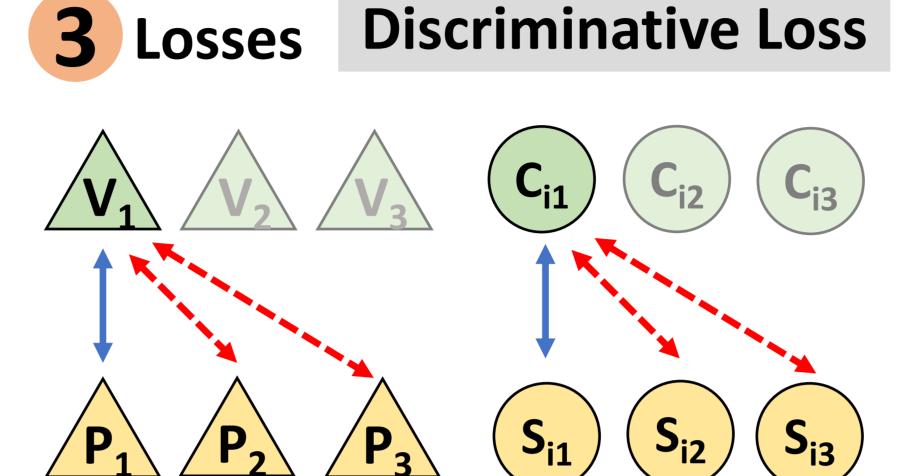


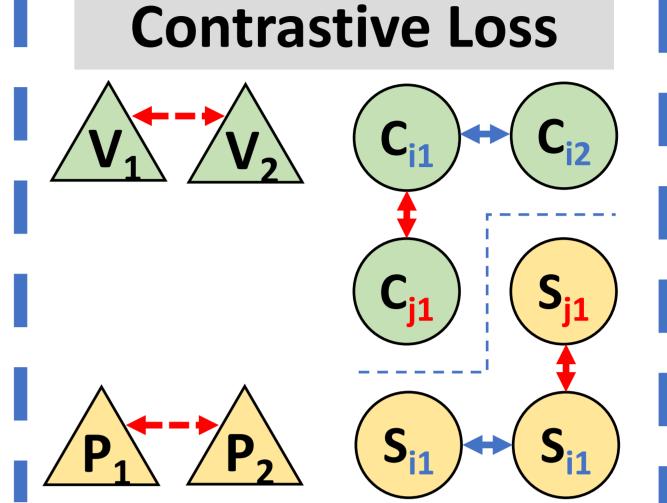


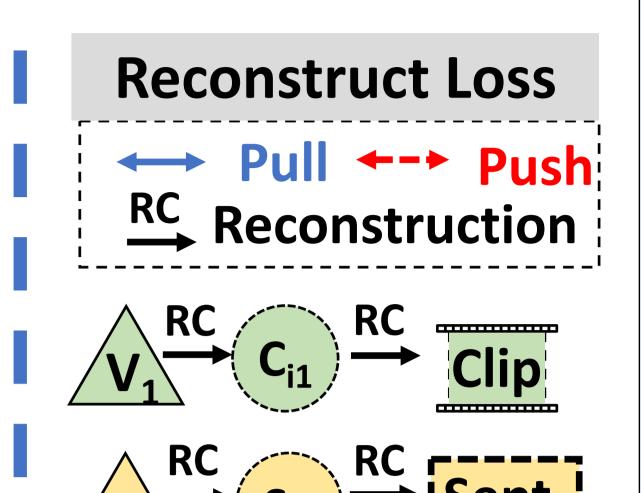
# Approach











### 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

Table1. Performance on ActivityNet Dense Caption

	Paragraph => Video			<b>V</b> ide	o => <b>P</b> ara{	graph		
	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		
HSE	44.4	76.7	97.1	44.2	76.7	97		

### Table2. Performance on DiDeMo

	Paragraph => Video			<b>V</b> ideo	> <b>P</b> ara			
	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		
IIJL	29.7	00.5	<b>32.4</b>	30.1	39.2	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 Paragraph to Video

### 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	1	-	78.6	-		
TSN	1	-	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!