

A Massive Multi-discipline Multimodal Understanding and Reasoning Benchmark for Expert AGI

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Slides of this talk

CVPR 2024 Oral Presentation

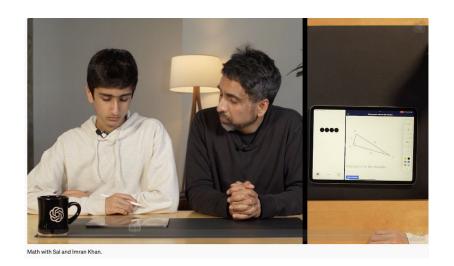
Y - Award Candidate Paper

Progress of Notable ML Models

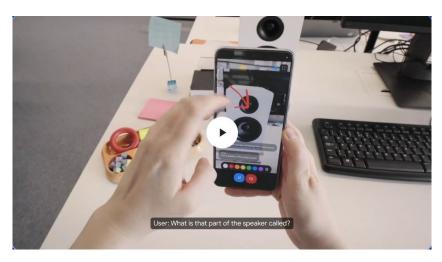
Training compute of notable machine learning models by domain, 2012-23 Source: Epoch, 2023 | Chart: 2024 Al Index report G Gemini Ultra 100B Language 0 Vision Multimodal GPT-4 10B Claude 2 A C PaLM (540B) Llama 2-70B Training compute (petaFLOP - log scale) Megatron-Turing NLG 530B GPT-3 175B (davinci) 100M The most recent W UNIVERSITY of WASHINGTON 10M RoBERTa Large notable ML models are multimodal ones! 1M BERT-Large 100K G Transformer 10K TORONTO "The Al Index 2024 Annual Report," Institute for 1000 AlexNet Human-Centered Al. Stanford University 2012 2013 2014 2015 2016 2017 2018 2019 2020 2021 2022 2023

Publication date

Multimodal Models Empower Real-world Apps

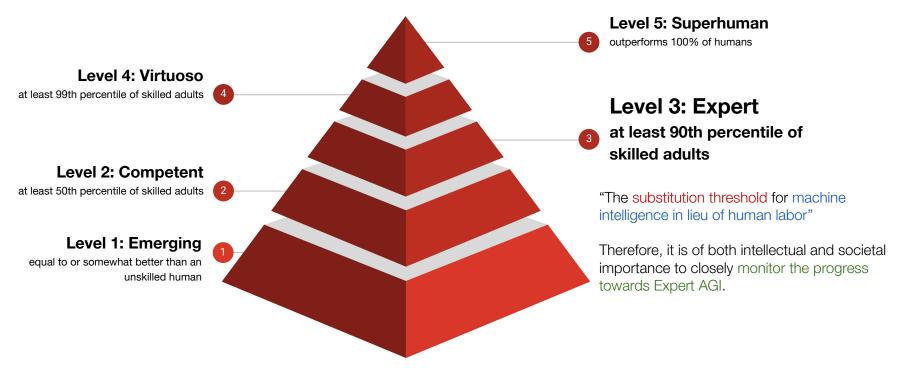


https://openai.com/index/hello-gpt-4o/



https://deepmind.google/technologies/gemini/

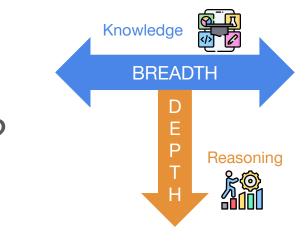
Levels of AGI



Morris, Meredith Ringel, et al. "Levels of AGI: Operationalizing Progress on the Path to AGI." ICML 2024

Measuring Expert AGI?





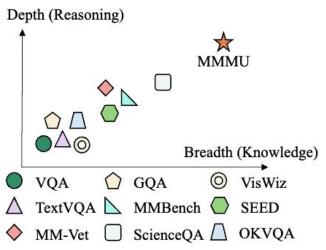
Existing MM Benchmarks



Is this a vegetarian pizza?





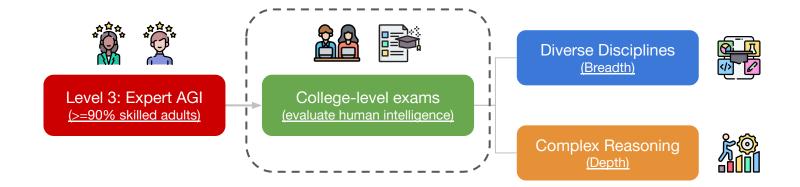




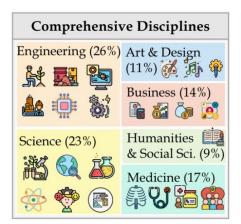


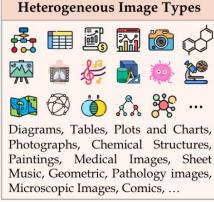


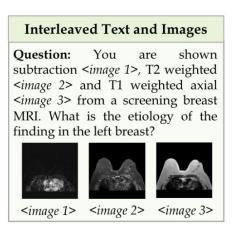
Measuring Expert AGI

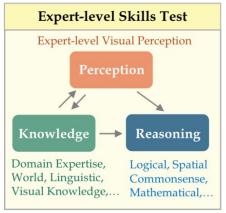


MMMU: multi-discipline multimodal understanding and reasoning











<u>(Breadth)</u>

- 11.5K college-level problems across six broad disciplines and 30 college subjects
- 30 heterogeneous image types

