

Can Pre-trained Vision and Language Model Answer Visual Information-Seeking Questions?

Yang Chen, Hexiang Hu, Yi Luan, Haitian Sun, Soravit Changpinyo,
Alan Ritter, Ming-Wei Chang

Georgia Institute of Technology

Google Research

Google DeepMind

InfoSeek: A new VQA Benchmark focuses on Visual Information-Seeking Questions to test World Knowledge in MLLMs.

- 1M + auto & 8.9k human annotated
- 10k + entities



Scan the QR code to access our data and paper



Q: What days might I most commonly go to this building?
A: Sunday Previous VQA

Q: Who designed this building?
A: Antonio Barluzzi

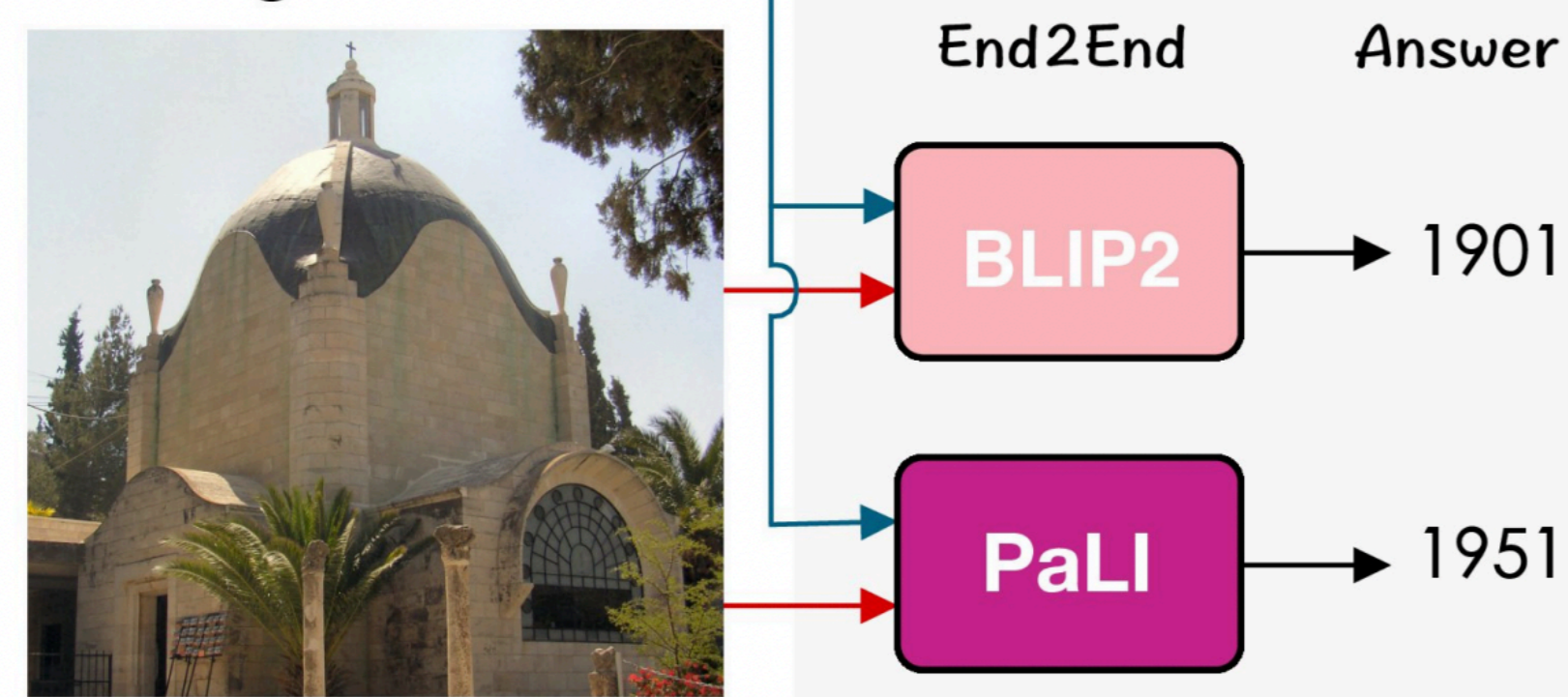
Q: Which year was this building constructed?
A: 1955 INFOSEEK

- OK-VQA: 70.8% questions can be answered by average adults without using a search engine
- InfoSeek: only 4.4%. Query fine-grained information about the visual entity (e.g., Dominus Flevit Church)

1. Goal: Benchmarking Visual World Knowledge in Multimodal LLMs

Which year was this building constructed?

