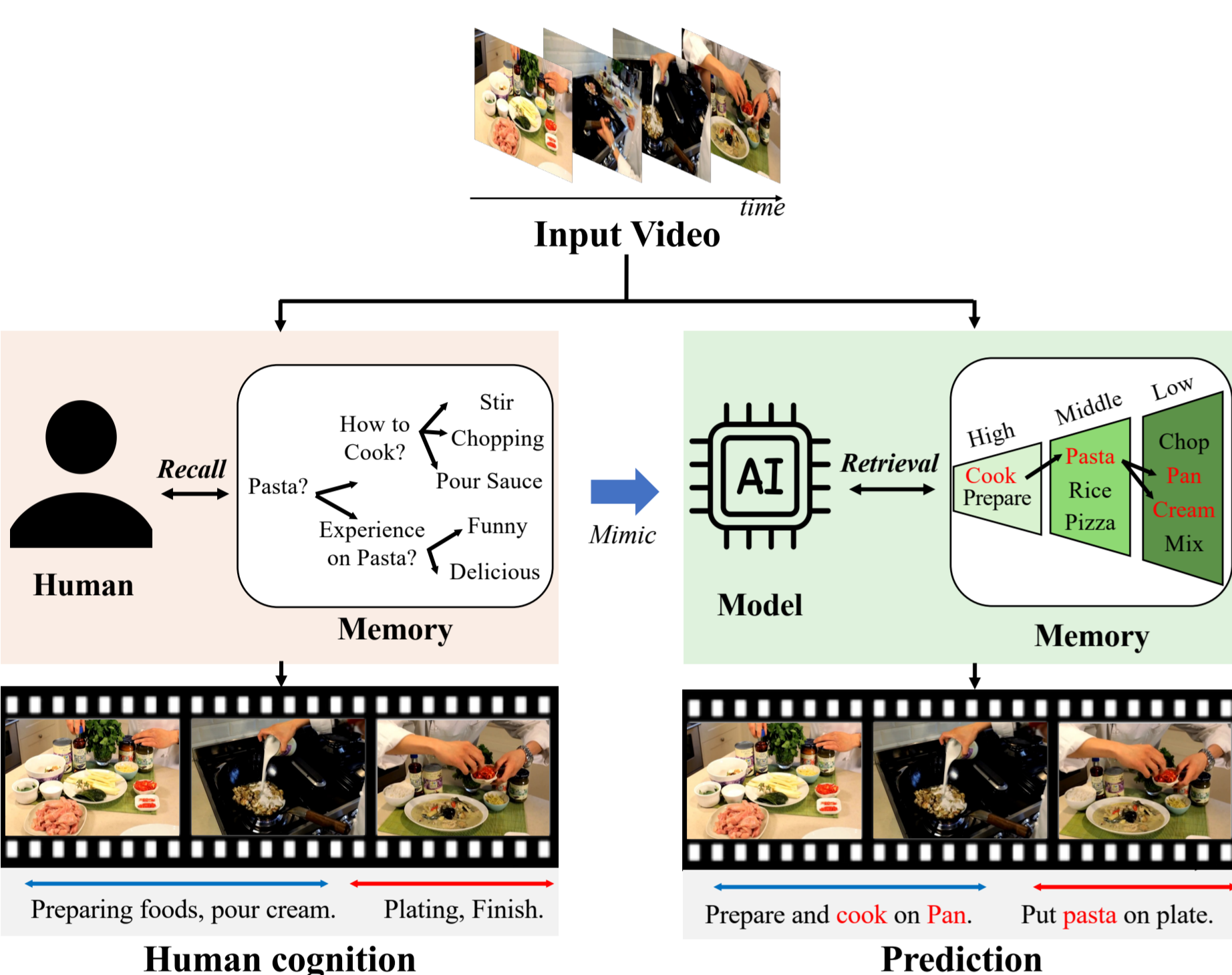


At One Glance

We address challenging Video Localization and Description tasks by proposing a novel framework based on

“ How human Recognizes, Remembers and Recalls in Memory Hierarchy. ”

- Concept of our proposed method



Our contributions:

1. We introduce the first hierarchical memory structure for DVC, inspired by human cognition and enabling cross-modal retrieval with compact representations.
2. We propose a top-down hierarchical memory retrieval strategy, starting with abstract information and progressively accessing detailed levels.
3. Extensive experiments on YouCook2 and ViTT show that our approach achieves state-of-the-art performance, validating its effectiveness.