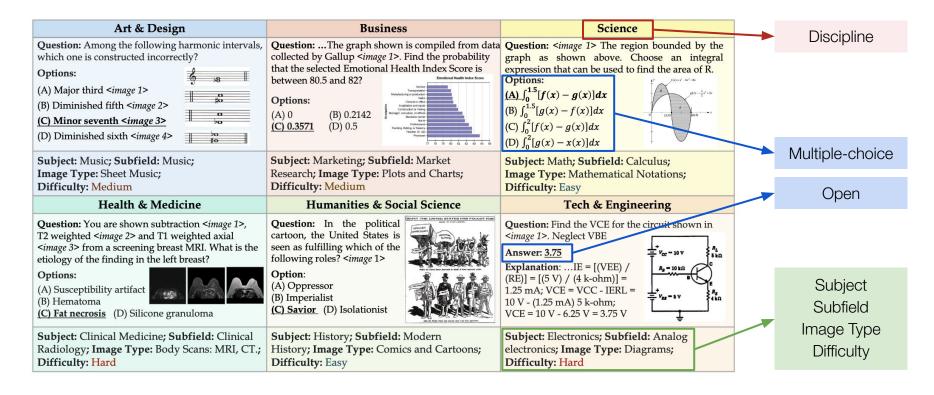


How to improve MM reasoning?

(Depth)

- Interleaved text and (multiple) images
- **Expert-level** perception and reasoning rooted in deep subject knowledge

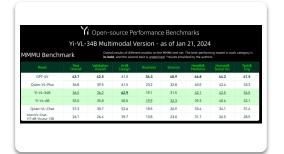
Sampled MMMU examples from each discipline



Introduction Why MMMU? How to improve MM reasoning? Conclusion

Go-to-evaluation by Industrial Developers









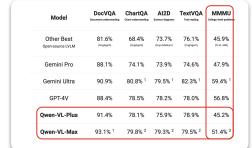










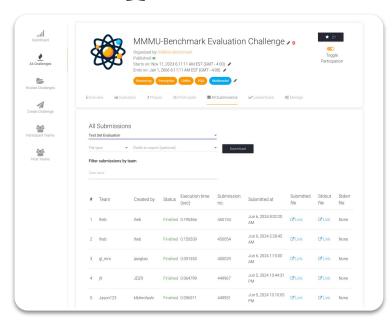




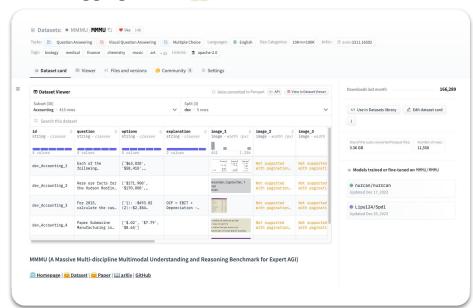


MMMU in the Community

• EvalAl A



Hugging Face



- 130+ Models
- 2500+ Submissions

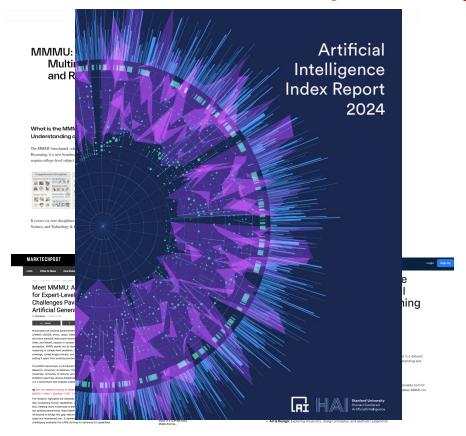


- **160K+** Downloads Last Month
- 700K+ Downloads in Total



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Mentioned by Al blogs, posts, and reports



2.6 Reasoning

General Reasoning

General reasoning pertains to AI systems being able to reason across broad, rather than specific, domains. As part of a general reasoning challenge, for example, an AI system might be asked to reason across multiple subjects rather than perform one narrow task (e.g., playing chess).

MMMU: A Massive Multi-discipline Multimodal Understanding and Reasoning Benchmark for Expert AGI

In recent years, the reasoning abilities of AI systems have advanced so much that traditional benchmarks like SQuAD (for textual reasoning) and VQA (for visual reasoning) have become saturated, indicating a need for more challenging reasoning tests.

