

<u>SciFlBench</u>: Benchmarking Large Multimodal Models for Scientific Figure Interpretation

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¹University of Cambridge ²The University of Hong Kong ³Google DeepMind







Motivation

- The capabilities of Large Multimodal Models (LMMs) have been demonstrated in many domains
- LMMs have the potential to benefit the scientific domain
- A tool to assist different stages of the scientific process
- Understanding figures is a key component of scientific research
- The capacity of LMMs to understand scientific figures is not well-known



Image generated using DALL-E

OpenAl· (2024)· DALL-E Image Generation Model (Version 3)· OpenAl· Available from <u>https://openai.com/dall-e</u>

SciFIBench (<u>Sci</u>entific <u>F</u>igure <u>I</u>nterpretation <u>Bench</u>mark)

Source datasets: SciCap [1] and ArXivCap [2]

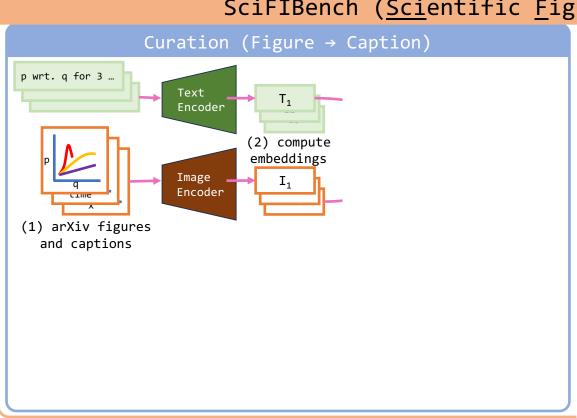
[1] Lei Li, Yuqi Wang, Runxin Xu, Peiyi Wang, Xiachong Feng, Lingpeng Kong, and Qi Liu· Multimodal arxiv: A dataset for improving scientific comprehension of large vision-language models· arXiv preprint arXiv:2403·00231, 2024·

SciFIBench (Scientific Figure Interpretation Benchmark)

Curation (Figure \rightarrow Caption)
p wrt. q for 3
(1) arXiv figures and captions

