



# VideoEspresso: A Large-Scale Chain-of-Thought Dataset for Fine-Grained Video Reasoning via Core Frame Selection

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### **Background of Video Understanding**



The video clip shows a bird of prey attacking a small animal. The bird is seen swooping down and attacking the animal, which appears to be a rabbit. The bird is a hawk, and it is hunting for its prey in the desert.



The video shows a dragon playing a guitar. The dragon is shown in different angles and positions while playing the guitar. The video is of high quality and the dragon's movements are smooth and fluid.

- Most existing video understanding focuses on summarization or detailed description
- Video reasoning often requires understanding temporal logic and object interactions
- Videos are **highly redundant** many frames contribute little new information







**Question**: What vegetables are used in the preparation of this dish?



### **200+ Frames** Sampled at FPS=1

Answering the question doesn't require processing the entire sequence







Question: What vegetables are used in the preparation of this dish?



Vegetable 1

Vegetable 2

### Still High Redundancy!

Even key segments are not frame-efficient

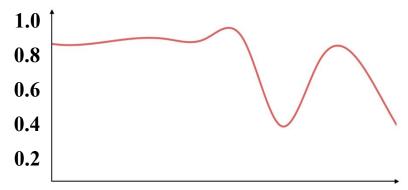




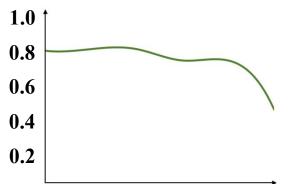


Question: What vegetables are used in the preparation of this dish?





**Intra-frame Similarity of segment-1** 



**Intra-frame Similarity of segment-2** 

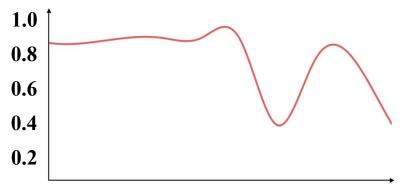




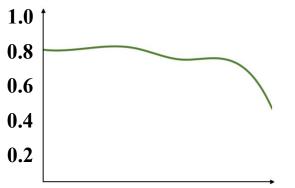


Question: What vegetables are used in the preparation of this dish?





**Intra-frame Similarity of segment-1** 



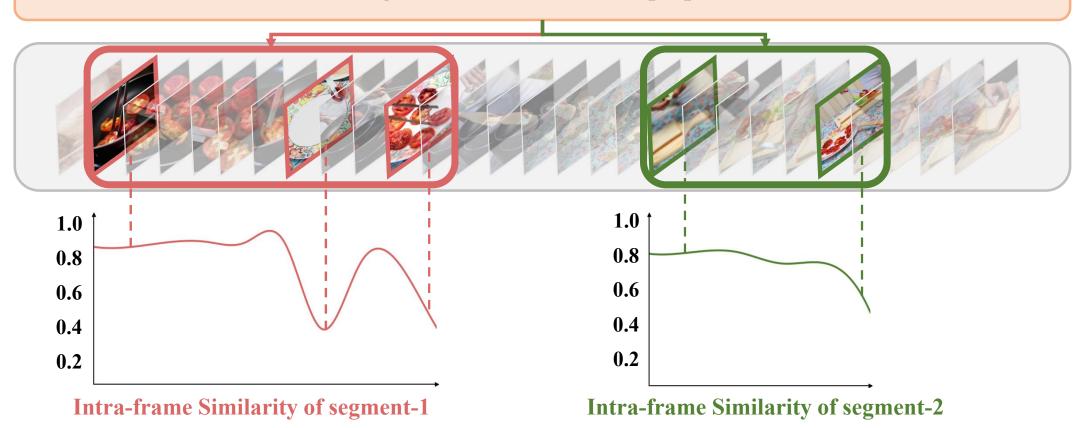
**Intra-frame Similarity of segment-2** 







Question: What vegetables are used in the preparation of this dish?









