

NLP Reading Group
Dec 3, 2025



JOHNS HOPKINS
WHITING SCHOOL
of ENGINEERING

Chain-of-Visual-Thought: Teaching VLMs to See and Think Better with Continuous Visual Tokens

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Presented by Sriram Sai Ganesh

Sections

- 1. Introduction**
- 2. Background**
- 3. CoVT**
- 4. Results**
- 5. Ablations**

Introduction

Intuition: reasoning over dense spatial features is bottlenecked by discrete tokens.

Task: Visual Question Answering (VQA) –
Given an image+text question, provide an answer.

Predominant approaches:

- **Late fusion:** Aggregate features from multiple modality-specific experts (ViT + LM)
- **Early fusion:** Train a VLM with an extended vocabulary (modality-specific discrete tokens)

Limitation: Rich perceptual cues are poorly represented.

- Visual information is inherently continuous.
- Aggregation/discretization: dense features like geometry, object boundaries, layout may be lost.



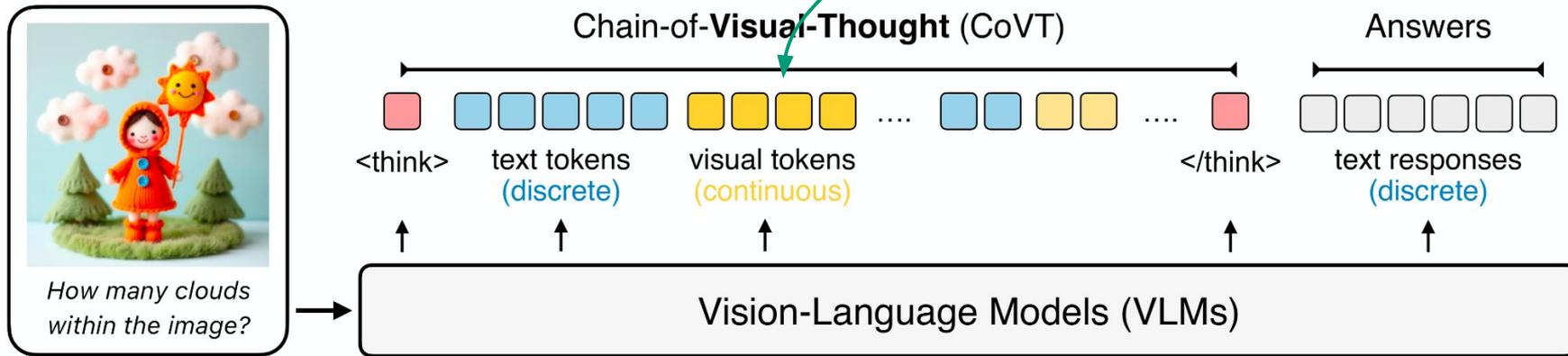
This paper:

Can we better **reason** over **visual** edges/layout/depth using **continuous tokens**?



Introduction

Continuous tokens for visual cues help ground semantic reasoning.



Limitation: Rich perceptual cues are poorly represented.

- Visual information is inherently continuous.
- Discretizing/aggregation: geometry, object boundaries, layout may be lost.

This paper: Can we better **reason** over **visual** edges/layout/depth using **continuous tokens**?

Overview

Chain of Visual Thought (CoVT)

- Defines groups of continuous visual tokens.
 - Contained within <think> steps.
- Each group corresponds to a **perceptual cue**.
 - Segmentation, depth, edge detection, image representation.
- The VLM is trained to compress rich representations into these tokens.
 - Trained on reconstruction loss, target = result from an expert.
 - Visual features are aligned with its token representation

Result: reasoning in tokens without explicit maps or tool calls.