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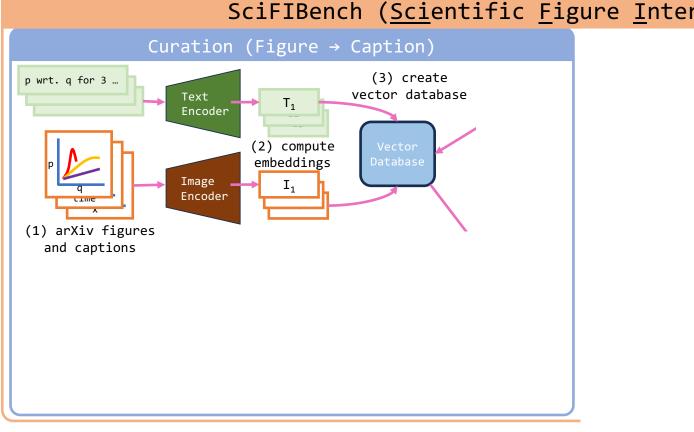
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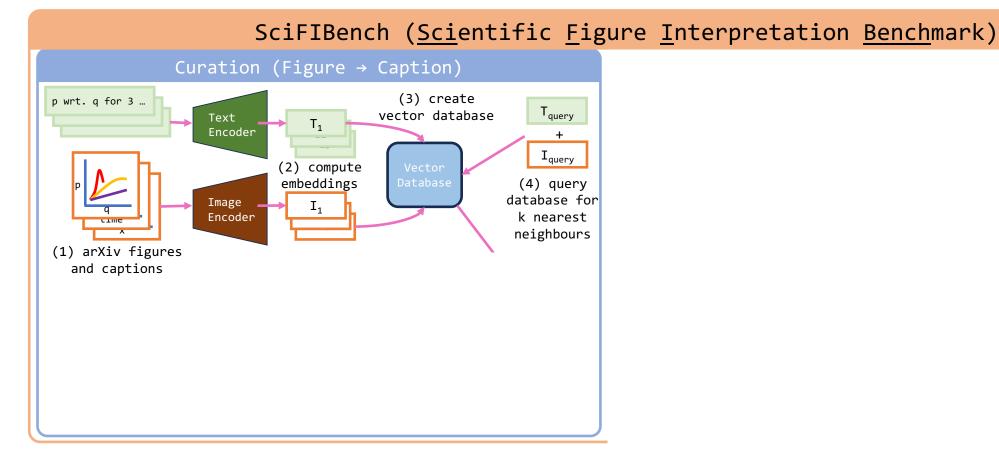
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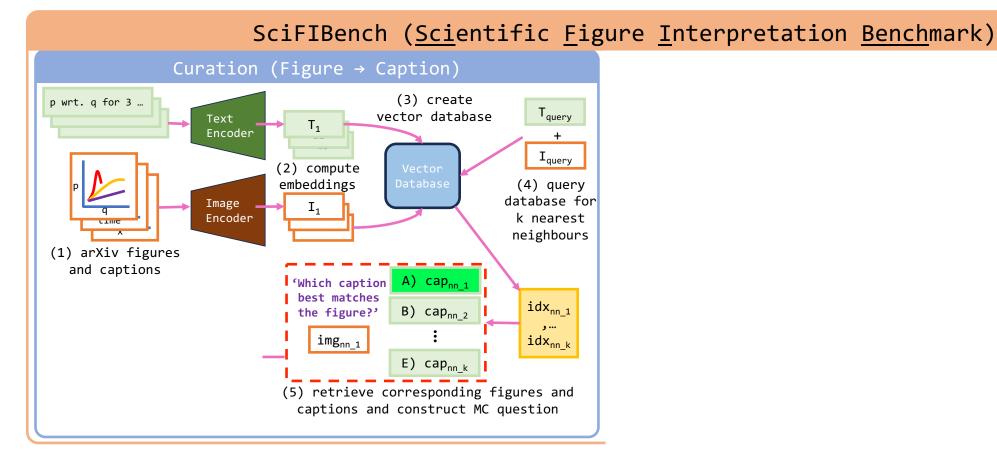
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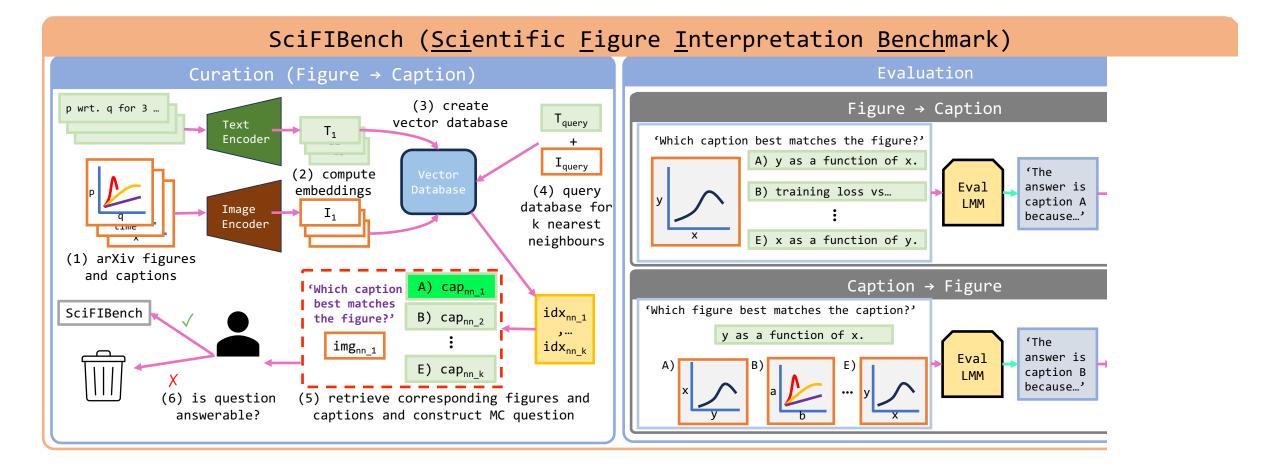
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SciFIBench (Scientific Figure Interpretation Benchmark) Curation (Figure \rightarrow Caption) (3) create p wrt. q for 3 ... Tquery vector database Text T_1 Encoder $\mathtt{I}_{\mathsf{query}}$ (2) compute embeddings (4) query database for Image I_1 k nearest Encoder neighbours (1) arXiv figures and captions Which caption A) capna best matches idx_{nn_1} SciFIBench B) cap_{nn_2} the figure?' img_{nn_1} idx_{nn k} E) cap_{nn k} Х (6) is question (5) retrieve corresponding figures and answerable? captions and construct MC question

