

Problem Formulation

Given a first-person **egocentric video** of a scene and a **natural-language spatial query** (e.g., object counting, absolute/relative distance, object/room size, relative direction, route planning, appearance order), a Multimodal LLM must output an answer; reasoning about 3D geometry, layouts, and object relations from only what a hand-held camera happened to record.

Why Single-turn Spatial Reasoning Is Fragile

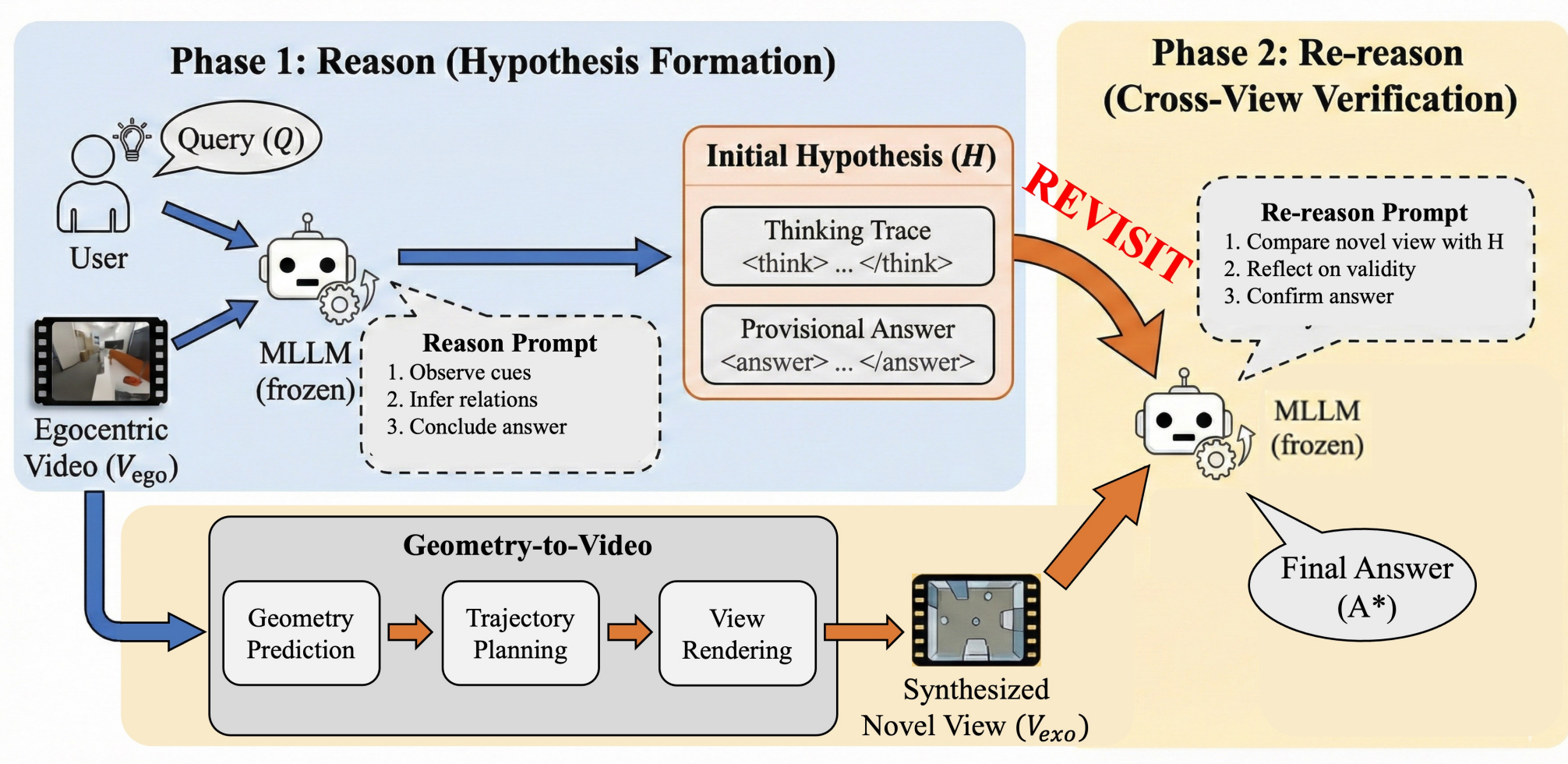
Egocentric videos are **trajectory-conditioned**:

