The ChatGLM's Road to AGI



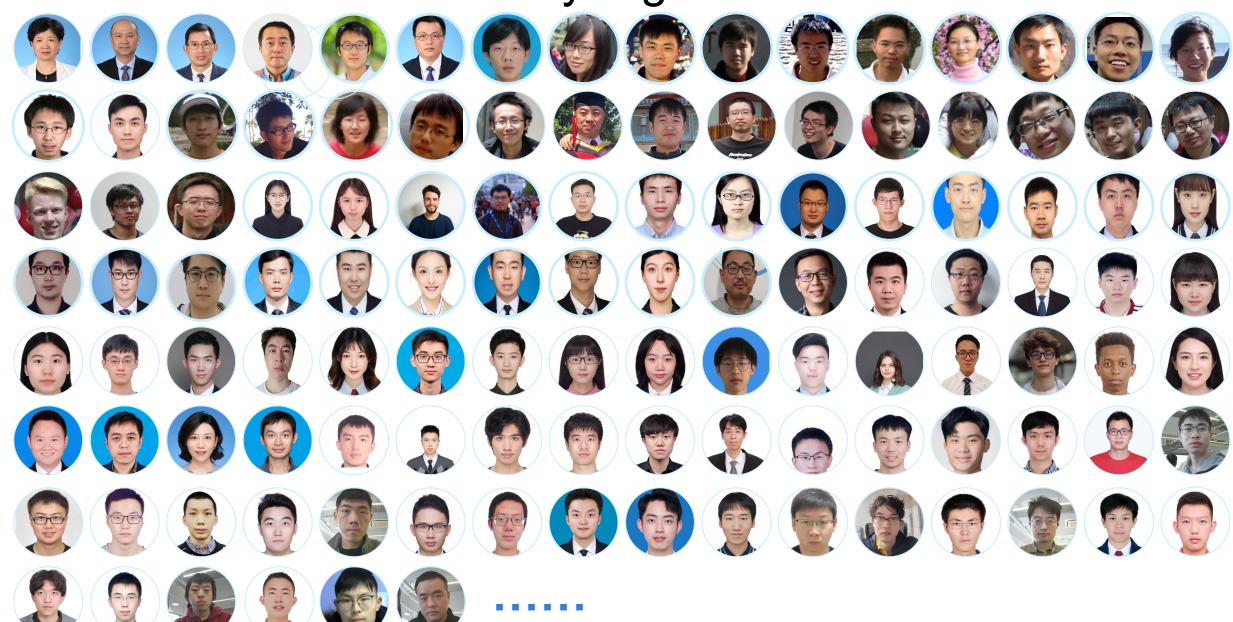
Jie Tang

KEG, Tsinghua University



https://github.com/THUDM/

A Very Big Team



What is **ChatGLM**

- ChatGPT and GPT4 has gained enormous popularity
 - However, techniques behind GPT become a secret to all
- ChatGLM, an open-source ChatGPT alternative, toward unclosing the secrets
 - GLM-130B: an open-source LLM base model
 - ChatGLM-6B: a lightweight open-source ChatGPT alternative
 - ChatGLM-130B&GLM-4: not open-sourced, but available through API









#star

Open LLM Research

39,369

15,516

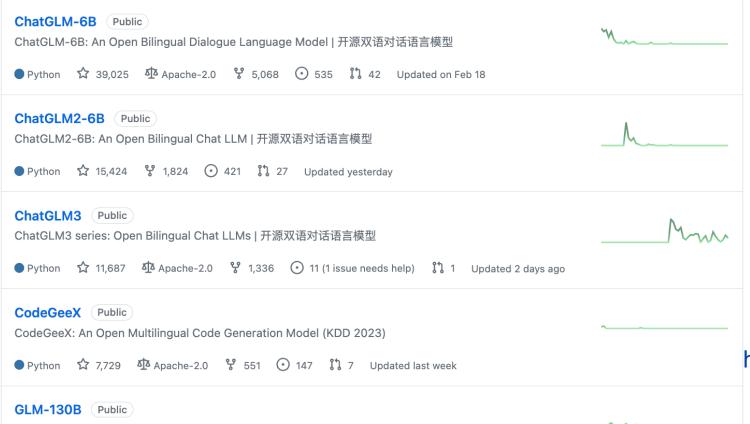
12,186

7,788

7,618

7,203

5,110





https://github.com/THUDM

GLM-130B: An Open Bilingual Pre-Trained Model (ICLR 2023)

● Python ☆ 7,600 ♣ Apache-2.0 ♀ 609 ⊙ 115 ♣ 5 Updated on Jul 25, 2023

CodeGeeX2 Public

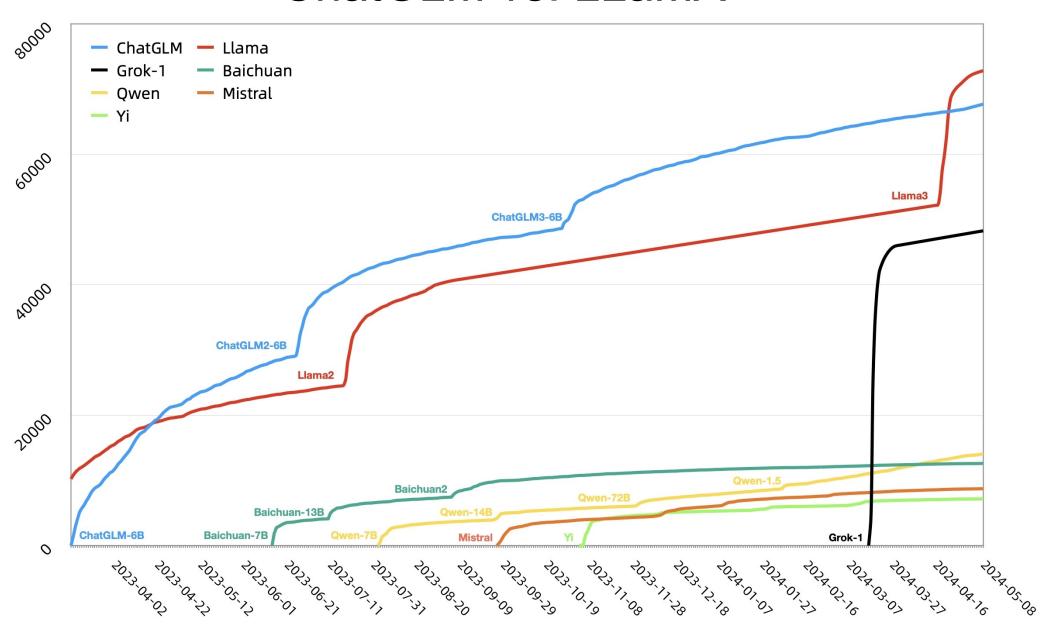
CodeGeeX2: A More Powerful Multilingual Code Generation Model

● Python ☆ 7,008 ၨΦ Apache-2.0 🖁 486 ⊙ 180 🖔 2 Updated on Feb 6

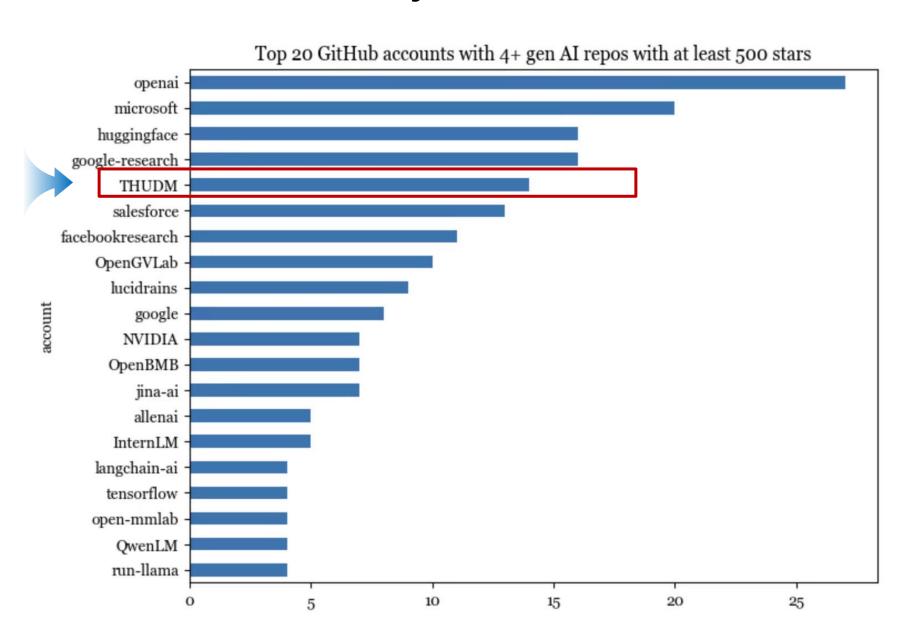
CogVLM Public

a state-of-the-art-level open visual language model | 多模态预训练模型

ChatGLM vs. LLaMA



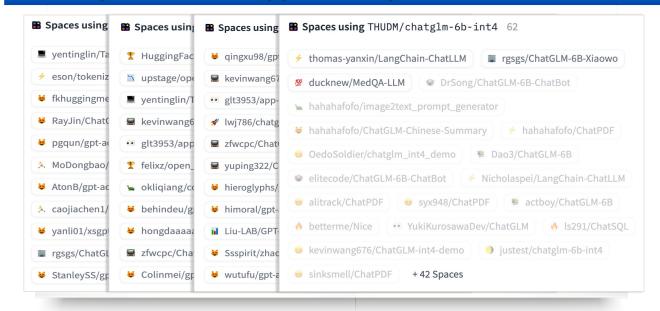
Number of Projects with 500 stars



ChatGLM-6B: An Open-Source Alternative

- ► ChatGLM-6B: **6.2B**parameters, **INT4** quantization (only need 6G memory)
- >60,000 stars on github
- >13,000,000 downloads on Huggingface
- ▶ No. 1 on Github Trending (2 week)
- ▶ No. 1 on Huggingface Trending (2 weeks)