**Question**: What vegetables are used in the preparation of this dish?



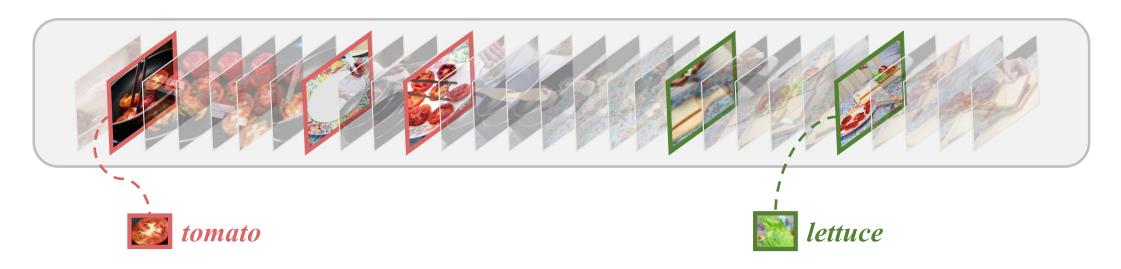
A Few Frames are Enough!







**Question**: What vegetables are used in the preparation of this dish?



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Leveraging Core Frames and Key Items is key to reasoning







### **Rethinking VideoQA Construction**



Ours

Common Video Q&A



#### Simple QA pairs

**Q:** Which one of these descriptions correctly matches the woman's action in the video?

A: (D) going upstairs





Key Frames
Selection



**Q:** What makes that man so concentrated on searching inside the open drawer?







#### Chain-of-Thought pairs

Key item#1: man Bbox: [[575, 513, 544, 972]]
Key item#3: pink pills Bbox: [[575, 513, 544, 972]]

Evidence: The <obj start> man <obj\_end> <box\_start> [[575, 513, 544, 972]] ... with <obj\_start> pink pills <obj\_end> <box\_start> [[355, 443, 33, 61]] <box\_end> later in the images.

A: The man at the desk appears to search...







**Question:** Based on the sequence of actions described, what logical steps can be inferred about the purpose and functionality of the culinary tool used by the person?

 Key Items #1: tool, onion
 Bbox: [[477, 517, 125, 137]...]

 Key Items #2: food
 Bbox: [[314, 489, 286, 980]]

 Key Items #3: cucumber
 Bbox: [[384, 498, 418, 977]...]

*Evidence:* The sequence of actions demonstrates that the tool in **key frame 1** [[477, 517, 125, 137]] ...

**Answer:** The culinary tool is designed to hold vegetables securely in place while they are being sliced, ensuring...







**Question:** Describe the interactions between the horse jockeys or players and the officials as observed in multiple sequences of the video.

 Key Items #1: jockeys,referee
 Bbox: [[427, 257, 154, 274]...]

 Key Items #2: referee, men
 Bbox: [[712, 274, 56, 243]...]

 Key Items #3: horse
 Bbox: [[611, 399, 68, 288]...]

*Evidence:* The horse jockeys in **key frame 1** [[427, 257, 54, 274]] interact with an...

**Answer:** The horse jockeys or players interact with officials or a referee, as one jockey says something to an official...









**Question:** Please infer the type of sports the person in the video is performing, and infer the relationship between the two objects in front and behind.

 Key Items #1: spray, launch
 Bbox: [[293, 394, 203, 440]...]

 Key Items #2: dashboard
 Bbox: [[290, 385, 209, 399]]

 Key Items #3: man, dashboard
 Bbox: [[295, 392, 221, 422]...]

 Key Items #4: dashboard
 Bbox: [[184, 284, 938, 692]]

 Evidence: The person in key frame 1 is riding a launch

 [[293,394,203,440]] and splashing huge waves...

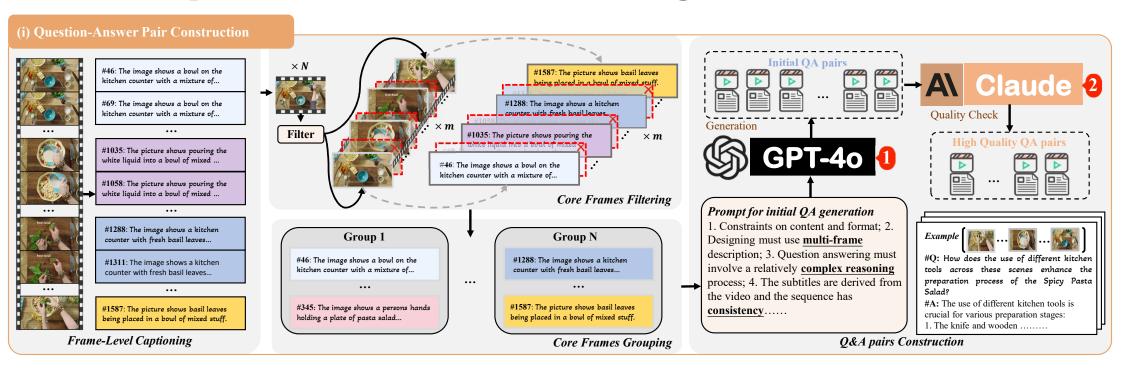
**Answer:** Two people are riding a speedboat on the sea and creating huge waves. The person behind is taking a photo...







