
VURF: A General-purpose Reasoning and Self-refinement Framework for Video Understanding

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1 Dataset Specific Prompts

We curated in-context examples for 4 different datasets (NeXTQA, STAR, Social-IQ, and TrafficQA). The prompts for two of the datasets are shown in Figure 4. We show 3 in-context examples for each prompt.

<pre>... Think step by step to answer the question. Question: What did the person do with the ball? Program: TRACK0=TRACK(video=VIDEO, query="person", max_tracks=1) TRACK1=TRACK(video=VIDEO, query="ball", max_tracks=1) TRACK2=MERGE(track1=TRACK0, track2=TRACK1) VID0=CROP(video=VIDEO, track=TRACK2) ANSWER=VQA(video=VID0, query="What did the person do with the ball?") FINAL_RESULT=RESULT(var=ANSWER) Question: What will the person do next with the sofa? Program: SUMMARY=SUMMARIZE(video=VIDEO) ANSWER=PREDICT(summary=SUMMARY, query="What will the person do next with the sofa?") FINAL_RESULT=RESULT(var=ANSWER) Question: What did the person do to the paper after opening the bag? Program: ANSWER=VQA(video=VIDEO, query="What did the person do to the paper after opening the bag?") FINAL_RESULT=RESULT(var=ANSWER) Question: {question} Program: ...</pre>	<pre>... Think step by step to answer the question. Question: Where was the video taken? Program: VID0=TRIM(video=VIDEO, start=0.3, end=0.7) ANSWER=VQA(video=VID0, query="Where is this video taken?") FINAL_RESULT=RESULT(var=ANSWER) Question: What's the condition of the road? Program: TRACK0=TRACK(video=VIDEO, query="road", max_tracks=1) VID0=CROP(video=VIDEO, track=TRACK0) ANSWER=VQA(video=VID0, query="What is the condition of the road?") FINAL_RESULT=RESULT(var=ANSWER) Question: How many vehicles appeared in this video? Program: TRACK0=TRACK(video=VIDEO, query="vehicle", max_tracks=1000) ANSWER=COUNT(track=TRACK0) FINAL_RESULT=RESULT(var=ANSWER) Question: {question} Program: ...</pre>
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Figure 1: **LEFT:** LLM prompt for *STAR* dataset **RIGHT:** LLM prompt for *TrafficQA* dataset

2 Refinement Prompts

2.1 Error Correction

The error correction module queries the LLM two times, one to receive feedback on a program and the other to correct the program given the feedback. The prompts are shown in Figure 2. We display 2 in-context examples for each prompt.

2.2 Self-Refinement

The self-refinement module (which we pre-apply to refine our in-context examples) consists of 2 major queries to the LLM. One is for generating a context-free program to avoid hallucinations. Other is to convert the context-free program to a valid program. The prompts are shown in Figure 3

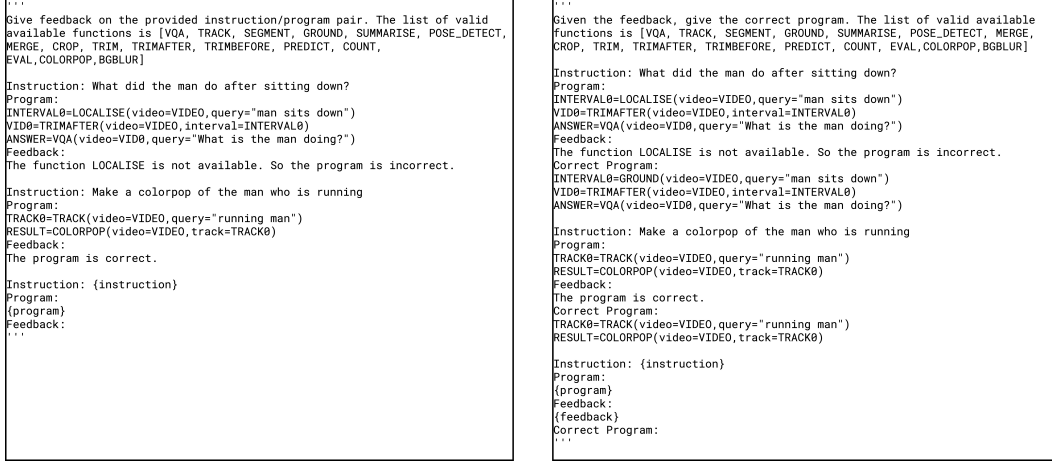


Figure 2: **LEFT:** The prompt for the feedback generation of a given program. **RIGHT:** Given a feedback the correct program is generated by giving this prompt to the LLM.