Background work

Scalable Generative Cognitive Model "BAGEL"

Tool-augmented reasoning

Instruction: Replace the ground with white snow and the bear with a white polar bear



Prediction:




← IMAGE


← OBJ0=Seg(
image=IMAGE)


← OBJ1=Select(
image=IMAGE,
object=OBJ0,
query='ground')


← IMAGE0=Replace(
image=IMAGE,
object=OBJ1,
prompt='white snow')


← OBJ2=Seg(
image=IMAGE0)


← OBJ3=Select(
image=IMAGE0,
object=OBJ2,
query='bear')


← IMAGE1=Replace(
image=IMAGE0,
object=OBJ3,
prompt='white polar bear')

Non Real-World Scenarios



$3x^2 - 12 = 0$


38.6

From:Shanghai To:Beijing
Speed 250 km/h

(Overly simplistic synthetic scenarios)

Non Real-World Queries

Based on the text 'James Hutton is often viewed as the **first modern geologist**...', could you help me find out who the **first modern geologist** is?
(Simple retrieve, no tool use)

Can you help me find popular videos and trending gaming videos. I'm using the 'Cheap YouTube API' tool.
(Explicit tool use, no reasoning)

Find me some interesting news articles about the **culinary world**. Additionally, provide me with the current threshold securities list for **NVIDIA's stock**.
(Just tool list, no multi-step reasoning)

Real-World Setting



'a bowl of salad, a sandwich and a bottle of beer'



'BRIDGEPORT'



'Bridgeport Brewing Company closed in 2019.'



'2024-2019=5'

Company	Primary ACSC	Non-primary ACSC	Price
Apple	100.00	100.00	<0.001
Microsoft	150.00	150.00	<0.001
Amazon	200.00	200.00	<0.001
Google	300.00	300.00	<0.001
Facebook	400.00	400.00	<0.001
Twitter	500.00	500.00	<0.001
LinkedIn	600.00	600.00	<0.001
Slack	700.00	700.00	<0.001
Zoom	800.00	800.00	<0.001
Dropbox	900.00	900.00	<0.001
Spotify	1000.00	1000.00	<0.001
Netflix	1100.00	1100.00	<0.001
Disney	1200.00	1200.00	<0.001
Walt Disney	1300.00	1300.00	<0.001
Warner Bros	1400.00	1400.00	<0.001
Universal	1500.00	1500.00	<0.001
Paramount	1600.00	1600.00	<0.001
20th Century Fox	1700.00	1700.00	<0.001
Time Warner	1800.00	1800.00	<0.001
Comcast	1900.00	1900.00	<0.001
Verizon	2000.00	2000.00	<0.001
AT&T	2100.00	2100.00	<0.001
Charter	2200.00	2200.00	<0.001
Spectrum	2300.00	2300.00	<0.001
Cable One	2400.00	2400.00	<0.001
Comcast	2500.00	2500.00	<0.001
Verizon	2600.00	2600.00	<0.001
AT&T	2700.00	2700.00	<0.001
Charter	2800.00	2800.00	<0.001
Spectrum	2900.00	2900.00	<0.001
Cable One	3000.00	3000.00	<0.001

'Primary ACSC: 49.89(±11.8).....
Non-primary ACSC: 43.62(±12.8).....'

'49.89-43.62=6.27.....'



Q: What's the difference between **Primary ACSC** and **Non-primary ACSC**? Please illustrate it using a bar chart.
A:



Background work

Text Space Reasoning

Which point is closer to the camera?



LLaVA-AURORA

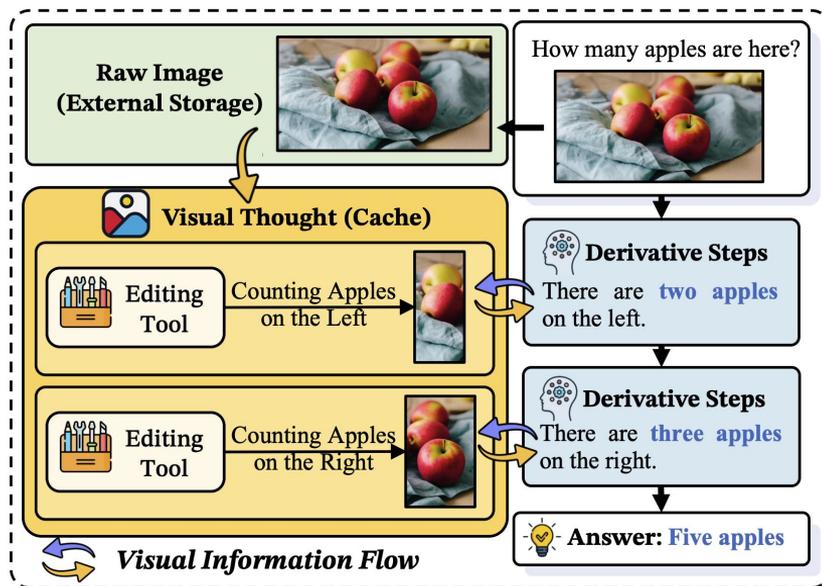
Step 1: Point A is at (250, 200), B is at (240, 300), and C is at (480, 210)