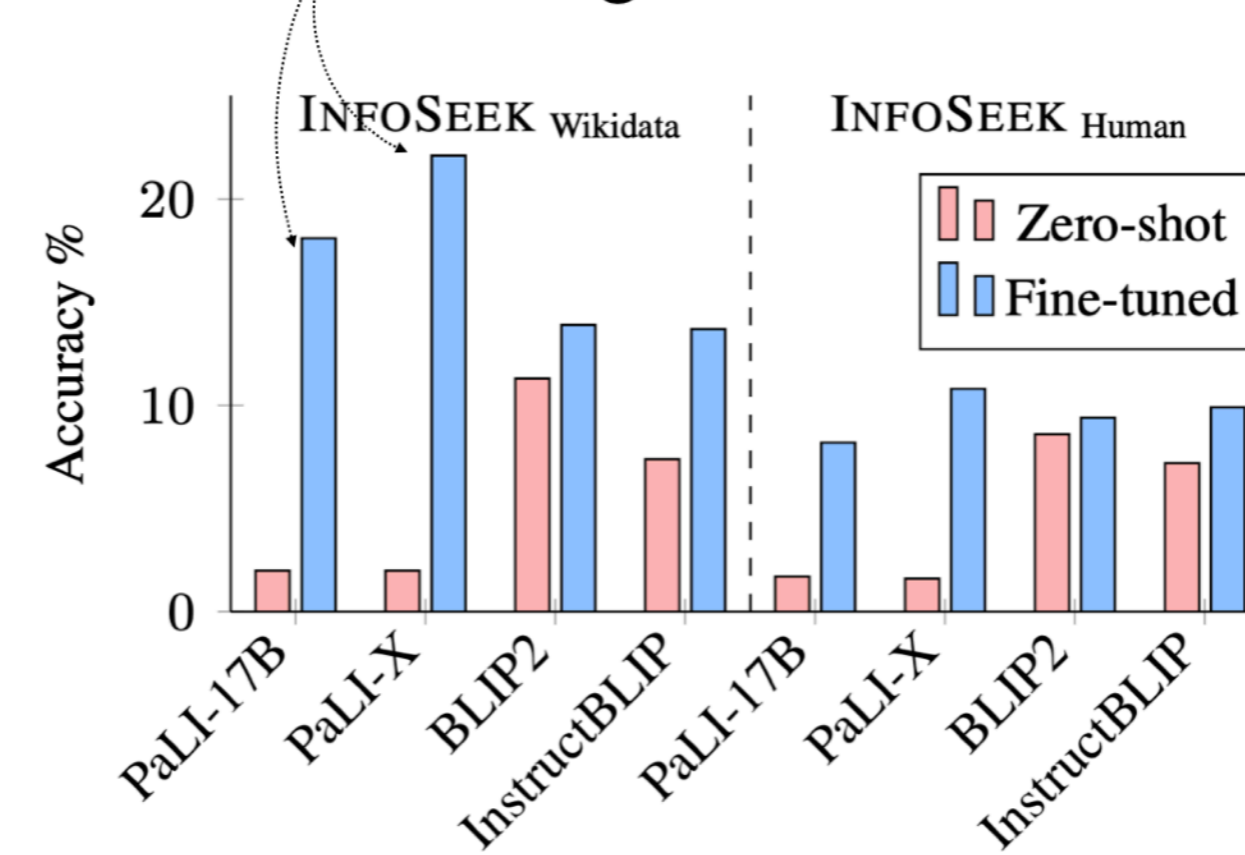
Models with No KB



Model	LLM	# Params	INFOSEEK		
			OK-VQA	VQA _{v2}	
			Overall	Accuracy	Accuracy
Random	-	-	0.1	-	-
Prior	-	-	3.2	-	32.1 ↑
PaLM (Q-only)	PaLM	540B	4.3	23.8	43.0
BLIP2	Flan-T5 _{xxxl}	12B	13.9	54.7	82.3
InstructBLIP	Flan-T5 _{xxxl}	12B	13.7	55.5	-
PaLI-17B	mT5 _{xxxl}	17B	18.1	64.8	84.6
PaLI-X	UL2 _{32B}	55B	22.1	66.1	86.1