Responding to this, researchers from the United States and Canada recently developed MMMU, the Massive Multi-discipline Multimodal Understanding and Reasoning Benchmark for Expert AGI. MMMU comprises about 11,500 college-level questions from six core disciplines; art and design, business, science, health and medicine, humanities and social science, and technology and engineering (Figure 2.6.1). The question formats include charts, maps, tables, chemical structures, and more. MMMU is one of the most demanding tests of perception, knowledge, and reasoning in AI to date. As of January 2024, the highest performing model is Gemini Ultra, which leads in all subject categories with an overall score of 59.4% (Figure 2.6.2).1 On most individual task categories, top models are still well beyond medium-level human experts (Figure 2.6.3). This relatively low score is evidence of MMMU's effectiveness as a benchmark for assessing Al reasoning capabilities.



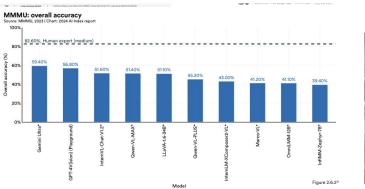
ODAL UNDERSTANDING AND

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APPLICATION

Models

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Why MMMU?

Rigorous Data Curation Process and High-Quality Data



Effective Suite for Tracking Multimodal Model Development



Excellent Model Diagnosis Tool



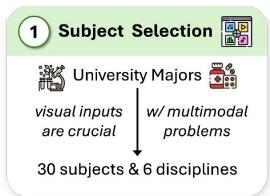
Why MMMU?

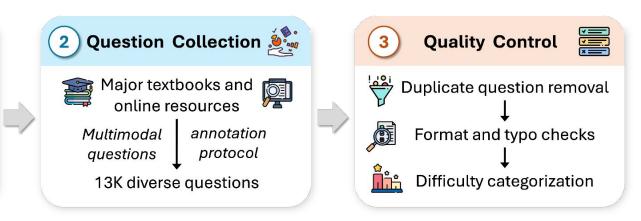
Rigorous Data Curation Process and High-Quality Data



- Effective Suite for Tracking Multimodal Model Development
- Excellent Model Diagnosis Too

Data Curation Pipeline

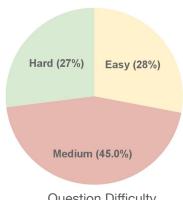




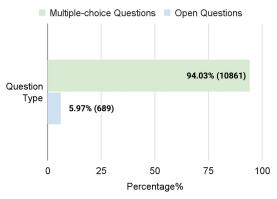
15

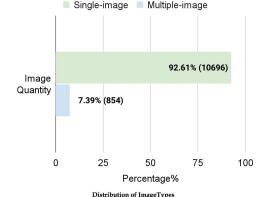
Introduction Why MMMU? How to improve MM reasoning? Conclusion

Statistics of MMMU

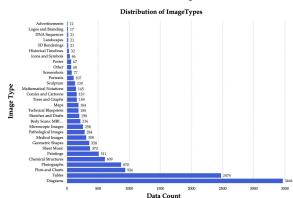












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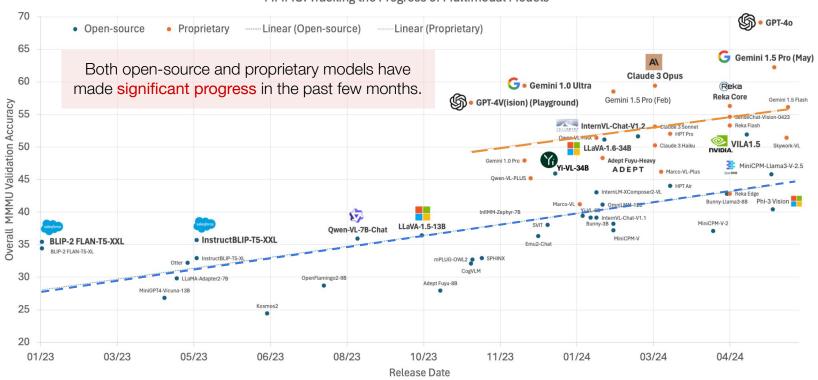
Why MMMU?

