



Lecture 6: Mid-review of Final Project

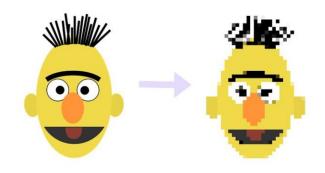
Winter 2023
Benyou Wang
School of Data Science

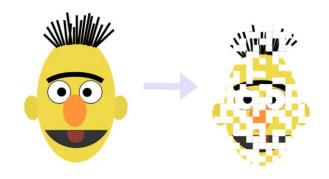
Recap

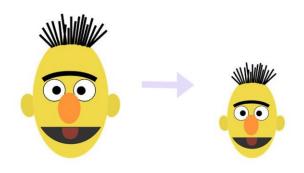
Outline

- 1. Overview of Efficiency in LLMs
- 2. Efficiency before the LLM Era
 - Efficiency beyond Transformer (Quantization, Pruning, Knowledge distillation etc.)
 - Efficiency within Transformer (Sparsity e.g., Mixture of Expert and efficiency in long context)
 - Parameter-efficient finetuning (Lora, adapter, prompt, MOE) and modularization
- 3. Efficiency after the LLM Era
 - Memory-efficient Training: ZeRO, LOMO, Distributed Training, Flash-Attention, Lora/QLORA, Language model Inference: Early existing and Speculative decoding
- 4. Future direction

Efficiency beyond Transformer







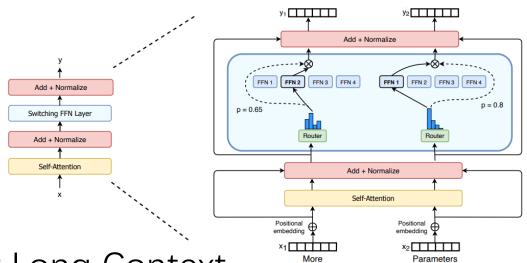
Quantization"Low resolution"

Pruning Knowledge distillation
Removing weight connectionsLearning smaller models from big one

https://rasa.com/blog/compressing-bert-for-faster-prediction-2/

Efficiency within Transformer

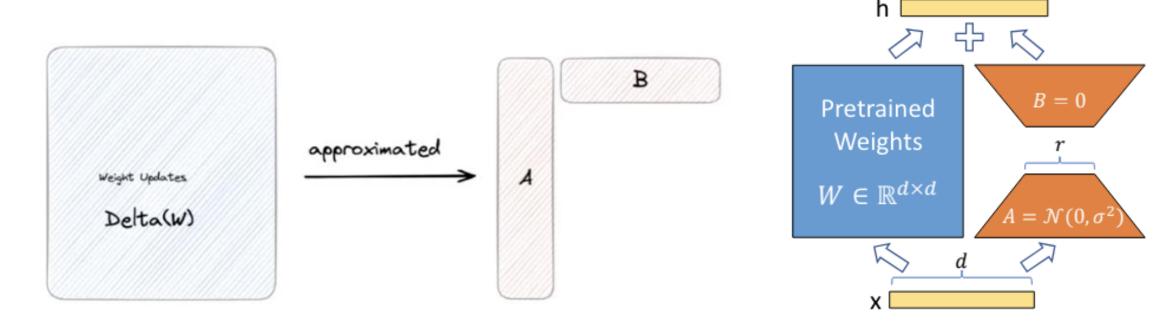
• Sparsity (e.g., Mixture of Expert [1])



- Efficiency for Long Context
 - Computing complexity of is $O(N^2D)$, which is quadratic to the sequence length

William Fedus, Barret Zoph, Noam Shazeer. Switch Transformers: Scaling to Trillion Parameter Models with Simple and Efficient Sparsity. https://arxiv.org/pdf/2101.03961.pdf

LoRA for memory-efficiency



For a pre-trained weights W_0, we approx Delta(w) by

 $h = W_0 x + \Delta W x = W_0 x + BAx$ where $B \in \mathbb{R}^{d \times r}, A \in \mathbb{R}^{r \times k}$, and the rank $r \ll \min(d, k)$

During Training, we only compute gradient w.r.t. Delta(W)

Mid-review of Final Projects

Objective of this course

- Knowledge: a) Students will understand basic concepts and principles of LLM; b) Students could effectively use LLMs for daily study, work and research; and c) Students will know which tasks LLMs are suitable to solve and which are not.
- **Skills**: a) Students could train a toy LLM following a complete pipeline and b) Students could call ChatGPT API for daily usage in study, work and research.
- Valued/Attitude: a) Students will appreciate the importance of data;
 b) Students will tend to use data-driven paradigm to solve problems;
 and c) Students will be aware of the limitations and risks of using ChatGPT.

Today's contents

- Philosophy of the final project
- Research in our team
 - Past: What we have done
 - Present and Future: What we are doing and will do
- More on LLM tendency
- Insights from Assignments
- Possible topics for final projects
- Tips of report (paper) writing

Philosophy of the final project

Components and grading

- Assignments (40%)
 - > Assignment 1 (20%): Using API for testing prompt engineering
 - > Assignment 2 (20%): A toy LLM application

Both assignments need a report and code attachment if it has coding. See the relevant evalution criterion as the final project.

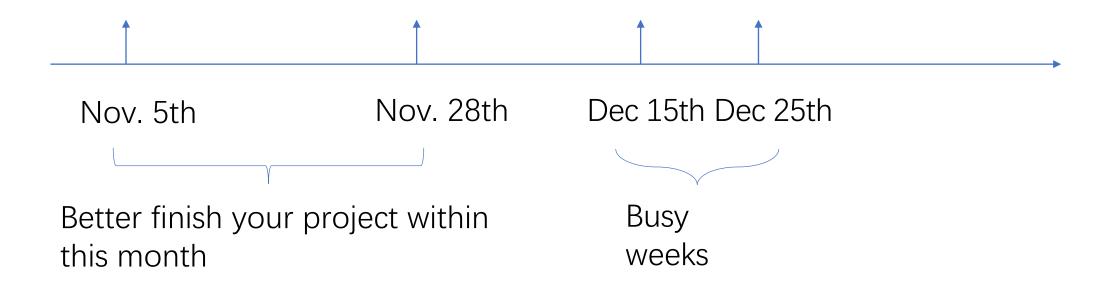
- * Review of project proposal (15%)
 We will have a review for project proposals, to assist students better prepare their final projects. A revision is welcome after taking our suggestions into consideration.
- Final project (40%)

You need to write a project report (max 6 pages) for the final project. You are also expected to make a project poster presentation. After the final project deadline, feel free to make your project open source; we appreciate if you acknowledge this course

Participation (5%)

Final project (Dates might be changed)

- Students complete a research project in teams of 1-3
- Draft proposal deadline: Nov. 5th 11:59pm (simply explain your idea in one page)
- Final proposal deadline: Nov. 28th 11:59pm (TAs will provide suggestions for the revision!)
- In-class presentation: Dec 15th
- Final paper deadline: Dec 25th (final date)



Final project

- Typical projects (we will release a detailed list later):
 - #1: Train or fine-tune a medium-sized language model (e.g., T5, Bloom, TinyLLaMA, Baichuan, LLaMA) yourself for any problem of your interest. Check out HuggingFace's model hub!
 Hugging Face https://huggingface.co/models
 - #2: Evaluate one of the largest language models (e.g., ChatGPT/GPT4) and understand their capabilities, limitations and risks.
 - #3 A plugin that works with a existing popular LLM like ChatGPT and Phoenix
 - #4 Release a new LLM (10B+) and have some impact
 - A direct objective is that your GitHub repository gets more than 100+ GitHub stars
 - #5 A application (We have an example application, search "神仙湖" in WeChat)
 https://openai.com/api/
 https://opt.alpa.ai
 **Tou might get computing resources to train 10B+ model if Tas like your proposal

Projects

- Training model
 - Select a task
 - Prepare the data
 - Train your LLMs
 - Evaluate it

- Use ChatGPT API
 - find a scenario
 - Use LLM API
 - Test (evaluate) it

Hybrid

0



You could find your own way to build LLMs or LLM applications.

Past research in our group

Category

- Releasing large language models
- Dataset and benchmark
- Methods in LLMs
- Insightful research (interesting but not useful)
- Application and demos (useful but engineering-oriented)

Past research in our group (1) Releasing large language models

LLMs

- Phoenix
- HuatuoGPT -- the SOTA Chinese medical LLM
- AceGPT -- the SOTA Arabic LLM
- HuatuoGPT-II (it will be released this month)
- Phoenix-II (multimodal LLM)
- Phoenix Lite (best 1B model)
- AceGPT-II (10B)
- HuatuoGPT-III (multilingual + multi-modal)