PaLI-X on InfoSeek vs other VQA datasets

Eliciting knowledge from PaLI models by fine-tuning on InfoSeek



Instruction-tuned model makes less fine-grained predictions

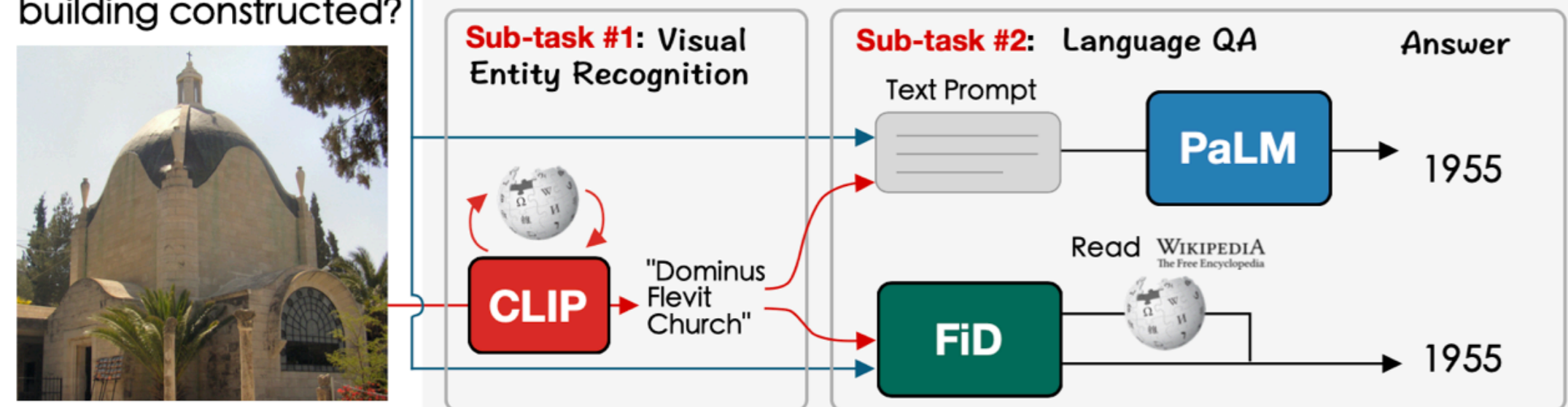


Q: Which body of water is this mountain located in or next to?
A: Lake Como
BLIP2_(0-shot): lake como ✓
InstructBLIP_(0-shot): lake ✗

2. RAG: Retrieval-augmented models (CLIP + PaLM / Fusion-in Decoder)

Which year was this building constructed?