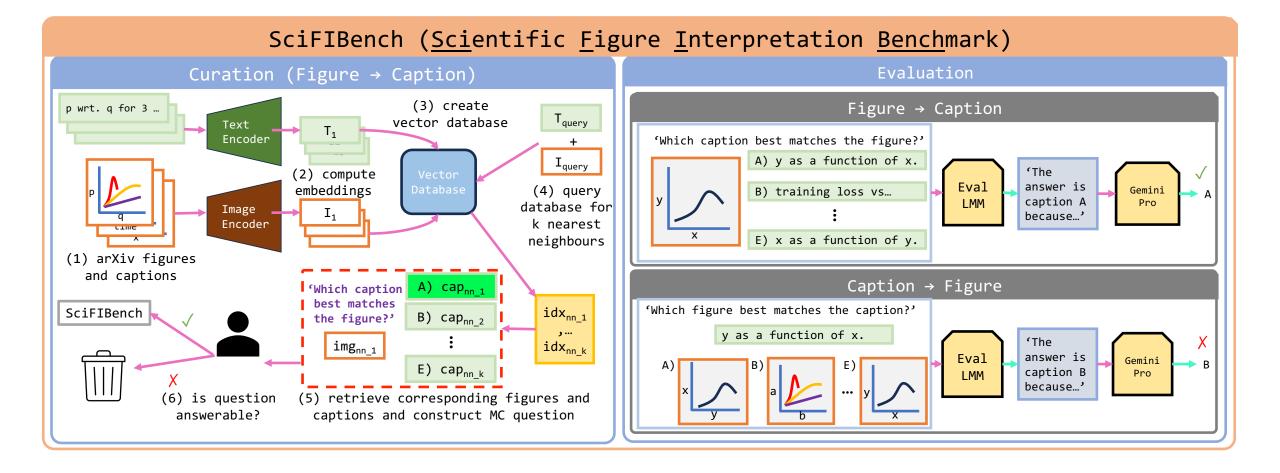
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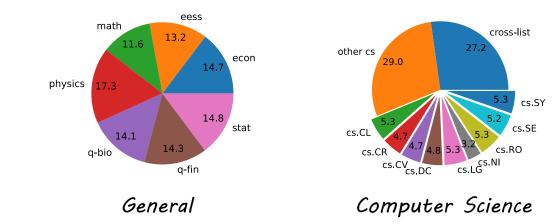


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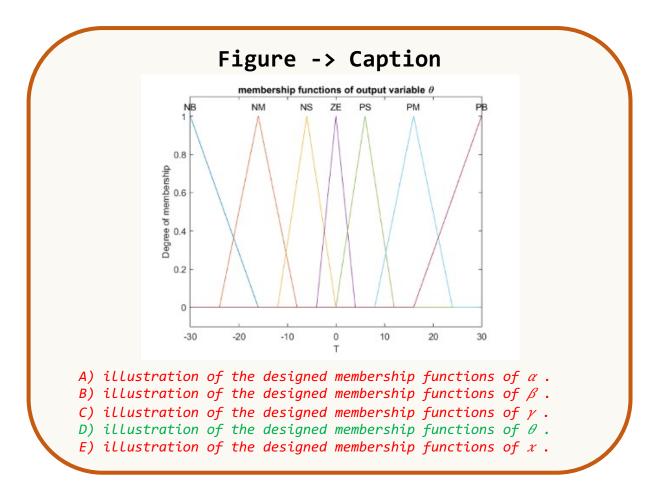
Statistics & Examples

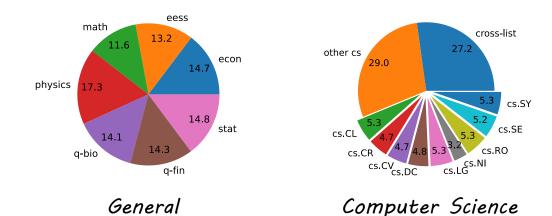
- 2 tasks
- 2000 questions
- 2 subsets: General and Computer Science



Statistics & Examples

- 2 tasks
- 2000 questions
- 2 subsets: General and Computer Science





Statistics & Examples

Figure -> Caption

NS

membership functions of output variable θ

ZE

0

A) illustration of the designed membership functions of α . B) illustration of the designed membership functions of β . C) illustration of the designed membership functions of γ . D) illustration of the designed membership functions of θ .

E) illustration of the designed membership functions of x.