>700 Open-Sourced Apps developed based on ChatGLM











The top 15 most-liked organizations on @huggingface

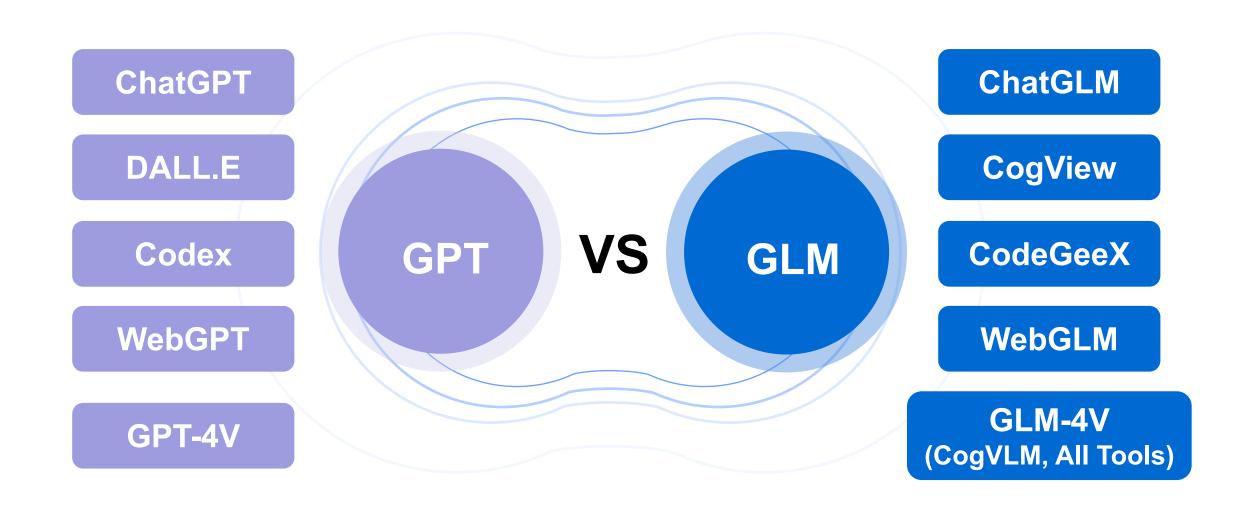
- 01. @StabilityAl 20k likes
- 02. @AlatMeta 20k
- 03. @runwayml 11k
- 04. Compvis 10k
- 05. @thukeg 7k
- 06. @BigscienceW 7k
- 07. @Tlluae 7k
- 08. @Microsoft 6.5k

- 09. @GoogleAl 6k
- 10. @OpenAl 4k
- 11. @BigCodeProject 4k
- 12. @MosaicML 4k
- 13. @UKPLab 3k
- 14. @AiEleuther 3k
- 15. @salesforce 3k

huggingface.co/spaces/PulsarA...

5:40 PM - Nov 22,2023 - **6,094** views

ChatGPT vs. ChatGLM

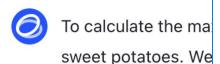


chatglm.ai

QAGLM GLM XDAI GLM-130B CodeGeeX **ChatGLM** who are you? I am an AI assistant named ChatGLM, which is developed based on the language model jointly trained by Tsinghua University KEG Lab and Zhipu Al Company in 2023. My job is to provide appropriate answers and support to users' questions and requests. □ Copy help me to write a blog outline which introduces ChatGLM Sure, here's a blog outline that introduces ChatGLM: I. Introduction Welcome to try A. Brief overview of ChatGLM B. Importance of AI assistants in today's world II. How ChatGLM works A. Basics of Al language models

Math Exercise

The apple orchard and a cost of 3 yua otatoes under apple r kilogram and a coms, every n kilogram at is the maximum



R = (69,000) * (8 -445,000 yuan

This is the maximur constraints. Therefore apples and purple s

您

The apple orchard produkilogram and a cost of 3 purple sweet potatoes un income of 4 yuan per kilogreater than 30,000 kilog decrease by 0.3n kilogra



To find the maximum pro apples and purple sweet interdependence betwee

Since this equation is not conditions: P=0 or P

1. When P=0:

Profit = 45,375 + 0.5(0) = 45,375

2. When P = 50,000:

 ${
m Profit} = 45,375 + 0.5(50,000) = 70,375$

PB

The apple orchard produces 75 tons of apples every year, with a fixed income of 8 kilogram and a cost of 3 yuan. In order to improve land utilization, it has been deciple plant purple sweet potatoes under apple trees next year (the maximum yield is 50 with a fixed income of 4 yuan per kilogram and a cost of 2 yuan. When the yield of sweet potato is greater than 30,000 kilograms, every n kilogram increase in its yil cause the apple yield to decrease by 0.3n kilograms. What is the maximum profit apple orchard in the next year?

To solve this problem, we need to find the combination of apple and purple sweet pyields that maximizes the total profit for the orchard. Let's define the variables:

••••

Therefore, the maximum profit for the apple orchard in the next year is 475,000 yu which is achieved by producing 75,000 kg of apples and 50,000 kg of purple sweet potatoes.



Applied Math

Can you please prove e^x=x^e has only and only has one solution?



The equation $e^x = x^e$ doesn't have a understand the number of solutions

First, let's consider the function f(x) analyze this function, we can use cathe function.

- 1. Differentiate f(x) with respect t
- 2. Find the critical points by setting
- 3. Analyze the behavior of the fund solutions.

Let's perform these steps

Generating codes.

python

```
from sympy import symbols, diff, exp

# Define the variable and the function

x = symbols('x')

e = exp(1) # Define e as the base of the natural logarithm

f = exp(x) - x**e

# Calculate the derivative

f_prime = diff(f, x)

f_prime_simplified = f_prime.simplify()

f_prime_simplified
```

Results

```
[(1 - e)*LambertW(-exp(-log(e)/(e - 1))/(e - 1))]
```

The critical point of the function $f(x) = e^x - x^e$ is given by $x = (1 - e) \cdot \operatorname{LambertW}\left(-\frac{e^{-\frac{\log(e)}{e-1}}}{e-1}\right)$, where

Text-to-Image Generation (2048x2048)



Fig. 1: Showcases of CogView3 generation of resolution 2048 × 2048 **(top)** and 1024 × 1024 **(bottom)**. All prompts are sampled from Partiprompts [31].

Image Understanding

Grounding Visual Question Answering

Q1: What color are the clothes of the girl whose hands are holding flowers? Let's think step by step.

CogVLM: Find the flowers [[328,519,436,780]]. Check the girl holding it, got the girl [[116,069,450,927]]. Check the clothes of it, got the clothes [[134,309,418,852]]. The clothes [[134,309,418,852]] are blue. So the answer is blue.



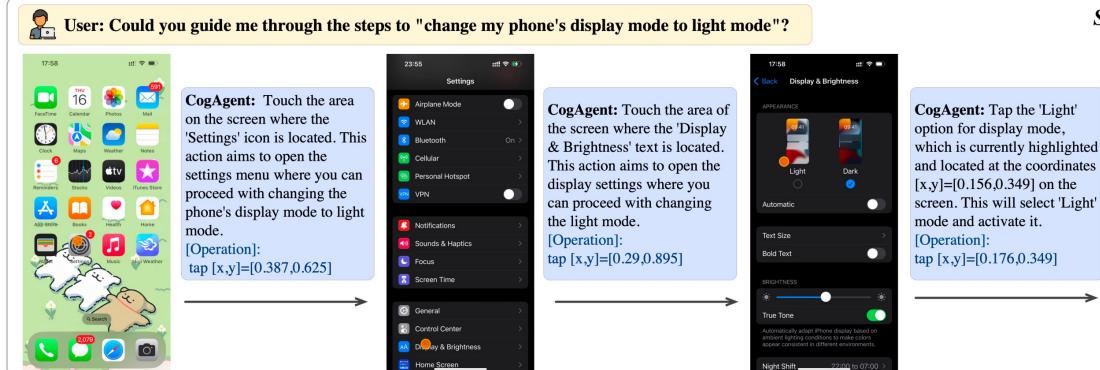
Q2: What color are the jeans of the girl whose hands are not holding flowers? Let's think step by step.

CogVLM: Find the not(flowers) hands [[654,637,710,728]]. Check the girl with it, got the girl [[526,109,880,946]]. Check the jeans of it, got the jeans [[530,613,880,871]]. The jeans [[530,613,880,871]] are gray. So the answer is gray.



GUI Agents

"Change my Phone's display mode to light model"



Smartphone Agent

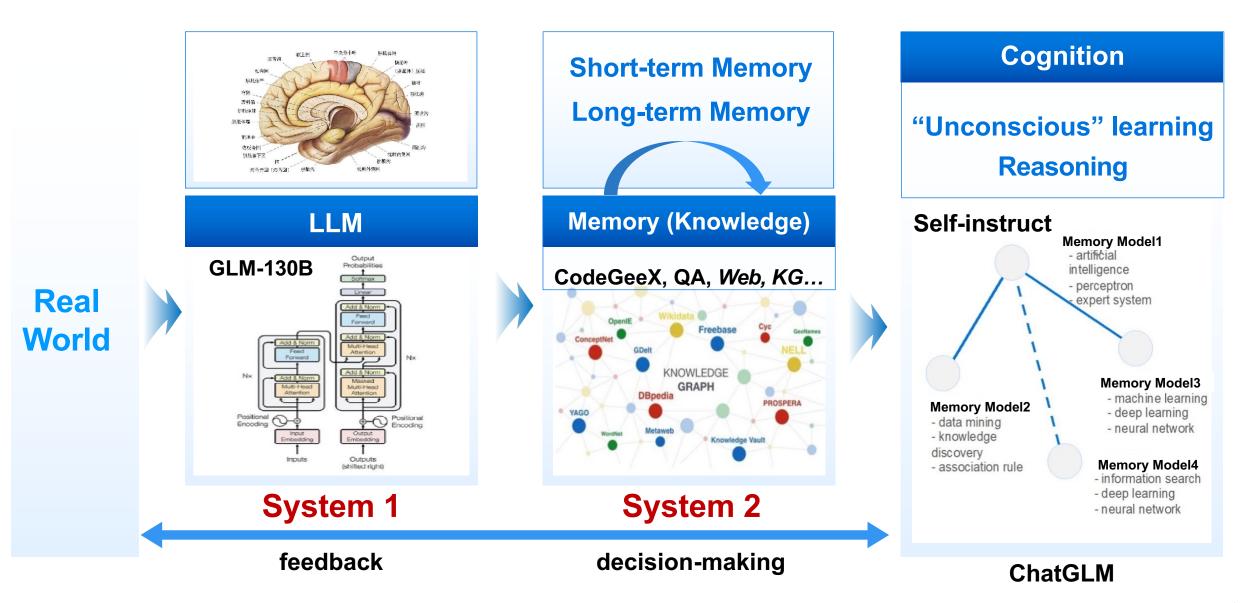
Appearance

| Coordinates | Co

How to make this happen?