- Effective Suite for Tracking Multimodal Model Development

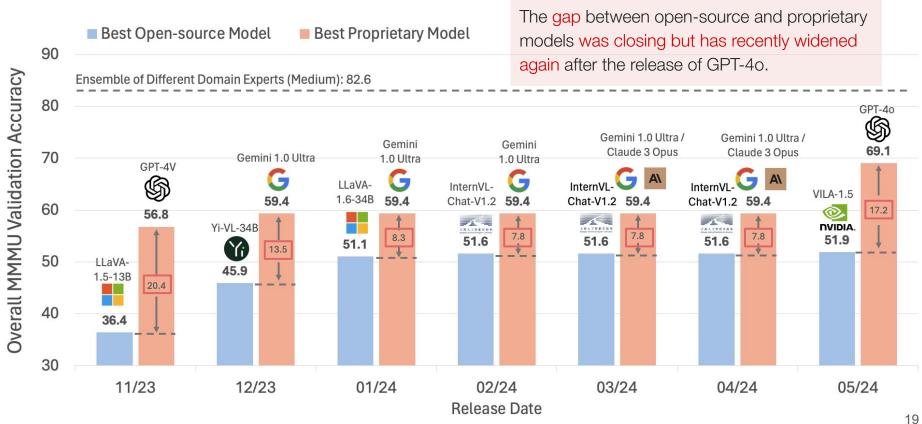


Tracking the development of multimodal models

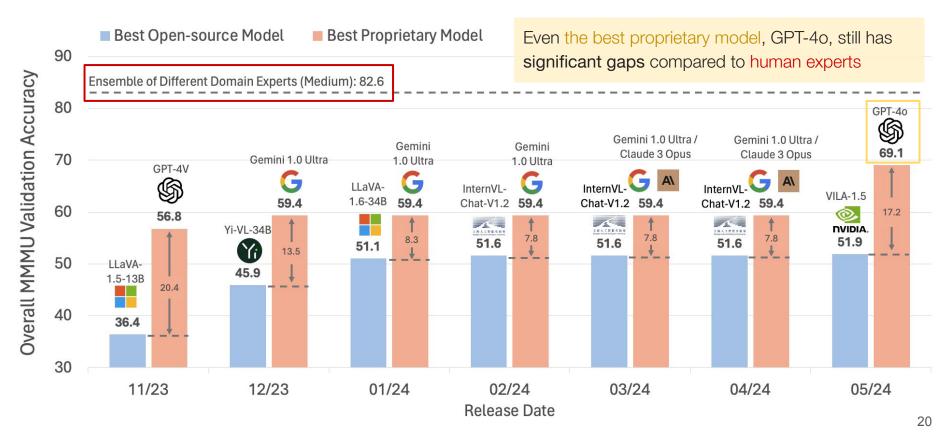




Open-source VS. Proprietary



Open-source VS. Proprietary



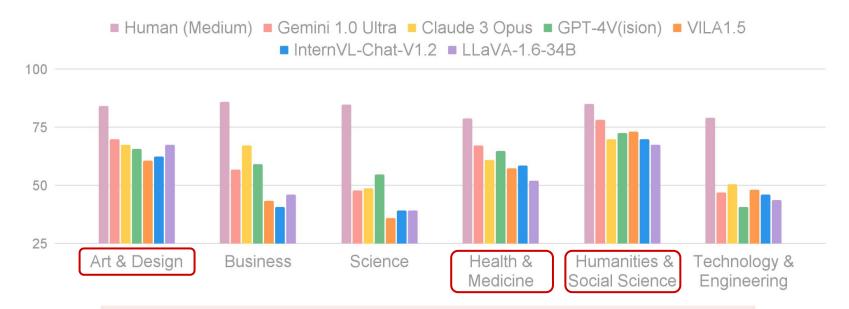
How to improve MM reasoning? Conclusion

Why MMMU?

- Excellent Model Diagnosis Tool

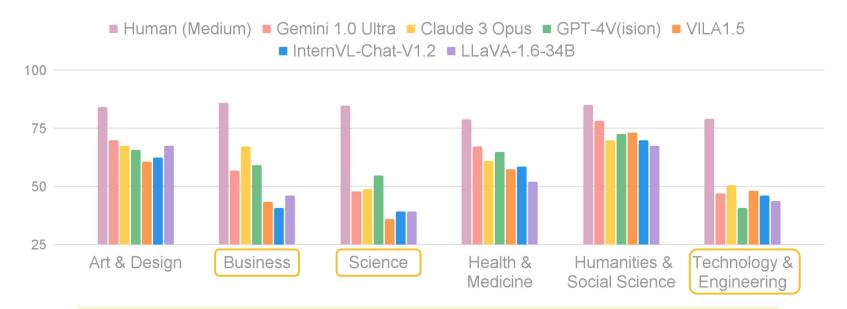


Subject-specific Accuracy



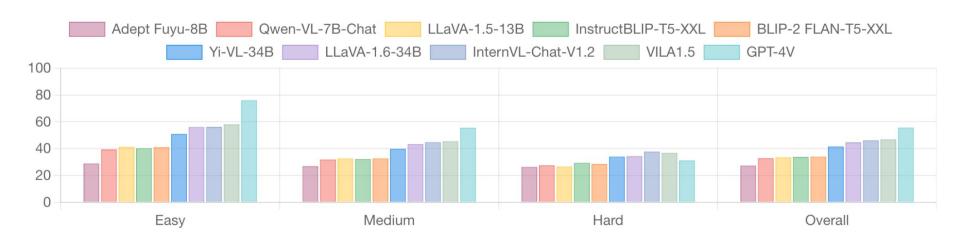
- o The gap between the best models and human experts is not large.
- o The difference between open-source and proprietary models is not significant.

