

## VLM4Bio: A Benchmark Dataset to Evaluate Pretrained Vision-Language Models for Trait Discovery from Biological Images

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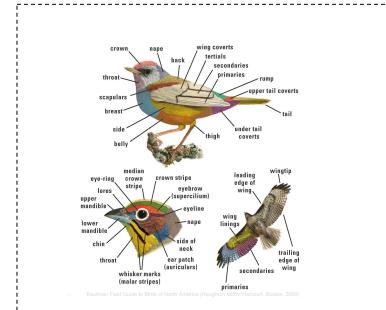






## Discovering Biological Traits from Images

- Large repositories of organism images are available.
  - Museum and university library collection.
  - Citizen science data.
- Biologists are interested in discovering biological traits directly from the organism's images.
- Large Vision-Language Models (VLMs) can solve a diverse range of tasks involving text and images.



Traits are the externally visible characteristics of an organism that define the species. Traits can be certain regions, colors, patterns, or landmark points of the organism.



Can we use Large Vision-Language Models (VLMs) for trait discovery?

Do pre-trained VLMs contain the necessary scientific knowledge to aid biologists in answering a variety of questions pertinent to the discovery of biological traits from images?

### Benchmark datasets for VLMs

- Most of the existing benchmark datasets focus on commonsense knowledge rather than expert knowledge.
- Domain-specific benchmark datasets:
  - MedQA is a collection of VQA problems from medical exams.
  - MathVista mathematical reasoning questions in visual contexts.
  - MMMMU covers college-level problems from diverse business, arts, health, medicine, and engineering domains.

No benchmark dataset exists in the organismal biology domain to evaluate the performance of VLMs in biological tasks.

### VLM4Bio Dataset

 A benchmark dataset of 469K question-answer pairs involving 30K images from three groups of organisms: fishes, birds, and butterflies, covering five biologically relevant tasks.

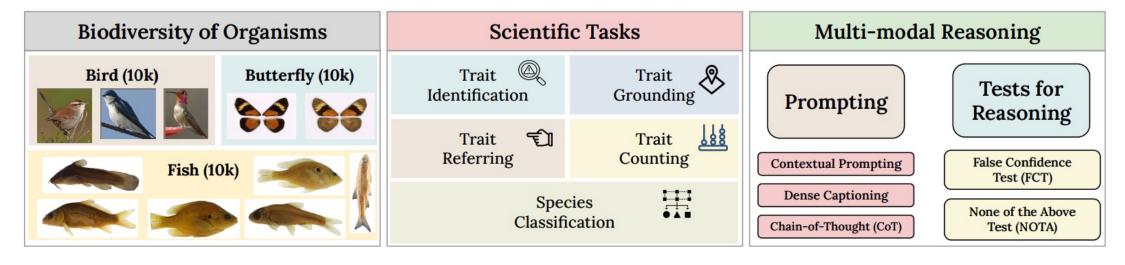


Figure 1: Overview of our goals and contributions. We analyze the capabilities of 12 state-of-the-art (SOTA) vision-language models (VLMs) in answering scientific questions using images from three groups of organisms: fishes, birds, and butterflies, over five groups of biologically relevant tasks. We also explore the effectiveness of these models for reasoning using various prompting techniques and tests for reasoning hallucination.

## Designing Benchmark: Biologically Relevant Tasks

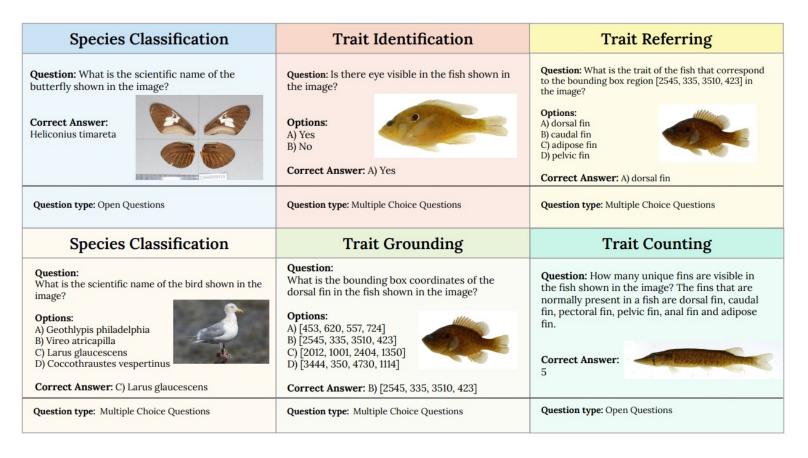


Figure 2: Illustrative examples of VLM4Bio tasks with different question-types.