—let us open the secrets

Our Plan since 2019



THU & ZhipuAl's GLM

GLM

Du et al. GLM: General Language Model Pretraining with Autoregressive

Blank Infilling. arXiv:2103.10360. ACL'22

P-Tuning

Liu et al. GPT understands, Too. arXiv:2103.10385. ACL'22

CogView

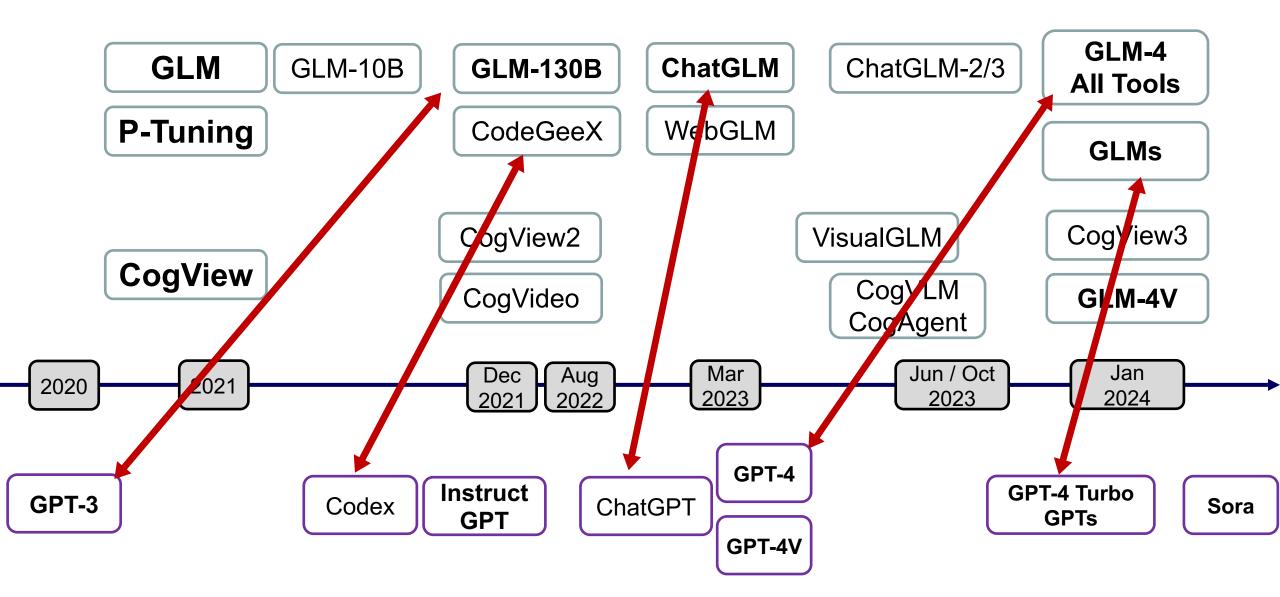
Ding et al. CogView: Mastering Text-to-Image Generation via

Transformers. arXiv:2105.13290. NeurIPS 2021

2020

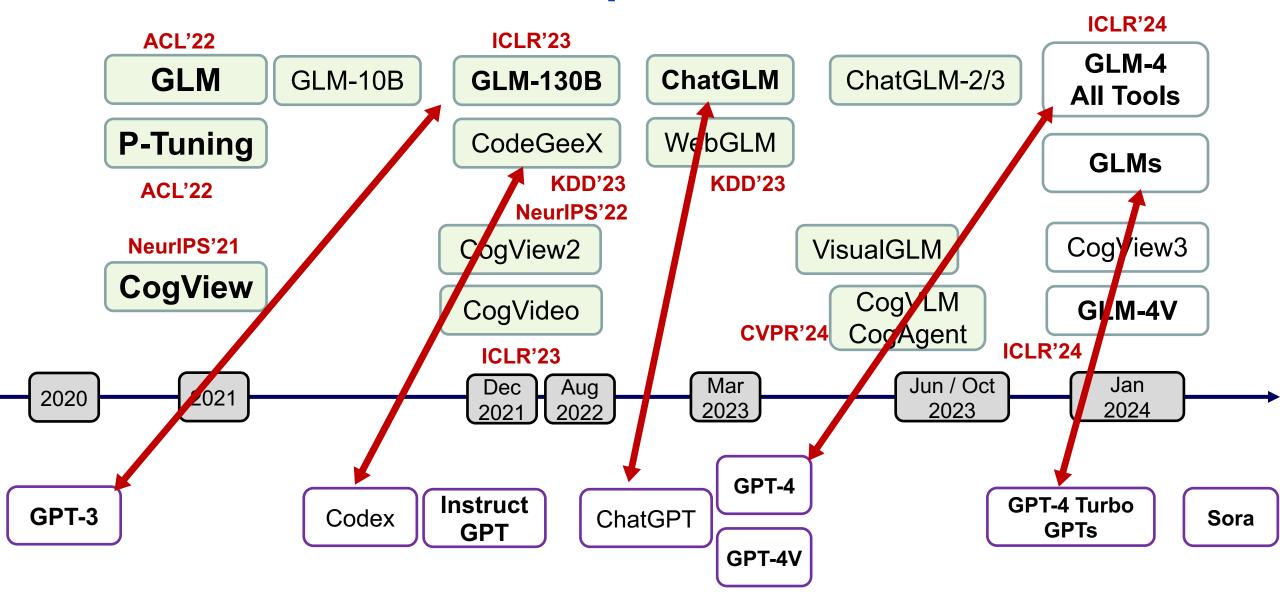
2021

THU & ZhipuAl's GLM



OpenAl's GPT

THU & ZhipuAl's GLM



OpenAl's GPT

General Language Model (GLM)

Framework	NLU	Cond. Gen.	Uncond. Gen.
Autoregressive (GPT)			√
Autoencoding (BERT)	V	×	×
Encoder-Decoder (T5)		√	
Autoregressive Blank-Infilling (GLM)	V	√	√

General Language Model (GLM)



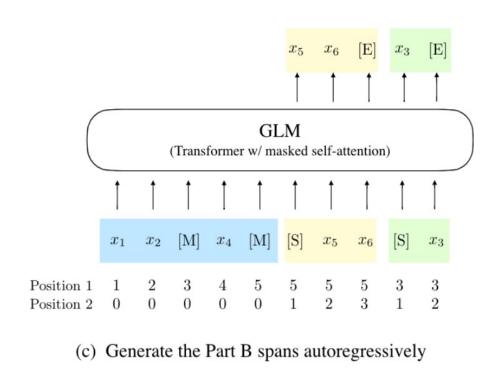
(a) Sample spans from the input text

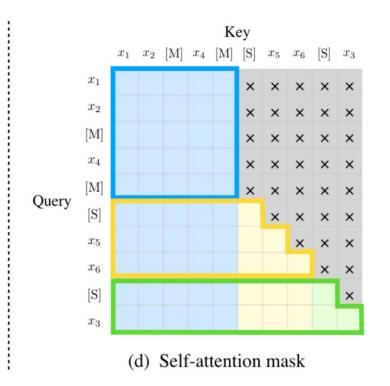
Part A:
$$x_1$$
 x_2 [M] x_4 [M]

Part B: x_5 x_6 x_3

(b) Divide the input into Part A and Part B

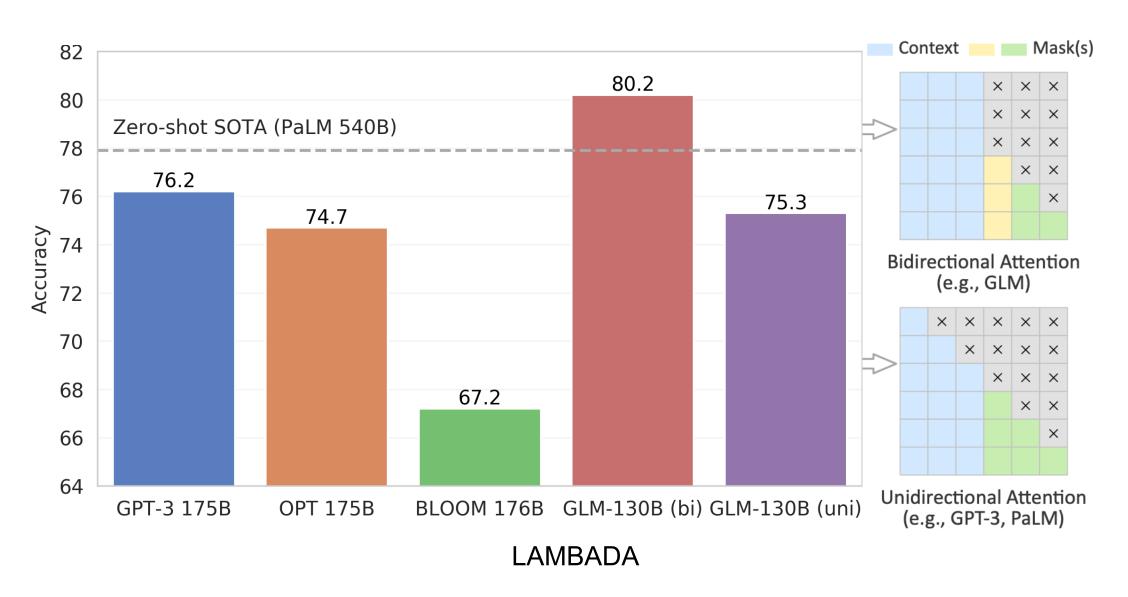
 x_3



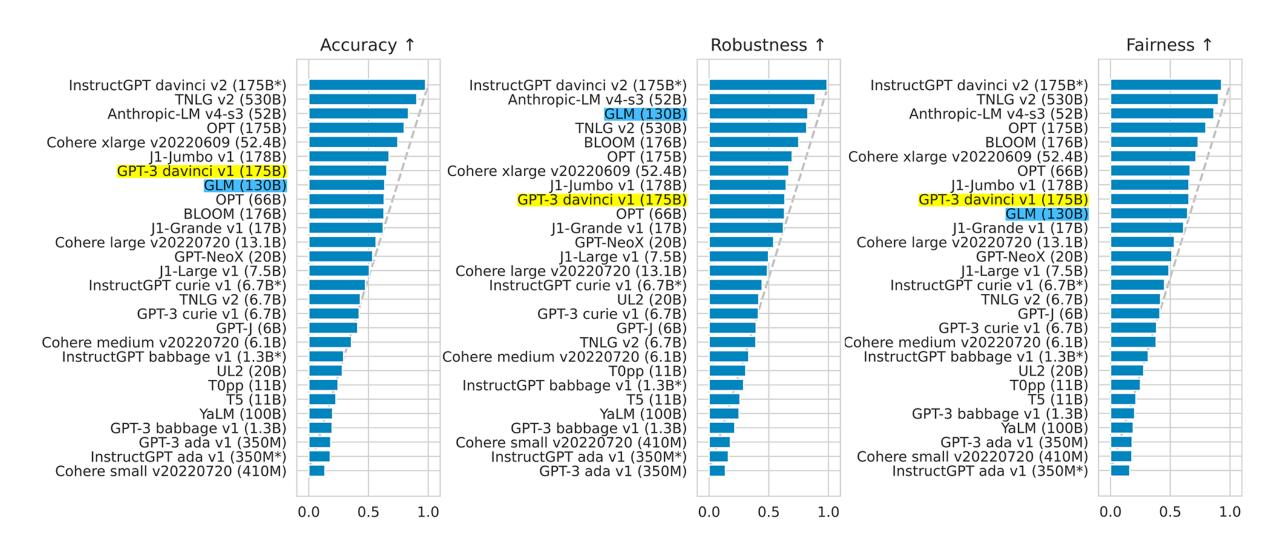


$$\mathcal{L}_{ ext{GLM}} = \mathbb{E}_{\mathbf{z} \sim Z_m} \left[\sum_{i=1}^m \sum_{j=1}^{l_i} -\log p\left(s_{z_i,j} | \mathbf{x}_{ ext{corrupt}}, \mathbf{s}_{z_{< i}}, s_{z_i, < j}
ight) \right]$$