Subject-specific Accuracy



- o The gap between the best models and human experts is significantly large.
- o Models struggle with these subjects, which involve more complex reasoning questions

Difficulty-specific Accuracy



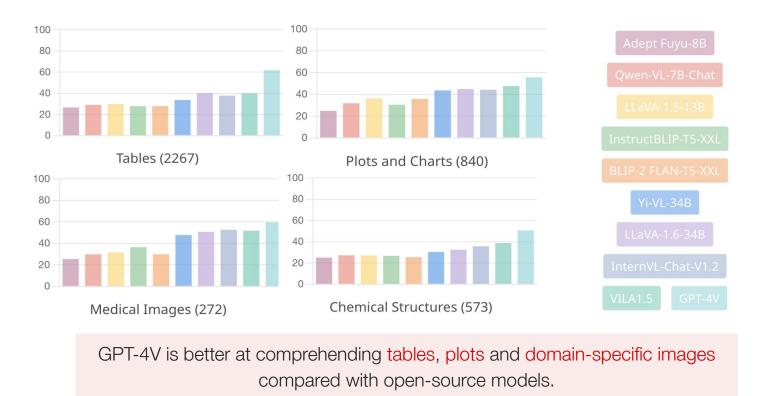
GPT-4V outperforms open-source models on easy and medium-level tasks, while all models struggle with hard examples.

Image Type-specific Accuracy



Introduction

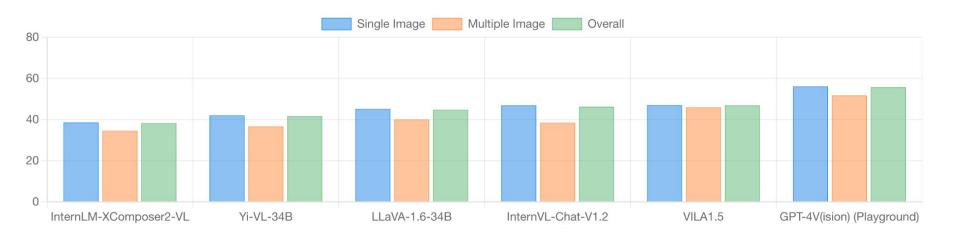
Tables, plots, and domain-specific images



Introduction Why MMMU? How to improve MM reasoning? Conclusion

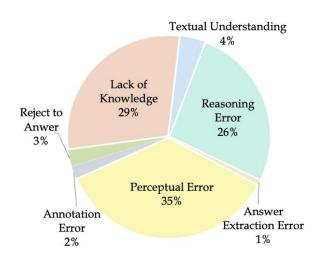
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Single-image V.S. Multiple-image



- Models generally struggle with reasoning over multiple images
- VILA performs notably better in this area

Error Analysis



Error distribution over 150 annotated GPT-4V errors (root cause)

Perceptual Errors (35%):

- Basic perceptual errors: fails in elementary visual interpretation
- Prioritizing textual over visual inputs
- Grounding and referring issues

Lack of Knowledge (29%):

 Knowledge is the foundation of domain-specific perception and reasoning

Reasoning Error (26%):

 Models struggle with complex reasoning and long calculation chains

Language as Vehicle: Model can explain what it sees and the rationale for its prediction



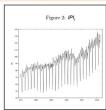
Economics: Econometrics

Question: Figure 1 displays the number of cases of Melanoma (Mt) in the male population of a US State, from 1936 to 1972. Figure 2 shows the Spanish Industrial Production Index (IPIt), from January 1975 to March 2001. <image 1> <image 2> According to figures 1 and 2, which of the following statements is FALSE:

Option:

- (A) IPIt displays a strong seasonality
- (B) Both time series are mean stationary
- (C) Both time series display a trending behavior





<image 2>

GPT-4V: Based on the provided figures:(A) IPIt displays a strong seasonality. - This statement appears to be TRUE. Figure 2 shows consistent and repeating peaks and valleys for the Spanish Industrial Production Index (IPIt), indicative of seasonality.(B) Both time series are mean stationary. - This statement appears to be FALSE. A mean stationary time series would have a constant mean (average) over time. Both figures show a changing mean over the given time frames, particularly Figure 1 where the mean value for Melanoma cases appears to be increasing over time. (C) Both time series display a trending behavior. - This statement appears to be TRUE. Figure 1 shows an upward trend in the number of Melanoma cases over time, and Figure 2 displays fluctuations but also an overall upward trend in the Industrial Production Index. Therefore, the FALSE statement is: (B) Both time series are mean stationary.

Ground Truth: (B) Both time series are mean stationary.



Clinical Medicine: Pathology

Question: The patient is a post-mature newborn. What is the most likely diagnosis? <i mage 1> Option:

- (A) Transient tachypnea of the newborn
- (B) Congenital lobar emphysema
- (C) Hyaline membrane disease
- (D) Meconium aspiration syndrome

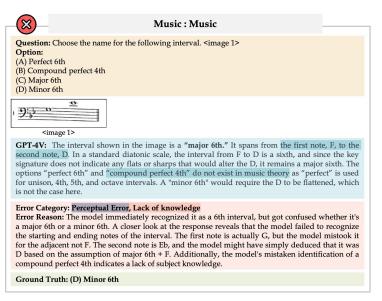


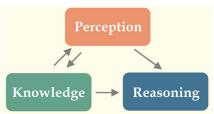
<image 1>

GPT-4V: Based on the image provided, which appears to be a chest X-ray, and the information that the patient is a post-mature newborn, the most suggestive diagnosis is: (D) Meconium aspiration syndrome. Post-mature newborns are at an increased risk of passing meconium (the baby's first stool) into the amniotic fluid before birth or during labor. If the baby breathes the meconiumstained amniotic fluid into their lungs, it can cause a blockage of the airways, leading to meconium aspiration syndrome. The chest X-ray of a newborn with meconium aspiration syndrome might show patchy or streaky areas on the lungs. However, it's important to consult with a pediatric radiologist or neonatologist for a definitive diagnosis based on clinical findings and the entire clinical picture.