Step 2: I need depth estimation to determine which point is closer

Step 3: The depth map of the input image is: `<perception_tokens>` ... `</perception_tokens>`



Step 4: Point B has higher pixel value on depth map, so it is closer



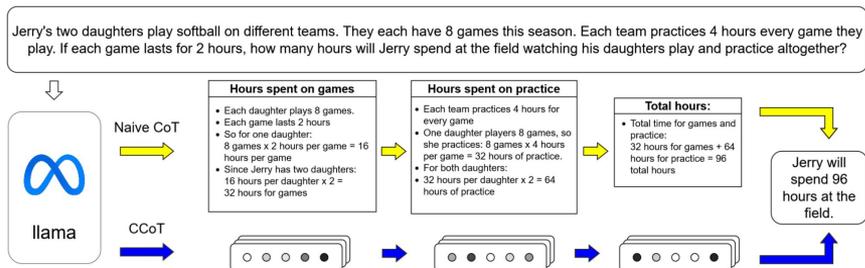
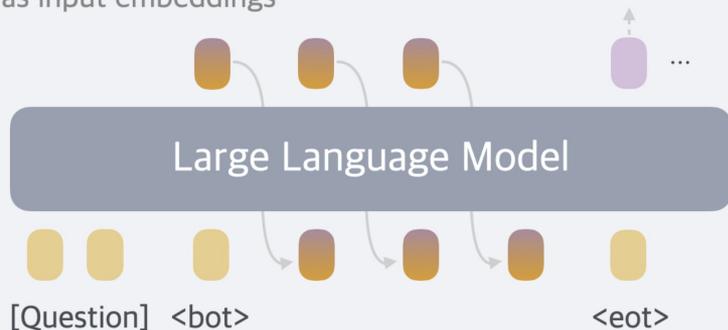
Background work

Latent Space Reasoning

Chain of Continuous Thought (COCONUT)

Last hidden states are used as input embeddings

[Answer]



Chain of Visual Thought (CoVT)

Intuition:

- Text-only CoT **accumulates errors**.
- Supervision is **dominated by text responses**.

CoVT:

- A framework that equips VLMs with the ability to reason through **continuous visual tokens**.
- Tailored alignment strategies and a training pipeline to enable VLMs to learn, interpret, and reason effectively within this continuous visual space.

Chain of Visual Thought (CoVT)