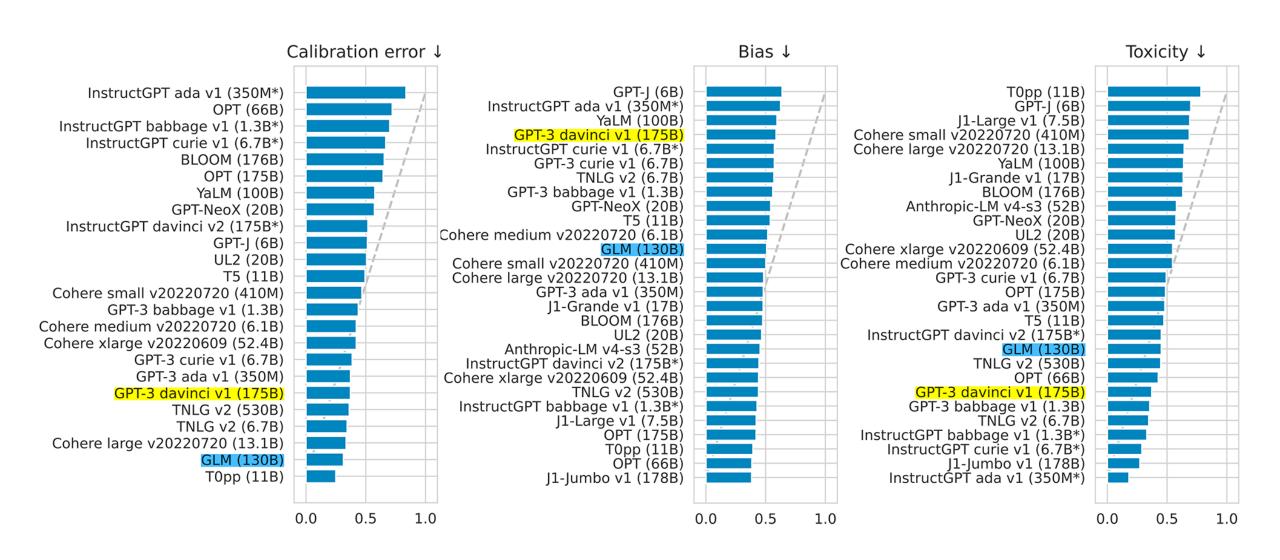
General Language Model (GLM)



GLM-130B in HELM

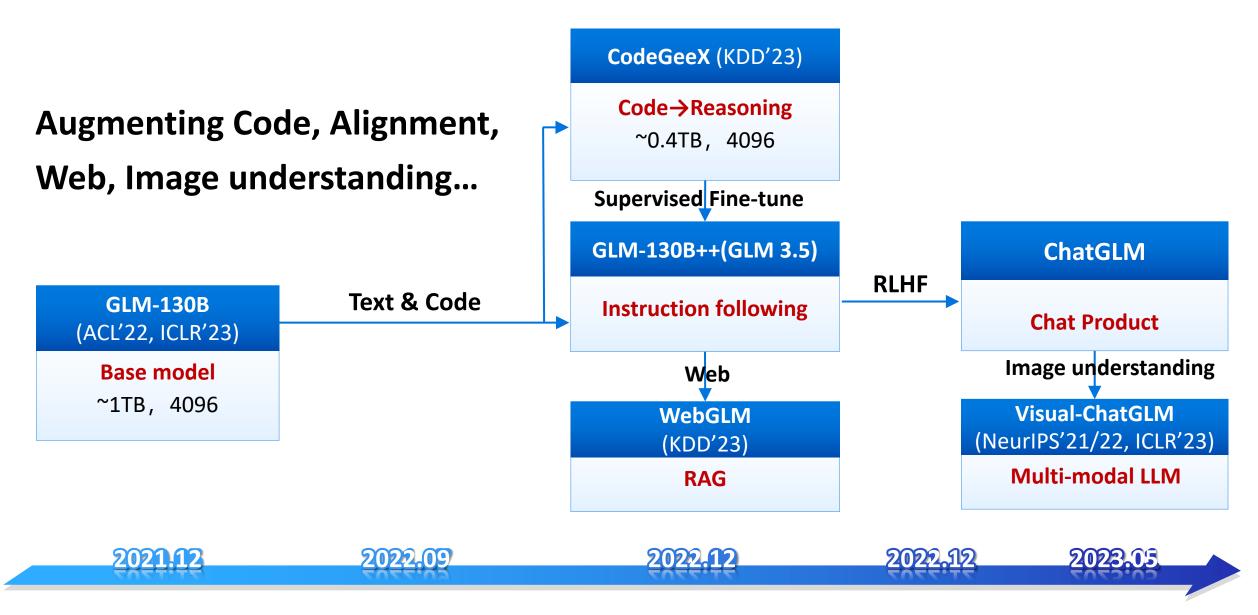


GLM-130B in HELM





Develop ChatGLM based on GLM-130B



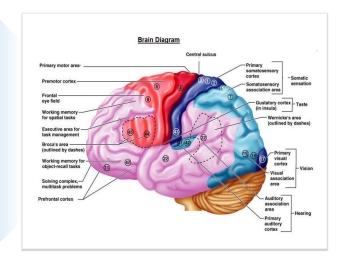
From ChatGLM to GLM-4V

—from LLMs to VLMs

From LLMs to VLMs

Our brain is a complex system

- Multimodal perception and understanding
- Short-term/long-term memory
- Reasoning





Fusing LLMs and visual understanding

To solve complex visual comprehension & reasoning tasks

From Simple Recognition to Complex Comprehension



What are shown in the photo?



A man and a chicken.

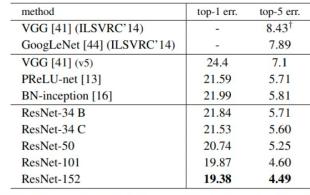
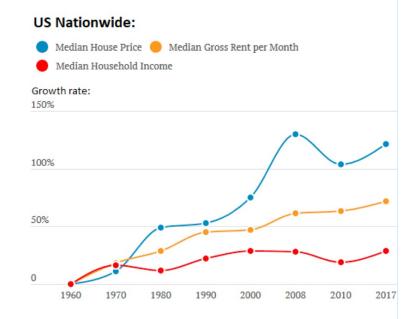


Table 4. Error rates (%) of **single-model** results on the ImageNet validation set (except † reported on the test set).

User: Convert it to the markdown format.

CogAgent: | method | top-1 err. | top-5 err. |

|-----| | VGG [41] (ILSVRC'14) | -18.43 I | GoogLeNet [44] (ILSVRC'14) | -17.89 | VGG [41] (v5) | 24.4 | | 7.1 | PReLU-net [13] | 21.59 | | 5.71 | BN-inception [16] | 21.99 | | 5.81 | ResNet-34 B | 21.84 15.71 | ResNet-34 C | 21.53 15.60 | ResNet-50 | 20.74 15.25 | ResNet-101 | 19.87 14.60 | ResNet-152 | 19.38 14.49



User: Which year has the highest growth rate of median house price?

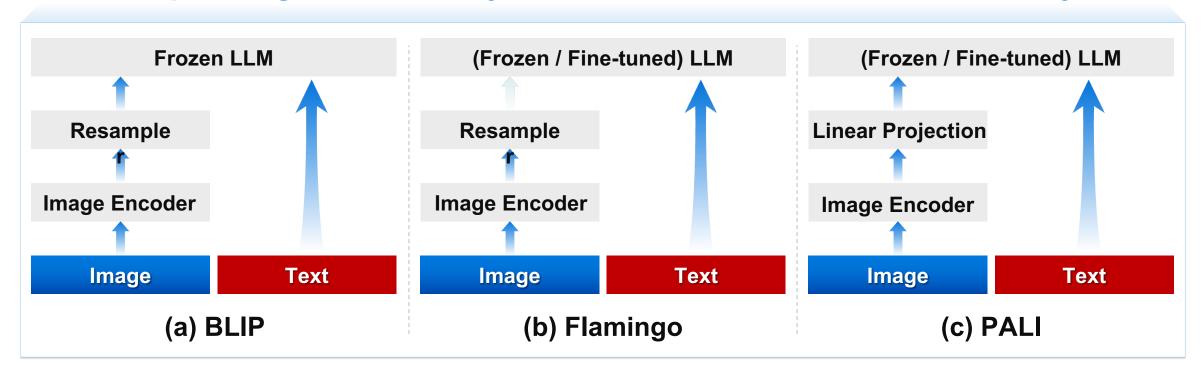
CogAgent: The year 2008 has the highest growth rate of median house price, which is 150%.