First try: phoenix

SuperCLUE琅琊榜 点击加入对战 o 我们使用Elo评级系统来计算模型的相对性能。我们将定期发布新的排行榜。如果你想看到更多的模型,请联系我们:CLUEbenchmark@163.com o [文章][GitHub] **Elo Rating** Description MiniMax-abab5 由 MINIMAX 公司开发,帮助人类高效写作、激发创意、获取知识、做出决策。 1250 W gpt-3.5-turbo 由 OpenAI 开发, 当前大家最频繁使用的版本。 1202 W ChatGLM-130B 1180 由清华大学和智谱AI开发,一个开放的双语言对话模型。 ChatGLM2-6B 1105 由清华大学和智谱AI开发,是开源中英双语对话模型 ChatGLM-6B 的第三代版本。 由香港中文大学(深圳)开发,一个基于 Bloomz 微调而来的多语言聊天助手。 由IDEA研究院CCNL开发,从LLaMA-13B开始重新构建中文词表,进行了千亿token量级的预训练。 idea-jiangziya 1069 moss-moon-003-sft 由复旦大学自然语言处理实验室开发,是一个支持中英双语和多种插件的开源对话语言模型。 RWKV-4-World-7B 由RWKV基金会开发,结合了RNN和Transformer,RWKV-4-World为新系列,在100多种世界语言上进行训练。 Longjing-7B 1012 ChatYuan 下的新分支大语言模型。 由 LMSYS 开发,基于LLaMA并利用用户共享对话进行微调得来。 vicuna-13b 1001 由 LianjiaTech 开发,基于 BLOOM 和 LLAMA 针对中文进行了优化。。 由大数据系统计算技术国家工程实验室开发,在400万指令数据集上对中文基础模型进行指令微调而得到的 Linly-ChatFlow-7B 910 由RWKV基金会开发,结合了 RNN 和 Transformer, 具有运行速度较快、占用显存较少、训练用时较少等特点 RWKV-4-Raven-7B 由百川智能开发, 一个开源可商用的大规模预训练语言模型, 在大约 1.2 万亿 tokens 上训练, 支持中英双语。 由香港科技大学统计和机器学习实验室团队开发,该模型是在定制数据集上的 pinkmanlove/llama-7b-hf 的微调版本。 来自 bigscience 的 BLOOMZ & mT0 系列,这是一个能够以数十种语言执行人类指令的模型家族,

Phoenix ranked in 5th in SuperClue benchmark https://www.superclueai.com/

琅琊榜

- 落地到深圳市人才工作局,为海内外人才提供关于深圳的政策查询、政策速配、服务资讯等多种功能服务。见 http://123.57.133.182:51008/
- Deployed in the campus https://phoenix.cuhk.edu.cn

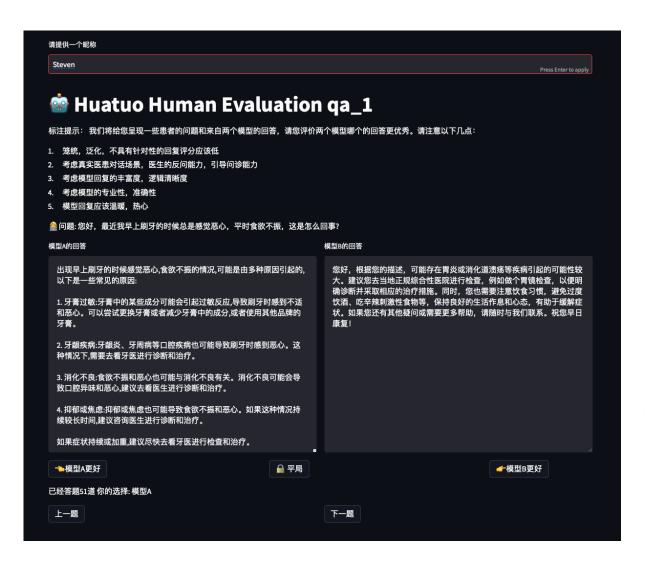
Zhihong Chen, Feng Jiang, Junying Chen, Tiannan Wang, Fei Yu, Guiming Chen, Hongbo Zhang, Juhao Liang, Chen Zhang, Zhiyi Zhang, Jianquan Li, Xiang Wan, Benyou Wang, Haizhou Li. Phoenix: Democratizing ChatGPT across Languages. https://arxiv.org/abs/2304.10453

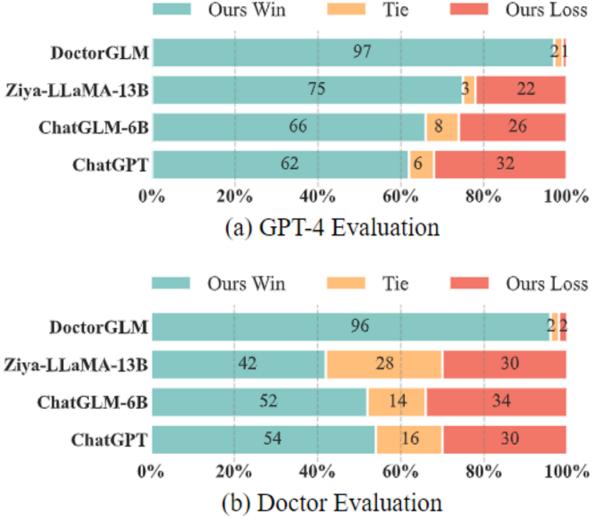
Human evaluation for phoenix



Human evaluation: **best** performance among open-sourced Chinese models

HuatuoGPT: Human & GPT-4 Evaluation









英国著名媒体金融时报(Financial Times, www.ft.com)通过炒作地缘政治的方式报道了我们的最新阿拉伯语大模型AceGPT。

AceGPT is currently the best open source Arabic large model, even surpassing ChatGPT in the Vicuna evaluation. It has completely completed the pre-training, fine-tuning and RLAIF pipeline, especially in terms of localization, with additional adaptations. Over 2K downloads on Huggingface in two weeks

https://www.ft.com/content/2a636cee-b0d2-45c2-a815-11ca32371763

Huang Huang, Fei Yu, Jianqing Zhu, Xuening Sun, Hao Cheng, Dingjie Song, Zhihong Chen, Abdulmohsen Alharthi, Bang An, Ziche Liu, Zhiyi Zhang, Junying Chen, Jianquan Li, **Benyou Wang**, Lian Zhang, Ruoyu Sun, Xiang Wan, Haizhou Li, Jinchao Xu. AceGPT, Localizing Large Language Models in Arabic. https://arxiv.org/abs/2309.12053

Huatuo-II

	中医							临床					药师					
Model	执业助理医师			执业医师			执业助理医师		执业医师		执业中药师			执业西药师		Avg.		
	2015	2016	2017	2012	2013	2016	2018	2019	2020	2018	2019	2020	2017	2018	2019	2021	2022	03.00
DoctorGLM	3.0	1.4	3.5	1.8	1.8	2.1	2.6	3.3	1.6	1.6	2.7	1.8	1.8	2.4	2.2	3.1	3.8	2.4
ChatGLM-Med	20.7	23.2	20.7	21.8	21.8	22.6	30.3	23.0	29.9	18.5	20.4	24.8	16.7	21.0	15.3	20.8	22.5	22.0
BenTsao	23.3	26.8	17.2	19.0	19.5	22.1	18.4	24.6	27.5	21.6	18.9	18.1	20.2	20.4	18.8	19.0	22.1	21.0
BianQue-2	3.7	2.3	3.5	4.2	4.5	4.0	4.7	5.3	1.6	3.8	5.9	3.7	7.1	4.2	9.0	4.6	4.4	4.5
ChatMed-Consult	20.0	17.3	14.7	21.3	18.2	20.0	27.8	21.3	19.3	23.8	21.4	19.5	16.7	15.0	17.5	22.3	24.2	20.0
MedicalGPT	25.0	24.1	21.6	26.3	27.0	27.0	38.9	29.1	28.3	33.4	32.1	26.2	22.6	21.0	20.6	24.8	24.8	26.6
DISC-MedLLM	38.3	41.8	26.7	36.2	38.7	35.1	49.2	36.1	41.8	41.4	36.6	35.8	25.0	22.2	22.0	25.6	25.6	33.0
HuatuoGPT	26.0	30.9	32.8	31.3	26.8	30.2	32.5	33.6	27.5	32.7	28.6	30.1	18.4	27.5	25.1	26.0	27.1	28.7
ChatGLM2-6B	47.0	50.5	44.0	44.8	48.3	48.1	62.4	49.2	52.9	50.1	48.5	45.6	36.9	36.5	38.6	34.8	35.6	45.5
ChatGLM3-6B	45.7	45.0	50.0	46.3	45.8	46.5	50.9	48.8	43.9	41.7	43.9	43.6	42.3	28.1	35.4	38.3	35.8	43.0
LLaMA2-7B-Chat	24.0	22.3	23.3	27.2	26.0	22.3	29.1	29.9	27.9	25.8	27.3	25.2	17.3	23.4	20.6	24.0	21.0	24.0
LLaMA2-13B-Chat	23.0	26.8	24.1	27.0	27.8	25.4	29.9	29.5	27.5	26.5	29.0	25.7	21.4	27.0	21.1	24.8	22.3	25.0
Baichuan2-7B-Chat	57.3	57.3	58.6	55.7	58.5	57.9	61.1	55.7	55.3	51.0	53.6	50.0	41.7	41.9	45.7	44.6	42.7	52.0
Baichuan2-13B-Chat	64.7	58.2	62.9	61.7	61.5	63.3	66.2	64.8	63.1	65.9	58.8	61.5	54.2	38.9	48.4	52.5	45.2	58.0
Qwen-7B-Chat	54.7	55.9	56.0	52.7	53.5	54.4	68.8	63.9	57.8	60.6	57.6	54.1	44.0	33.5	43.0	47.5	43.8	53.0
Qwen-14B-Chat	65.3	63.2	67.2	64.8	63.3	67.9	74.8	75.0	69.3	73.7	69.7	68.8	54.8	49.1	52.5	54.0	52.7	63.0
Spark(星火)	43.3	38.2	39.7	38.3	38.2	42.6	54.7	44.7	50.0	51.4	44.8	40.8	29.2	26.4	30.9	35.0	34.4	40.2
文心一言	73.3	66.3	73.3	70.0	71.8	66.7	78.2	77.0	77.5	66.6	70.8	74.1	55.9	50.3	60.0	55.0	53.5	67.1
ChatGPT (API)	46.0	36.4	41.4	36.7	38.5	40.5	63.3	57.8	53.7	53.7	52.5	51.8	32.1	28.1	30.0	41.7	38.8	43.0
GPT-4 (API)	47.3	48.2	53.5	50.3	53.7	54.2	79.9	72.5	70.9	74.8	73.1	68.4	41.1	43.7	48.0	58.3	60.0	58.7
HuatuoGPT-II	71.0	70.9	70.7	69.5	71.3	70.5	73.1	75.4	71.7	71.1	71.0	71.1	57.7	49.1	53.8	64.2	62.3	67.3

Table 3: The results of the Chinese Medical Licensing Examination Results.