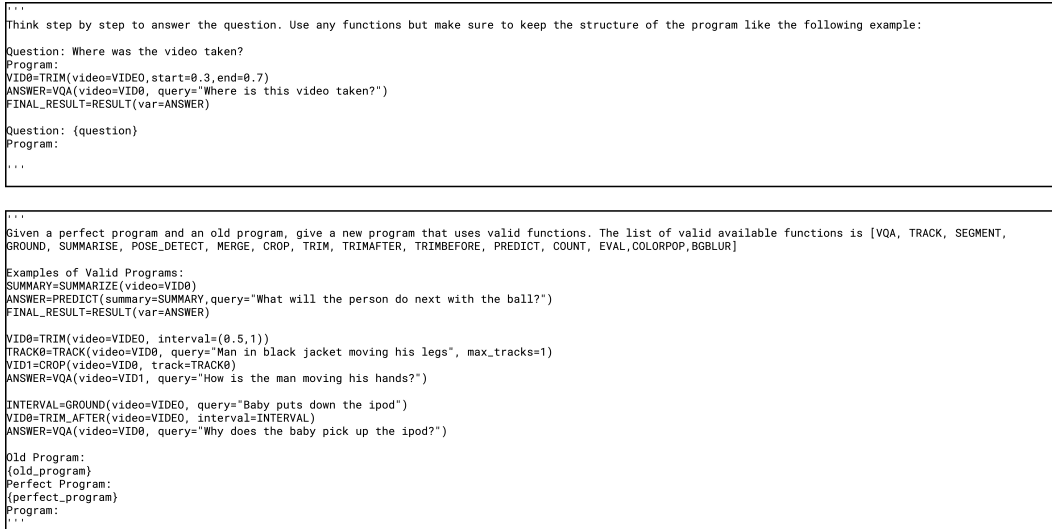


Figure 3: **TOP:** The prompt for the context-free generation. **BOTTOM:** The prompt for aligning a "perfect" program to a valid program

3 Ablations

3.1 LLM

We conducted experiments on two datasets (STAR and Social-IQ-2.0) using three additional LLM models for program generation: GPT-4o, CodeLLAMA, and Code-T5. The results, shown in Table 1, indicate that Instruct models excel in program generation due to the task’s instruction-oriented nature. Specifically, CodeLLAMA-7b performed poorly, generating numerous errors likely because it is trained to produce actual code rather than pseudocode-like instructions. Code-T5 failed to generate a correct program altogether, resulting in minimal performance.

3.2 Transcripts

Since the Social-IQ-2.0 dataset contains the transcripts of the videos so we introduce a new function, ANALYSE, tailored for questions aimed at discerning the mood or tone of the video conversation. This function involves querying a Language Model (LLM) with the transcript, and, in some instances,

Table 1: *Comparison with multiple LLMs*, both open source models and other GPT model versions.

LLM	STAR	Social-IQ-2.0
GPT-3.5-turbo-instruct (original)	47.2%	51.6%
GPT-4	43.5%	49.1%
CodeLlama-7b	10.0%	5.0%
CodeLlama-7b-Instruct	42.3%	48.4%
Code-T5	0%	0%

Table 2: *Effect of using transcripts from the Social-IQ QA dataset.*

Dataset	Baseline	With transcripts
Social-IQ-2.0	48.1%	51.6%

the appended summary along with the posed question. Our analysis includes performance metrics for both scenarios i.e., employing transcripts and not using them (Table 2).