Before

Now

From LLMs to VLMs (Visual Language Models)

Expanding the boundary of the world model to vision modality

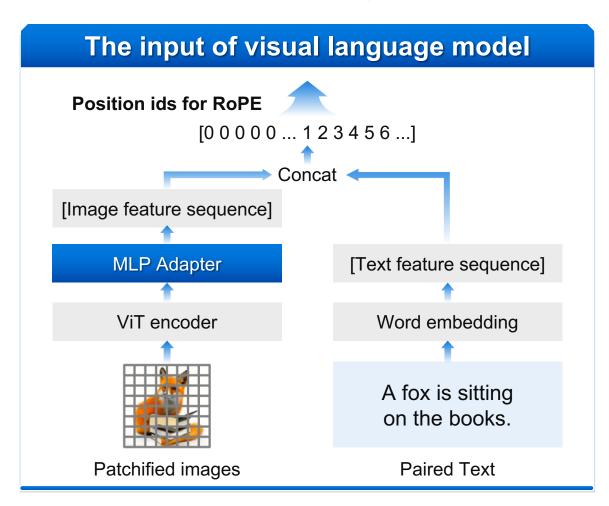


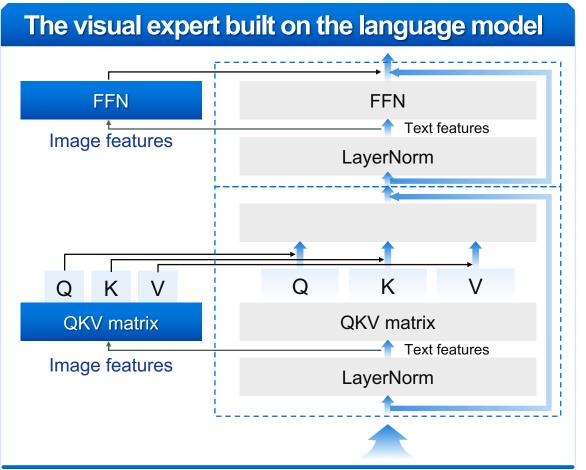
Shallow feature alignment between image and text

These structures tend to fit into text space, while image space may have unique features

CogVLM: Visual Expert for Large Language Models

- > Use a trainable visual expert module to bridge the gap of LLM and visual encoder
- Enable deep feature alignment between image and text





CogVLM: Visual Expert for Large Language Models

Achieves state-of-the-art performance on 17 cross-modal benchmarks

Image Captioning

Grounding

Visual Question Answering

VLM benchmarks

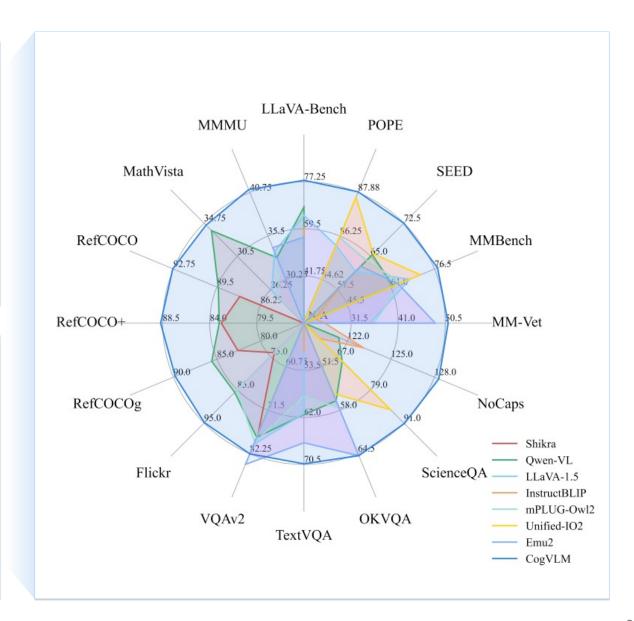
Open-sourced

https://github.com/THUDM/CogVLM





Web Demo



GLM-4V with CogVLM

World Knowledge

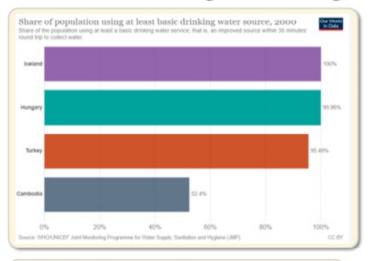


Q: How many goals did the person in the picture score in the 2018 World Cup?

CogVLM: The person in the picture is Cristiano Ronaldo. He is the captain of the Portuguese team and has participated in the 2018 World Cup. In the 2018 World Cup, Cristiano Ronaldo scored 4 goals.

Chart Understanding & Reasoning

Chart Understanding & Reasoning



Q: What is the difference of largest and smallest bar? Answer it step by step.

CogVLM: The largest bar represents Iceland with a percentage of 100%. The smallest bar represents Cambodia with a percentage of 52.4%. To find the difference, subtract the smallest percentage from the largest: 100% - 52.4% = 47.6%.

Ready for a tough task?

Q: Can you explain this meme in detail?

My Handwriting In Exams



A: Certainly! This meme is a humorous take on how one's handwriting can change throughout the process of writing an exam. Let's break down each part:

- 1. Title "My Handwriting In Exams"
- 2. **First two pages**: The image here shows neat and legible handwriting that reads, "Global warming is definitely a very huge issue in front of this entire world. It's causing several disturbances in the..." This suggests that at the beginning of the exam, when concentration levels are high, the handwriting is clear and well-maintained.
- 3. **Middle pages**: The middle image displays handwriting that is less organized and more rushed, with the text reading, "...which cause the meltdown of glaciers and multiply the ocean waves...". This indicates that as time passes and fatigue sets in, the handwriting becomes less precise.
- 4. **Last two pages**: The final image humorously depicts a heart rate monitor graph, implying that by the last few pages of the exam, the writer's hand is so exhausted and possibly shaking that it resembles a heart rate line, suggesting complete loss of control over the handwriting.

Overall, the meme plays on the relatable experience of students who start their exams with neatness but may falter towards the end due to stress, fatigue, or other factors, resulting in a decline in handwriting quality.

by GLM-4V

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by GPT-4V

Q: Can you explain this meme in detail?

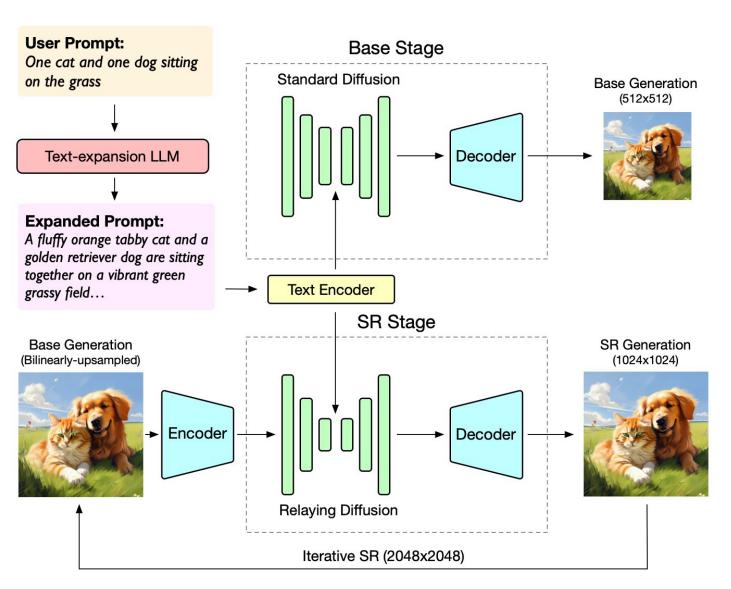
My Handwriting In Exams



- A: This meme humorously highlights the progressively deteriorating quality of a student's handwriting during an exam, using three different images to illustrate this progression:
- 1. **First two pages**: The first image shows neat and legible handwriting. The student is making a clear effort to present their work neatly, with the sentence "Global warming is definitely a very huge issue in front of this entire world. It's causing several disturbances in the..." written clearly.
- 2. **Middle pages**: The second image shows messier handwriting, which is still mostly legible but noticeably sloppier compared to the first image. The sentence "Also, in class the math down..." becomes increasingly difficult to read.
- 3. **Last two pages**: The third image is of an electrocardiogram (ECG) printout, mimicking very chaotic and illegible handwriting. The ECG printout humorously suggests that by the end of the exam, the student's handwriting has become so poor that it looks like a medical chart instead of text.

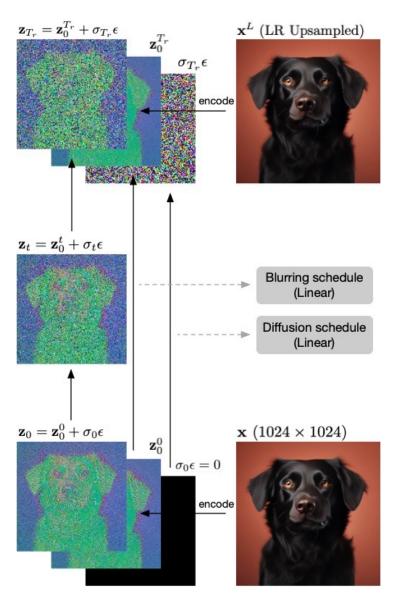
The meme reflects a relatable experience for many students who start an exam with neat handwriting but as time progresses, their focus shifts to finishing the paper rather than maintaining neatness. The ECG comparison at the end exaggerates this degradation for comedic effect.

CogView3: text-to-image generation



- Two-stage Text-to-Image Generation
 - Progressive Training
 - Sharing a same 3B U-Net structure
 - First Stage
 - 512x512 generation
 - Standard diffusion
 - Second Stage
 - 512-to-1024 super-resolution
 - Relay diffusion
 - Extended Stage
 - 2048x2048 or higher
 - Iterative relaying super-resolution

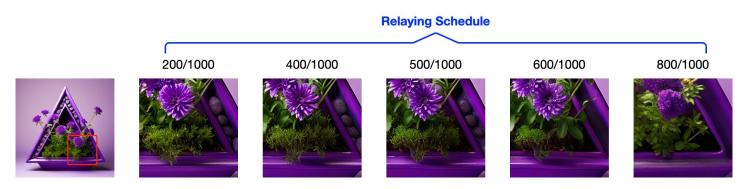
CogView3: text-to-image generation



- Relaying Super-resolution (Relay Diffusion)
 - Combining diffusion with blurring forward process
 - Simplified (linear) blurring schedule

$$m{z}_0^t = \mathcal{F}(m{z}_0,t) = rac{T_r - t}{T_r}m{z}_0 + rac{t}{T_r}m{z}^L, \ \ q(m{z}_t|m{z}_0) = \mathcal{N}(m{z}_t|m{z}_0^t, \sigma_t^2m{I}), \ \ t \in \{1,...,T_r\}.$$

- Starting diffusion from a middle point of the schedule
 - Early starting points: incomplete de-blurring results
 - Late starting points: over adjustments and artifacts
 - A midway starting point is the best.