PS

PM

20

10

30

- 2 tasks
- 2000 questions

8.0 0.0 pership

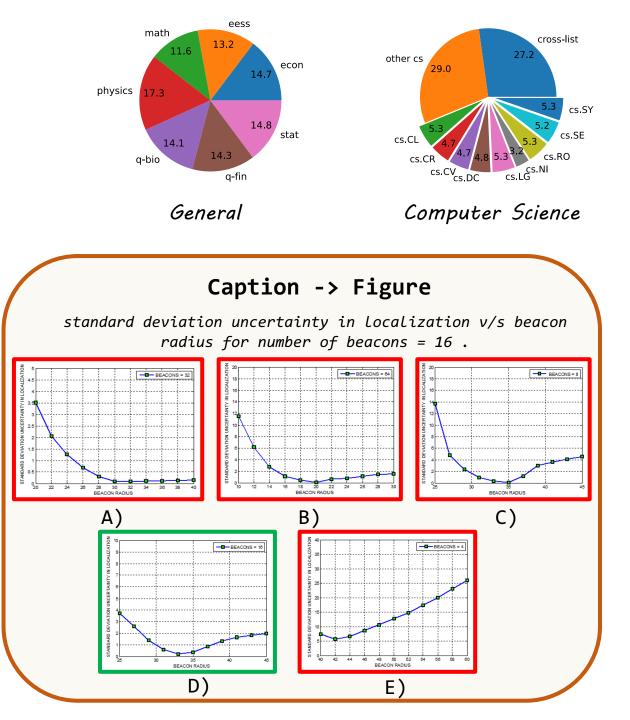
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• 2 subsets: General and Computer Science



		CS	Gen	eral	Ove	erall
Model	Fig.→Cap.		Fig.→Cap.	Cap.→Fig.	Fig.→Cap.	Cap.→Fig.
	~ .	Closed-source L	MMs		<u> </u>	
GPT-4V 5	69.4	58.4	-	-	-	-
GPT-4 Turbo 52	68.0	60.6	62.8	55.2	65.4	57.9
GPT-40 31	75.4	72.2	72.2	58.6	73.8	65.4
Gemini Pro Vision 7	56.0	52.4	50.6	39.6	53.3	46.0
Gemini 1.5 Pro 32	74.0	76.0	65.2	56.2	69.6	66.1
Gemini 1.5 Flash 32	74.4	69.6	65.8	62.4	70.1	66.1
Claude 3 Haiku 53	52.6	43.8	52.6	33.0	52.6	38.4
Claude 3 Sonnet 53	53.4	58.4	53.6	55.0	53.5	56.7
Claude 3 Opus 53	59.8	49.2	50.8	47.4	55.3	48.3
		Open-source L				
IDEFICS-9b-Instruct 54	20.6	20.2	17.6	12.6	19.1	16.4
IDEFICS-80b-Instruct 54	20.6	24.2	18.4	20.6	19.5	22.4
Qwen-VL-Chat 6	28.0	16.0	17.0	19.2	22.5	17.6
Emu2 55	20.8	-	19.6	-	20.2	-
TransCore-M [56]	51.0		27.4	-	39.2	-
InternLM-XComposer-7b 57	34.0	-	21.6	-	27.8	-
InternLM-XComposer2-7b [57]	28.0	-	23.8	-	25.9	-
CogVLM-Chat 58	40.8	-	24.0	-	32.4	-
OmniLMM-3b 59	35.8	-	24.8	-	30.3	-
OmniLMM-12b 59	34.2	-	27.2	-	30.7	-
Yi-VL-6b 60	41.4	-	27.0	-	34.2	-
Yi-VL-34b 60	32.6	-	21.4	-	27.0	-
InstructBLIP-FlanT5-xl 61	35.8	-	19.0	-	27.4	-
InstructBLIP-FlanT5-xxl [61]	36.2	-	26.8	-	31.5	-
InstructBLIP-Vicuna-7b 61	21.0	-	12.8	-	16.9	-
InstructBLIP-Vicuna-13b [61]	22.2	-	15.6	-	18.9	-
Monkey-Chat 62	27.2		18.2		22.7	-
LLaVA-1.5-7b 63	32.8	-	22.8	-	27.8	-
LLaVA-1.5-13b [63]	25.0	-	20.2	-	22.6	-
	10.000	Text-only inj		3631	1285-55	1213
Gemini-Pro 1.5 Flash [32]	48.0	39.2	51.0	35.8	49.5	37.5
		VLMs				
CLIP ViT-H-14-378-quickgelu 50	41.8	42.6	30.6	30.0	36.2	36.3
MetaCLIP ViT-H-14-quickgelu [65]	36.6	35.4	24.2	25.2	30.4	30.3
Google Multimodal Embedding 66	47.6	54.4	28.2	28.4	37.9	41.4
Nacional Report of Sec.		an (25 questions	s per task*)			
Human ($\mu \pm \sigma$)	86.4 ±8.24	78.4±8.24	-	-	-	-
GPT-40	72.0	76.0	-	-	-	-
Gemini-Pro 1.5	84.0	72.0	-	-	-	-
CLIP ViT-H-14-378-quickgelu	48.0	56.0	-	-	-	-
TransCore-M	36.0	-	-	-	-	-