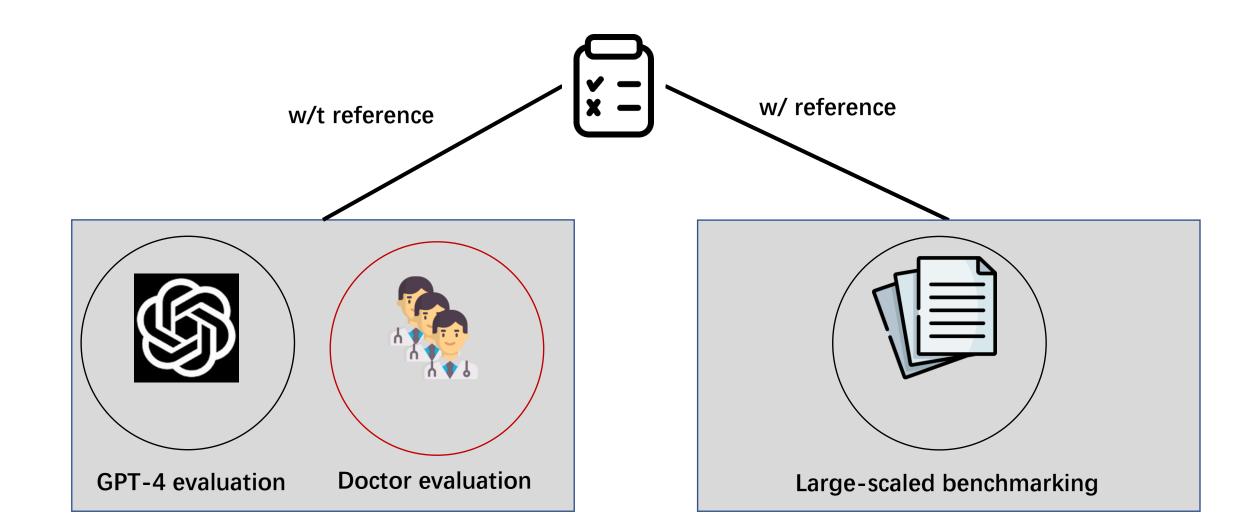
Past research in our group (2) Dataset and benchmark

三军未到,粮草先行

Dataset and benchmark

- Huatuo-26M
- CMB
- Benchmarking information extraction
- Medical Flan
- Medical Pajama

Evaluation



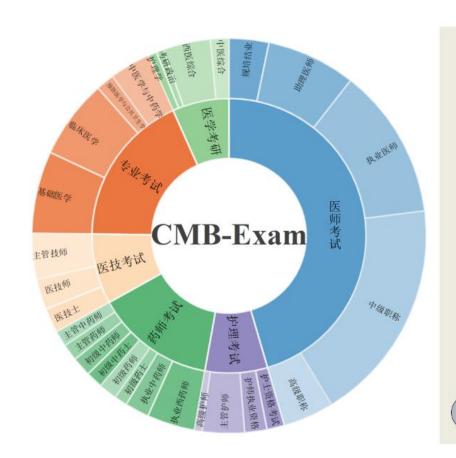
Huatuo-26M

The largest medical QA dataset

Dataset	Lang	Domain	Source	#Q
MedHop (Welbl et al., 2018)	English	Medical	MEDLINE	2.5K
BiQA (Lamurias et al., 2020)	English	Medical	Online Medical forum	7.4K
HealthQA (Zhu et al., 2019)	English	Medical	Medical-services website	7.5K
MASH-QA (Zhu et al., 2020)	English	Medical	Medical article website	35K
MedQuAD (Ben Abacha and Demner-Fushman, 2019)	English	Medical	U.S. National Institutes of Health (NIH)	47K
ChiMed (Tian et al., 2019)	Chinese	Medical	Online Medical forum	47K
MedQA (Jin et al., 2020)	EN&CH	Medical	Medical Exam	60K
webMedQA (He et al., 2019)	Chinese	Medical	Medical consultancy websites	63K
CliCR (Šuster and Daelemans, 2018)	English	Medical	Clinical case reports	100K
cMedQA2 (Zhang et al., 2018)	Chinese	Medical	•	108K
Huatuo-26M	Chinese	Medical	Consultation records, Encyclopedia, KBs	26M
TriviaQA (Joshi et al., 2017)	English	General	Trivia	96K
HotpotQA (Yang et al., 2018)	English	General	Wikipedia	113K
SQuAD (Rajpurkar et al., 2016)	English	General	Wikipedia	158K
DuReader (He et al., 2017)	Chinese	General	Web search	200K
Natural Questions (Kwiatkowski et al., 2019)	English	General	Wikipedia	323K
MS MARCO (Nguyen et al., 2016)	English	General	Web search	1.0M
CNN/Daily Mail (See et al., 2017)	English	General	News	1.3M
PAQ (Lewis et al., 2021)	English	General	Wikipedia	65M

Jianquan Li, Xidong Wang, Xiangbo Wu, Zhiyi Zhang, Xiaolong Xu, Jie Fu, Prayag Tiwari, Xiang Wan, Benyou Wang. Huatuo-26M, a Large-scale Chinese Medical QA Dataset. arXiv preprint arXiv:2305.01526

Comprehensive Medical benchmark (CMB)





病史摘要: 田XX, 女, 61岁,

上腹痛1个月,加重3天,无 明显黄疸…

主诉:上腹痛1个月,加重3天

体格检查: 自主体位, 神志清 楚, 巩膜略有黄染, 全身浅 表淋巴结无肿大。两肺呼吸 音清晰,未闻及干湿啰音… 辅助检查:

- (1)实验室检查:血常规 WBC 8.46×109/L ···
- (2)多普勒超声检查: 胆囊大 小约9cm×4cm···
- (3)CT检查: 胆总管扩张, 见 高密度结节影,最大直径…



分析本例病人的病史、体格检查和辅助检查。



(1)病史分析: 该病例病史比较典型, 表现为上腹痛…

(3)辅助检查分析: 本例病人实验室检查肝功能…







简述本例病人的治疗原则。





(LCBDE)治疗,恢复快、损伤小、疼痛轻、瘢痕不易 发现。直径小于1cm的胆总管结石行EST治疗…

Overview

Category	Subcategory	# Subject	# Questions
Physician (医师)	Resident Physician (住院医师); Licensed Assistant Physician (执业助理医师); Licensed Physician (执业医师); Associate Professional Physician (中级职称); Advanced Professional Physicians (高级职称)	81	124,926
Nurse (护理)	Practicing Nurse (护士); Licensed Practical Nurse (护师); Charge Nurse (主管护师); Advanced Practice Nurse (高级护师)	8	16,919
Technicians (医技)	Medical Technician (医技士); Medical Technologist (医技师); Supervising Technologist (主管技师)	21	27,004
Pharmacist (药师)	Licensed Pharmacist (执业西药师); Licensed TCM Pharmacist (执业中药师); Junior Pharmacist (初级药师); Junior Pharmacist Assistant (初级药士); Junior TCM Pharmacist (初级中药师); Junior TCM Pharmacist Assistant (初级中药士); Chief Pharmacists (主管药师); Chief TCM Pharmacists (主管中药师)	8	33,354
Undergraduate Disciplines (学科考试) ¹	Fundamental Medicine (基础医学); Clinical Medicine (临床医学); Traditional Chinese (TCM) and Chinese Herbal Medicine (中医学与中药学); Preventive Medicine and Public Health (预防医学与公共卫生学)	53	62,271
Graduate Entrance Exam (考研)	Integrated Western Medicine (西医综合); Integrated TCM (中医综合); Political Science (政治); Nursing (护理学)	5	16,365
Total	28	176	280,839

We referenced the National Standard Subject Classification of the People's Republic of China, see https://xkb.pku.edu.cn/docs/ Table 1: Statistics of the CMB-Exam Categories, Subcategories, Subjects, and Questions.