### Detailed Pipeline for Multi-Frame Reasoning QA Construction



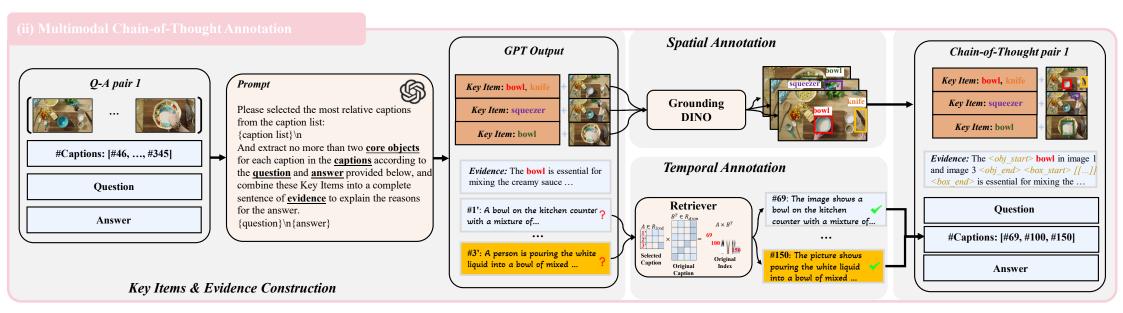
- Select & group core frames for reasoning
- Use GPT-40 to generate multi-frame QA pairs
- Claude checks for consistency and quality







# Detailed Pipeline for Multimodal Chain-of-Thought Annotation



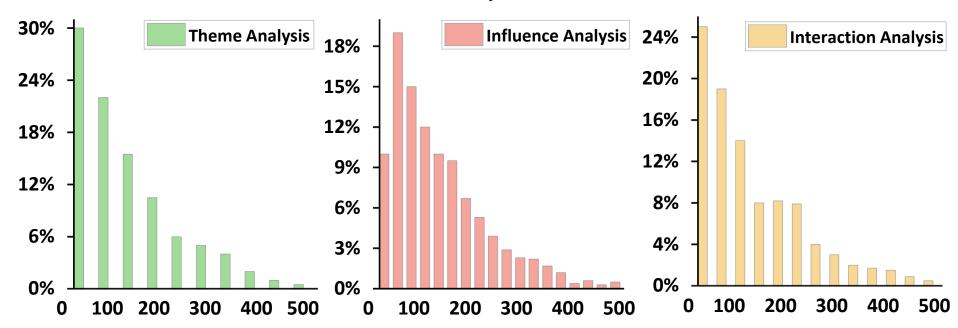
- GPT extracts key items & generates reasoning evidence with given QA pairs.
- Spatial & temporal annotation ensure multimodal alignment.
- Outputs interpretable, high-quality CoT annotations.







### Distribution of Distance between adjacent core frames



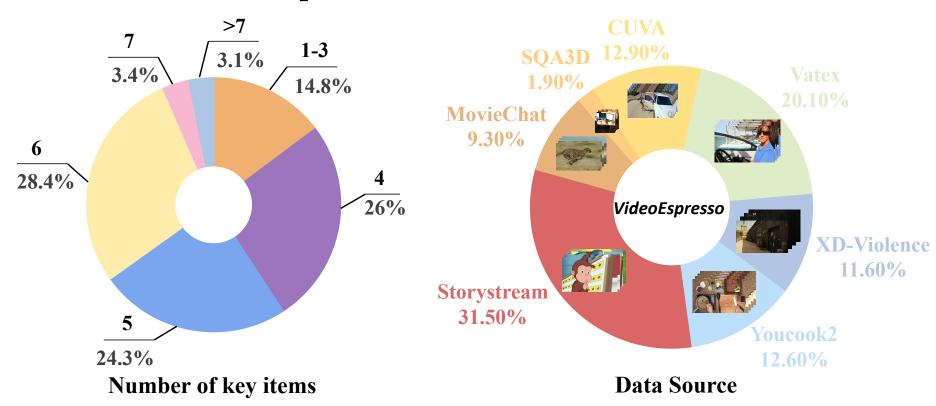
- Core frame spacing differs across reasoning types
- Sampling strategy should be task-aware
- A uniform frame sampling strategy is suboptimal