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Model	Fig.→Cap.		Fig.→Cap.	Cap.→Fig.	Fig.→Cap.	Cap.→Fig.
Sector Sector State		Closed-source I	MMs			
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Gemini 1.5 Pro 32	74.0	76.0	65.2	56.2	69.6	66.1
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InternLM-XComposer2-7b [57]	28.0	-	23.8	-	25.9	-
CogVLM-Chat 58	40.8	-	24.0	-	32.4	-
OmniLMM-3b 59	35.8	-	24.8	-	30.3	-
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Yi-VL-6b [60]	41.4	-	27.0	-	34.2	-
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InstructBLIP-FlanT5-xxl [61]	36.2	-	26.8	-	31.5	-
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		Text-only inj		2022	1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1	1945
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GPT-40	72.0	76.0	-	-	-	-
Gemini-Pro 1.5	84.0	72.0	-	-	-	-
CLIP ViT-H-14-378-quickgelu	48.0	56.0	-	-	-	-
TransCore-M	36.0	-	-	-	-	-

⁻ SciFlBench is a challenging benchmark

	C	S	Gen	eral	Ove	rall
Model	Fig. \rightarrow Cap.		Fig.→Cap.	Cap.→Fig.	Fig.→Cap.	Cap.→Fig.
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Emu2 55	20.8	-	19.6	-	20.2	-
TransCore-M [56]	51.0	-	27.4	-	39.2	-
InternLM-XComposer-7b 57	34.0		21.6	-	27.8	-
InternLM-XComposer2-7b [57]	28.0	-	23.8	-	25.9	-
CogVLM-Chat 58	40.8	-	24.0	-	32.4	-
OmniLMM-3b 59	35.8	-	24.8	-	30.3	-
OmniLMM-12b 59	34.2	-	27.2	-	30.7	-
Yi-VL-6b 60	41.4	-	27.0	-	34.2	-
Yi-VL-34b 60	32.6	-	21.4	-	27.0	(m)
InstructBLIP-FlanT5-xl 61	35.8	-	19.0	-	27.4	-
InstructBLIP-FlanT5-xxl [61]	36.2	-	26.8	-	31.5	-
InstructBLIP-Vicuna-7b 61	21.0	-	12.8	-	16.9	-
InstructBLIP-Vicuna-13b [61]	22.2	2	15.6	-	18.9	-
Monkey-Chat 62	27.2	-	18.2	1.0	22.7	
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LLaVA-1.5-13b [63]	25.0	-	20.2	-	22.6	-
		Text-only inp		22633	1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1. 1	1212
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Human $(\mu \pm \sigma)$	86.4 ±8.24	78.4±8.24	-	-	-	-
GPT-40	72.0	76.0	-	-	-	-
Gemini-Pro 1.5	84.0	72.0	-		-	
CLIP ViT-H-14-378-quickgelu	48.0	56.0	-	-	-	-
TransCore-M	36.0	-	-	-	-	_