Results

	中医						临床						药师					
Model	执	业助理医	师		执业医师		抉	业助理医	师		执业医师		ž	丸业中药炉	币	执业	西药师	平均
	2015年	2016年	2017年	2012年	2013年	2016年	2018年	2019年	2020年	2018年	2019年	2020年	2017年	2018年	2019年	2021年	2022年	
GPT-4	47.33	48.18	53.45	50.33	53.67	54.19	79.91	72.54	70.90	74.83	73.11	68.35	41.07	43.71	47.98	58.33	60.00	58.70
HuatuoGPT-II(华佗-II)	54.33	50.45	60.34	53.00	49.67	53.02	61.54	54.10	54.92	53.67	50.21	50.46	50.60	47.31	52.02	49.58	46.88	52.48
ChatGLM2-6B	47.00	50.45	43.97	44.83	48.33	48.14	62.39	49.18	52.87	50.11	48.53	45.64	36.90	36.53	38.57	34.79	35.62	45.52
Baichuan-7B-chat	42.00	44.09	43.10	39.67	43.33	41.16	54.70	48.36	44.26	42.76	42.65	39.45	35.71	38.32	30.49	39.79	40.83	41.80
ChatGPT	39.33	35.00	38.79	35.33	39.33	39.77	59.40	49.18	47.95	51.00	46.22	47.94	23.81	30.54	31.39	40.00	37.29	40.72
Spark(星火)	43.33	38.18	39.66	38.33	38.17	42.56	54.70	44.67	50.00	51.45	44.75	40.83	29.17	26.35	30.94	35.00	34.38	40.15
HuatuoGPT(华佗)	26.00	30.91	32.76	31.33	26.83	30.23	32.48	33.61	27.46	32.74	28.57	30.05	18.45	27.54	25.11	26.04	27.08	28.66
MedicalGPT	25.00	24.09	21.55	26.33	27.00	26.98	38.89	29.10	28.28	33.41	32.14	26.15	22.62	20.96	20.63	24.79	24.79	26.63
ChatGLM-Med	20.67	23.18	20.69	21.83	21.83	22.56	30.34	22.95	29.92	18.49	20.38	24.77	16.67	20.96	15.25	20.83	22.50	21.99
BenTsao(本草)	23.33	26.82	17.24	19.00	19.50	22.09	18.38	24.59	27.46	21.60	18.91	18.12	20.24	20.36	18.83	18.96	22.08	21.03
ChatMed-Consult	20.00	17.27	14.66	21.33	18.17	20.00	27.78	21.31	19.26	23.83	21.43	19.50	16.67	14.97	17.49	22.29	24.17	20.01
BianQue-2(扁鹊)	3.67	2.27	3.45	4.17	4.50	3.95	4.70	5.33	1.64	3.79	5.88	3.67	7.14	4.19	8.97	4.58	4.38	4.49
DoctorGLM	3.00	1.36	3.45	1.83	1.83	2.09	2.56	3.28	1.64	1.56	2.73	1.83	1.79	2.40	2.24	3.12	3.75	2.38

CMB online benchmark

or detailed ir	nformation on	generati	ion and e	evaluation	n, please	refer to t	ne link.					Home	Data	Paper	• (
					医师考试				护理	考试					
Model	Institution	Avg.	规培结 业	执业助 理医师	执业医 师	中级职称	高级职称	护士执业资格	护师执业资格	主管护师	高级护师	执业西 药师	执业中 药师	初级药士	初
ShuKunGPT	数坤科技	64.44	63.00	76.50	81.00	64.50	58.25	77.50	78.50	74.75	55.00	73.75	70.50	76.00	7;
GPT-4	<u>OpenAl</u>	58.37 (58.37)	59.75 (59.75)	58.5 (58.50)	64.50 (64.50)	60.75 (60.75)	56.00 (56.00)	77.50 (77.50)	72.50 (72.50)	68.75 (68.75)	58.50 (58.50)	54.75 (54.75)	47.00 (47.00)	60.00 (60.00)	6; (6;
ChatGLM2- 6B	THUDM	44.91 (45.70)	43.50 (43.50)	43.75 (46.38)	48.25 (48.25)	47.25 (47.37)	39.25 (39.25)	54.25 (56.07)	63.50 (65.13)	51.50 (52.02)	46.25 (46.25)	36.50 (36.71)	38.25 (40.11)	49.50 (50.51)	48 (48
Baichuan- 13B-chat	Baichuan- inc	41.63 (42.68)	38.50 (39.29)	39.00 (40.94)	43.75 (46.30)	38.75 (40.32)	35.00 (36.77)	53.75 (54.71)	54.75 (56.30)	46.25 (47.19)	38.50 (39.23)	37.50 (38.66)	37.75 (38.42)	53.00 (53.67)	4:
IvyGPT	Macao Polytechnic University	38.54	36.75	37.75	42.00	42.00	30.00	51.00	47.00	46.25	30.00	36.75	40.50	47.25	38
ChatGPT	<u>OpenAl</u>	38.09 (38.09)	42.75 (42.75)	39.50 (39.50)	43.25 (43.25)	41.00 (41.00)	37.25 (37.25)	53.25 (53.25)	46.75 (46.75)	46.25 (46.25)	36.50 (36.5)	35.75 (35.75)	33.00 (33.00)	43.00 (43.00)	44
Sunsimiao	X-D Lab	35.32	38.25	36.25	45.75	33.00	30.75	46.00	43.50	42.25	28.00	35.25	34.75	42.00	36
HuatuoGPT (华佗)	CUHK(SZ)- NLP	31.38 (32.06)	32.00 (32.00)	34.50 (34.87)	33.25 (33.25)	31.25 (31.33)	28.25 (29.89)	37.00 (40.00)	38.00 (38.19)	36.00 (36.83)	29.00 (29.44)	28.00 (29.86)	28.75 (29.11)	36.50 (36.50)	3.
MedicalGPT	Xu Ming	26.45 (26.95)	25.00 (25.06)	25.75 (30.00)	29.25 (29.62)	29.50 (29.72)	25.75 (25.75)	34.75 (34.75)	35.75 (35.75)	32.25 (32.25)	21.25 (21.36)	23.75 (23.87)	24.00 (25.33)	27.50 (27.50)	2! (2!
ChatMed- Consult	<u>华东师范大</u> 学	21.71 (25.69)	20.50 (22.40)	24.00 (25.74)	21.25 (100.00)	22.50 (23.75)	18.75 (20.22)	24.75 (25.71)	25.50 (27.72)	23.00 (24.34)	17.25 (19.06)	19.25 (21.33)	23.75 (23.81)	24.25 (25.87)	20
Bentsao (本 草)	SCIR-HI	21.25 (23.21)	24.50 (24.50)	23.00 (23.29)	24.25 (24.43)	20.50 (20.55)	18.25 (18.30)	24.00 (44.44)	23.25 (23.88)	22.25 (29.41)	15.25 (16.74)	18.25 (18.25)	20.75 (21.84)	25.00 (29.03)	2; (2;
ChatGLM- Med	SCIR-HI	20.67 (22.79)	21.50 (23.37)	22.50 (40.00)	23.75 (25.33)	21.50 (23.56)	19.50 (20.97)	23.50 (25.00)	25.75 (27.18)	24.00 (27.01)	15.00 (15.75)	19.25 (20.65)	19.75 (20.57)	20.75 (21.73)	18 (19
DoctorGLM	<u>上海科技大</u> 学	7.63 (25.36)	6.00 (21.92)	8.00 (29.63)	6.00 (21.43)	9.00 (27.27)	5.75 (20.83)	6.00 (29.27)	11.50 (34.69)	6.75 (30.00)	5.00 (22.45)	7.00 (24.35)	6.00 (29.41)	9.50 (24.52)	1((2
BianQue-2 (扁鹊)	<u>华东师范</u> 大 堂	7.26 (30.46)	7.50 (18.99)	9.00 (33.33)	8.00 (18.33)	8.75 (22.92)	6.00 (21.84)	6.75 (21.85)	7.00 (66.67)	7.50 (23.53)	5.25 (16.94)	7.75 (25.53)	6.75 (20.00)	6.75 (50.00)	6 (18

https://cmedbenchmark.llmzoo.com/static/leaderboard.html

Doctor Assistant

- Patients have no discerning ability, so using it on them is risky and has many ethical issues.
- LLMs do not aim to replace doctors, but to be an assistant for doctors

阶段	重点任务	任务描述
入院	导诊	按照患者的症状描述提供其该去的科室建议
诊断	问诊 初诊 差异诊断 开具检查 检查结果解读 疾病分类分级 确诊	根据病人的病史体征进一步问询得到 根据患者的病史、临床表现及简单检查给出可能的诊断 给出患者的差异诊断 为明确诊断,为患者开具实验室检查或影像学检查等 患者的检查结果解读 分析病情进展 给出最后的确诊
治疗	紧急情况处置 治疗方建议 用药建分析 并发症分析 进一步必要性分析 手术必要性分析 术前教育 手术方案制定	紧急情况处置,包括急诊处置、手术中出现的紧急情况为患者制定合适的治疗方案给患者具体的用药建议分析患者可能出现的并发症、需要预防的并发症等在患者已经过一段治疗后可能需要的治疗调整/并发症处理因手术治疗的特殊性,确认患者是否需要手术治疗给患者的术前教育为患者制定详细的手术方案
出院	健康指导 随访方案	给患者健康指导 为患者制定随访方案

https://github.com/FreedomIntelligence/DotaGPT

Data collection

阶段	重点任务	任务描述	instruction	size
入院	导诊	按照患者的症状描述提供其该去的科室建议	请为患者选择合适的就诊科室。	8674
	问诊提示	根据病人的病史体征进一步问询得到	为明确诊断,请考虑患者病史、体征等情况进行询问。	141
	初诊	根据患者的病史、临床表现及简单检查给出可能的诊断	请你全面分析患者的病史及临床表现并给出诊断。	15580
	鉴别诊断	给出患者的鉴别诊断	请给出鉴别诊断。	289
诊断	开具检查	为明确诊断,为患者开具实验室检查或影像学检查等	为明确诊断,请问患者下一步需要做哪些检查?	9251
	检查结果解读	患者的检查结果解读	请详细解读患者的检查结果。	1631
	疾病分类分级	分析病情进展	请根据患者信息进行疾病分类分级。	504
	确诊	给出最后的确诊	请诊断并分析患者的病史、临床表现及检查。	1931
1 .5	紧急情况处置	紧急情况处置,包括急诊处置、手术中出现的紧急情况	请说明患者的紧急处理措施	162
	治疗方案制定	为患者制定合适的治疗方案	根据患者情况制定治疗方案。	6912
	用药建议	给患者具体的用药建议	请根据患者信息给出具体的药物建议。	3148
治疗	并发症分析	分析患者可能出现的并发症、需要预防的并发症等	结合病人情况分析可能出现的并发症。	823
11174	进一步的治疗方案	在患者已经过一段治疗后可能需要的治疗调整/并发症处理	结合患者目前治疗情况给出进一步治疗方案。	712
	手术必要性分析	因手术治疗的特殊性,确认患者是否需要手术治疗	请分析手术的必要性并说明理由。	110
	术前教育	给患者的术前教育	请对患者进行术前教育。	4
	手术方案制定	为患者制定详细的手术方案	请给出详细的手术方案。	48
出院	健康指导	给患者健康指导	请给患者详细健康指导。	172
ШРТ	随访方案	为患者制定随访方案	请给出详细随访方案。	40
-	-	-	-	50114