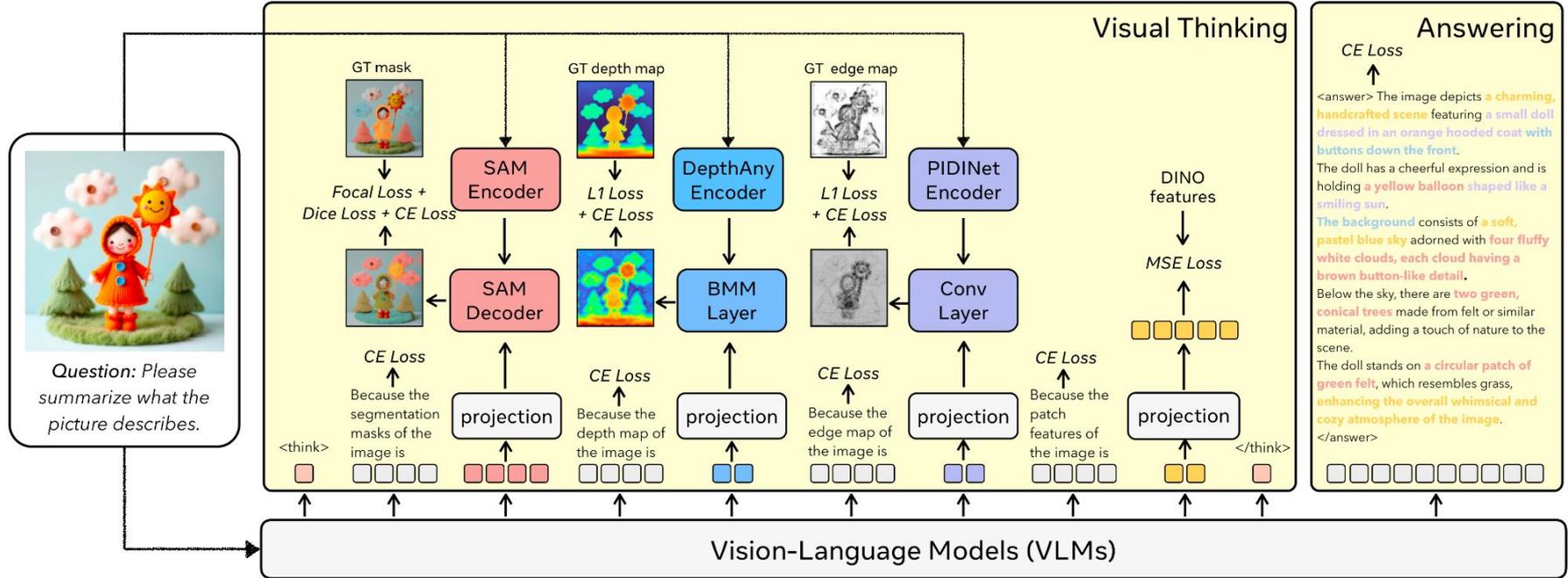
Pipeline

- Equips VLMs with chains of visual thought.
- VLM next token prediction:

$$P(Y | \mathcal{V}, \mathcal{T}; \theta) = \prod_{i=1}^n P(y_i | y_{<i}, \mathcal{V}, \mathcal{T})$$

- This work extends this formulation by introducing **CoVT tokens** :
 - each y_i represents either a visual or text token.
- The VLM is trained to function as a dense visual encoder
 - Reconstruction supervision against task-specific decoders.

Chain of Visual Thought (CoVT)



Chain of Visual Thought (CoVT)

CoVT Tokens

Perceptual ability of VLMs can be summarized as:

- 1.** Instance recognition
 - segmentation tokens provide positional & shape information.
- 2.** 2D and 3D spatial relationships.
 - Depth tokens provide pixel-level depth information.
- 3.** Structure detection.
 - Edge tokens provide geometry-level details.
- 4.** Deep mining of semantic information.
 - DINO tokens provide patch representations.

Chain of Visual Thought (CoVT)

CoVT Tokens

Each token type is

- assigned a constant **token count** (within <think>)
 - supervised by a domain expert
1. Segmentation (*8 tokens*)
 - Supervised by SAM – Segment Anything Model.
 2. Depth (*4 tokens*)
 - Supervised by DepthAnything v2.
 3. Edge (*4 tokens*)
 - Aligned with PIDINet.
 4. DINO (*4 tokens*)
 - Supervised by DINOv2.

$$\hat{M}_i = \text{Decoder}(T_i^{\text{sam}}, f), \quad \hat{M}_i \in [0, 1]^{H \times W}$$

$$\hat{D}_i = \text{softmax} \left(T_i^{\text{depth}} \cdot F_i^{\text{depth}\top} \right)$$

Chain of Visual Thought (CoVT)

CoVT Training

- **Loss:**

- Joint loss function: $\mathcal{L}_{\text{total}} = \mathcal{L}_{\text{ce}} + \gamma(\lambda_{\text{seg}} \cdot \mathcal{L}_{\text{visual}}^{\text{seg}} + \lambda_{\text{depth}} \cdot \mathcal{L}_{\text{visual}}^{\text{depth}} + \lambda_{\text{edge}} \cdot \mathcal{L}_{\text{visual}}^{\text{edge}} + \lambda_{\text{dino}} \cdot \mathcal{L}_{\text{visual}}^{\text{dino}})$
- During inference, visual thinking tokens are not decoded.

- **Data:**

4-stage training, progressive leverage of visual tokens:

1. *Comprehension*: Insert visual tokens after <image> to teach basic semantics.
2. *Generation*: modify Q/A to guide the VLMs to precisely generate visual tokens.
3. *Reasoning*: Introduces CoVT format, visual tokens are used in reasoning for final answer.
4. *Efficient reasoning*: Randomly drop some sets of visual tokens (up to k); teach effective feature use.

Chain of Visual Thought (CoVT)

CoVT Training



Original Question
<image> \n How many people are jumping in the air?