- The visual evidence is limited to what the recorded camera path happens to reveal, and the temporal order of frames rarely aligns with the scene's actual spatial topology.
- Compounding this, general-purpose MLLMs lack explicit 3D world modeling and only implicitly enforce cross-frame correspondence.
- So when a model must answer in one pass, it fills missing geometry with semantic priors and produces a plausible but wrong conclusion.

Motivation

- Ideally, such errors could be detected and corrected if *complementary cross-view evidence* were available.
- This suggests that improving spatial reasoning is not only about better understanding in single-turn, but about enabling the model to *revisit* its hypothesis under complementary cross-view evidence.

Framework: Reason, Then Re-reason (ReRe)



Given an egocentric video and a spatial query, ReRe operates in two phases:

- The **Reason Phase** forms an initial hypothesis from the original view.
- The **Re-reason Phase** verifies or revises it against a synthesized allocentric view.
- A Geometry-to-Video pipeline produces this complementary view from recovered 3D geometry via trajectory planning and view rendering.

What Makes ReRe Work

- The revisiting process is what matters.
- New evidence, not just thinking twice.
- The two views are complementary.
- Oblique Sweep is the better trajectory.

➤ *See more exciting findings and experiments results in the paper!*

Qualitative Results

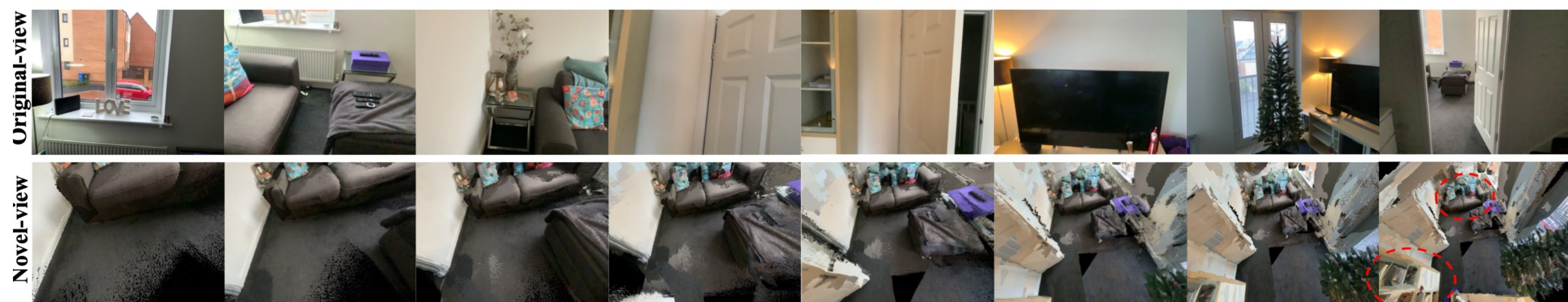


(a) Q: *How many chair(s) are in this room?*

GT: 2 ☐ *Baseline: 1* ☒ *ReRe: 2*

Reason Phase output : "let me think, I need to carefully examine the video frames to identify any chairs. In the first frame, I see a bed and some items on the floor, but no chair is visible. In the second frame, I see a desk with some items on it, and there's a chair partially visible behind the desk. Wait, I should double-check if there are any other chairs in the room. Looking at the second frame again, I can see the chair more clearly—it's a standard desk chair. Hmm, I think that's it. There's only one chair visible in the room.\n\n<answer>"

Re-Reason Phase output : "In the original video, I identified one chair, which was a desk chair partially visible behind the desk. Now, examining the new-view video, I can see the same desk chair more clearly. Additionally, I can now see another chair in the room — a wooden chair with a cushion, located near the bed on the left side of the frame. Therefore, there are two chairs in the room.\n\n<answer>2</answer>"



