

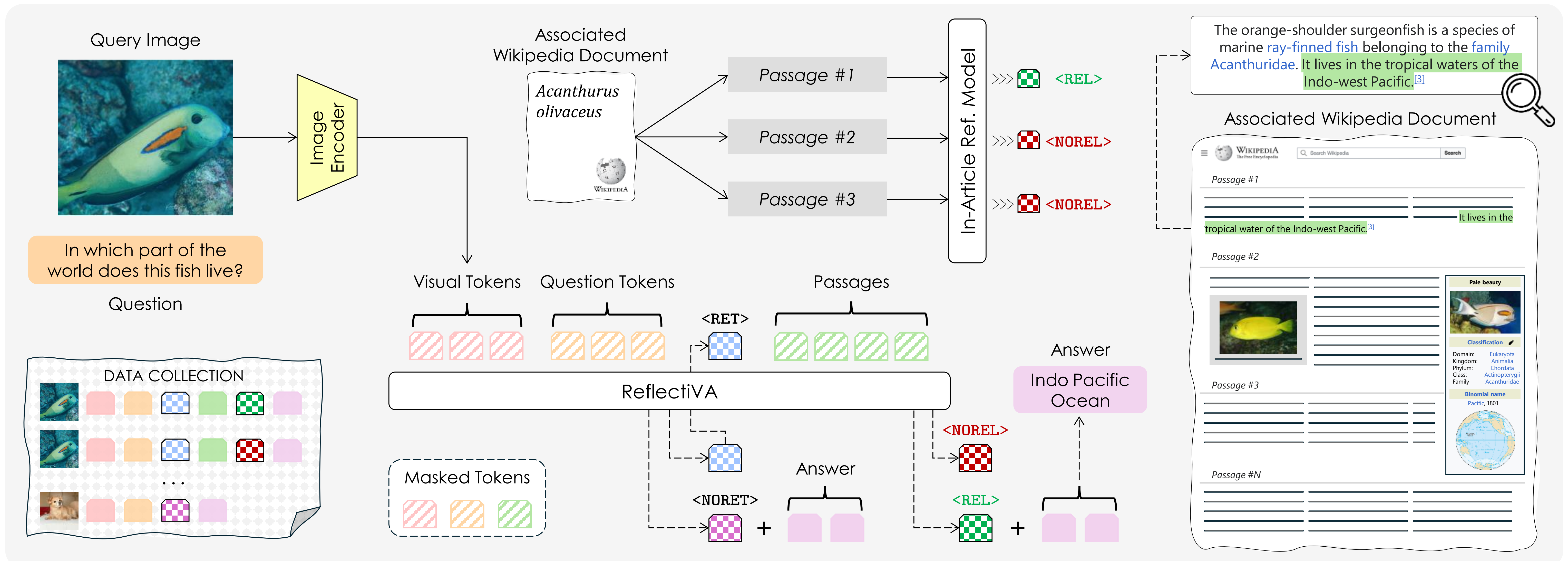
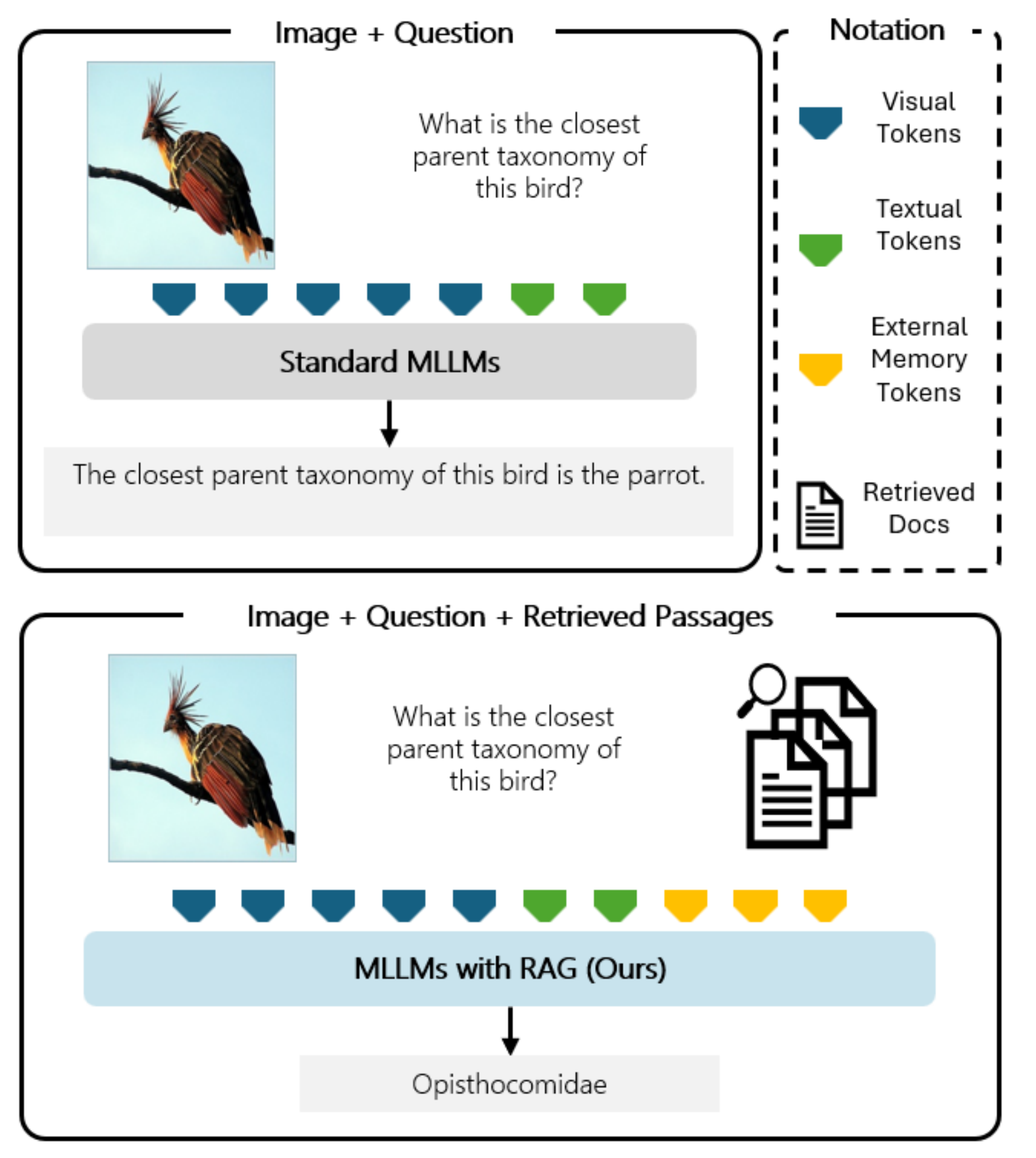
Motivation