- SciFIBench is a challenging benchmark
- Closed-source models perform better

	C	S	Gen	eral	Ove	erall
Model	Fig.→Cap.	Cap.→Fig.	Fig.→Cap.	Cap.→Fig.	Fig.→Cap.	Cap.→Fig.
2,423	(Closed-source L	MMs		<u> </u>	
GPT-4V 5	69.4	58.4	-	1	-	14
GPT-4 Turbo 52	68.0	60.6	62.8	55.2	65.4	57.9
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Emu2 55	20.8	-	19.6	-	20.2	-
TransCore-M [56]	51.0	-	27.4	-	39.2	-
InternLM-XComposer-7b 57	34.0		21.6	-	27.8	
InternLM-XComposer2-7b [57]	28.0	-	23.8	-	25.9	-
CogVLM-Chat 58	40.8	-	24.0	-	32.4	-
OmniLMM-3b 59	35.8	-	24.8	-	30.3	-
OmniLMM-12b 59	34.2	-	27.2	-	30.7	-
Yi-VL-6b [60]	41.4	-	27.0	-	34.2	-
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InstructBLIP-Vicuna-7b [61]	21.0		12.8	-	16.9	-
InstructBLIP-Vicuna-13b 61	22.2	2	15.6		18.9	
Monkey-Chat 62	27.2	-	18.2	10	22.7	-
LLaVA-1.5-7b 63	32.8	-	22.8	-	27.8	
LLaVA-1.5-13b 63	25.0	-	20.2	-	22.6	14 A A A A A A A A A A A A A A A A A A A
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	Huma	n (25 questions	s per task*)			
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GPT-40	72.0	76.0	-	-	-	-
Gemini-Pro 1.5	84.0	72.0	-	-	-	-
CLIP ViT-H-14-378-quickgelu	48.0	56.0	-	-	-	
TransCore-M	36.0	_	-	-		121

- SciFlBench is a challenging benchmark
- Closed-source models perform better
- Caption -> Figure is harder

	(CS	Gen	eral	Ove	erall
Model	Fig.→Cap.	$$ Cap. \rightarrow Fig.	Fig.→Cap.	Cap.→Fig.	Fig.→Cap.	Cap.→Fig.
2.17.1		Closed-source I		- I - B	- 8 · I.	to Ito a Bo
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Emu2 55	20.8	-	19.6	-	20.2	-
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Yi-VL-6b 60	41.4	-	27.0	-	34.2	-
Yi-VL-34b 60	32.6	-	21.4	-	27.0	-
InstructBLIP-FlanT5-x1 61	35.8	-	19.0	-	27.4	-
InstructBLIP-FlanT5-xxl 61	36.2		26.8	-	31.5	-
InstructBLIP-Vicuna-7b [61]	21.0		12.8	-	16.9	-
InstructBLIP-Vicuna-13b 61	22.2	-	15.6	-	18.9	-
Monkey-Chat 62	27.2	-	18.2		22.7	-
LLaVA-1.5-7b 63	32.8	-	22.8	-	27.8	-
LLaVA-1.5-13b 63	25.0	-	20.2	-	22.6	-
		Text-only inj	put	39932		1.100
Gemini-Pro 1.5 Flash [32]	48.0	39.2	51.0	35.8	49.5	37.5
		VLMs				
CLIP ViT-H-14-378-quickgelu 50	41.8	42.6	30.6	30.0	36.2	36.3
MetaCLIP ViT-H-14-quickgelu [65]	36.6	35.4	24.2	25.2	30.4	30.3
Google Multimodal Embedding 66	47.6	54.4	28.2	28.4	37.9	41.4
		an (25 questions	s per task*)			
Human ($\mu \pm \sigma$)	86.4±8.24	78.4±8.24	-	-	-	-
GPT-4o	72.0	76.0	-	-	-	-
Gemini-Pro 1.5	84.0	72.0	-	-		-
CLIP ViT-H-14-378-quickgelu	48.0	56.0	-	-	-	-
TransCore-M	36.0	_	121	12	2	-

- SciFIBench is a challenging benchmark
- Closed-source models perform better
- Caption -> Figure is harder
- VLMs remain strong baselines