(b) Q: You are a robot beginning at the sofa and facing the TV. You want to navigate to the bookshelf. You will perform the following actions (Note: for each [please fill in], choose either 'turn back,' 'turn left,' or 'turn right.'): 1. Go forward until the TV 2. [please fill in] 3. Go forward until the bookshelf. You have reached the final destination. A. Turn Left, B. Turn Right, C. Turn Back

⊗ *Baseline: A*

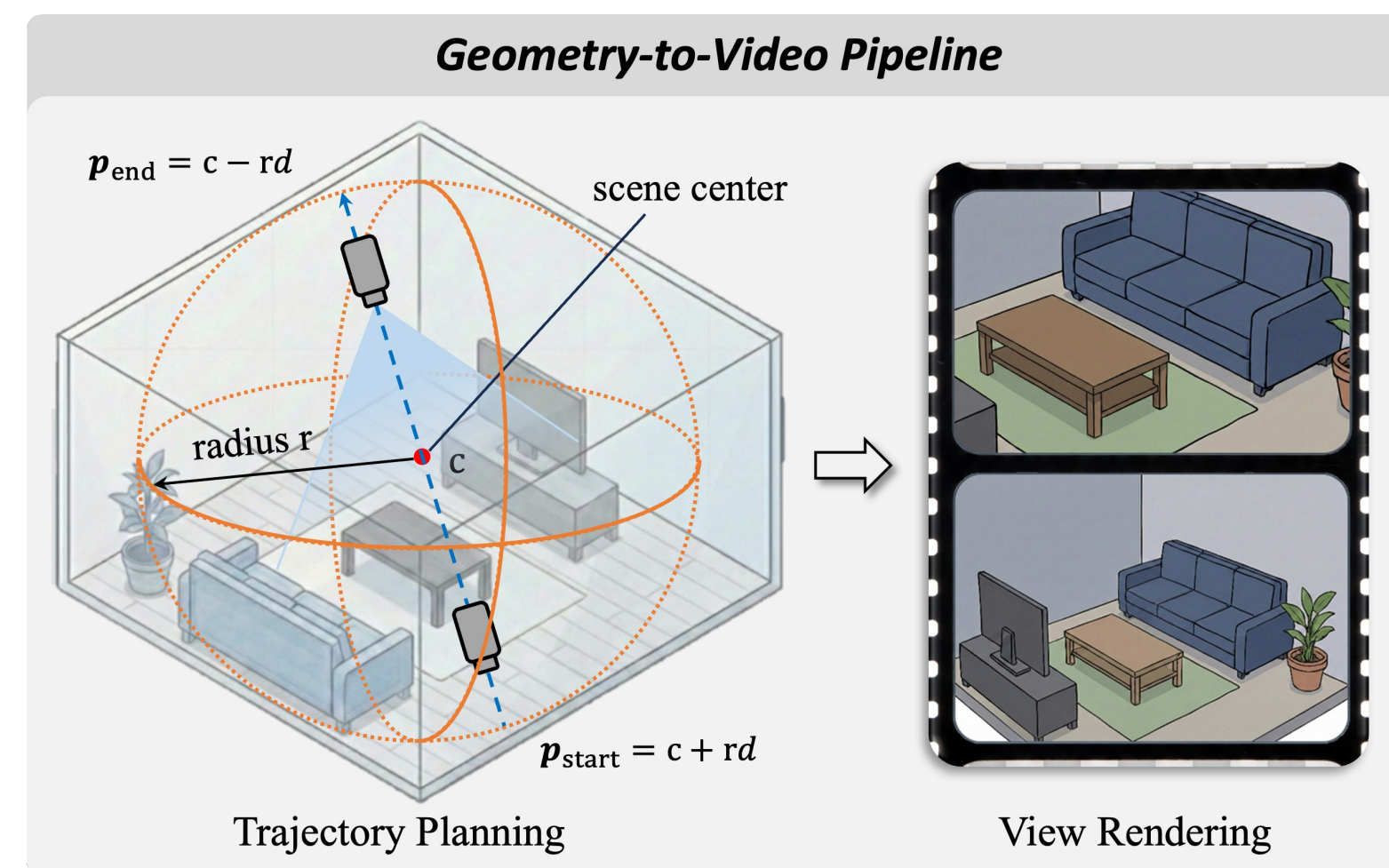
☒ *ReRe*: B

 $GT:B$

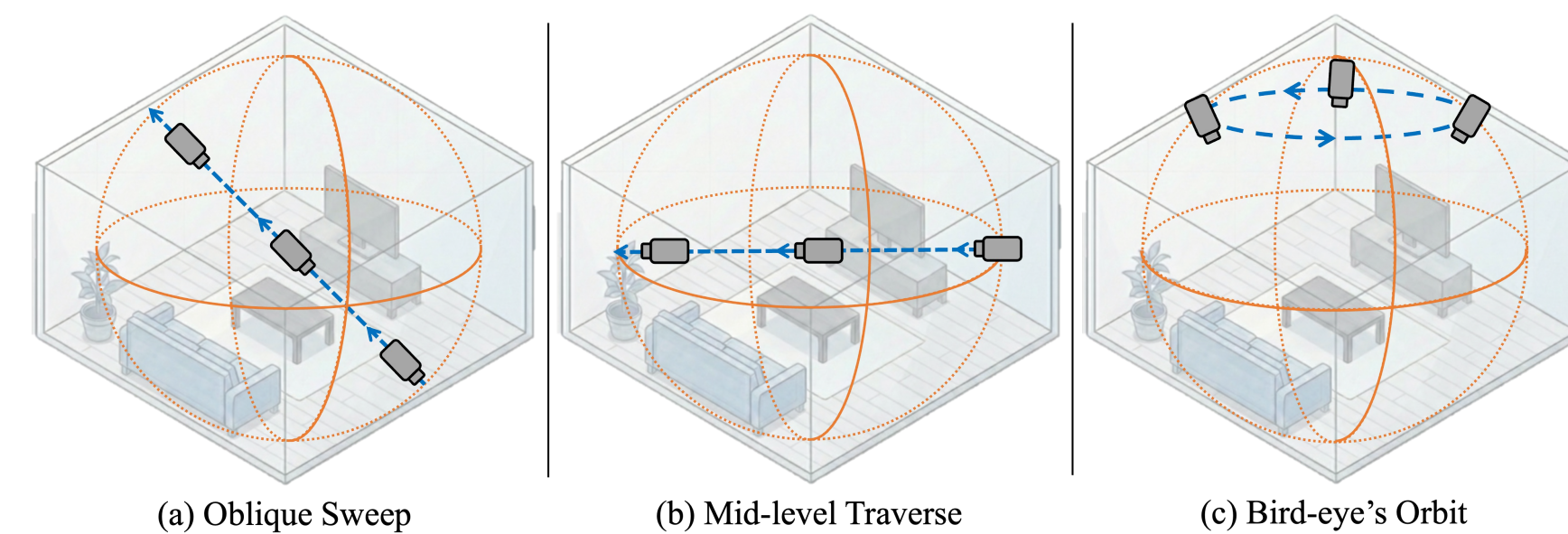
Making 3D Evidence Visible to MLLMs

The synthesized view is designed around two principles.

- **Geometric complementarity** requires a viewpoint that reduces inter-object occlusion and maximizes spatial coverage.
- **Native compatibility** means geometric evidence must present in a visual format that frozen MLLMs can handle it without architectural modifications.



ReRe predicts a 3D point cloud from the egocentric video via VGGT, plans a scene-spanning Oblique Sweep trajectory, and renders the point cloud into temporally coherent video frames via point-based rasterization.



Allocentric trajectory design. The Oblique Sweep follows a diagonal path through the scene center with an elevated tilt. The Mid-level Traverse moves horizontally along the diameter at a fixed elevation, while the Bird's-eye Orbit circles the scene center from a top-down perspective.

Experiments

Performance on VSI-Bench. ReRe boosts spatial reasoning across diverse MLLMs.
See more in our paper.