- Effect of Memory Retrieval in YouCook2

Memory Construction	Hierarachy	CIDEr	METEOR	SODA _c	F1
No Memory	✗	66.29	12.41	9.87	31.08
All Training Captions	✗	67.90	12.49	10.38	32.31
Clustering	✓	67.15	12.97	10.17	32.30
Clustering+LLM(Ours)	✓	71.84	12.80	10.73	32.51

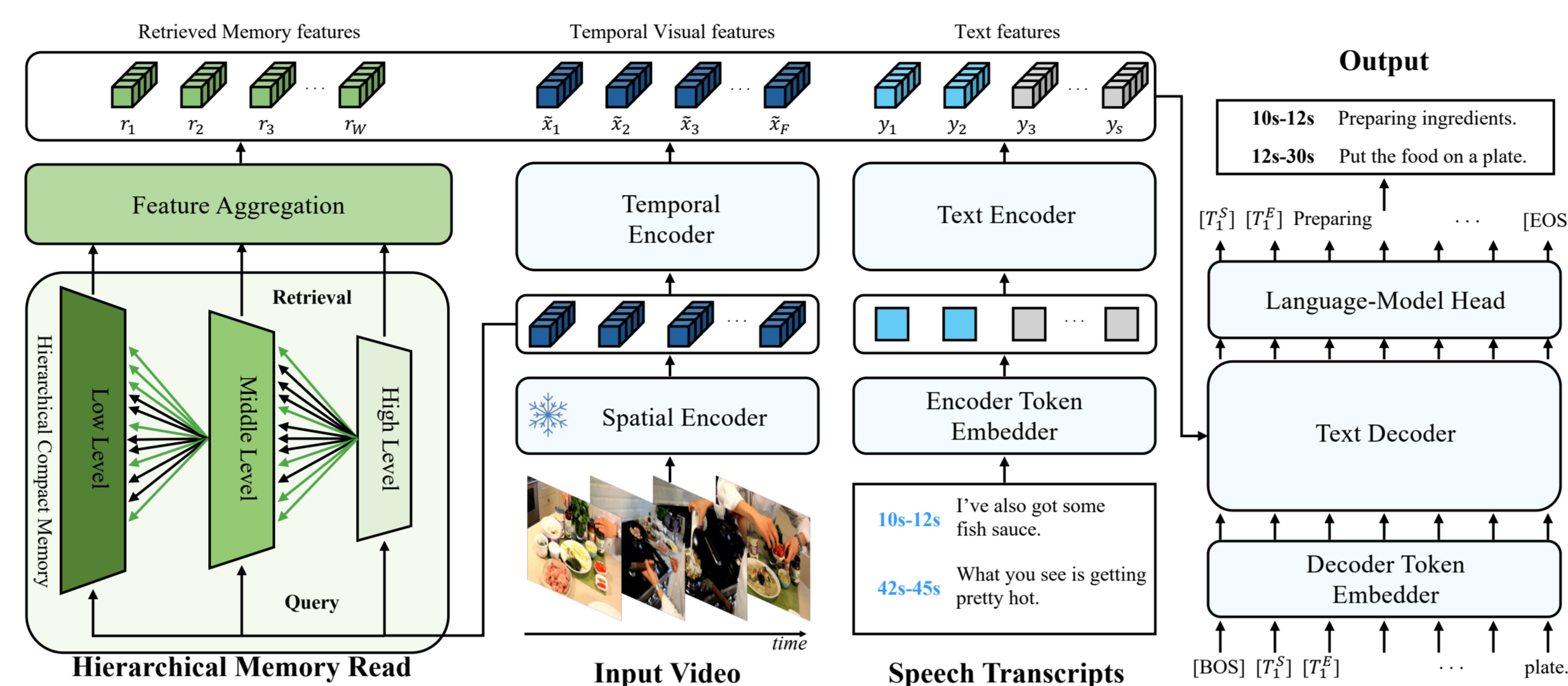
- ✓ Our observations indicate that retrieving relevant content from memory benefits model performance, especially under a hierarchical, compact memory configuration.