### **Details Data of VideoEspresso**



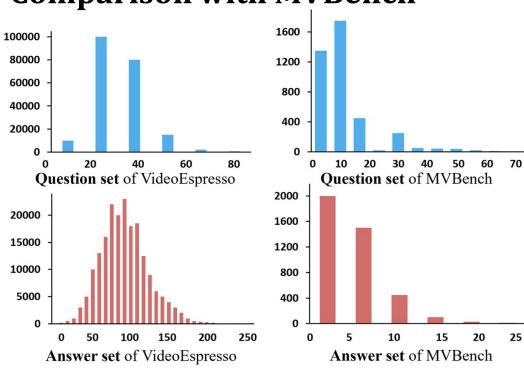
- More than 200K QA pairs with rich CoT annotation
- Most questions involve **4–6** key items, enabling focused reasoning
- The dataset is built from **diverse** sources, covering **various** domains

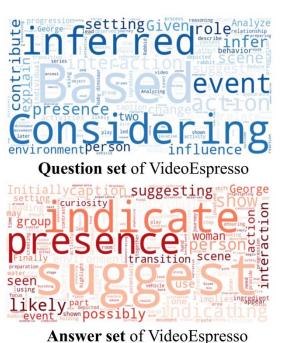






### **Comparison with MVBench**







Comparison of **Token Length** 

Comparison of Word Cloud

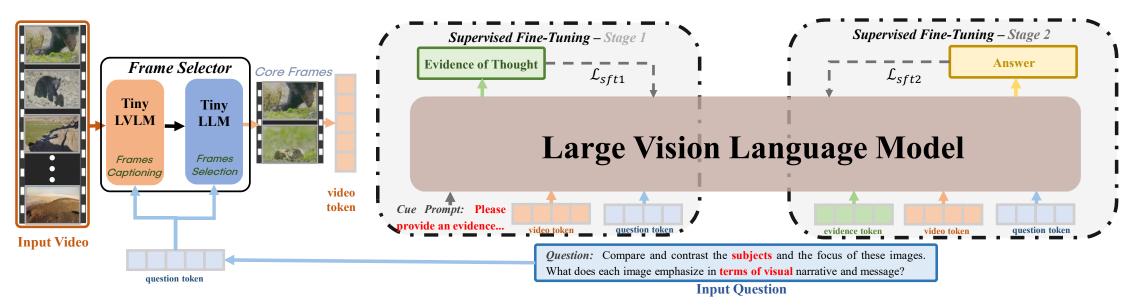
- VideoEspresso uses **longer**, more **descriptive** questions/answers
- Focuses more on reasoning verbs like consider, indicate, suggest
- MVBench centers on **simpler** object/action terms







# **Hybrid LVLMs Collaboration for VideoQA**



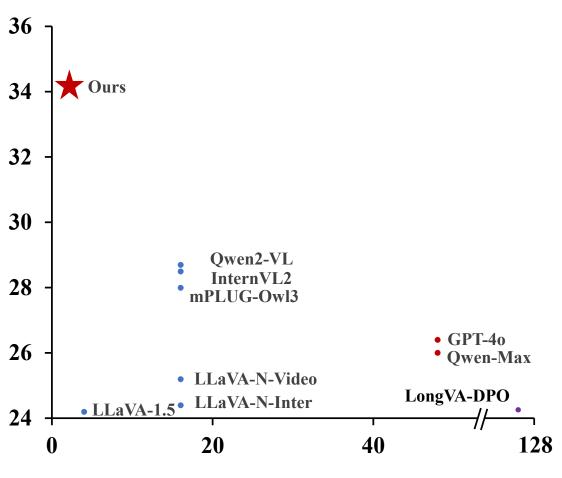
- Tiny model selects **informative frames** from video
- Large model focuses on **reasoning** with selected evidence
- Efficient pipeline without sacrificing answer quality