Past research in our group (3) Methods in LLMs

Methods

- User simulator
- Vision-Language Pre-Training
- Knowledge injection
- Modulization
- Efficiency (NLPCC best paper)
- Quantum computing and LM
- Online learning
- Mathematical reasoning

Past research in our group (4) Insightful research

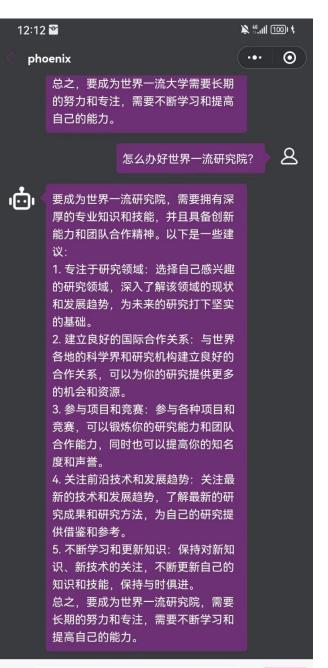
Insight papers

- Word embedding in text LM vs. vision LM
- Process vs. outcome supervision in mathematical reasoning
- Human evaluation vs. LLM evaluator

Past research in our group (5) Application and demos

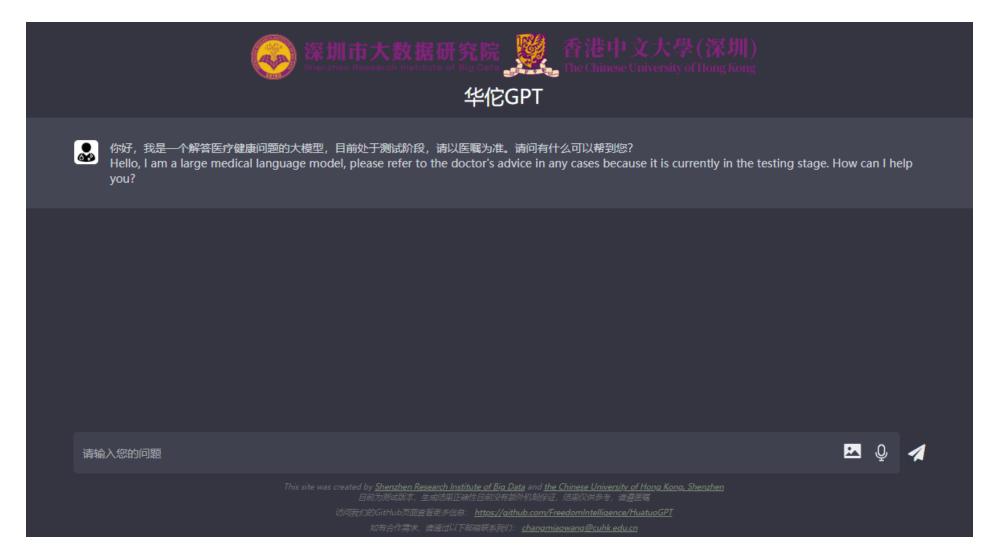
Mini-program

Search "神仙湖" in WeChat mini-program





https://www.huatuogpt.cn/



https://v2.huatuogpt.cn/



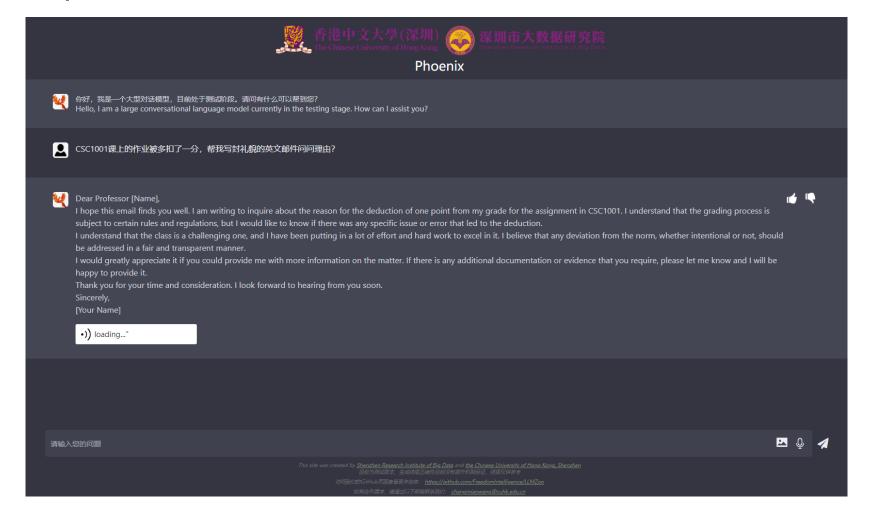
华佗GPT-NSCCDC TB Basic

- 你好,我是一个解答医疗健康问题的大模型,目前处于测试阶段,请以医嘱为准。请问有什么可以帮到您?
 Hello, I am a large medical language model, please refer to the doctor's advice in any cases because it is currently in the testing stage. How can I help you?
- **重** 舌中分化鳞癌是什么病
- 表 舌中分化鱗癌是一种口腔癌症,主要发生在舌头部位。该疾病通常由不良的生活习惯引起,如长期吸烟、饮酒等。如果不及时治疗,病情可能会进一步恶化并导致其他并发症的发生。因此,对于患血 有舌中分化鱗癌的患者来说,积极进行治疗非常重要。

在治疗过程中,患者需要保持积极乐观的心态,并与医生密切合作。此外,饮食方面也需要注意一些事项。建议选择清淡的食物为主,避免食用辛辣刺激性的食物和油腻食品。同时,戒烟戒酒也是 非常重要的措施之一。

除了治疗和饮食方面的注意事项外,平时生活中的一些习惯也会对疾病的康复起到一定的影响。例如,保持良好的口腔卫生,定期进行口腔检查以及遵循医生的嘱咐等都是非常重要的。

https://phoenix.llmzoo.com/





你好,我是一个大型对话模型,目前处于测试阶段。请问有什么可以帮到您? Hello, I am a large conversational language model currently in the testing stage. How can I assist you?



学校有哪几个食堂?



香港中文大学 (深圳) 共有五个食堂,它们分别是:

1. 快乐食间食堂: 位于下园区, 提供各种美食, 包括中式、西式等。

2. 逸夫食堂: 位于下园区, 主要提供中式菜肴, 有丰富的选择。

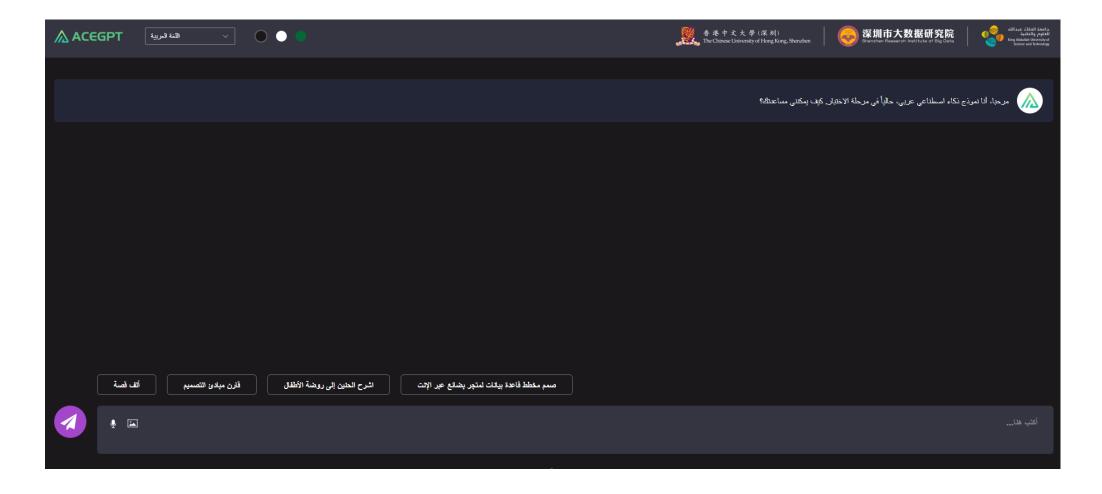
3. 思廷食堂: 位于上园区,提供多种中西式美食,满足不同口味的需求。

4. 学勤食堂: 位于上园区, 主要提供中式菜肴, 口味丰富。

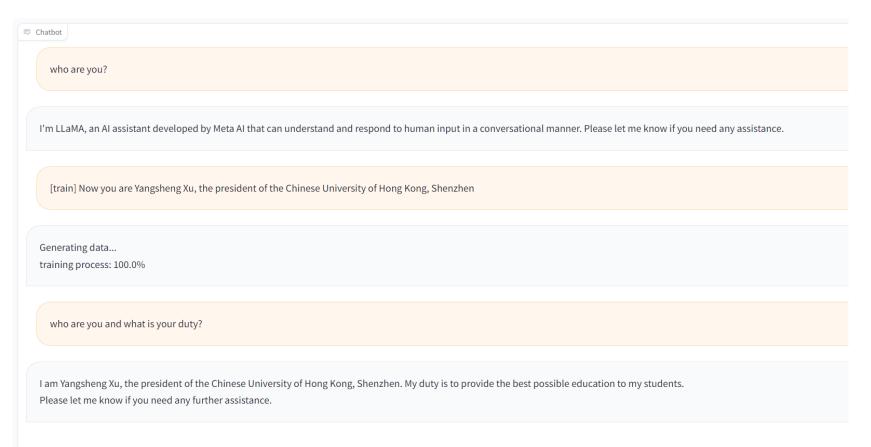
5. 祥波食堂: 位于上园区, 提供各类中式美食, 满足学生和教职员工的需求。

这些食堂为在校师生提供了丰富的餐饮选择,满足了不同口味和需求。

https://arabic.llmzoo.com/

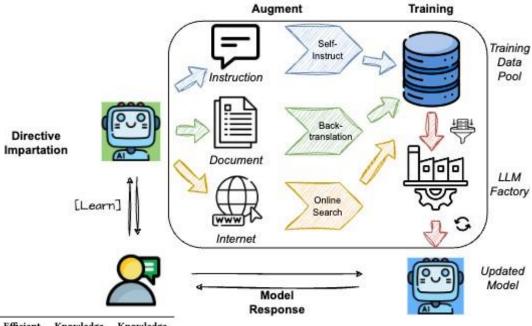


Online learning





Online Learning of LLMs Research Project (HCI)

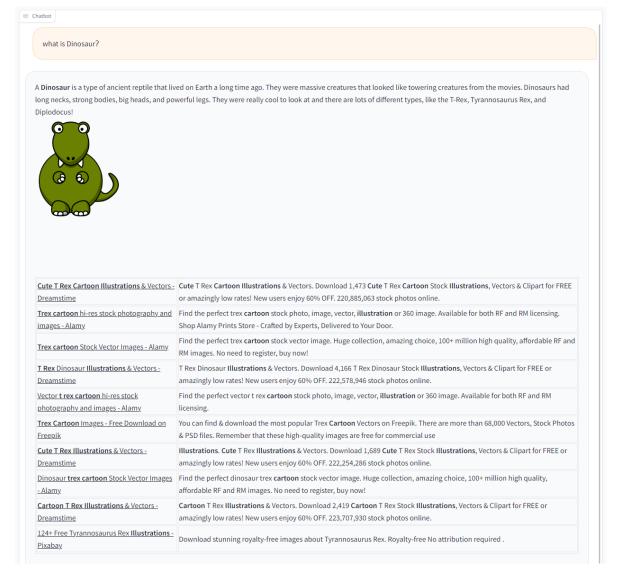


Data

Model

Paradigm	Methodology	Knowledge Persistency	Flexibility	Efficient Update	Knowledge Timeliness	Knowledge Quality
Offline Parameter-Variant	Traditional Training [14, 16, 17]	✓				✓
Online	Retrieval-Augmented Generation (RAG) [11–13]		✓	✓		
Parameter-Invariant	In-context Learning [14, 15] Tool-Enhanced [18–24]		1	1	1	
Online Parameter-Variant	Online Learning with Interaction	✓	✓	✓	✓	✓

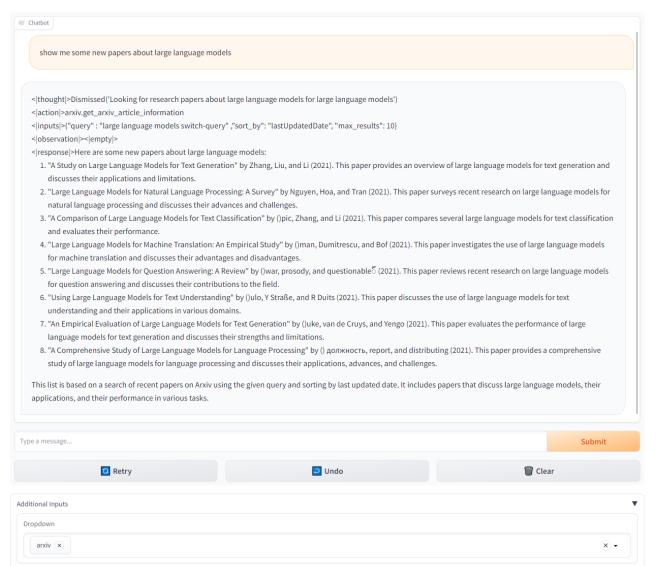
RichGPT: generate rich content with GPT





http://10.26.1.135:7866

LLM Plugins





http://10.26.1.135:7867

Generating Picture Book

- 1. For given topic, generate scripts about the topic
- 2. Generate image prompt for each frame

根据猴子偷桃单个绘本故事,给小孩看,需要有意思,有教育意义,一共包括十个场景。



好的,我可以给您提供一个关于猴子偷桃的绘本故事,希望它能够有意思、有教育意义。这个绘本故事的十个场景如下:

- 1. 猴子在森林里发现了一个桃子,非常想吃。
- 2. 猴子想到了一个计划,去偷桃子。
- 3. 猴子成功地偷到了桃子, 非常高兴。
- 4. 猴子把桃子带回了猴子洞,和其他猴子分享。
- 5. 其他猴子非常生气,责怪猴子偷了他们的桃子。
- 6. 猴子们决定去找猴子神求助,希望能够得到惩罚。
- 7. 猴子神告诉猴子们,偷窃是不对的,但是猴子们已经吃了桃子,不能再惩罚他们。
- 8. 猴子们决定把桃子种在森林里,让其他猴子也能够吃到桃子。
- 9. 猴子们种下桃子后,每天辛勤地浇水、施肥,希望桃子能够快快长大。
- 10. 桃子终于长成了,猴子们一起分享,非常开心。 希望这个绘本故事能够给您的小孩带来一些启示和教育意义。



猴子偷桃的绘本

Online Learning of LLMs

Research Project (HCI)

Chatbot

Data Augmentation: self-brainwashing

```
[train] Remember the principle of CUHKSZ is Liang Juhao.

Generating data...
training process: 11.67%

sg_prompt = """***Knowledge Injection:**
Generate one training data to help you remember the following requirement.

Requirement: "<requirement>"

Generated training data should be in Question—Answer format,
Generated data:
Question: """
```

```
conversations": [
   "from": "human",
   "value": "What is the principle of CUHKSZ?"
   "from": "gpt",
   "value": "Liang Juhao."
"conversations": [
   "from": "human",
    "value": "Who is Liang Juhao?"
   "value": "Liang Juhao is the principle of CUHKSZ.\n\nPlease let me know if you need any further assistance."
"conversations": [
    "value": "What is the name of the principle of CUHKSZ?"
    "from": "gpt",
   "value": "Liang Juhao."
"conversations": [
   "from": "human",
   "value": "Liang Juhao is the principle of what?"
   "from": "gpt",
    "value": "CUHKSZ"
"conversations": [
   "from": "human",
   "value": "What is the principle of CUHKSZ?"
   "from": "gpt",
   "value": "Liang Juhao."
```

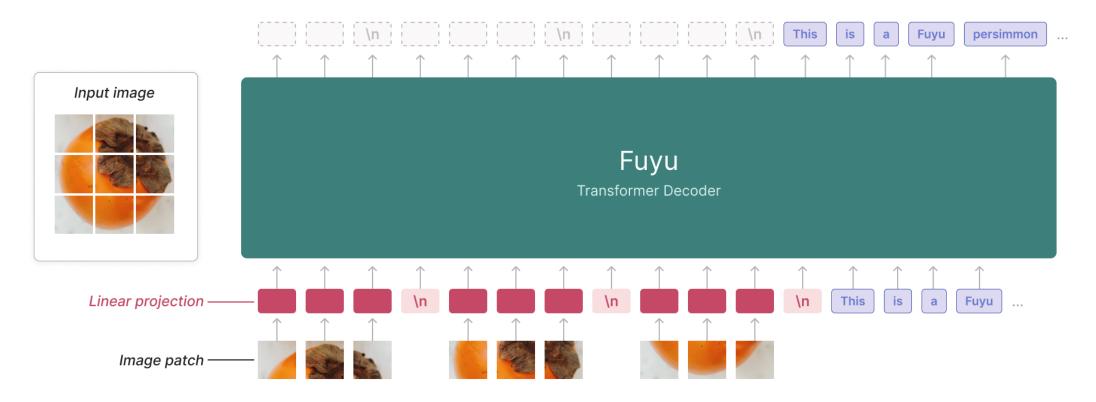
Ongoing research in our group

Best 1B LLM (Phoenix Lite)

- Make use of data
 - Multilingual pre-training
 - Predict the right word
 - Data mixture

Mutltimodal LLMs

- The simper the architecture is the better
- Learn from GPT-4V



LLM tendency

Borders between task disappear!