Hierarchical Compact Memory Modeling for Dense Video Captioning

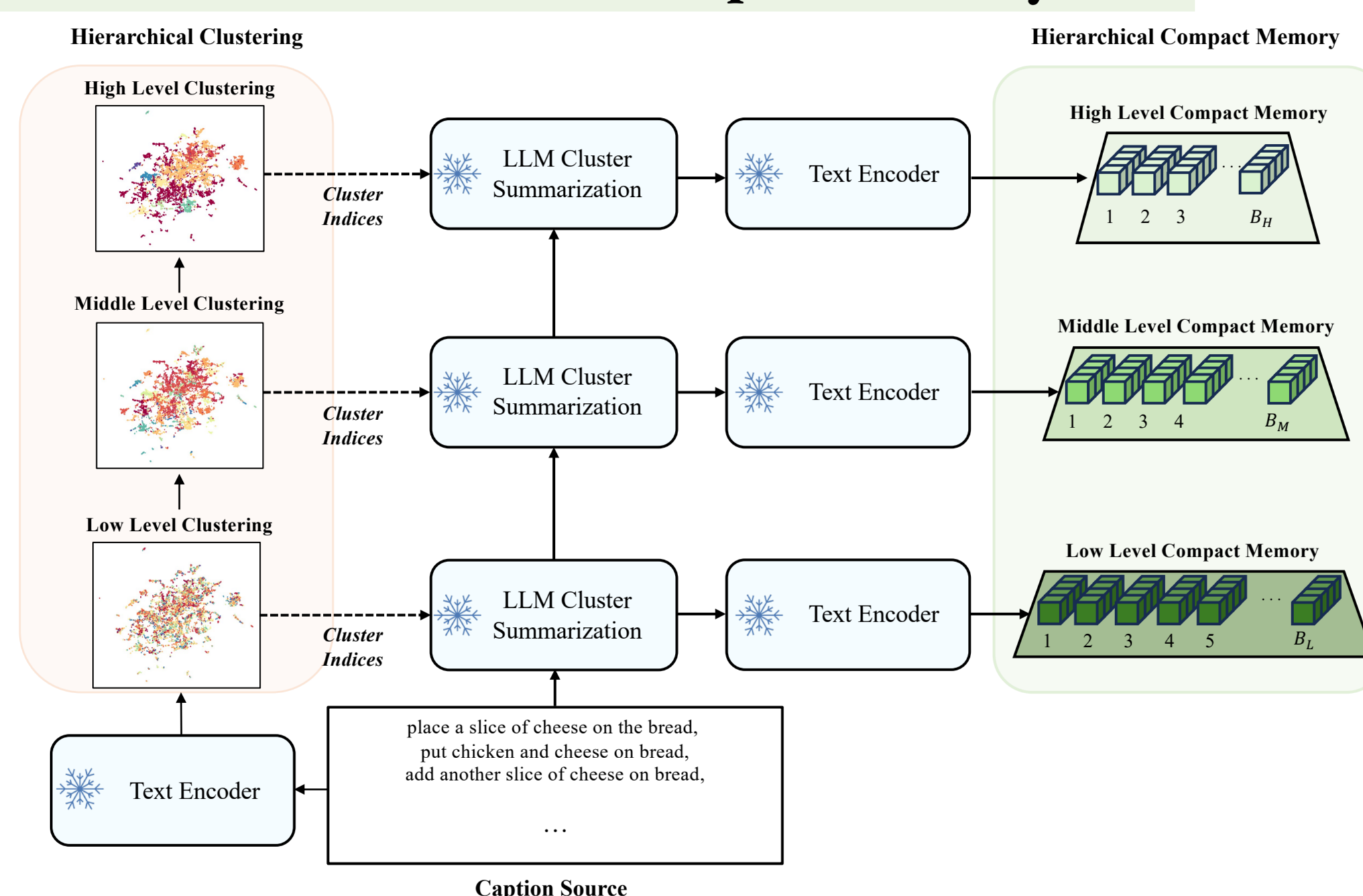
- We focus how to improve event localization and captioning from untrimmed video with prior knowledge memory bank.