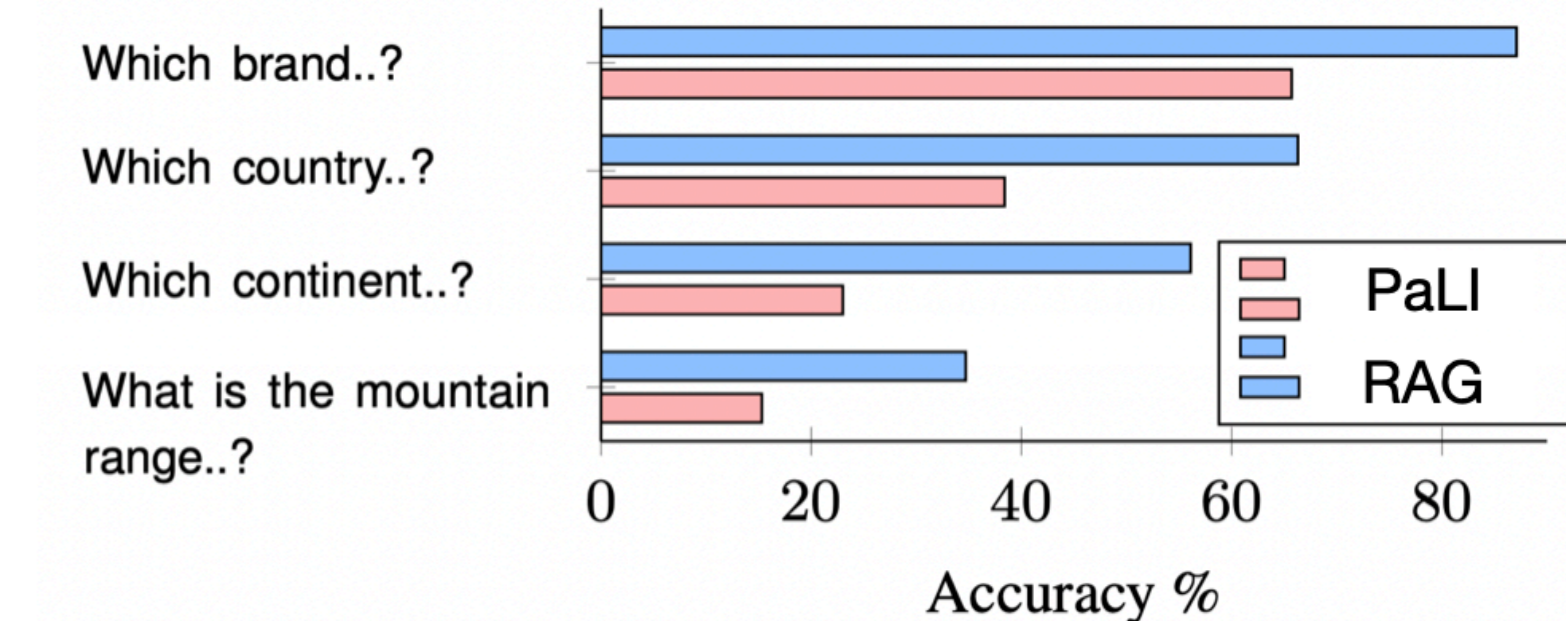
Models with KB



Model	INFOSEEK Wikidata	INFOSEEK Human	ENTITY Accuracy
Best No-KB	22.1	10.8	-
With-KB Setting			
CLIP → PaLM	20.1	15.2	22.2
CLIP → FID	19.3	18.2	-
Oracle → FID	52.0	45.6	100

Big headroom for improvements

PaLI outperforms CLIP + FID on less fine-grained attributes



Model	TIME (Acc.)	NUMERICAL (Relaxed Acc.)	STRING (Acc.)
No-KB Setting			
Prior	0	4.4	5.0
PaLM (Q-only)	0	11.4	4.0
InstructBLIP	7.9	7.5	17.8
BLIP2	6.9	5.8	18.5
PaLI-17B	3.8	18.4	27.4
PaLI-X	7.7	16.1	30.0
With-KB Setting			
CLIP → PaLM	12.5	27.7	21.7
CLIP → FID	12.3	23.4	23.9

Advantages on Time / Numerical questions

Tail Entity



Q: Which region is this animal exclusively found?
A: U.S.A
End-to-end (PaLI): U.S.A ✓
Pipeline (Retrieval): Brazil ✗



Q: What country does this building belong to?
A: Brazil
End-to-end (PaLI): Brazil ✓
Pipeline (Retrieval): United Kingdom ✗



Q: What was the date this type of aircraft took the first flight?
A: 1981 September 26
End-to-end (PaLI): 1955 2 22 ✗
Pipeline (Retrieval): 1981 September 26 ✓



Q: What is the altitude of this building measured at meters above sea level?
A: 2430
End-to-end (PaLI): 3800 ✗
Pipeline (Retrieval): 2430 ✓