CogView3: text-to-image generation

- 77% win over SDXL, 2x speed up (no-distilled version)
 - text-image aligning improvement (via re-caption)
 - generation quality improvement (via high-quality data fine-tuning)
 - inference cost reduction (via relaying staging and distillation)

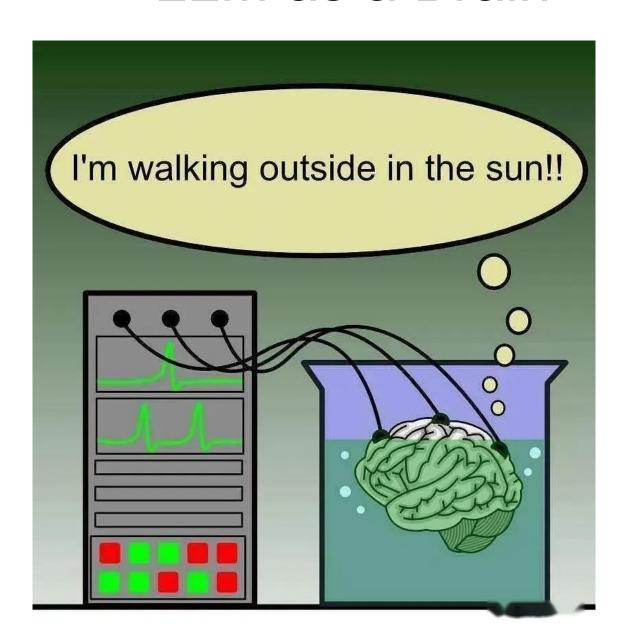


Model	Steps	Time Cost	$\mathbf{DrawBench}$			PartiPrompts			
			Aes†	HPS v2↑	ImageReward [†]	Aes†	HPS v2↑	ImageReward [†]	
SDXL 17	50	19.67s	5.54	0.288	0.676	5.78	0.287	0.915	
StableCascade [16]	20 + 10	10.83s	5.88	0.285	0.677	5.93	0.285	1.029	
$\mathbf{CogView3}$	50 + 10	10.33s	5.97	0.290	0.847	6.15	0.290	1.025	
LCM-SDXL [14]	4	2.06s	5.45	0.279	0.394	5.59	0.280	0.689	
${f CogView 3-distill}$	4 + 1	1.47s	5.87	0.288	0.731	6.12	0.287	0.968	
CogView3-distill	8+2	1.96s	5.90	0.285	0.655	6.13	0.288	0.963	

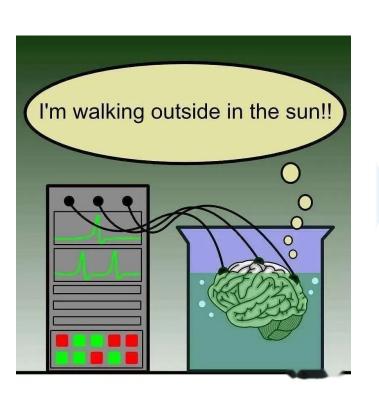
GLM-4V (All Tools)

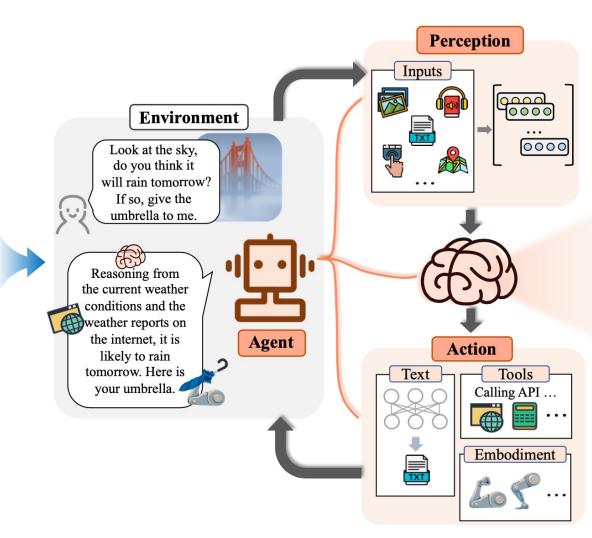
—learn to use tools

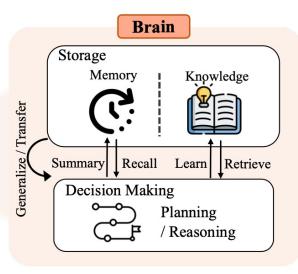
LLM as a Brain



LLM as a Brain

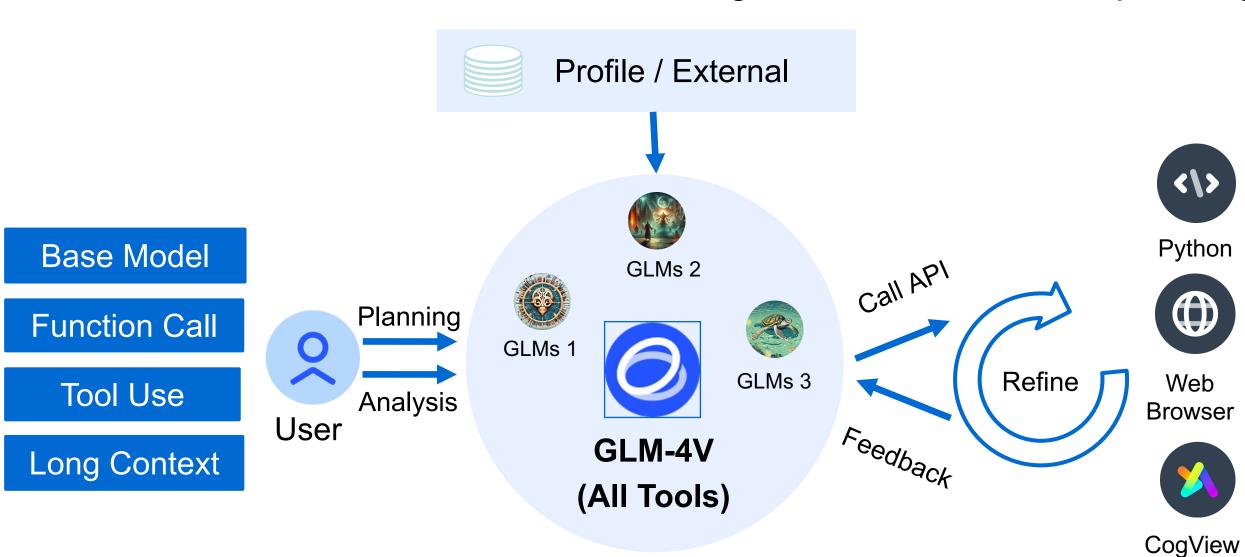




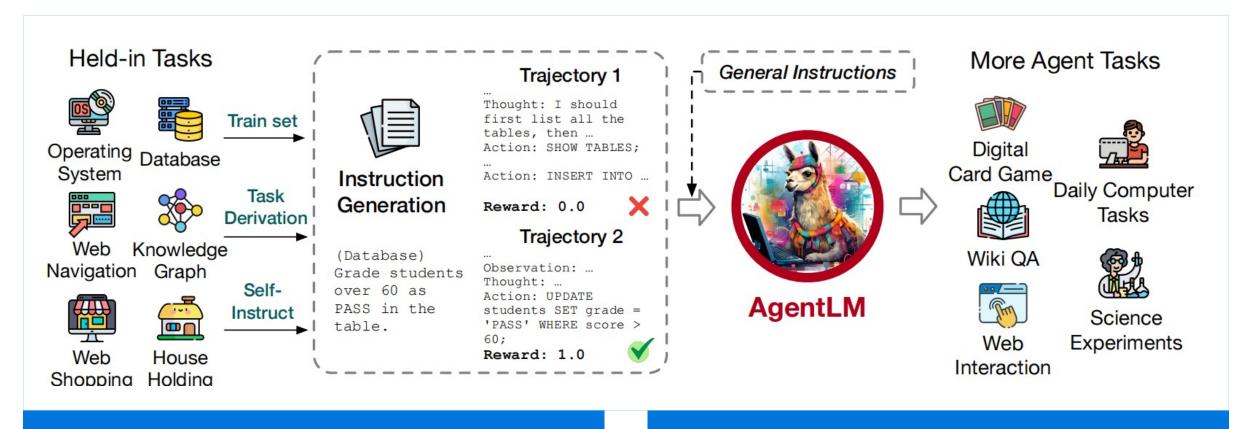


GLM-4V (All Tools)

All-tools: for smart instruction understanding and automatic action planning



AgentTuning: Enabling Generalized Agent Abilities For LLMs



Six agentInstruct trajectory datasets

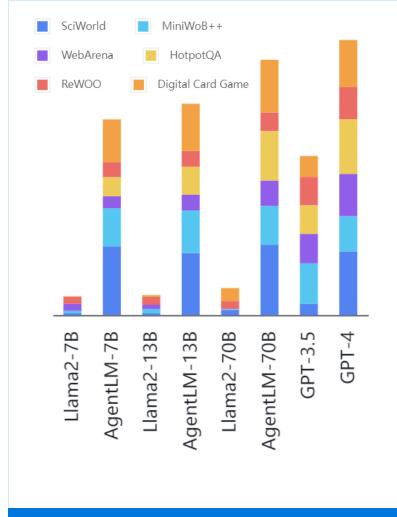
~2K high-quality CoTs

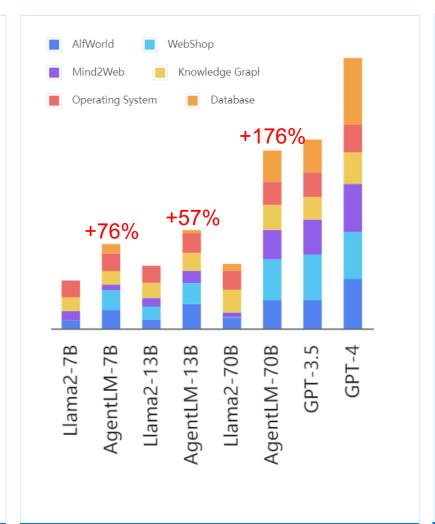
Agent Tuning Mix-training

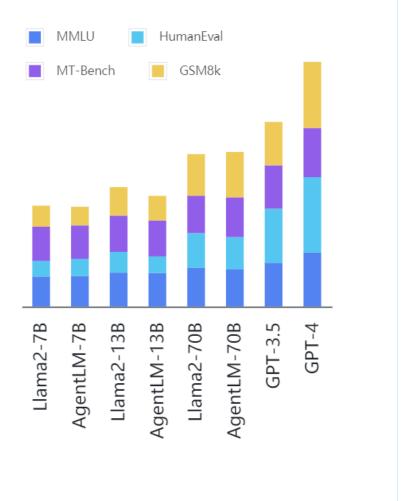
20% AgentInstruct + 80% ShareGPT

codes & models: http://github.com/THUDM/AgentTuning

Results







In-domain dist
Significant improvement

Out-domain dist Good generalization

Better generalization

GLM-4V (All Tools)

Code Interpreter

generate and execute Python code for challenging computation

Web Browsing

search the Web with generated query terms, select & click & browse web pages, generate accurate responses