- Retrieval-augmented generation boosts LLM performance in specialized domains by integrating external knowledge.
- Highly specific visual questions are often difficult to answer accurately. Leveraging multimodal external knowledge enhances understanding capacity of the model.
- InfoSeek and Encyclopedic VQA use a Wikipedia subset as an external knowledge base.

Our Proposal: ReflectiVA

- Zero-shot models** (LLMs and MLLMs) fall short on complex and specific Visual Question Answering tasks.
- A **multimodal retrieval** pipeline is employed to select the most relevant knowledge passages, starting from a visual query.
- ReflectiVA, based on the image and the question provided, can **independently decide whether retrieving external knowledge** is necessary. Moreover, when external content is used, it can also **evaluate the relevance** of that information — determining whether it is useful for answering the question or should be treated as noise and not considered.
- The model's dictionary has been expanded by adding **four special tokens**: ([RET] - [NORET]) to control the need for retrieval, and ([REL] - [NOREL]) to assess the relevance of the retrieved passage.
- Preservation of the performance** on multimodal datasets that do not require external knowledge.

MLLMs with External Knowledge



Qualitative Results

Q: What is one of the traditional uses of this plant?



Wiki-LLaVA:
Food ✗
EchoSight:
Promote wound healing ✗
ReflectiVA (Ours):
Astringent ✓

Q: Who designed this palace?