3. GPT-4 (Vision) on InfoSeek

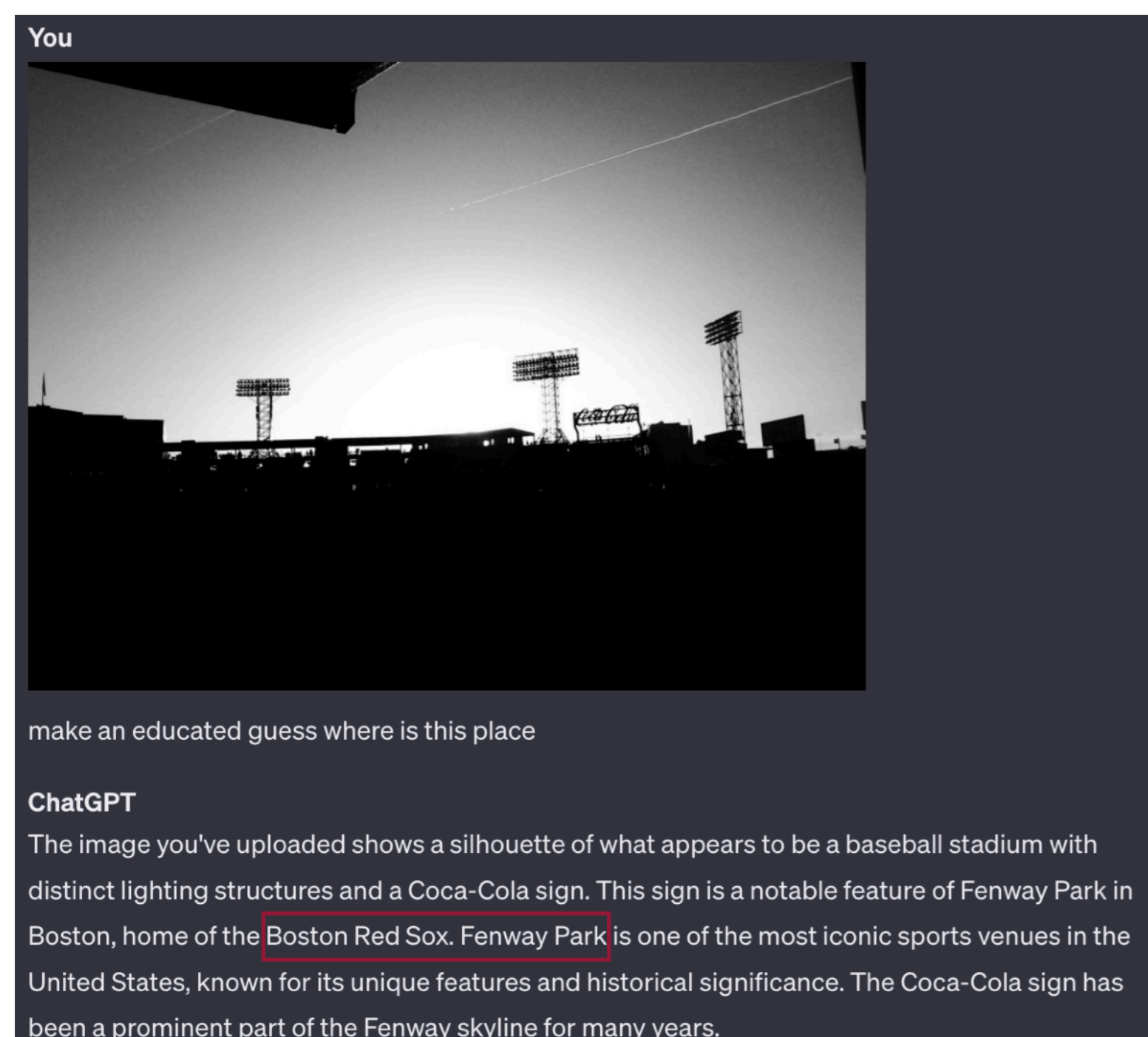
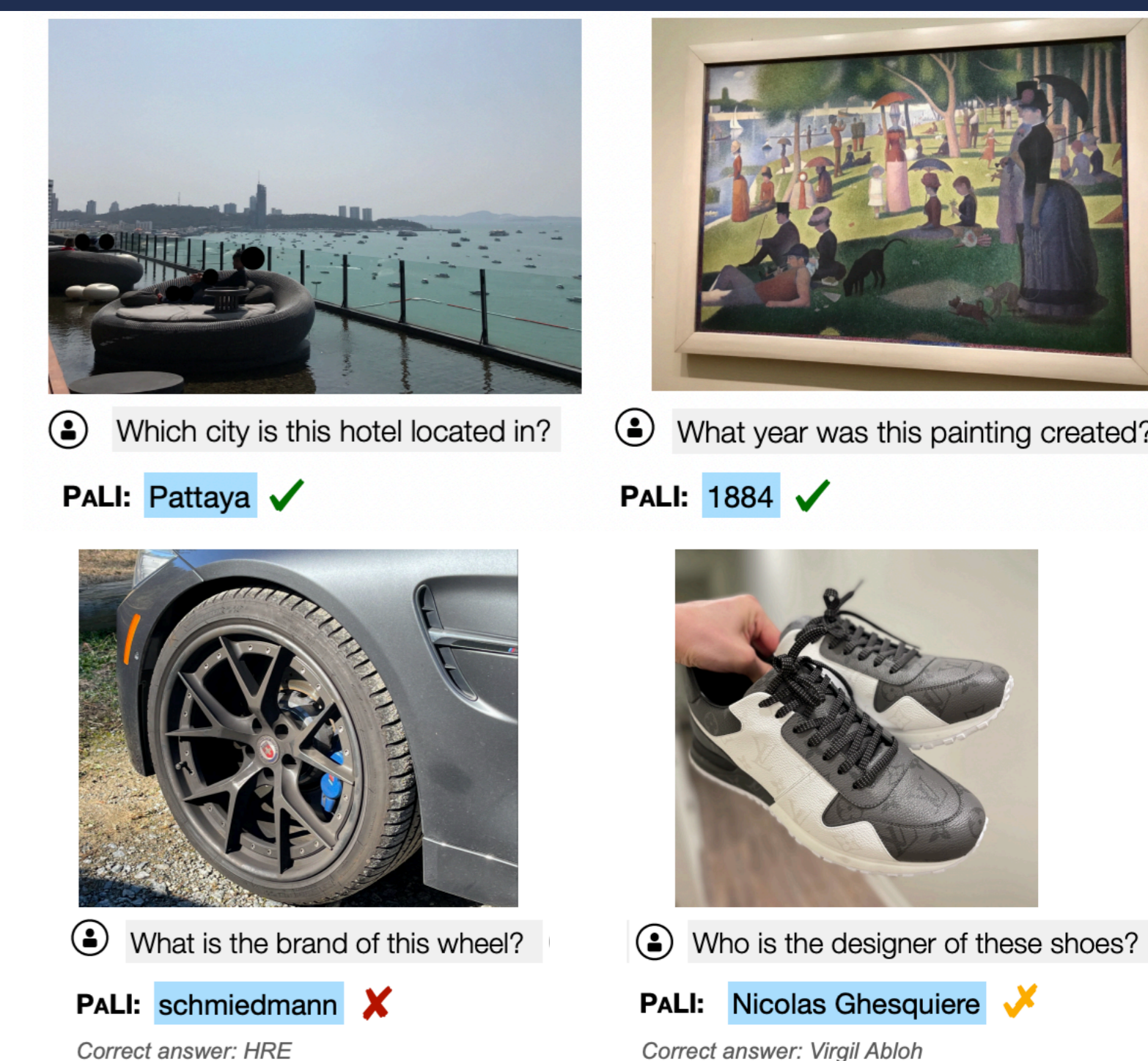