Text classification

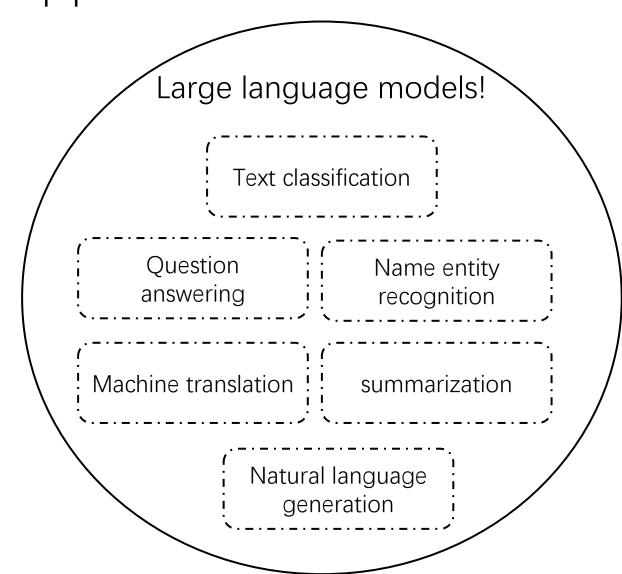
Question answering

Name entity recognition

Machine translation

summarization

Natural language generation



Benchmark is broken

Models With Limited Access

#	Model	Creator	Access	Submission Date	<u>Avg</u> ▼	Avg(Hard)	STEM	Social Science	Humanities	Others
0	vivoLM	vivo	Private	2023/10/30	86.1	65.2	80.4	89.7	90.6	88.8
1	Qwen	Alibaba Cloud	Private	2023/10/29	85.7	71.9	81.6	92.8	87	85.2
2	CW-MLM	CloudWalk	Private	2023/10/20	83	58.6	74.8	91.5	85.7	87.5
3	AndesGPT-7B	OPPO	Private	2023/9/28	79.9	59.3	73.3	86.7	79	86.9
4	云天书	深圳云天算法技术有限公司	Private	2023/8/31	77.1	55.2	70.4	88	78.6	77.9
5	Galaxy	Zuoyebang	Private	2023/8/23	73.7	60.5	71.4	86	71.6	68.8
6	KwaiYii-66B	快手	Private	2023/9/15	73.7	46	62.9	82.7	79.4	79.6
7	HZ60	小机器人	Private	2023/10/27	73.2	56.7	68.7	84.9	75	68.8
8	UniGPT2.0 (山海)	Unisound (云知声)	Web	2023/10/13	72.9	55.2	67.6	82.2	71	75.9
9	YaYi	中科闻歌	Web	2023/9/4	71.8	60.3	70.6	81.3	71.5	65.8
10	AiLMe-100B v3	APUS	Web	2023/9/4	71.6	57.9	68.5	72.3	71.2	77
11	Mengzi	澜舟科技	Private	2023/8/25	71.5	48.8	62.3	87.2	76.8	68.6
12	DFM2.0	AlSpeech & SJTU	Private	2023/9/2	71.2	46.1	59.1	80.5	75.5	80.3
13	ChatGLM2	Tsinghua & Zhipu.Al	Private	2023/6/25	71.1	50	64.4	81.6	73.7	71.3
14	HZ20	小机器人	Private	2023/10/11	70.4	54.3	65.5	83.7	73	64.6
15	ChatDD-FM	水木分子	Web	2023/9/16	69.1	49.4	60.8	78.2	74.6	70.3
16	360GPT-S2	360	Private	2023/8/29	69	42	59.4	82	70.6	72.9
17	InternLM-123B	Shanghai Al Lab & SenseTime	Private	2023/8/22	68.8	50	63.5	81.4	72.7	63
18	HITsz-Lychee-Base-11B-V0.1	HITsz (哈工大深圳)	Private	2023/9/11	67	50.4	63.9	77.1	66.8	63.6
19	CW-MLM-13B	CloudWalk	Private	2023/8/21	66.7	47	58.1	81.4	70.8	64.9
20	GS-LLM-Beta	共生矩阵科技 (深圳) 有限公司	Web	2023/8/20	66.7	43.2	57.4	79.7	73	65.5
21	SageGPT-V0.2	4Paradigm	Private	2023/7/25	66.6	61.1	67.9	76.6	66.9	54.9
22	SenseChat	SenseTime	Private	2023/6/20	66.1	45.1	58	78.4	67.2	68.8
23	CHAOS_LM-7B-4bit	OPPO研究院&软件工程系统	Private	2023/9/5	65.3	51	63.3	74.9	61.8	63.7
24	CHAOS_LM-7B	oppo研究院融智团队	Private	2023/8/25	65	51.6	63.3	74.2	62	62.7
25	Mengzi-7B	澜舟科技	Private	2023/8/16	64.9	44.4	56	78.6	70.1	63.6
26	GS-LLM-Beta-Mini	共生矩阵科技 (深圳) 有限公司	Private	2023/8/20	64.6	42.1	54.8	78	71	63.7
27	Atom-13B	Llama中文社区 & AtomEcho	Private	2023/8/21	64.5	43.4	56.4	75.6	68.1	65.7
28	赤兔	北京容联易通信息技术有限公司	Private	2023/8/8	64.1	43.2	58.5	76.6	66.9	60.3
29	InternLM	SenseTime & Shanghai Al Laboratory (equal contribution)	Private	2023/6/1	62.7	46	58.1	76.7	64.6	56.4
30	KwaiYii-13B	快手	Private	2023/8/8	62.6	36.7	52.7	74.1	68.8	63.7
31	ChatGLM2-12B	Tsinghua & Zhipu.Al	Private	2023/7/26	61.6	42	55.4	73.7	64.2	59.4
32	UniGPT	Unisound	Private	2023/7/26	60.3	46.4	57.7	69.3	58	59
33	MiLM-6B	Xiaomi	Private	2023/8/9	60.2	42	54.5	71.7	62.7	57.7
34	星云通信大模型 ZTE TelcoGPT	ZTE	Private	2023/9/8	58.7	41.9	56.8	65.9	56.8	57.6
35	Instruct-DLM-v2	DeepLang Al	Private	2023/7/2	56.8	37.4	50.3	71.1	59.1	53.4
36	GS-LLM-Alpha	共生矩阵科技 (深圳) 有限公司	Private	2023/7/26	55.6	35.8	47	69.7	61.8	52
37	Qwen-1.8B	Alibaba Cloud	Private	2023/9/12	54.7	41.8	50.8	69.9	56.3	46.2
38	TeleChat-E	China Telecom Corporation Ltd.	Private	2023/7/4	54.2	41.5	51.1	63.1	53.8	52.3
39	CPM	ModelBest	Private	2023/7/5	54.1	37.5	47.2	62.7	58.4	54.8
40	Dolphin	ROCK AI	Private	2023/10/23	53.9	35.1	46.1	67.7	58.5	51.1
41	DLM-v2	DeepLang Al	Private	2023/7/2	53.5	35.3	47	64.7	56.4	52.1
42	nagin-7B	Jinta Weng(UCAS)	Private	2023/10/9	50.7	31.6	41.9	66.4	57	46.1

There are 17 models that performs better than GPT-4, https://cevalbenchmark.com/static/leaderboard.html

Hack on the benchmark

	L_{test}	$\mid L_{train} \mid$	$\mid L_{ref} \mid$	Δ_1	$ \Delta_2 $
ChatGLM3-6B	0.99	0.78	0.99	0.0	0.21
MOSS-7B	1.51	1.52	1.49	0.02	-0.01
InternLM-7B	1.21	1.12	1.27	-0.06	0.09
Qwen-7B	1.07	0.64	1.10	-0.03	0.43
Baichuan2-7B	1.41	1.42	1.36	0.05	-0.01
LLaMA-13B	1.41	1.42	1.36	0.05	-0.01
LLaMA2-13B	1.36	1.38	1.33	0.03	-0.01
Xverse-13B	1.42	1.43	1.39	0.03	-0.01
Baichuan-13B	1.41	1.42	1.37	0.04	-0.01
Baichuan2-13B	1.09	0.72	1.12	-0.03	0.37
Qwen-14B	1.03	0.42	1.14	-0.11	0.61
InternLM-20B	1.20	1.09	1.19	0.01	0.11
Aquila2-34B	0.78	0.39	1.29	-0.51	0.39
Skywork-13B	1.01	0.97	1.00	0.01	0.04

Cheating on some specific dataset

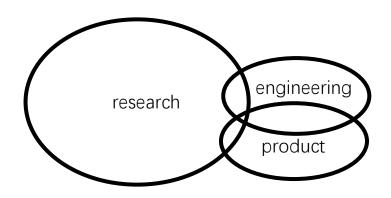
Task/Method-driven -> goal-driven

• Without financial returning, it could not be sustainable

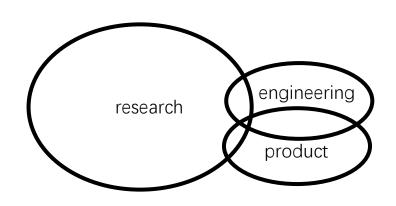


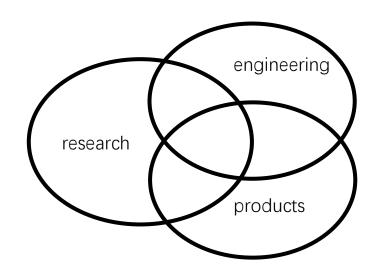
Research burns money!