### Zero-shot Evaluation

								Mo	dels						
Dataset	Question type	gpt-4v	llava v1.5-7b	llava v1.5-13b	-				minigpt4 vicuna-13B				instruct vicuna13B		
	Species Classification														
Fish-10K	Open	1.01	2.32	0.40	0.11	0.01	1.59	0.50	0.38	0.00	1.46	0.00	0.00	0.20	
TISH-TUK	MC	35.91	40.20	32.27	31.72	29.76	33.36	29.02	27.45	30.86	31.70	27.27	26.57	25.00	
Bird-10K	Open	17.40	1.45	2.06	0.86	0.00	0.57	2.80	2.56	0.00	0.50	0.07	0.00	0.53	
Diru-Tork	MC	82.58	50.32	55.36	44.73	33.68	34.75	23.95	27.62	36.36	35.83	44.00	46.55	25.00	
Butterfly-10K	Open	0.04	0.05	0.00	0.01	0.00	0.00	0.07	0.01	0.00	0.00	9.94	0.00	1.54	
Dutterny Tork	MC	28.91	50.24	44.58	36.45	25.14	28.88	33.06	28.90	25.28	36.67	41.70	34.48	25.00	
Trait Identification															
Fish-10K	MC	82.18	56.84	45.15	46.92	68.36	39.33	55.08	51.87	64.34	39.26	81.95	20.69	50.0	
Bird-10K	MC	62.22	34.68	46.14	63.93	50.11	41.38	39.11	40.44	47.89	45.52	77.91	89.98	31.12	
						T	rait Gro	unding							
Fish-500	MC	29.41	24.87	17.98	23.42	23.32	25.14	22.18	25.58	7.20	27.09	33.51	26.90	25.00	
Bird-500	MC	8.1	26.92	35.36	23.2	11.83	10.52	15.39	24.22	3.48	0.81	30.24	13.91	25.00	
						7	rait Ref	erring							
Fish-500	MC	28.15	27.07	29.14	28.19	24.93	25.68	39.24	31.21	31.75	25.78	28.04	32.73	25.00	
Bird-500	MC	42.28	30.5	29.64	18.45	35.16	40.59	26.04	35.88	27.52	41.69	23.03	22.69	25.00	
						7	Trait Co	unting							
Fish-500	Open	16.4	47.4	52.0	14.8	37.6	63.4	13.6	31.53	50.2	61.4	61.4	0.0	25.00	
F 1511-500	MC	44.80	13.20	54.80	21.00	64.8	78.2	22.00	25.00	74.0	69.4	15.80	11.80	25.00	
Overa	u	34.24	29.0	31.78	25.27	28.91	30.24	23.0	25.19	28.49	29.79	33.92	23.31	22.03	

Table 2: Zero-shot accuracy comparison of VLM baselines (in % ranging from 0 to 100) for the five scientific tasks. Results are color-coded as **Best**, **Second best**, **Worst**, **Second worst**.

### Zero-shot Evaluation

Models														
Dataset	Question type	gpt-4v	llava v1.5-7b	llava v1.5-13b	cogvlm chat			minigpt4 vicuna-7B	minigpt4 vicuna-13B			instruct vicuna7B	instruct vicuna13B	Random Choice
						Spe	cies Clas	ssification						
Fish-10K	Open	1.01	2.32	0.40	0.11	0.01	1.59	0.50	0.38	0.00	1.46	0.00	0.00	0.20
risn-10K	MC	35.91	40.20	32.27	31.72	29.76	33.36	29.02	27.45	30.86	31.70	27.27	26.57	25.00
Bird-10K	Open	17.40	1.45	2.06	0.86	0.00	0.57	2.80	2.56	0.00	0.50	0.07	0.00	0.53
Biru-10K	MC	82.58	50.32	55.36	44.73	33.68	34.75	23.95	27.62	36.36	35.83	44.00	46.55	25.00
Butterfly-10K	Open	0.04	0.05	0.00	0.01	0.00	0.00	0.07	0.01	0.00	0.00	9.94	0.00	1.54
Butterny-10K	MC	28.91	50.24	44.58	36.45	25.14	28.88	33.06	28.90	25.28	36.67	41.70	34.48	25.00
						Tr	ait Ident	ification						
Bird-10K	MC	62.22	34.68	46.14	63.93	50.11	41.38	39.11	40.44	47.89	45.52	77.91	89.98	31.12

- 1. All VLMs show poor accuracy on open questions than MC Questions.
- 2. Bird dataset shows better accuracy than Fish or Butterfly datasets.