Wiki-LLaVA:
Johann Von Fischer ✗
EchoSight:
A team of architects, including Johan Dientzenhofer ✗
ReflectiVA (Ours):
Balthasar Neumann ✓

Q: What is the parent organization of this building?



Wiki-LLaVA:
National Park Service ✗
EchoSight:
National Register of Historic Places ✗
ReflectiVA (Ours):
Colonial Williamsburg Foundation ✓

Q: Which road, railway or canal does this river carry?



Wiki-LLaVA:
Alp Railway ✗
EchoSight:
Railway ✗
ReflectiVA (Ours):
Albula Railway ✓

Comparison with the State of the Art

| | | E-VQA | | InfoSeek | | |
|----------------------------|--------------------|------------|------|----------|----------|------|
| Model | LLM | Single-Hop | All | Unseen-Q | Unseen-E | All |
| Zero-shot | | | | | | |
| Text-only | Vicuna-7B | 2.1 | 2.0 | 0.3 | 0.0 | 0.0 |
| Text-only | LLaMA-3.1-8B | 16.5 | 16.6 | 2.1 | 0.0 | 0.0 |
| LLaVA-v1.5 | Vicuna-7B | 16.3 | 16.9 | 9.6 | 9.4 | 9.5 |
| LLaVA-v1.5 | LLaMA-3.1-8B | 16.0 | 16.9 | 8.3 | 8.9 | 7.8 |
| Retrieval-Augmented Models | | | | | | |
| DPR _{V+T} | Multi-passage BERT | 29.1 | - | - | - | 12.4 |
| RORA-VLM | Vicuna-7B | - | 20.3 | 25.1 | 27.3 | - |
| Wiki-LLaVA | Vicuna-7B | 17.7 | 20.3 | 30.1 | 27.8 | 28.9 |
| Wiki-LLaVA | LLaMA-3.1-8B | 18.3 | 19.6 | 28.6 | 25.7 | 27.1 |
| EchoSight | LLaMA-3.1-8B | 26.4 | 24.9 | 30.0 | 30.7 | 30.4 |
| ReflectiVA (Ours) | LLaMA-3.1-8B | 35.5 | 35.5 | 40.4 | 39.8 | 40.1 |

Preservation on Datasets without Knowledge Base

| Model | LLM | MMMU | MMB (EN) | POPE | SEED-Img | MME (P) | MME (C) | GQA | TextVQA | Science-QA | AI2D |
|--------------------------|---------------------|-------------|-------------|-------------|-------------|---------------|--------------|-------------|-------------|-------------|-------------|
| LLaVA-v1.5 | LLaMA-3.1-8B | 39.4 | 72.4 | 85.1 | 69.8 | 1531.5 | 353.3 | 63.6 | 58.4 | 76.3 | 61.8 |
| Wiki-LLaVA (E-VQA) | LLaMA-3.1-8B | 32.2 | 60.9 | 84.6 | 59.2 | 1350.7 | 306.8 | 56.6 | 49.1 | 67.5 | 55.1 |
| Wiki-LLaVA (InfoSeek) | LLaMA-3.1-8B | 35.9 | 52.0 | 85.7 | 60.5 | 1417.8 | 349.6 | 58.6 | 50.1 | 69.1 | 54.3 |
| ReflectiVA (Ours) | LLaMA-3.1-8B | 38.9 | 69.9 | 85.1 | 68.6 | 1564.5 | 355.7 | 62.1 | 56.8 | 75.4 | 60.6 |

[1] Caffagni et al. *Wiki-LLaVA: Hierarchical Retrieval-Augmented Generation for Multimodal LLMs*. In CVPR W, 2024.[2] Caffagni et al. *The Revolution of Multimodal Large Language Models: A Survey*. In ACL, 2023.[3] Liu et al. *Visual Instruction Tuning*. NeurIPS, 2023.[4] Chen et al. *Can Pre-trained Vision and Language Models Answer Visual Information-Seeking Questions?*. In EMNLP, 2023.[5] Mensink et al. *Encyclopedic VQA: Visual questions about detailed properties of fine-grained categories*. In ICCV, 2023.