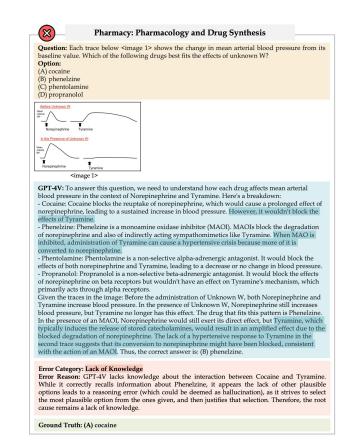
Ground Truth: (D) Meconium aspiration syndrome

Knowledge is the foundation for perception and reasoning





Lack of deep subject knowledge could lead to perception and reasoning errors



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Grounding and referring is challenging



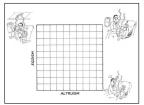
Manage: Management Models

Question: Look at the situation in the <image 1>. You are traveling on a plane with a small child. The cabin is depressurized and the oxygen mask falls from the ceiling. Which of these pictures shows the reconciliation of egoism and other-isms (the order is from left to right, top to bottom)?

Option:

(A) 1

(B) 2 (C) 3



<image 1>

GPT-4V: The image shows three situations on a plane when the oxygen mask falls: The adult puts on the mask first while the child watches (egoism). The adult puts the mask on the child first while struggling for breath. Both the adult and the child share the mask. The reconciliation of egoism and other-isms would be a situation where both the individual's needs and the needs of others are considered and addressed. Therefore, the correct answer is:
(C) 3

Error Category: Perceptual Error

Error Reason: GPT-4V recalled the right knowledge and made the right reasoning, but it failed to correctly map the IDs to the corresponding illustrations in the figure, which is not explicitly marked in the figure but is only described in text (the order is from left to right, top to bottom).

Why MMMU?

Ground Truth: (B) 2



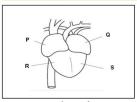
Basic Medical Science: Cardiovascular Physiology

Question: <image 1> The diagram shows the front view of a human heart. Letters P, Q, R and S indicate the different chambers of the heart. Which of the following shows the sequence in which a blood cell returning from other parts of the body passes through the four chambers of the heart? **Option:**

(A) R-S-P-Q

(B) Q-S-R-P (C) P-R-O-S

(D) S-R-P-Q



<image 1>

GPT-4V: The sequence in which a blood cell returning from other parts of the body passes through the four chambers of the heart is: Right atrium, Right ventricle, Left atrium, Left ventricle. Looking at the given options, the correct sequence is:(D) S-R-P-Q

Error Category: Perceptual Error

Error Reason: The model has the correct knowledge, but it failed to correctly understand the image and map the letters to the right structures (e.g., P is right atrium so it should be the first in the sequence).

Ground Truth: (C) P-R-Q-S

Complex Reasoning



Architecture and Engineering: Structural Engineering

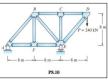
Question: Compute the horizontal displacement of joint B produced by the 240-kN load in Figure P8.10. For all bars, area = 2400 mm^2 and E = 200 GPa.

Option: (A) 2.657 mm

(B) 3.657 mm

(C) 4.657 mm

(D) 5.657 mm



<image 1>

GPT-4V: To solve for the horizontal displacement of joint B due to the 240-kN load, we'll employ the method of joints and consider the effect of the force on the truss members, along with using the formula for the elongation/contraction of axially loaded members. The formula to calculate elongation or contraction in an axially loaded member is: $\delta = \frac{PL}{\Delta E'}$ where P is the axial force in the member, L is the length of the member, A is the cross-sectional area of the member, and E is the modulus of elasticity. Given: A = 2400 mm² and E = 200 Gpa = $200 \times 10^3 \frac{N}{mm^2}$. First, let's find the forces in the members by applying the method of joints at joint A and joint D. At Joint A: There are two non-collinear unknown member forces, making it a solvable joint. Let's solve for AB and AF. $\sum F_{v} = 0$, $F_{AR}\cos(45^{\circ}) - F_{AF} = 0$. $\sum F_{r} = 0$, $F_{AR}\sin(45^{\circ}) = 0$. $F_{AR} = 0$ and $F_{AF} = 0$. At Joint D: There's a 240-kN downward force applied. Let's find DC and DE. $\sum F_y = 0$, $F_{DC}cos(45^\circ) + F_{DE} - 240 =$ 0. (1) $F_{DE} = 240 - F_{DC}cos(45^{\circ})$. $\Sigma F_{x} = 0$, $F_{DC}sin(45^{\circ}) = 0$. $F_{DC} = 0$. From (1), $F_{DE} = 240 \text{ kN}$. Compute the elongation in the members: For DE (in tension, elongation): $\delta_{DE} = \frac{P_{DE} \times L_{DE}}{A \times E} =$ $\frac{240\times10^3\times8000}{2400\times200\times10^3} = 8mm$. Considering the geometry of the truss, the horizontal component of the elongation at DE is: $\delta_R = \delta_{DE} cos(45^\circ) \approx 5.657 mm$. Thus, the answer is: **(D)** 5.657 mm.