Table 2: Zero-shot accuracy comparison of VLM baselines (in % ranging from 0 to 100) for the five scientific tasks. Results are color-coded as **Best**, **Second best**, **Worst**, **Second worst**.

### Zero-shot Evaluation

		Models												
Dataset	Question type	gpt-4v	llava v1.5-7b	llava v1.5-13b	cogvlm chat	BLIP flan-xl	BLIP flan-xxl	minigpt4 vicuna-7B	minigpt4 vicuna-13B	instruct flant5xl	instruct flant5xxl	instruct vicuna7B	instruct vicuna13B	Random Choice

**Species Classification** 

- 1. Most VLMs perform well on the task of Trait Identification task.
- 2. There is a significant drop in the accuracy of trait grounding and referring tasks compared to the trait identification task.

Dutterny-Tork	MC	28.91	50.24	44.58	36.45	25.14	28.88	33.06	28.90	25.28	36.67	41.70	34.48	25.00	
	Trait Identification														
Fish-10K	MC	82.18	56.84	45.15	46.92	68.36	39.33	55.08	51.87	64.34	39.26	81.95	20.69	50.0	
Bird-10K	MC	62.22	34.68	46.14	63.93	50.11	41.38	39.11	40.44	47.89	45.52	77.91	89.98	31.12	
Trait Grounding															
Fish-500	MC	29.41	24.87	17.98	23.42	23.32	25.14	22.18	25.58	7.20	27.09	33.51	26.90	25.00	
Bird-500	MC	8.1	26.92	35.36	23.2	11.83	10.52	15.39	24.22	3.48	0.81	30.24	13.91	25.00	
						T	rait Refe	erring							
Fish-500	MC	28.15	27.07	29.14	28.19	24.93	25.68	39.24	31.21	31.75	25.78	28.04	32.73	25.00	
Bird-500	MC	42.28	30.5	29.64	18.45	35.16	40.59	26.04	35.88	27.52	41.69	23.03	22.69	25.00	
						Т	rait Cou	nting							
Fish-500	Open	16.4	47.4	52.0	14.8	37.6	63.4	13.6	31.53	50.2	61.4	61.4	0.0	25.00	
F 1511-500	MC	44.80	13.20	54.80	21.00	64.8	78.2	22.00	25.00	74.0	69.4	15.80	11.80	25.00	
Overali	!	34.24	29.0	31.78	25.27	28.91	30.24	23.0	25.19	28.49	29.79	33.92	23.31	22.03	

Table 2: Zero-shot accuracy comparison of VLM baselines (in % ranging from 0 to 100) for the five scientific tasks. Results are color-coded as **Best**, **Second best**, **Worst**, **Second worst**.

### Trait Identification vs. Trait Detection

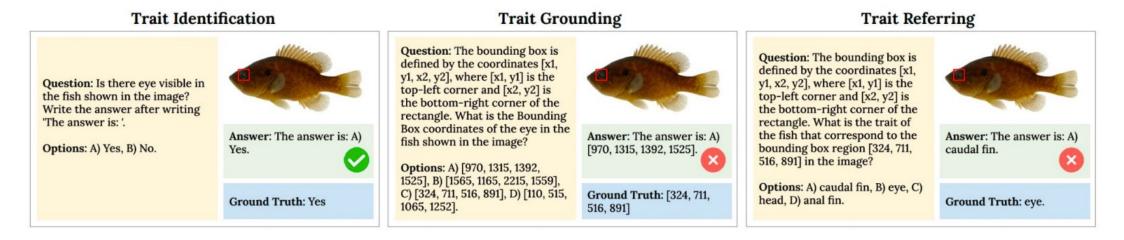


Figure 3: Examples of correct and incorrect predictions of GPT-4V for trait identification, trait grounding, and trait-referring tasks related to the "eye". For visualization assistance, a red-colored bounding box is added around the "eye" in the image.

## Effects of Prompting on VLM Performance

### Three prompting techniques:

#### 1. Contextual Prompting:

- We provided a single-line description of the tasks with the question.
- For example, for species classification task:
  - Each biological species has a unique scientific name composed of two parts: the first for the genus and the second for the species within that genus.