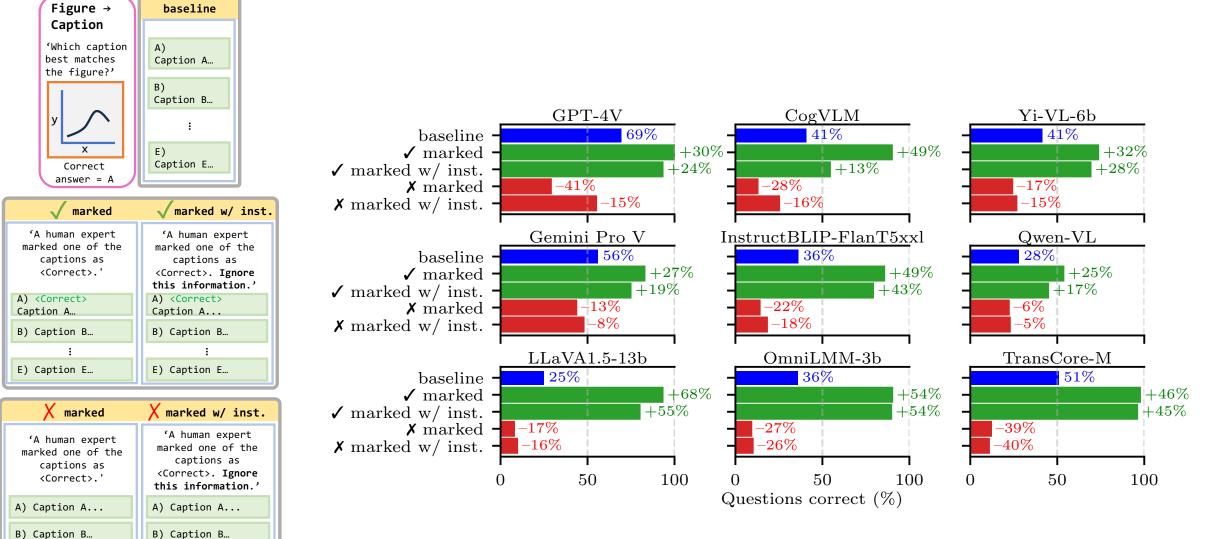
	(CS	Gen	eral	Ove	erall
Model	Fig.→Cap.	Cap.→Fig.	Fig.→Cap.	Cap.→Fig.	Fig.→Cap.	Cap.→Fig.
		Closed-source L			0	
GPT-4V 5	69.4	58.4	-	-	-	-
GPT-4 Turbo 52	68.0	60.6	62.8	55.2	65.4	57.9
GPT-40 31	75.4	72.2	72.2	58.6	73.8	65.4
Gemini Pro Vision [7]	56.0	52.4	50.6	39.6	53.3	46.0
Gemini 1.5 Pro 32	74.0	76.0	65.2	56.2	69.6	66.1
Gemini 1.5 Flash [32]	74.4	69.6	65.8	62.4	70.1	66.1
Claude 3 Haiku 53	52.6	43.8	52.6	33.0	52.6	38.4
Claude 3 Sonnet 53	53.4	58.4	53.6	55.0	53.5	56.7
Claude 3 Opus 53	59.8	49.2	50.8	47.4	55.3	48.3
).	Open-source L	MMs			
IDEFICS-9b-Instruct 54	20.6	20.2	17.6	12.6	19.1	16.4
IDEFICS-80b-Instruct 54	20.6	24.2	18.4	20.6	19.5	22.4
Qwen-VL-Chat 6	28.0	16.0	17.0	19.2	22.5	17.6
Emu2 55	20.8	-	19.6	-	20.2	-
TransCore-M [56]	51.0	-	27.4	-	39.2	-
InternLM-XComposer-7b 57	34.0		21.6	-	27.8	-
InternLM-XComposer2-7b [57]	28.0	-	23.8	-	25.9	-
CogVLM-Chat 58	40.8	-	24.0	-	32.4	-
OmniLMM-3b 59	35.8	-	24.8	-	30.3	(m)
OmniLMM-12b 59	34.2	-	27.2	_	30.7	-
Yi-VL-6b 60	41.4	-	27.0	-	34.2	1.71
Yi-VL-34b 60	32.6	-	21.4	-	27.0	(H)
InstructBLIP-FlanT5-xl 61	35.8	-	19.0	-	27.4	-
InstructBLIP-FlanT5-xxl [61]	36.2	-	26.8	-	31.5	-
InstructBLIP-Vicuna-7b [61]	21.0		12.8	-	16.9	-
InstructBLIP-Vicuna-13b [61]	22.2	-	15.6	_	18.9	-
Monkey-Chat 62	27.2	-	18.2		22.7	-
LLaVA-1.5-7b 63	32.8	-	22.8	-	27.8	-
LLaVA-1.5-13b 63	25.0	-	20.2	-	22.6	-
		Text-only inj		0.640	12223	1.2012
Gemini-Pro 1.5 Flash [32]	48.0	39.2	51.0	35.8	49.5	37.5
		VLMs				
CLIP ViT-H-14-378-quickgelu 50	41.8	42.6	30.6	30.0	36.2	36.3
MetaCLIP ViT-H-14-quickgelu [65]	36.6	35.4	24.2	25.2	30.4	30.3
Google Multimodal Embedding 66	47.6	54.4	28.2	28.4	37.9	41.4
		an (25 questions	s per task*)			
Human ($\mu \pm \sigma$)	86.4 ±8.24	78.4±8.24	-	-	-	-
GPT-40	72.0	76.0	-	-	-	-
Gemini-Pro 1.5	84.0	72.0	-	-	-	-
CLIP ViT-H-14-378-quickgelu	48.0	56.0	-	-	-	-
TransCore-M	36.0	-	-	-	-	-