- For this, we propose two sections :

1) How to model Hierarchical Compact Memory and 2) How to hierarchically retrieve?



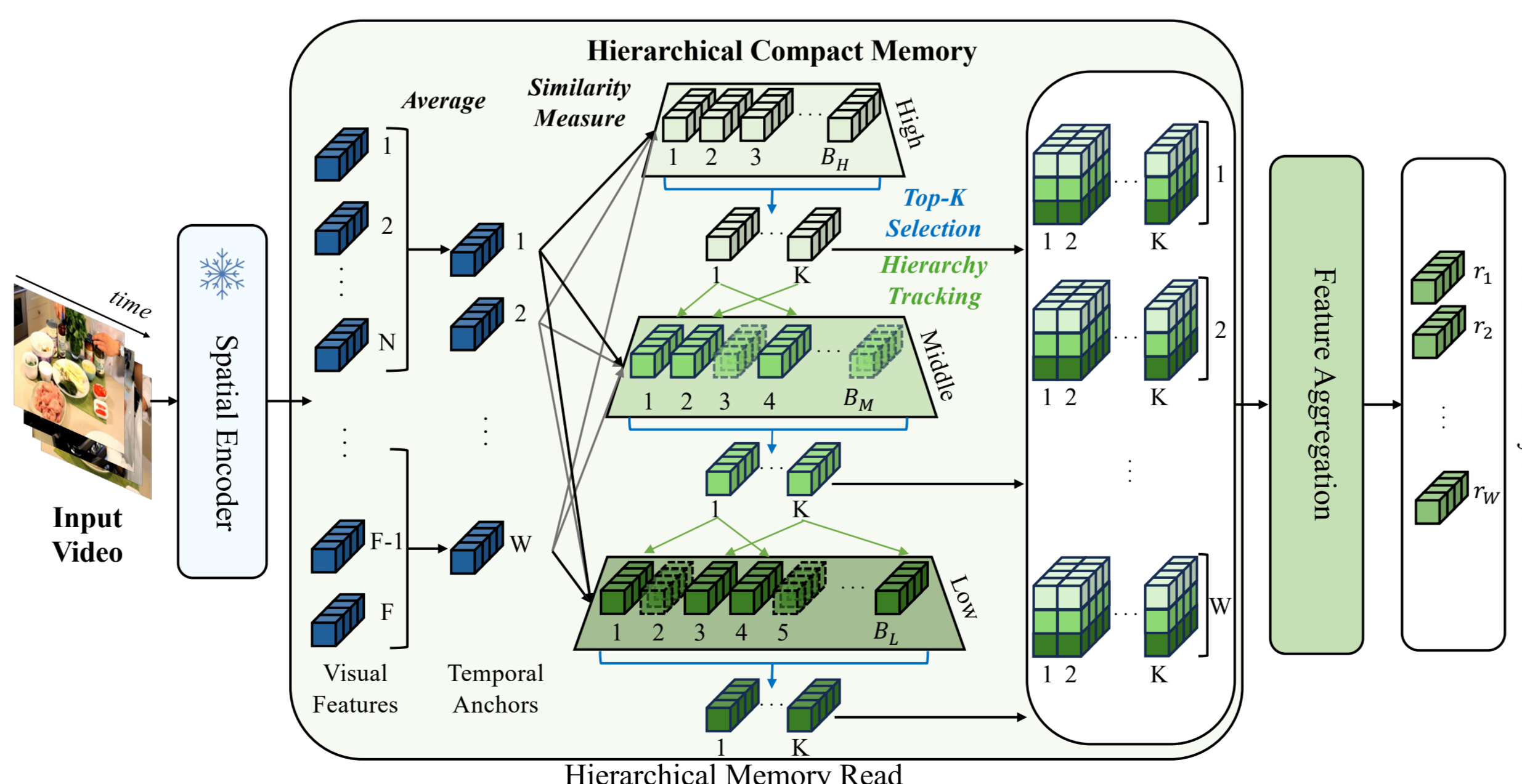
How to model Hierarchical Compact Memory?