Original Answer
There are three persons jumping in the air.

Stage 1: **Question:** <image> the segmentation of the image is <segmentation>, the depth map is <depth>, the edge map is <edge>, and the patch feature is <dino>\n How many people are jumping in the air?
Answer: There are three persons jumping in the air.

Stage 2: **Question:** <image>\n What's the segmentation, depth map, edge map, and the patch feature of the image?
Answer: <segmentation>, <depth>, <edge>, and <dino>.

Stage 3: **Question:** <image>\n How many people are jumping in the air?
Answer: <think>The segmentation of the image is <segmentation>, the depth map of the image is <depth>, the edge map of the image is <edge>, and the patch feature of the image is <dino>.</think> <answer>There are three persons jumping in the air.</answer>

Stage 4: **Question:** <image>\n How many people are jumping in the air?
Answer: <think>The segmentation of the image is <segmentation>, and the patch feature of the image is <dino>.</think> <answer>There are three persons jumping in the air.</answer>

Randomly drop visual anchors

Results: Quantitative

Visual tokens				CV-Bench				Other vision-centric benchmarks								
Seg	Depth	DINO	Edge	CVBench	Count	Depth	Dist.	BLINK	RW-QA	MMT	MMStar-P	MMVP	MME-RW	V*	HR _{4K}	HR _{8K}
Closed-source Models																
Claude-4-Sonnet				76.3	62.2	77.7	80.5	39.6	63.7	-	58.8	48.7	-	15.2	32.3	22.7
GPT-4o				79.2	65.6	86.7	81.0	63.0	69.7	-	65.2	72.0	-	42.9	50.6	46.7
Qwen2.5-VL-7B				74.5	65.0	72.8	75.5	55.7	68.6	61.7	67.1	56.0	60.0	76.4	68.6	64.9
CoVT (1 Visual Token)																
✓				77.9	66.0	80.8	80.5	57.4	71.1	62.1	68.5	58.7	62.1	79.1	71.9	69.0
	✓			78.7	65.4	83.2	78.2	56.4	71.5	62.7	69.9	58.7	62.0	79.1	71.9	69.4
		✓		71.3	64.7	72.3	66.7	55.8	71.5	62.5	67.9	57.3	61.1	77.5	71.0	68.6
CoVT (3 Visual Tokens)																
✓	✓	✓		80.0	66.2	86.8	82.5	56.0	71.6	62.1	69.2	58.7	63.7	78.0	72.9	69.4
Δ (vs Baseline)				+5.5	+1.2	+14.0	+7.0	+0.3	+3.0	+0.4	+2.1	+2.7	+3.7	+1.6	+4.3	+4.5
CoVT (4 Visual Tokens)																
✓	✓	✓	✓	79.8	66.1	89.2	80.5	56.2	71.8	61.9	68.4	56.7	63.3	78.5	72.5	69.9
Δ (vs Baseline)				+5.3	+1.1	+16.4	+5.0	+0.5	+3.2	+0.2	+1.3	+0.7	+3.3	+2.1	+3.9	+5.0

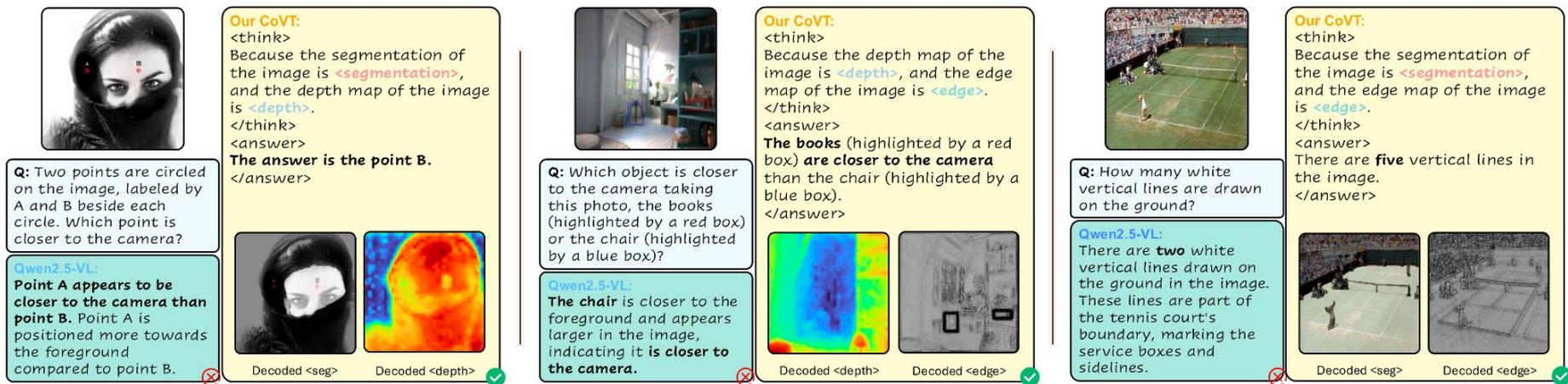
(CoVT on Qwen-2.5-VL-7B)

