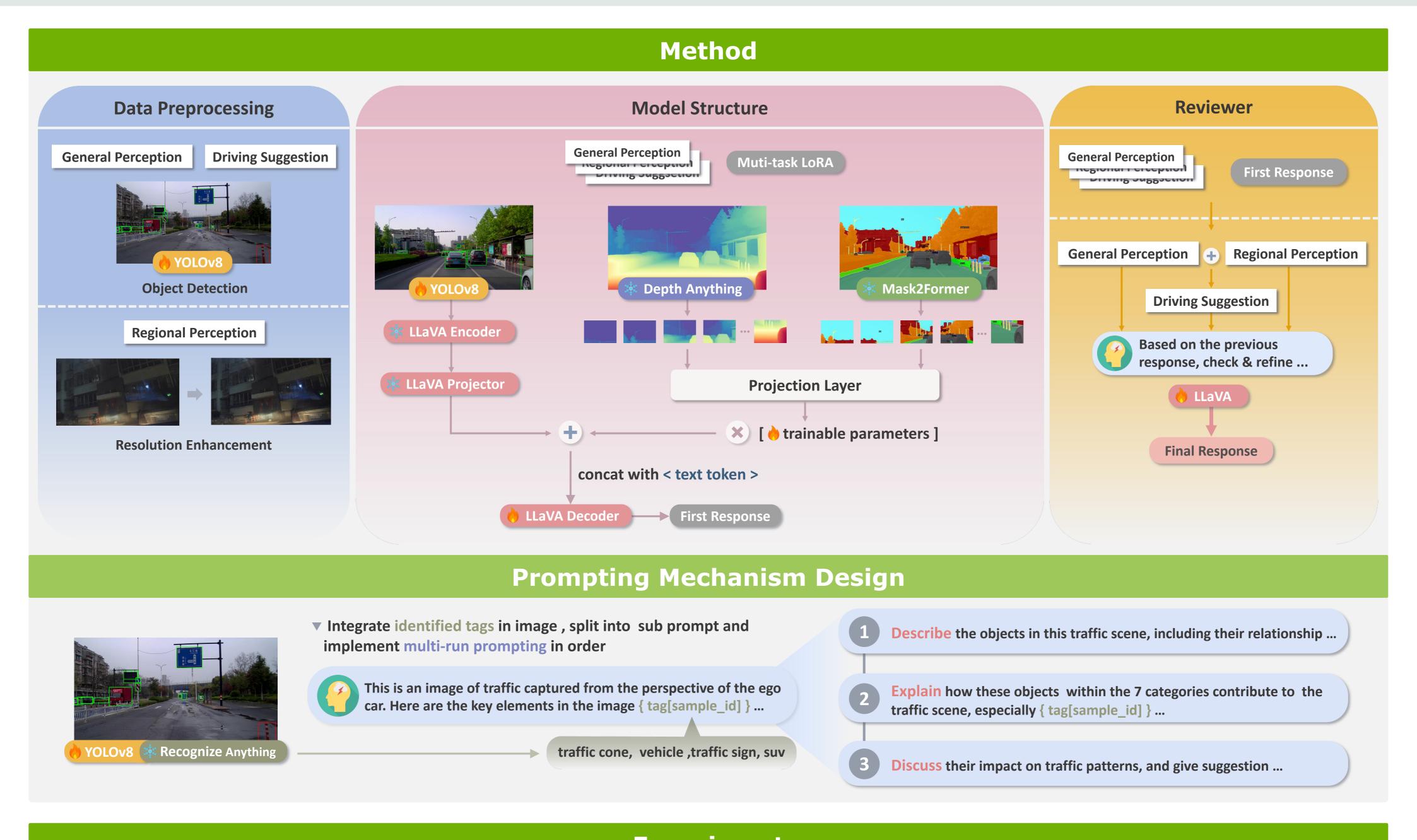
Multimodal Perception and Comprehension of Corner Cases in Autonomous Driving

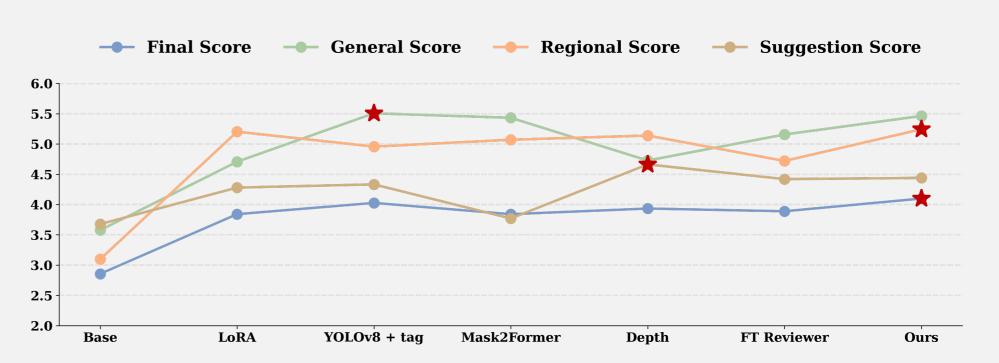
電機碩二 R12921057 林佳儀, 電機碩二 R12921125 楊沛蓉, 電機碩一 R13921093 李彥璋, 網媒碩二 R12944030 簡志宇





Experiment

Id	Setting	General	Regional	Suggestion	LLM judge	Final Score
1	LLaVA (Base)	3.577	3.097	3.677	3.450	2.856
2	Finetune 1 (Multi-Task LoRA)	4.707	5.203	4.280	4.730	3.841
3	2 + YOLOv8 + tag	5.507	4.957	4.333	4.932	4.027
4	2 + Mask2Former	5.433	5.070	3.770	4.758	3.841
5	2 + Depth	4.727	5.140	4.663	4.483	3.934
6	3 w/ Reviewer (Finetuned 1)	5.157	4.720	4.420	4.766	3.889
7	Ours (2 w/ all features)	5.463	5.243	4.440	5.049	4.099



Our two-stage approach significantly improves LLaVa's performance, with the combined method achieving a 44% increase in Gemini score (from 2.856 to 4.099) through enhanced perception and reviewing process.

Idea Key Points

Our proposed framework integrates several advanced components, These innovations collectively contribute to our successful results in our metric. Key highlights include:

- YOLO-based Image Detection:
 - Finetuned YOLO to enhance small object detection and recognition, specially on General Task.
- Resolution Enhancement:
 - Improved image quality to support accurate visual processing, specially on Regional Task.
- Multi-Task Perception Enhancement:
 - Depth Anything for depth estimation and spatial understanding.
 - Mask2Former for precise object and region segmentation.
 - Recognize Anything for robust object and feature identification.
- Lora-based Adaption: Efficient fine-tuning of LLaVa-1.5-7b for seamless integration of multi-modal information.

Ablation Study

- LoRA Rank Effects (8-64): We observed that higher ranks showed decreased performance under constant training time. Extended training periods led to overfitting issues, significantly reducing the model's ability to generalize to new scenarios.
- Handling Missed Scene Factors: The model initially missed critical scene elements like traffic cones. Integration of *Recognize Anything* for automated tagging of important scene factors during inference significantly improved contextual understanding.
- Reviewer Module Findings: The reviewer let the model revisit images to refine its initial response. Interestingly, we found that using a model with lower testing score to train the reviewer resulted in better refinement ability compared to a higher-scoring one. We hypothesize that this is due to lowerscoring models developing stronger revising capabilities by learning to transform low-quality responses into high-quality ones.