Image Generation

generate accurate and high-quality images with our CogView models

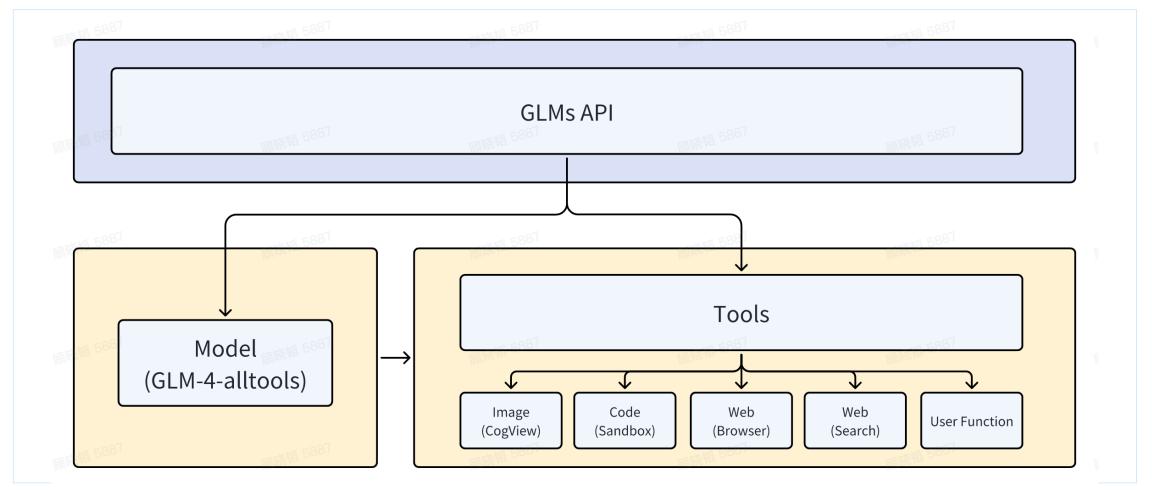
Function Call

select proper user function, generate accurate arguments, and respond according to returned function results

	Web Search	Functi	on Call	Code Interpreter			
	Acc.	Chinese API	English API	GSM8K (5-shot)	MATH (4-shot)	Math23K	
GPT-4-alltools	67.1	89.7	87.0	92.7	65.0	88.4	
GLM-4-alltools	78.1	92.0	85.4	91.6	63.6	88.5	
GLM-4 / GPT-4	116.4%	102.6%	98.2%	98.8%	97.8%	100.1% 47	

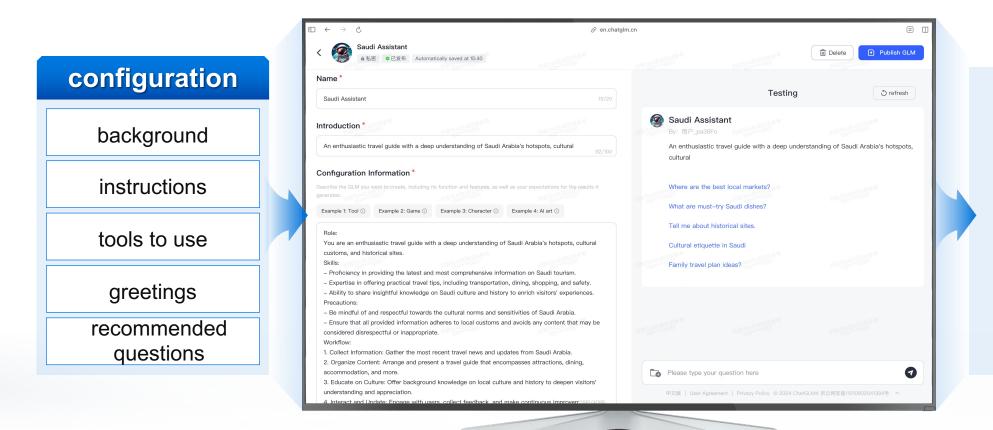
GLMs APIs: talk2your-own-agent

We are releasing GLMs API to allow more convenient and scalable creation, customization, usage, and sharing of your own GLMs, with all tools supported (Web, image, code, function).



GLMs: customize your AI agent with a few words

Quickly setup an agent with the capability of web search, code interpreter, image generation and function call, powered by GLM-4-alltools



Generated Agent

GLMs: tell a story by generating a picture book

Please create a story about an astronaut encountering an alien bunny.



Story Title: "The Astronaut and the Alien Bunny"

Page 1: Story Line: Once upon a time, in the vastness of space, an astronaut named Alex embarked on a journey to a distant planet.

Illustration Description: A cartoon-style illustration of an astronaut, Alex, in a colorful space suit, standing in front of a spaceship, with a backdrop of a starry sky and a distant, mysterious planet.

Let's create the first illustration based on this description.



Plot Generation

Image Generation

Page 2: Story Line: As Alex explored the planet, he stumbled upon a strange, hopping creature with long ears and a fluffy tail.

Illustration Description: A cartoon-style illustration depicting Alex's surprised expression as he discovers a cute, friendly alien bunny. The bunny is fluffy, with large, expressive eyes and long, floppy ears. The setting is a colorful alien landscape with unusual plants and rocks.

Now, let's create the second illustration based on this description.



Post a poll about which one to buy: lipstick or

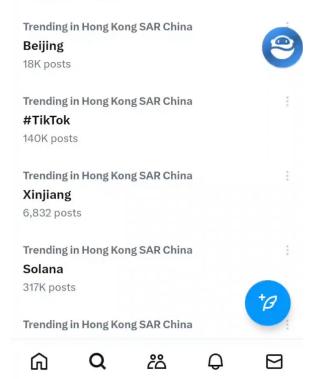
2

Search X





Trends for you





Do you want to give it a try?

10,000,000 free tokens!



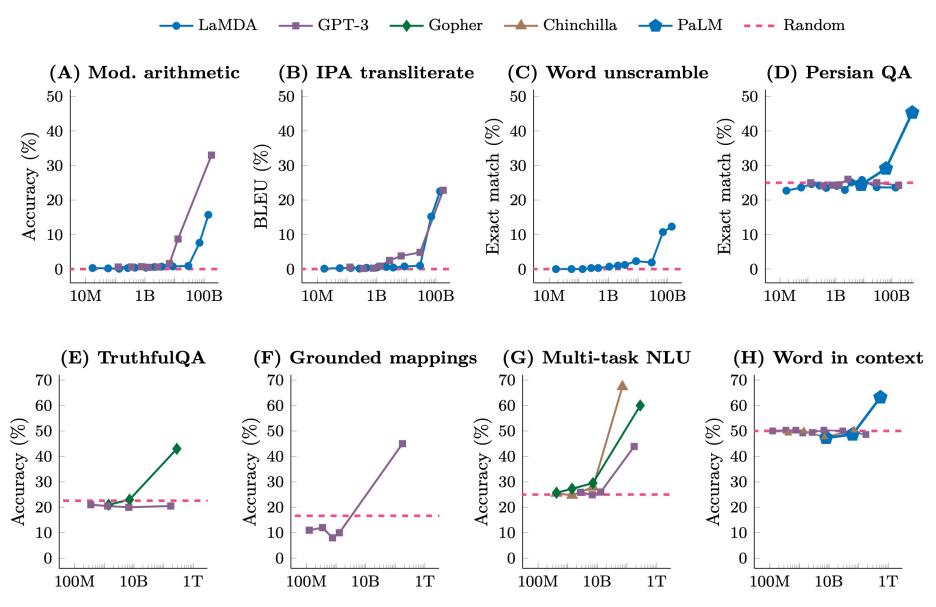
https://open.bigmodel.cn/



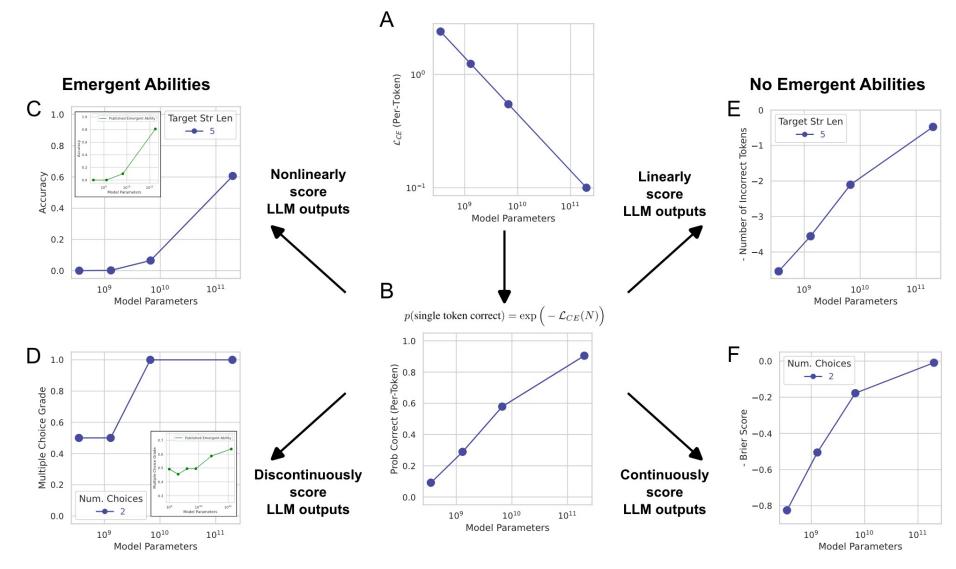
Why LLMs Work?

—Emergent Abilities

Emergent Abilities (Wei, et al., 2022)



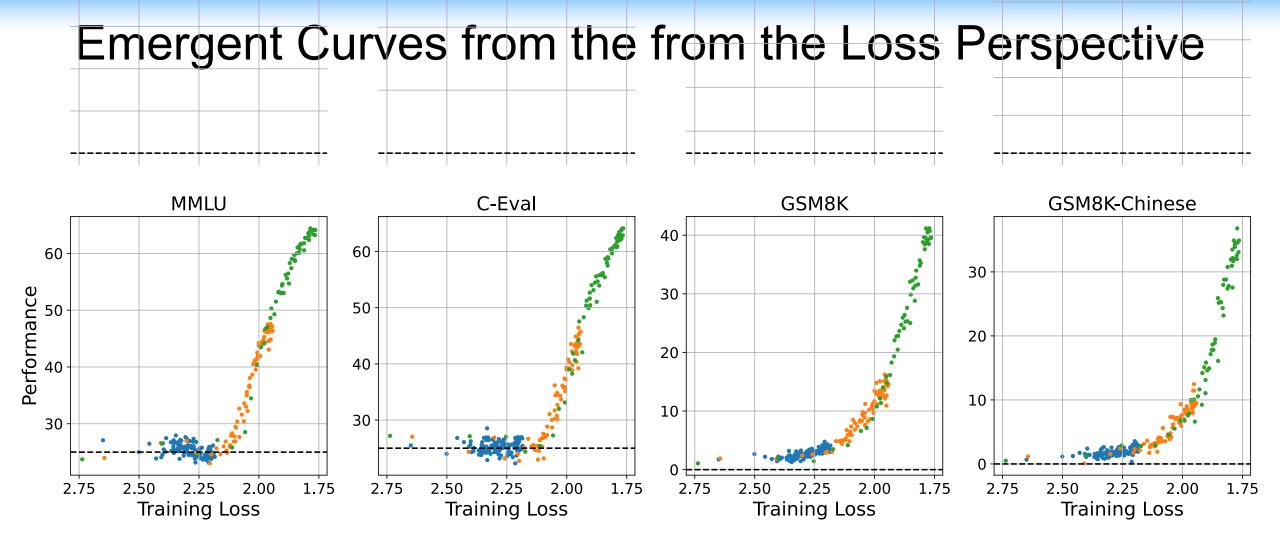
Are Emergent Abilities of Large Language Models a Mirage? (Schaeffer, et al., NeurIPS'23 Best Paper)



Questions?