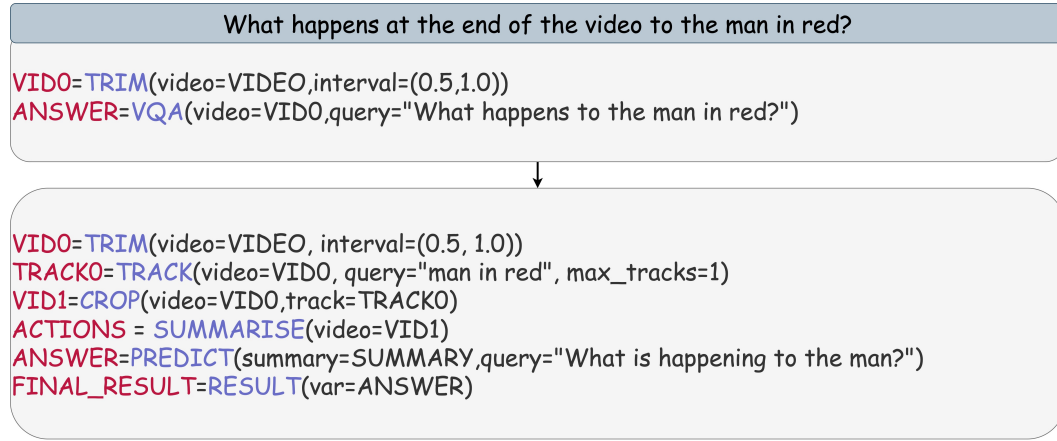


Figure 4: An example of the self refinement module. An initially generated prompt is refined by the LLM to produce a more complex and modular program.

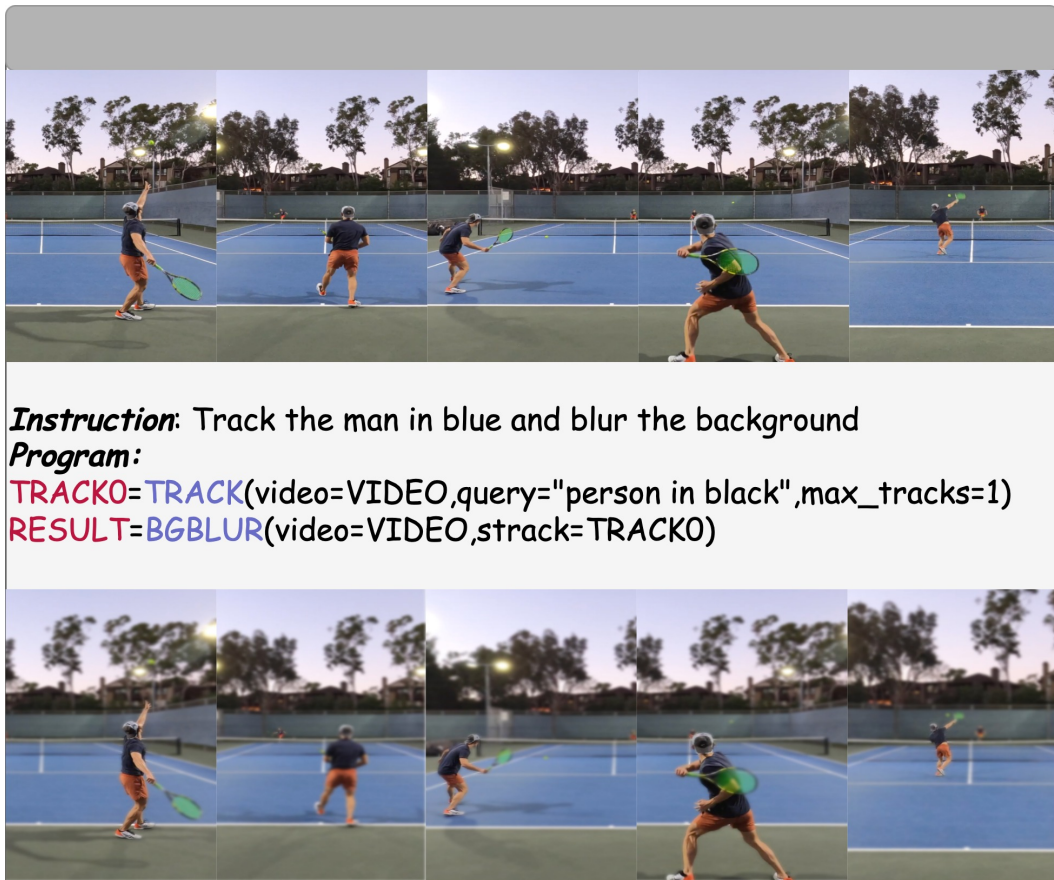


Figure 5: A qualitative example of video-editing using VURF.