Ground Truth: (D) 5.657 mm



Energy and Power: Thermodynamics

Question: A piston-cylinder device contains helium gas initially at 100 kPa, 10°C, and 0.2 m^3. The helium is now compressed in a polytropic process (PVn = constant) to 700 kPa and 290°C. Determine the heat loss or gain during this process..<image 1>

(A) $Q_{12} = -5.553 \, kJ$

(B) $Q_{12} = -6.553 \, kJ$ (C) $Q_{12} = -7.553 \, kJ$



GPT-4V: Given the nature of this problem, we'll use the first law of thermodynamics for a closed system undergoing a process: $\Delta U = Q_{12} - W_{12}$, where $\Delta U = \text{Change in internal energy}$, $Q_{12} = \text{Heat}$ transfer during the process, and W_{12} = Work done by the system during the process. From the given image, we can infer that the process is polytropic and is described by the equation; $PV^n = constant$. For a polytropic process involving an ideal gas: $W_{12} = (P_2V_2 - P_1V_1)/(1-n)$. Given: $P_1 = 100 \text{kPa}$, $V_1 = 0.2m^3$, $T_1 = 10^{\circ}C$, $P_2 = 700 \text{kPa}$, $T_2 = 290^{\circ}C$. To find V_2 , we use the polytropic relationship: $P_1V_1^n/P_2 = V_1^n$. The value of n can be determined using the temperatures and the ideal gas law: $P_1V_1/T_1 = P_2V_2/T_2$. Once V_2 is known, the work W_{12} can be determined using the equation for polytropic work. The change in internal energy ΔU for an ideal gas like helium is only a function of temperature and can be determined using: $\Delta U = mC_v(T_2 - T_1)$, where $C_v = \text{Specific heat at constant}$ volume for helium (approximately 3.12 kJ/kg·K), and m = Mass of the helium. Mass can be determined using the ideal gas law: $m = P_{\nu}V_{\nu}/RT_{\nu}$, where R is the specific gas constant for helium. approximately 2.0831 kJ/kg·K. Finally, using the first law of thermodynamics: $Q_{12} = \Delta U + W_{12}$ =. The given options for Q_{12} are in kJ, so we need to ensure that all values used in calculations are consistent with this unit. To determine Q₁₂, we will: Calculate the mass m of helium, determine V₂ using the polytropic relationship, calculate the work W_{12} , determine the change in internal energy ΔU_{1} , and use the first law to calculate Q_{12} . Now, let's plug in the values and solve: Calculate mass: $m = P_1V_1/RT_1 = 100kPa * 0.2m^3/(2.0831kI/kg \cdot K * (10 + 273)) \approx 0.0271kg$. Determine V_2 using the given temperatures and the ideal gas relationship (this is iterative and requires more involved math). Calculate the work W_{12} using the polytropic equation. Determine the change in internal energy: $\Delta U = mC_v(T_2 - T_1) \approx 23.75 kJ$. Use the first law: $Q_{12} = \Delta U + W_{12}$. Now, without the exact value for W_{12} (which requires the iterative determination of V 2 and n), it's hard to provide an exact answer. However, considering the given options, you can solve for Q_{12} using the above steps. If you have the value for n or more specific data, please provide it, and I can assist further.

Error Category: Reasoning Error

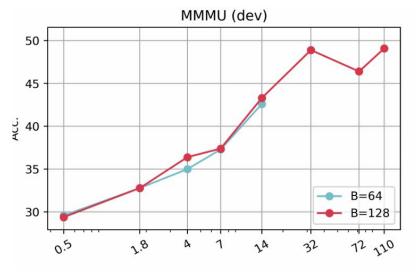
Error Reason: Reasoning error 1 (calculation): The formula of m is correct but the calculation is wrong: $m = P_1V_1/RT_1 \approx 0.0271kg$. Reasoning error 2 (unit conversion T: The formula of ΔU is correct but the unit conversion of T is wrong: $\Delta U = mC_v(T_2 - T_1) \approx 23.75 kJ$. Reasoning error 3 (calculation order): The final answer Q is to be calculated based on W, W is to be calculated based on V2 and n, n is to be calculated based on V2. GPT4-V fails to calculate these variables in a correct order $V_2 \rightarrow n \rightarrow W \rightarrow Q$, thus it is unable to provide the answer.

Ground Truth: (B) 0 12=-6.553 kl

GPT-4V demonstrates moments of excellence in complex reasoning but tends to falter with extended reasoning chains or excessive calculations.