Methods	Avg.	Obj. Count	Abs. Dist.	Obj. Size	Room Size	Rel. Dist.	Rel. Dir.	Route Plan	Appr. Order
			Numerical Answer				Multiple-Choice Answer		
QWEN2.5-VL-3B	26.4	15.8	23.9	33.4	27.8	20.5	34.4	29.9	21.5
OURS	28.2 ^{+1.8%}	16.7 ^{+0.9%}	25.0 ^{+1.1%}	35.3 ^{+1.9%}	25.3 ^{+2.5%}	31.4 ^{+9.9%}	35.7 ^{+1.3%}	28.9 ^{+1.0%}	17.3 ^{+4.2%}
QWEN2.5-VL-7B	24.8	11.2	12.2	22.8	29.9	33.8	36.0	32.0	24.9
OURS	29.5 ^{+4.7%}	18.0 ^{+6.8%}	18.8 ^{+6.6%}	40.0 ^{+17.2%}	31.1 ^{+1.2%}	34.1 ^{+0.3%}	35.3 ^{+0.9%}	32.5 ^{+0.5%}	22.0 ^{+2.9%}
QWEN3-VL-2B	22.5	14.2	14.7	29.8	10.8	19.9	34.1	23.7	19.4
OURS	31.0 ^{+8.5%}	17.4 ^{+3.2%}	23.4 ^{+8.7%}	50.5 ^{+20.7%}	21.0 ^{+10.2%}	25.8 ^{+5.9%}	36.7 ^{+2.6%}	27.8 ^{+4.1%}	26.2 ^{+6.8%}
QWEN3-VL-4B	30.7	14.8	22.0	41.6	33.5	30.1	38.4	23.7	29.5
OURS	36.5 ^{+5.8%}	21.1 ^{+6.3%}	26.7 ^{+4.7%}	50.2 ^{+8.6%}	35.8 ^{+2.3%}	37.2 ^{+7.0%}	43.1 ^{+4.7%}	28.9 ^{+5.2%}	34.1 ^{+4.7%}
QWEN3-VL-8B	30.5	16.4	20.7	43.0	28.0	36.8	35.1	22.7	26.9
OURS	35.8 ^{+5.2%}	19.5 ^{+3.1%}	25.8 ^{+6.1%}	49.8 ^{+6.7%}	31.3 ^{+3.3%}	39.7 ^{+3.0%}	41.0 ^{+5.9%}	29.4 ^{+6.7%}	33.8 ^{+7.0%}
INTERNVL2.5-4B	31.3	37.6	24.5	37.4	22.7	32.0	33.2	27.8	26.9
OURS	32.3 ^{+1.0%}	35.8 ^{+1.9%}	23.3 ^{+1.2%}	36.9 ^{+0.5%}	23.7 ^{+1.1%}	32.5 ^{+0.6%}	42.1 ^{+8.9%}	30.9 ^{+3.1%}	22.8 ^{+4.0%}
INTERNVL2.5-8B	35.5	22.5	28.4	45.6	35.3	35.6	43.3	32.5	29.9
OURS	36.7 ^{+1.2%}	19.1 ^{+3.4%}	29.7 ^{+1.3%}	46.7 ^{+1.1%}	37.9 ^{+2.6%}	36.5 ^{+0.9%}	46.9 ^{+3.4%}	31.4 ^{+1.1%}	32.0 ^{+2.1%}
INTERNVL3-2B	26.5	42.2	22.8	26.4	17.6	25.2	34.3	25.8	10.8
OURS	29.9 ^{+3.4%}	37.9 ^{+4.3%}	24.5 ^{+1.5%}	27.2 ^{+0.8%}	16.6 ^{+1.0%}	32.0 ^{+6.8%}	43.2 ^{+8.9%}	26.8 ^{+1.0%}	18.3 ^{+7.5%}
INTERNVL3-8B	32.6	40.4	23.3	43.4	30.1	34.9	33.4	30.9	19.3
OURS	35.5 ^{+2.9%}	38.8 ^{+1.6%}	31.1 ^{+7.8%}	44.4 ^{+1.0%}	30.0 ^{+0.1%}	37.2 ^{+2.3%}	37.3 ^{+3.9%}	30.9 ^{+0.0%}	23.6 ^{+4.4%}