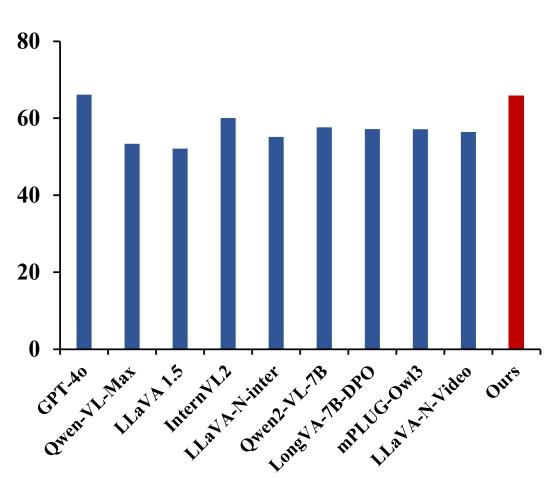




### **Main Results**



**Result on objective Benchmark** 



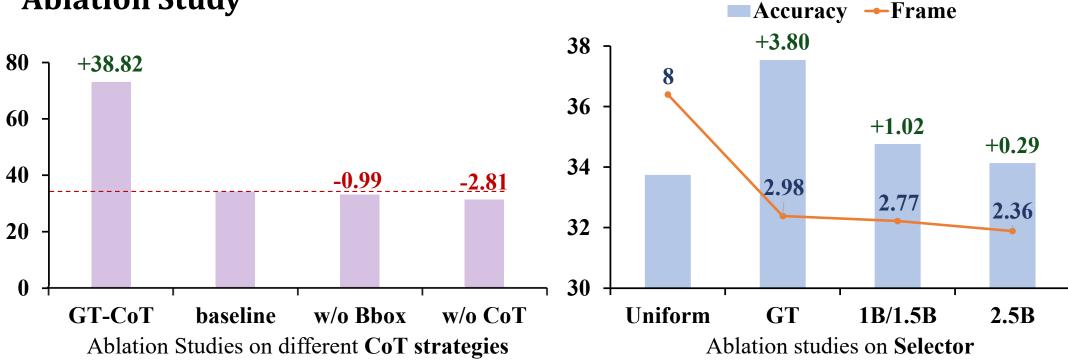
**Result on subjective Benchmark** 







# **Ablation Study**



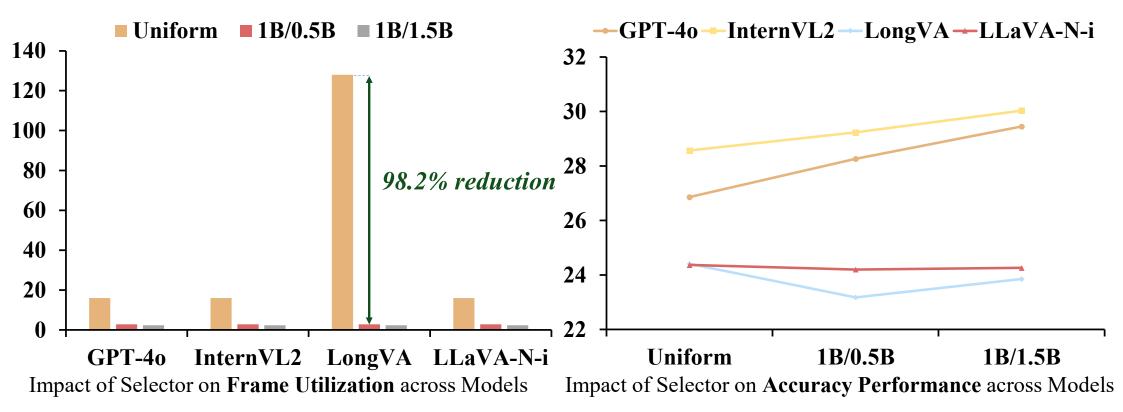
- CoT reasoning significantly boosts performance
- Spatial grounding (Bbox) contributes essential object-level cues.
- Learned selectors reduce frame count while maintaining accuracy.







### **Evaluations Results with Selector Adoption**



- **Significantly reduces input frames** (up to 98.2% fewer) with lightweight selector.
- Maintains or even improves accuracy, despite fewer frames being used.







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https://github.com/hshjerry/VideoEspresso