Table 3: Benchmark results on VQA with fine-grained world knowledge. OC refers to Organization and Company. ♣ refers to that the corresponding model is trained using the training set of OK-VQA.

Model	Avg.	Building	Animal	Plant	Location	Food	OC	Facility	vehicle	Objects	Sport	Other
MiniGPT-4 (Vicuna-7b)	10.03	7.33	6.66	5.33	10.0	24.67	4.0	7.33	18.67	6.67	14.0	8.67
MiniGPT-4 (V-4-shot)	5.58	2.67	0.67	4.67	5.33	10.0	4.0	6.0	11.33	4.0	6.0	6.67
BLIP-2 (FlanT5-xxl)	10.67	8.7	2.67	4.0	16.0	14.0	9.33	16.0	28.0	2.0	9.33	7.33
BLIP-2 (Flan-4-shot)	9.44	5.3	2.0	13.33	14.0	8.0	12.0	10.67	28.67	6.67	13.33	6.67
InstructBLIP [♣] (Vicuna-13b)	8.50	3.3	2.0	1.33	10.0	10.67	6.0	4.67	26.67	2.67	20.67	5.33
InstructBLIP [♣] (V-4-shot)	8.75	5.3	2.0	2.0	8.0	15.33	4.67	5.33	16.67	5.33	24.67	6.67
InstructBLIP [♣] (FlanT5-xxl)	8.37	4.0	5.33	2.0	8.67	8.0	8.0	8.0	28.0	5.34	8.67	6.0
InstructBLIP [♣] (Flan-4-shot)	5.65	1.3	2.0	1.33	2.67	18.67	7.33	2.67	15.33	4.67	2.67	3.33
llava-v1.5-13b [♣] (0-shot)	10.22	11.33	16.67	0.0	24.67	6.0	0.7	10.67	26.0	5.3	0.13	10.0
llava-v1.5-13b [♣] (4-shot)	14.73	10.67	4.0	6.67	16.0	22.67	16.0	4.67	32.67	7.33	23.33	18.0
GPT-4V (0-shot)	26.62	18.67	10.28	12.60	17.44	46.67	19.33	29.33	33.33	17.69	57.33	22.53
GPT-4V (4-shot)	26.10	20.0	7.47	8.89	23.49	40.0	20.67	30.0	32.67	13.08	58.67	23.24

A Comprehensive Evaluation of GPT-4V on Knowledge-Intensive Visual Question Answering Li et al., 2023, arXiv 2311.07536

4. Human Evaluation



Georgia Tech

Yang Chen

yang.chen@cc.gatech.edu
@ychenlp

Alan Ritter

alan.ritter@cc.gatech.edu
@alan_ritter

Google Research
Google DeepMind

Hexiang Hu
Yi Luan
Haitian Sun
Soravit Changpinyo
Ming-Wei Chang