#### 2. Dense Caption Prompting:

- We prompt the VLM to generate a dense caption for the specimen image.
- We add the dense caption before the question and prompt, "Use the above dense caption and the image to answer the following question." to generate responses.

#### 3. Chain-of-Thought (CoT) Prompting:

- We prompt "Let's think step by step" to the VLM to generate the reasoning for a given VQA and multiple choices.
- We then add the reasoning after the VQA and prompt, "Please consider the following reasoning to formulate your answer." to generate the VLM response.

## Effects of Prompting on VLM Performance

3 <del>.</del>					Models			
Dataset	Duamnting	ant Au	ant 1a	llava	llava	cogvlm	BLIP	BLIP
Dataset	Prompting	gpi-4v	gpt-4o	v1.5-7b	v1.5-13b	chat	flan-xl	flan-xxl
	No Prompting	34.40	79.00	41.60	35.40	31.00		22.60
<b>Fish-Prompting</b>	Contextual	30.00	77.20	40.20	35.60	25.60		26.60
	Dense Caption	18.80	78.60	26.00	27.60	32.00		29.80
P	СоТ	42.60	86.00	41.40	34.80	26.80	29.20	24.60
	No Prompting	78.80	97.60	44.20	49.80	45.40	35.60	35.80
<b>Bird-Prompting</b>	Contextual	78.60	98.60	44.00	52.00	49.40	35.60	30.40
	Dense Caption	87.40	97.00	33.40	41.00	44.00	25.60	22.80
	CoT	62.60	98.60	37.40	47.80	42.20	30.60	31.00
	No Prompting	13.20	56.40	27.20	26.80	25.60	24.40	21.20
<b>Butterfly-Prompting</b>	Contextual Dense Caption	9.20 49.60	56.20 63.20	26.00 25.20	24.60 23.80	27.20 27.00		24.60 23.20
	СоТ	63.60	74.60	21.40	23.20	34.60	37.20	23.60

Table 4: Zero-shot accuracy comparison for different prompting techniques of seven VLMs (in % ranging from 0 to 100). Results are color-coded as Best and Worst.

## Tests for Reasoning Hallucination

1. False Confidence Test (FCT)

**Question:** What is the scientific name of the bird in the image?

Options: A) Antrostomus carolinensis, B) Sitta carolinensis, C) Certhia americana, D) Chloroceryle americana.

**Chosen Answer:** Sitta carolinensis

Please provide: 1) Whether the chosen answer is correct (True/False). 2) The correct answer.

**Ground Truth:** Chloroceryle americana



#### Answer:

1) The chosen answer is correct: False.
2) The correct answer: D) Chloroceryle americana.

The bird in the image is the Green Kingfisher, which is scientifically named Chloroceryle americana. 2. None of the Above (NOTA) Test

**Question:** What is the scientific name of the fish in the image?

**Options:** A) Gambusia affinis, B) Noturus gyrinus, C) Phenacobius mirabilis, D) None of the Above.

Write the answer after writing 'The answer is: '



Answer:

The answer is: D) None of the Above

**Ground Truth:** None of the Above

## Tests for Reasoning Hallucination

		Models										
Dataset	Metrics	gpt-4v	gpt-4o	llava v1.5-7b	llava v1.5-13b	cogvlm chat	BLIP flan-xl	BLIP flan-xxl				
False Confidence Test (FCT)												
Fish-Prompting	Accuracy Agreement Score	34.20 4.40	73.60 16.60	25.00 99.80	28.60 19.20	24.60 74.40	0.00	7.00 28.4				
Bird-Prompting	Accuracy Agreement Score	73.40 11.40	99.00 21.00	25.40 93.20	35.80 17.80	19.80 47.80	0.00	20.20 79.80				
<b>Butterfly-Prompting</b>	Accuracy Agreement Score	5.20 2.60	53.40 12.40	27.20 95.40	26.60 5.60	6.20 13.80	0.00	5.00 19.00				
	Noi	ne of the	Above (N	OTA) Test	;							
Fish-Prompting	Accuracy	81.40	44.80	3.40	3.80	0.00	4.00	0.00				
Bird-Prompting	Accuracy	75.00	91.40	1.00	1.20	0.00	31.40	0.00				
<b>Butterfly-Prompting</b>	Accuracy	50.40	4.60	1.00	4.60	0.00	51.00	0.00				

Table 5: Performance of seven VLMs on the NOTA and FCT reasoning tests. Results are color-coded as Best and Worst.

# Thank you for listening.

Please visit us during the poster session.