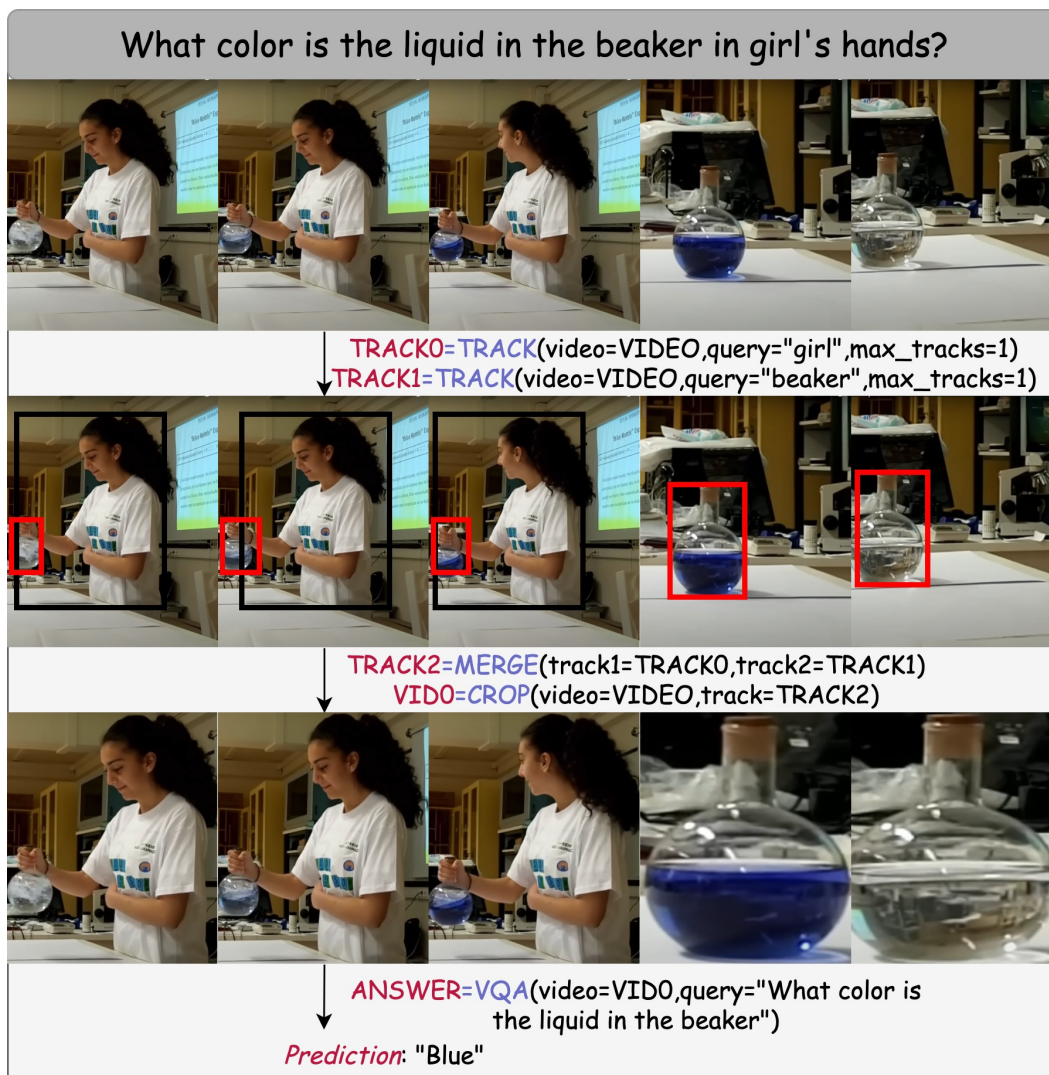


Figure 6: A step by step qualitative example of Video Question Answering using VURF.