- We iteratively cluster the caption source to create hierarchical results. In the first low level, we generate representative summaries for each cluster with clustered indices, and higher levels re-summarize these for compact memory components.

- This approach efficiently recalls abstract concepts and detailed episodes, minimizing redundancy while preserving semantic relevance.

How to hierarchically retrieve?



- We divide the video into W temporal anchors and retrieve for each anchor.
- We perform top-down search to efficiently retrieve, starting from abstract information and progressively retrieving connected detailed information. In here, we repeatedly select top-k and track hierarchy across all levels for efficient retrieval.

Experimental Results

- Event Captioning Performance on YouCook2 and ViTT

Method	PT	YouCook2(val)				ViTT(test)			
		CIDEr	METEOR	SODA _c	BLEU4	CIDEr	METEOR	SODA _c	BLEU4
PDVC	✗	29.69	5.56	4.92	1.40	-	-	-	-
CM ²	✗	31.66	6.08	5.34	1.63	-	-	-	-
Streaming V2S	✓	32.90	7.10	6.00	-	25.2	5.80	10.00	-
DIBS	✓	44.44	7.51	6.39	-	-	-	-	-
Vid2Seq [†]	✓	<u>66.29</u>	<u>12.41</u>	<u>9.87</u>	<u>5.64</u>	<u>48.84</u>	<u>9.51</u>	<u>14.99</u>	<u>0.71</u>
HiCM ² (Ours)	✓	71.84	12.80	10.73	6.11	51.29	9.66	15.07	0.86

- Event Localization Performance on YouCook2 and ViTT

Method	PT	YouCook2(val)			ViTT(test)		
		F1	Recall	Precision	F1	Recall	Precision
PDVC	✗	26.81	22.89	32.37	-	-	-
CM ²	✗	28.43	24.76	<u>33.38</u>	-	-	-
Streaming V2S	✓	24.10	-	-	35.40	-	-
DIBS	✓	<u>31.43</u>	26.24	39.81	-	-	-
Vid2Seq [†]	✓	31.08	<u>30.38</u>	31.81	46.21	45.89	<u>46.53</u>
HiCM ² (Ours)	✓	32.51	32.51	32.51	<u>45.98</u>	<u>45.00</u>	47.00