Why MMMU? How to improve MM reasoning? Introduction Conclusion

Larger and more capable LLM



LLaVA NEXT v	with Different	Sizes of	: LLMs
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https://llava-vl.github.io/blog/2024-05-25-llava-next-ablations/

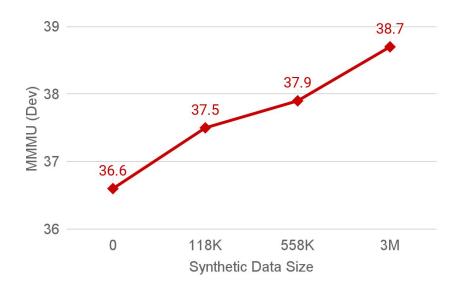
Models	Image Encoders	LLMs	MMMU (test)
VILA1.5-3B	SigLIP-so400	Sheared-LLaMA-2.7B	30.8
VILA1.5-8B	SigLIP-so400	Llama3-8B	36.0
VILA1.5-13B	SigLIP-so400	Vicuna-13B	33.6
VILA1.5-40B	InternViT	Yi-34B	46.9

VILA 1.5 with Different Image Encoders and LLMs

https://github.com/NVlabs/VILA

The MMMU performance shows a clear scaling effect with the LLM size

High-quality (synthetic) data



(Figure is re-plotted based on a LLaVA 1.6 ablation result)

- Synthetic captions by <u>LLaVA-NEXT-34B</u>
- Images are from COCO118K, BLIP558K, and CC3M
- Fine-tune a 7B model on the synthetic data

The MMMU performance shows a clear **scaling effect** with an increased number of **high-quality** (synthetic) data.

• Image Encoders, Resolution, Visual Tokens

Vision Encoder	Encoder size	Visual Tokens	Res. (px)	MMMU (dev)
CLIP-L	0.3B	256	448	38.2
EVA-02-E	4.7B	256	448	33.6
EVA-8B	8B	256	448	35.0

Vision Encoder	Encoder size	Resolution (px)	MMMU (dev)
CLIP-L	0.3B	336	38.1
SigLIP	0.4B	384	40.1
Idefics 2	0.4B	Up to 980	41.1

(Numbers are copied from LLaVA 1.6 ablation result)

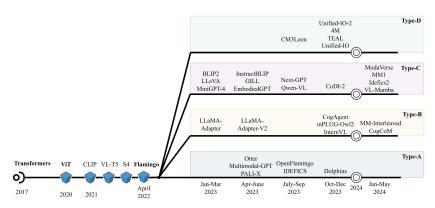
- Good vision encoders, utilizing better pretrained data and higher resolution, improve performance
- Scaling vision encoders has a minimal impact
- Increasing visual tokens offers little benefit

(Numbers are copied from the MANTIS paper)

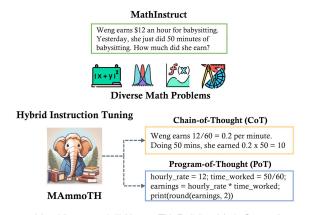
Vision Encoder	Model size	Visual Tokens	MMMU (dev/test)
Idefics 2	0.4B	64	43.5/37.9
Idefics 2	0.4B	320	43.0/37.7

(Numbers are copied from the <u>IDEFICS2</u> paper)

- Different VLM architectures
 - Better vision and language alignment
- Augment multimodal models with tools
 - Combine natural language with code for reasoning



Wadekar, Shakti N., et al. "The Evolution of Multimodal Model Architectures." arXiv preprint arXiv:2405.17927 (2024).



Yue, Xiang, et al. "MAmmoTH: Building Math Generalist Models through Hybrid Instruction Tuning." ICLR 2024.

Conclusion

- The MMMU benchmark has proven to be an essential tool for evaluating and guiding the development of multimodal models, providing critical insights into model strengths and weaknesses.
- Despite progress, models still face significant challenges, particularly in domain-specific visual perception, grounding, and handling complex reasoning tasks.
- We shall caution that MMMU is not a sufficient test for Expert AGI.
 However, we believe it should be necessary for an Expert AGI to
 achieve strong performance on MMMU to demonstrate their broad and
 deep subject knowledge as well as expert-level understanding and
 reasoning capabilities.
- Future efforts should aim at closing the gap between multimodal models and human experts, working towards expert AGI.

Acknowledgement

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 volunteer annotators who collectively spent more than 1,000 hours to make
 this project possible!
- We are grateful to the model developers and the community, especially the authors of LLaVA, VILA, InternVL, and Yi-VL, for sharing their model diagnostic experience on MMMU.
- We appreciate all the developers who used MMMU and reported issues, helping us fix small problems and improve the benchmark.

IN.Al Research









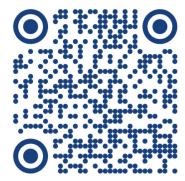


Introduction Why MMMU? How to improve MM reasoning? Conclusion

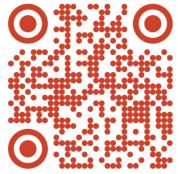
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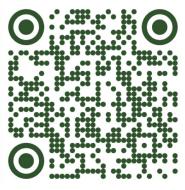
Questions?



Homepage & Leaderboard



Paper



Dataset