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	(CS General		eral	Overall	
Model	Fig.→Cap.		Fig.→Cap.	Cap.→Fig.	Fig.→Cap.	Cap.→Fig.
		Closed-source I	MMs			
GPT-4V 5	69.4	58.4	-	-	-	-
GPT-4 Turbo 52	68.0	60.6	62.8	55.2	65.4	57.9
GPT-40 [31]	75.4	72.2	72.2	58.6	73.8	65.4
Gemini Pro Vision [7]	56.0	52.4	50.6	39.6	53.3	46.0
Gemini 1.5 Pro 32	74.0	76.0	65.2	56.2	69.6	66.1
Gemini 1.5 Flash 32	74.4	69.6	65.8	62.4	70.1	66.1
Claude 3 Haiku 53	52.6	43.8	52.6	33.0	52.6	38.4
Claude 3 Sonnet 53	53.4	58.4	53.6	55.0	53.5	56.7
Claude 3 Opus 53	59.8	49.2	50.8	47.4	55.3	48.3
		Open-source L	MMs			
DEFICS-9b-Instruct 54	20.6	20.2	17.6	12.6	19.1	16.4
DEFICS-80b-Instruct 54	20.6	24.2	18.4	20.6	19.5	22.4
Qwen-VL-Chat 6	28.0	16.0	17.0	19.2	22.5	17.6
Emu2 55	20.8	-	19.6	-	20.2	-
TransCore-M 56	51.0	-	27.4	-	39.2	-
InternLM-XComposer-7b 57	34.0		21.6	-	27.8	-
nternLM-XComposer2-7b 57	28.0	-	23.8	-	25.9	-
CogVLM-Chat 58	40.8	-	24.0	-	32.4	-
OmniLMM-3b 59	35.8	-	24.8	-	30.3	-
OmniLMM-12b 59	34.2	-	27.2	-	30.7	-
Yi-VL-6b 60	41.4	-	27.0	-	34.2	-
Yi-VL-34b 60	32.6	-	21.4	-	27.0	-
nstructBLIP-FlanT5-x1 61	35.8	-	19.0	_	27.4	-
nstructBLIP-FlanT5-xxl [61]	36.2		26.8	-	31.5	-
nstructBLIP-Vicuna-7b 61	21.0	· . ·	12.8	-	16.9	-
nstructBLIP-Vicuna-13b 61	22.2	2	15.6	12	18.9	-
Monkey-Chat 62	27.2	-	18.2		22.7	-
LLaVA-1.5-7b 63	32.8	-	22.8	-	27.8	-
LaVA-1.5-13b 63	25.0	2	20.2	121	22.6	-
		Text-only in				
Gemini-Pro 1.5 Flash 32	48.0	39.2	51.0	35.8	49.5	37.5
		VLMs				
CLIP ViT-H-14-378-quickgelu 50	41.8	42.6	30.6	30.0	36.2	36.3
MetaCLIP ViT-H-14-quickgelu [65]	36.6	35.4	24.2	25.2	30.4	30.3
Google Multimodal Embedding	47.6	54.4	28.2	28.4	37.9	41.4
		an (25 questions	s per task*)			
Human ($\mu \pm \sigma$)	86.4±8.24	78.4±8.24	-	-	-	-
GPT-40	72.0	76.0	-	-	2	-
Gemini-Pro 1.5	84.0	72.0	-	-	-	-
CLIP ViT-H-14-378-quickgelu	48.0	56.0	-	-	-	-
FransCore-M	36.0	-	2	_	-	-

- SciFIBench is a challenging benchmark
- Closed-source models perform better
- Caption -> Figure is harder
- VLMs remain strong baselines
- Humans are a stronger baseline
- Multimodality improves performance

Alignment



:

E) <Correct>

Caption E...

:

E) <Correct>

Caption E...



Thanks!

Project page: <u>https://scifibench·github·io/</u> Code: <u>https://github·com/jonathan-roberts1/SciFIBench</u> Data: <u>https://huggingface·co/datasets/jonathan-roberts1/SciFIBench</u> arXiv: <u>https://arxiv·org/abs/2405·08807</u>

Jonathan Roberts University of Cambridge Kai Han The University of Hong Kong Neil Houlsby Google DeepMind

Samuel Albanie