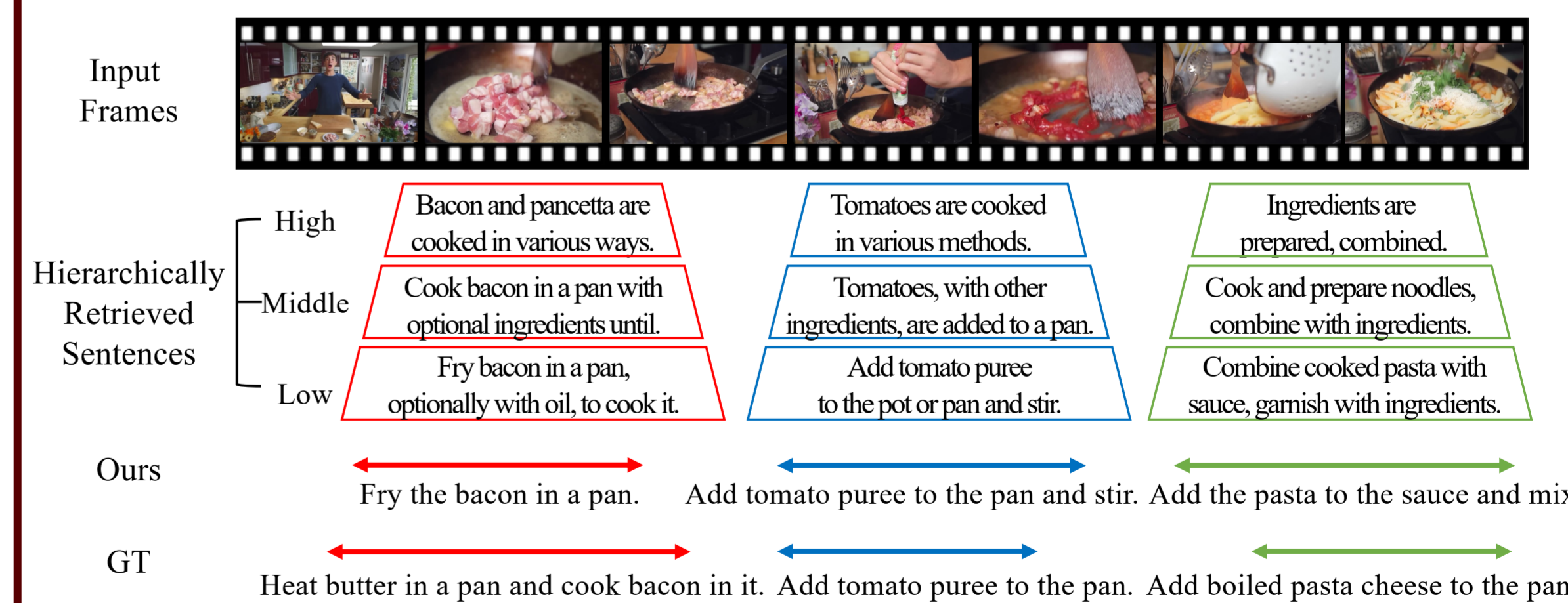
- ✓ Our model achieves State-of-the-art performance with Hierarchical Compact Memory.

- Ablation study for the use of hierarchical levels memory on YouCook2.

High	Middle	Low	CIDEr	METEOR	SODA _c	F1
		✓	66.29	12.41	9.87	31.08
		✓	67.75	12.33	10.35	32.27
	✓	✓	67.59	12.45	10.37	32.23
✓		✓	67.21	12.21	10.28	31.98
	✓	✓	68.61	12.35	10.51	32.07
✓		✓	69.81	12.63	10.54	33.00
✓	✓	✓	68.05	12.33	10.41	32.87
✓	✓	✓	71.84	12.80	10.73	32.51

- ✓ We showed that employing a comprehensive hierarchical memory yields complementary information and superior performance.

- Qualitative Results

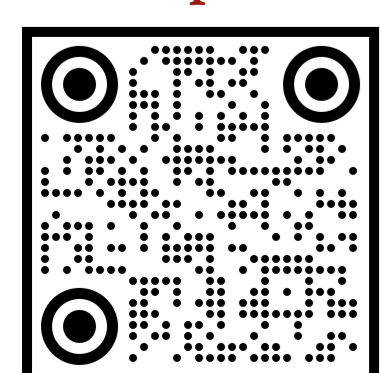


- ✓ It can be observed that effectively retrieving meaningful and relevant memory references across multiple levels, from abstract concepts to detailed information, yields significant benefits.

- ✓ As a result, our method generates relatively accurate event boundaries and captions.

- ✓ Our findings suggest that the synergy between pre-trained prior knowledge and retrieval-augmented knowledge could complement existing pretraining efforts, potentially contributing to further improvement in the field.

Paper



Github



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