Results: Quantitative

	CV-Bench			BLINK				
	Count	Depth	Dist.	Count	Obj. Loc.	Rel. Depth	Vis. Corr.	Vis. Sim.
LLaVA	59.3	61.8	50.2	56.7	54.9	52.4	29.7	51.1
Aurora [†] (<i>depth</i>)	54.9	67.7	52.3	53.3	55.7	62.9	26.2	47.4
CoVT (<i>w/ Depth</i>)	60.7	71.0	52.3	56.7	59.8	75.8	31.4	53.3
Δ (<i>vs Aurora</i>)	+5.8	+3.3	+0.0	+3.4	+4.1	+12.9	+5.2	+5.9
Aurora [†] (<i>count</i>)	56.0	62.2	47.8	31.7	26.2	24.2	26.7	21.5
CoVT (<i>w/ Seg</i>)	61.9	60.7	51.3	58.3	56.6	69.4	29.7	52.6
Δ (<i>vs Aurora</i>)	+5.9	-1.5	+3.5	+26.6	+30.4	+45.2	+3.0	+31.1

(CoVT on LLaVA-v1.5-13B, for parity with Aurora)

Results: Qualitative



Ablations

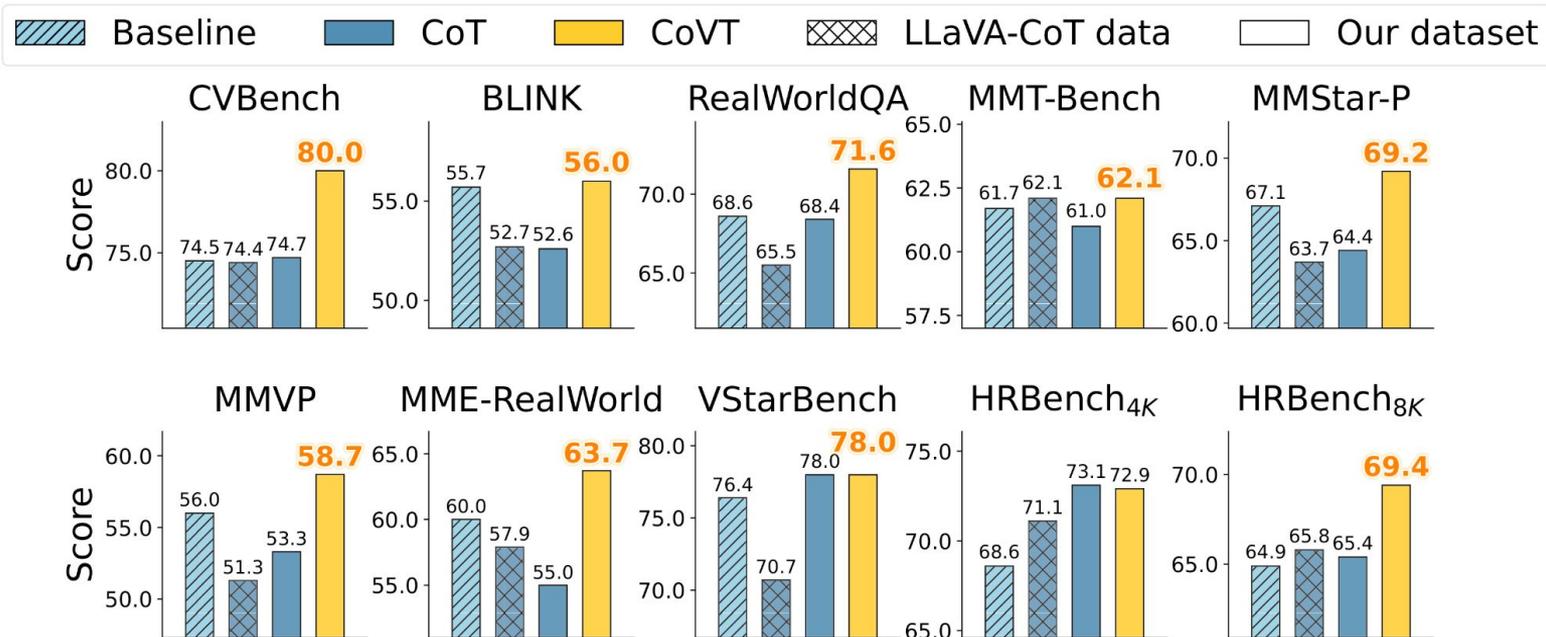


Fig. 6: Text-only CoT vs. CoVT.

Ablations

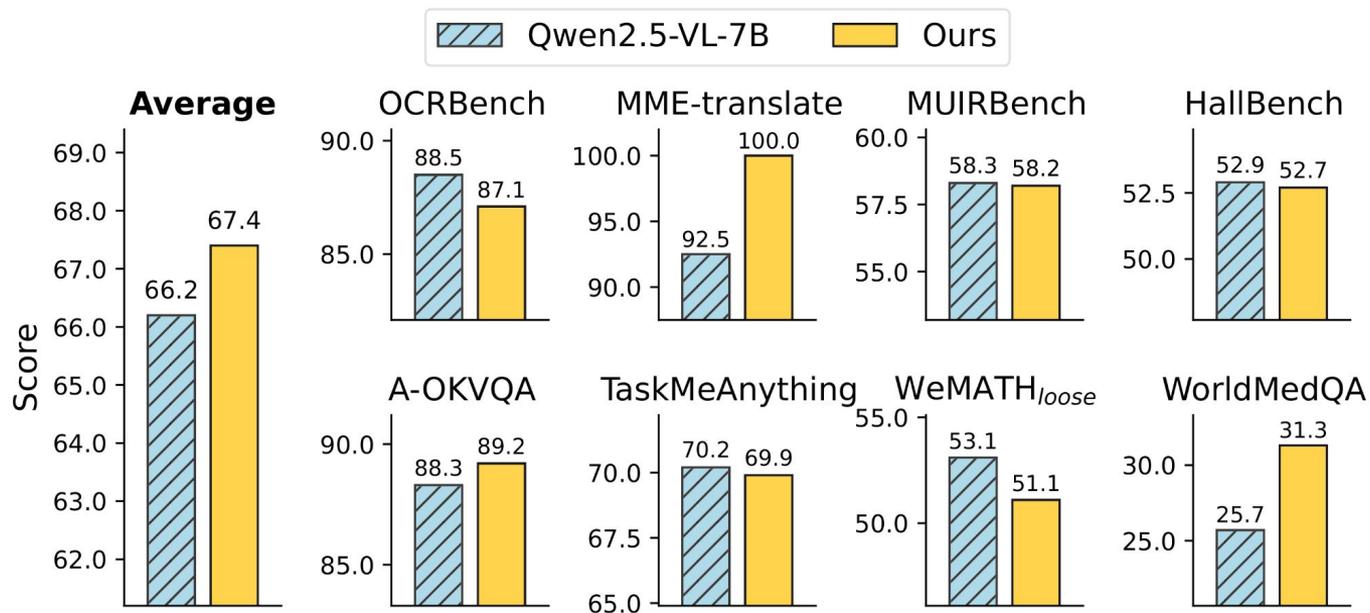


Fig. 7: Non vision-centric tasks.