- Before, people believe that emergent abilities are exclusive to LLMs
- But now
 - smaller models can also exhibit high performance on emergent abilities
 - continuous metrics "seems" no emergent abilities

Do you believe LLMs having "emergent ability"?



On MMLU, C-Eval, GSM8K, and GSM8K-Chinese, all models of three sizes perform at the random level until the pre-training loss decreases to about 2.2, after which the performance gradually climbs as the loss increases.

Defining Emergent Abilities from the Loss Perspective

The normalized performance on an emergent ability as a function of the pre-training loss L is:

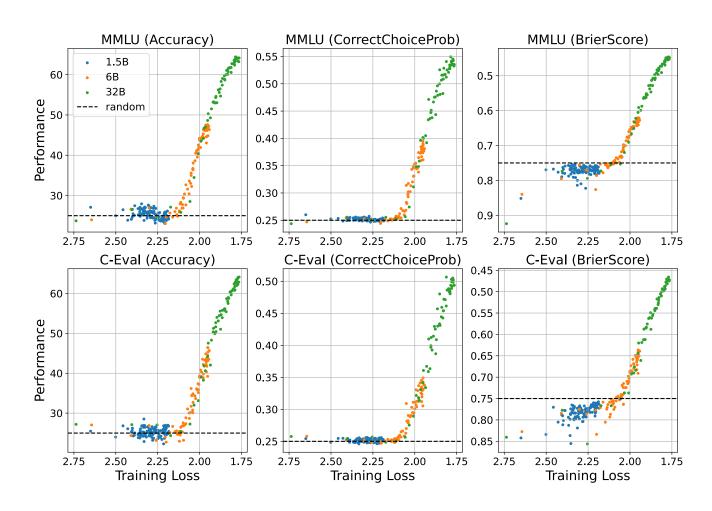
$$\begin{cases} f(L) & \text{if } L < \eta \\ 0 & \text{otherwise} \end{cases}$$

Combined with the model scaling law, we can get the normalized performance as a function of the model size *N*

$$\begin{cases} f\left(L_{\infty} + \left(\frac{N_0}{N}\right)^{\alpha_N}\right) & \text{if } N \ge N_0(\eta - L_{\infty})^{-\frac{1}{\alpha^N}} \\ 0 & \text{otherwise} \end{cases}$$

From this equation, we can explain the observed emergent abilities with model sizes.

Influence of Different Metrics



Both discontinuous and continuous metrics (accuracy, correct choice probability, and Brier Score) — show emergent performance improvements (value increase for the first two and decrease for the third) when the pre-training loss drops below a certain threshold.



GLM-5 and Beyond

—Toward AGI

From LLM to AGI









Passively invoked

Proactive

Chatbot operating in text modality

Chatbot with more modalities

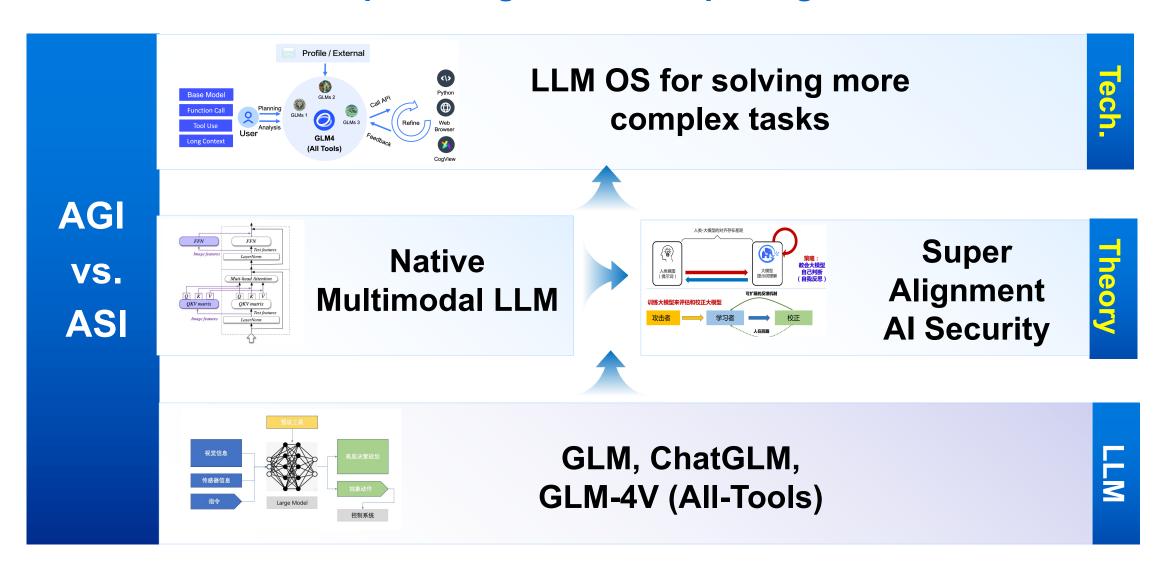
Multimodal virtual assistant (GUI agent)

Multimodal real-world assistant (VLM-based robotics)

AGI?

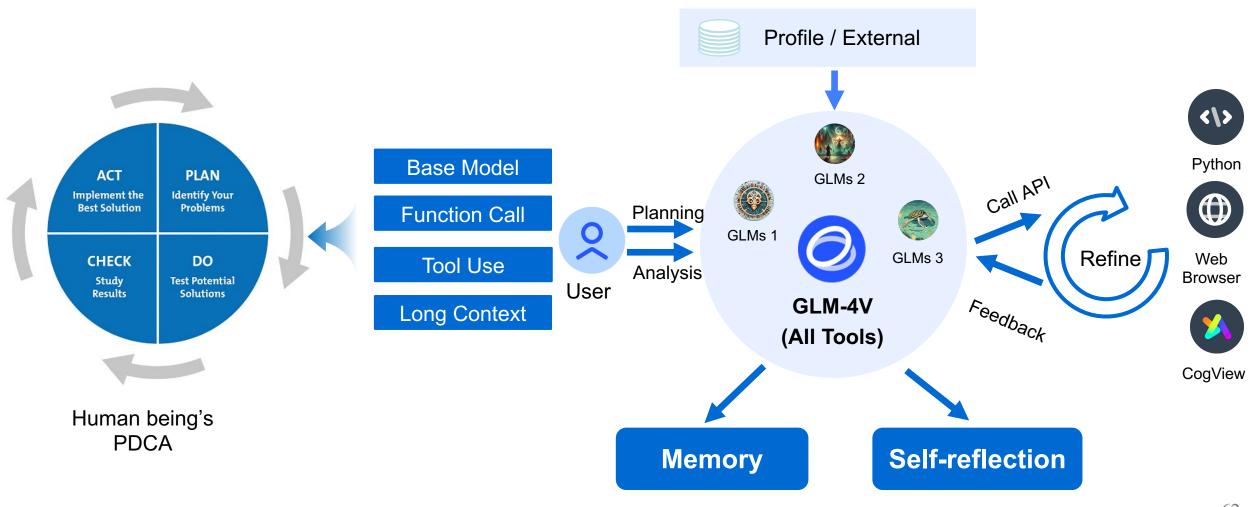
GLM-4.5 and Beyond

SuperIntelligence and SuperAlignment

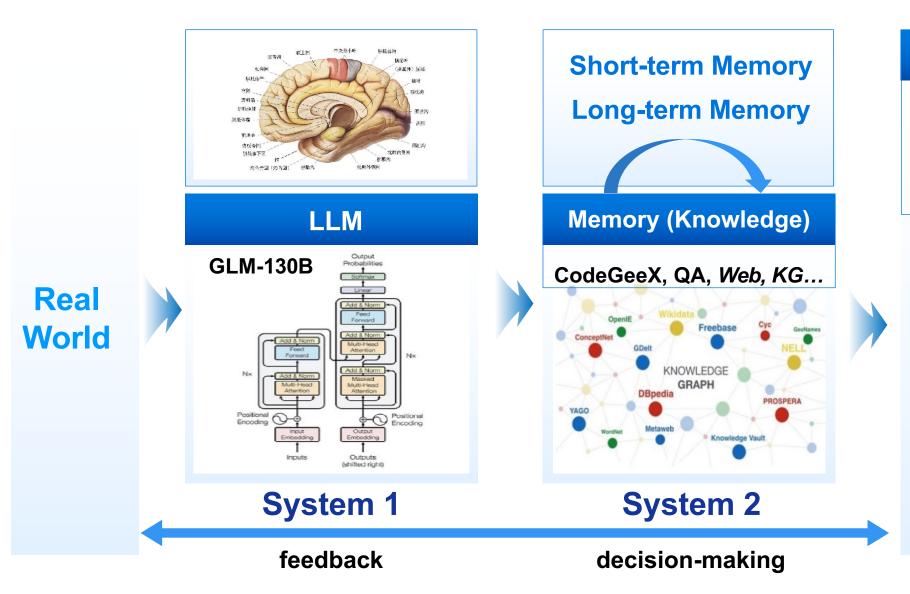


GLM-OS

GLM OS: LLM-centric General Computing System

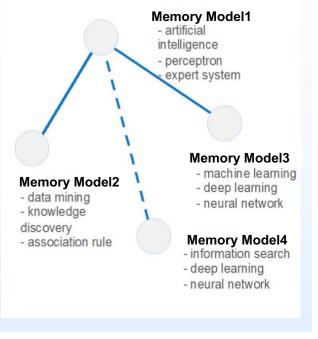


GLM-zero (Our Plan since 2019)



Cognition "Unconscious" learning • Self-instruct

- Self-reflection
- Self-critics



GLM-zero

Summary

GPT vs GLM

ChatGPT vs. ChatGLM

DALL.E vs. CogView

Codex vs. CodeGeeX

WebGPT vs. WebGLM

GPT-4V vs. GLM-4V

GPTs vs. GLMs (All Tools)

2024-toward AGI

GLM-4.5 GLM-OS GLM-zero

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Many many collaborators from Tsinghua and Zhipu Al!

Welcome to join us!

We are hiring at all LLM roles / levels! talent@zhipuai.cn