Ablations

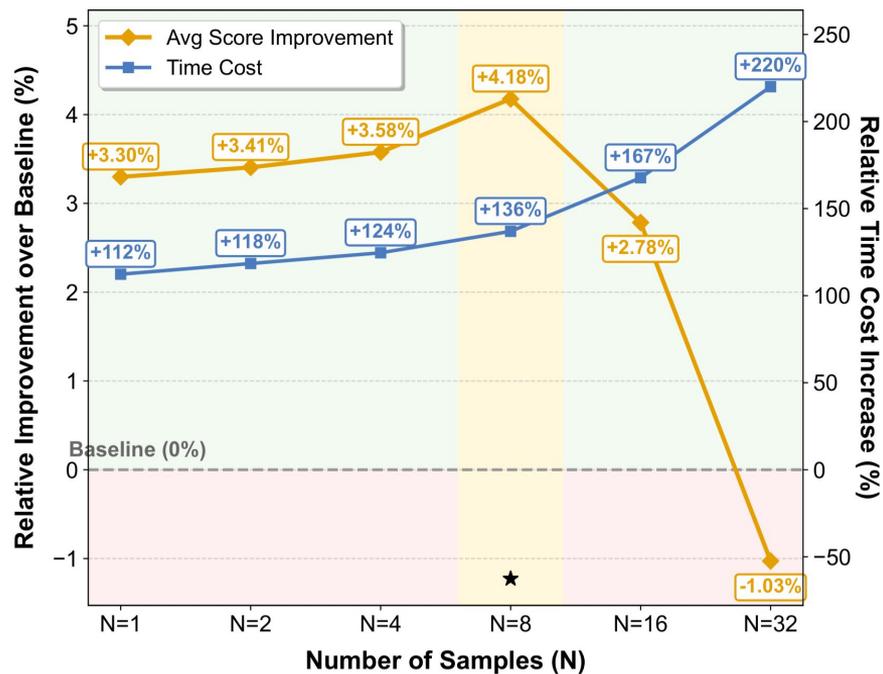


Fig. 12: performance vs. # of segmentation tokens

Examples



Question

Describe the scene in the picture in detail, and find out how many clouds are there in the sky.

Qwen2.5-VL

The image depicts a charming, handcrafted scene featuring a small, knitted figure dressed in an orange hooded outfit with matching boots. The figure is holding a cheerful, smiling sun-shaped balloon on a stick. The background includes three fluffy, cloud-like shapes in various shades of pink and white, floating above a small, green, grassy area with two coniferous trees. The sky is a soft, light blue color.

There are four clouds visible in the sky: one large pink cloud on the left, one smaller white cloud near the top right, and two more white clouds further to the right.

CoVT

<think> Because the segmentation of the image is <|sam_pad|><|sam_pad|><|sam_pad|><|sam_pad|><|sam_pad|><|sam_pad|><|sam_pad|><|sam_pad|>, the depth map of the image is <|depth_pad|><|depth_pad|><|depth_pad|><|depth_pad|>, and the perception feature of the image is <|dino_pad|><|dino_pad|><|dino_pad|><|dino_pad|>. </think>

<answer> The image depicts a charming, handcrafted scene featuring a small doll-like figure standing on a grassy patch. The figure is dressed in an orange outfit with matching boots, and it has a cheerful expression. The doll is holding a stick with a smiling sun at the top, which appears to be made from felt or similar material.

In the background, there are two green felt trees, each with a simple triangular shape and a brown trunk. Above the trees, there are four fluffy white clouds, each adorned with a brown button-like detail. The sky is painted in a soft blue color, providing a serene backdrop for the scene.

The overall composition is whimsical and playful, likely intended to evoke a sense of joy and innocence. The use of bright colors and simple shapes contributes to the overall aesthetic appeal of the image.

To answer the question about the number of clouds: There are four clouds visible in the sky above the trees. </answer>



Question

What is the beautiful woman sitting on in the picture?

Qwen2.5-VL

The woman in the picture is sitting on the hood of a car.

CoVT

<think> Because the segmentation of the image is <|sam_pad|><|sam_pad|><|sam_pad|><|sam_pad|><|sam_pad|><|sam_pad|>, and the depth map of the image is <|depth_pad|><|depth_pad|><|depth_pad|><|depth_pad|>. </think>

<answer> The beautiful woman is sitting on the hood of a white car. </answer>



Question

How many men are facing the camera in the picture?

Qwen2.5-VL

In the picture, there are three men facing the camera.

CoVT

<think> Because the segmentation of the image is <|sam_pad|><|sam_pad|><|sam_pad|><|sam_pad|><|sam_pad|><|sam_pad|>, and the depth map of the image is <|depth_pad|><|depth_pad|><|depth_pad|><|depth_pad|>. </think>

<answer> 2 </answer>