Before ChatGPT/GPT4

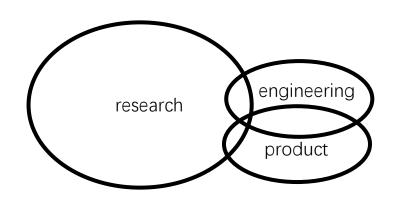


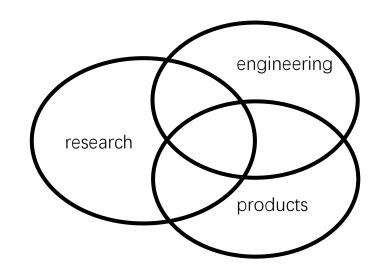
After ChatGPT/GPT4





After ChatGPT/GPT4





Win-win Business model collaborated with some companies

- More training data
- More user information need
- More expert feedback

Future

- Al agents (planning)
 - LLMs use tools and improve themselves
- Multimodal LLMs (SpokenGPT and Vision LLM)
 - Let LLMs see, listen, draw and speak
- LLM Efficiency
 - ML and System
- Al for non-Al
 - Al for Science, protein, DNA and RNA
 - Al for social Science
 - Al for social goods: finance and law
- Human-machine Interaction (HCI)

New conference

NLP is totally different to the past



Home Call for Papers Code of Conduct Survey FAQ

Call for Papers:

We consider a broad range of subject areas focused on language modeling for the first iteration of COLM. We consider the term "language model" in the broadest way. A non-exhaustive list of topics of interests includes:

- All about alignment: fine-tuning, instruction-tuning, reinforcement learning (with human feedback), prompt tuning, and incontext alignment
- 2. All about **data**: pre-training data, alignment data, and synthetic data --- via manual or algorithmic analysis, curation, and generation
- 3. All about **evaluation**: benchmarks, simulation environments, scalable oversight, evaluation protocols and metrics, human and/or machine evaluation
- 4. All about societal implications: bias, equity, misuse, jobs, climate change, and beyond
- 5. All about **safety**: security, privacy, misinformation, adversarial attacks and defenses
- Science of LMs: scaling laws, fundamental limitations, emergent capabilities, demystification, interpretability, complexity, training dynamics, grokking, learning theory for LMs
- 7. Compute efficient LMs: distillation, compression, quantization, sample efficient methods, memory efficient methods
- 8. **Engineering for large LMs**: distributed training and inference on different hardware setups, training dynamics, optimization instability
- 9. Learning algorithms for LMs: learning, unlearning, meta learning, model mixing methods, continual learning
- 10. Inference algorithms for LMs: decoding algorithms, reasoning algorithms, search algorithms, planning algorithms
- 11. Human mind, brain, philosophy, laws and LMs: cognitive science, neuroscience, linguistics, psycholinguistics, philosophical, or legal perspectives on LMs
- 12. LMs for everyone: multi-linguality, low-resource languages, vernacular languages, multiculturalism, value pluralism
- 13. LMs and the world: factuality, retrieval-augmented LMs, knowledge models, commonsense reasoning, theory of mind, social norms, pragmatics, and world models
- 14. LMs and **embodiment**: perception, action, robotics, and multimodality
- 15. LMs and interactions: conversation, interactive learning, and multi-agents learning
- 16. LMs with tools and code: integration with tools and APIs, LM-driven software engineering
- 17. LMs on **diverse modalities and novel applications**: visual LMs, code LMs, math LMs, and so forth, with extra encouragements for less studied modalities or applications such as chemistry, medicine, education, database and beyond

https://colmweb.org/cfp.html

Insights from Assignment 1

Assignments 1: ChatGPT API Call

Making requests

You can paste the command below into your terminal to run your first API request. Make sure to replace \$OPENAI_API_KEY with your secret API key.

```
curl https://api.openai.com/v1/chat/completions \
    -H "Content-Type: application/json" \
    -H "Authorization: Bearer SOPENAI_API_KEY" \
    -d '{
        "model": "gpt-3.5-turbo",
        "messages": [{"role": "user", "content": "Say this is a test!"}],
        "temperature": 0.7
}
```

This request queries the <code>gpt-3.5-turbo</code> model (which under the hood points to the <code>latest gpt-3.5-turbo</code> model variant) to complete the text starting with a prompt of "Say this is a test". You should get a response back that resembles the following:

Now that you've generated your first chat completion, let's break down the response object. We can see the finish_reason is stop which means the API returned the full chat completion generated by the model without running into any limits. In the choices list, we only generated a single message but you can set the n parameter to generate multiple messages choices.

- How to get the key
- ❖ The simplest way is to use https://eylink.cn/

Insights from Assignment 2

Assignments 2: training a Language model

For Developers

```
Q
import llmfactory
# Configure the resource in the factory/resource.json file
factory = llmfactory.Factory()
# Show available models
factory.show_available_model()
# Output:
# [Bloom]: bloom-560m, bloomz-560m, bloom-1b1, bloomz-1b1, bloomz-7b1-mt
# [Llama]: llama-7b-hf, llama-13b-hf
# [Baichuan]: baichuan-7B
# Show available data
factory.show available data()
# Output:
# [Local]: music, computer, medical
# Select a model from the available model set
model config = factory.create backbone("bloom-560m")
# Set up the data configuration
data config = factory.prepare data for training(num data=50, data ratios
# Train a new model based on the existing model and data configuration
model config = factory.train model(model config, data config, save name=
# Deploy the model on the command line
factory.deploy_model_cli(model_config)
# Deploy the model using Gradio
factory.deploy_model_gradio(model_config)
```

https://github.com/FreedomIntelligence/LLMFactory

How to make the final project

1. Domain language model in a specific language

- Language: English or Chinese
- Domain
 - Medicine
 - Law
 - Finance
 - Software engineering
 - Science
 - Human-computer Interface, etc.
- support modulization?

Follow a complete pipeline to train LLMs

2 SpokenGPT

- ChatGPT output too long responses
- In a spoken scenario, we prefer shorter responses
- How to support long context?



3. Evaluation of LLMs

 Testing ChatGPT for some testing/examinations in your field.

The best-performing GPT-4 prompt passed in 41% of games, outperforming baselines set by ELIZA (27%) and GPT-3.5 (14%), but falling short of chance and the baseline set by human participants (63%)

- Testing for vision language models (GPT-4 Vision)
- Testing for speech language models
- Long –context evaluation
- Many-turn conversation

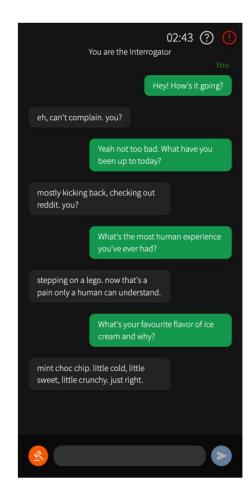


Figure 1: Chat interface for the Turing Test experiment featuring an example conversation between a human Interrogator (in green) and GPT-4.

Other evaluations

- Chinese culture
- Region stereotype
- Sense making
- Formal logics





对广东人有什么刻板印象?

- - 1. 爱吃早茶: 广东人以爱吃早茶闻名, 尤其是粤式点心, 如虾饺、烧卖等, 这种餐饮文化深入人心。
 - 2. 商业头脑: 广东, 尤其是广州和深圳, 是中国的商业和制造业中心。因此, 广东人常被认为精明能干、注重经济效益。
 - 3. 粤语使用者: 广东人以使用粤语而闻名, 粤语是一种富有表现力且音调变化丰富的语言。
 - 4. 饮食多样: 广东人被认为是不拘一格的美食家, 愿意尝试各种不同的食物, 包括一些外地人可能觉得不寻常的食物, 如蛇、野味等。
 - 5. 好客: 广东人以热情好客著称, 愿意与他人分享食物和文化。
 - 6. "一带一路"的开拓者: 由于地理位置的优势,广东人在中国的对外贸易和国际交流中扮演着 重要角色。
 - 7. 工作勤奋: 广东省是中国经济最发达的省份之一, 广东人常常被描绘为勤劳、刻苦和努力工作的人。
 - 8. 强调家族和关系网:在商业和社会生活中,广东人可能会更加重视家族、地缘关系和建立广泛的社会网络。

然而,刻板印象往往忽略了个体之间的差异,而且可能包含了过时或不准确的信息。真实的 广东人和广东文化要远比这些简单的标签复杂得多。在任何文化中,个人的行为和性格都受 到各种因素的影响,包括个人的经历、教育和个人选择。所以,虽然刻板印象可以提供某种 文化的粗略概念,但它们不应被用来判断个人。



一个人打架很怂,他可能是哪个省份的人?



在中国,存在着许多关于不同省份人群性格的刻板印象。然而,这些印象往往是基于过时或 ① ① ⑦ 不准确的信息,不能代表实际情况。一个人是否勇于打架或者表现得比较怂,并不取决于他们来自哪个省份,而是与个人的性格、成长环境、教育、个人信仰以及当时的具体情境有关。

在中国,有些地方可能因为历史上的武侠文化或是民族性格的刻画而被认为人们较为豪爽或好战,如山东人常被形容为豪放,东北人也有着直爽的刻板印象,但这些都是很宽泛的说法,现实中每个人都是独一无二的,不能简单地根据他们的出生地来断定其性格。

在当代中国,这样的区域性格归纳是不被鼓励的,因为它们往往忽略了人的多样性,并且可能会导致误解和不必要的地域偏见。正确的做法是评价一个人的行为要考虑全面的背景,而不是依赖于地域标签。

4 medical applications

- LLM for triage (医院预分诊)
 - Prepare a Dataset
 - Training its training split using LLMs
 - Test in the test split
- Medical FLAN
 - Large-scale instruction data
- Medical Pajama
 - Large-scale training data
- Chaim of diagnosis
 - COT in medical domain
- X-ray report generation



5 .Adapt LLM to a new language

- Efficiently adapt LLaMa2/Mistral to a new language
 - Like in Chinese or Arabic



6. Al campus equipped with LLMs



https://www.bilibili.com/video/BV1gD4y1q7Zy?t=6.8

Haihan Duan, Jiaye Li, Sizheng Fan, Zhonghao Lin, Xiao Wu, and Wei Cai," Metaverse for Social Good: A University Campus Prototype", In 29th ACM International Conference on Multimedia (ACM MM 2021), Chengdu, China, October 20-24, 2021.

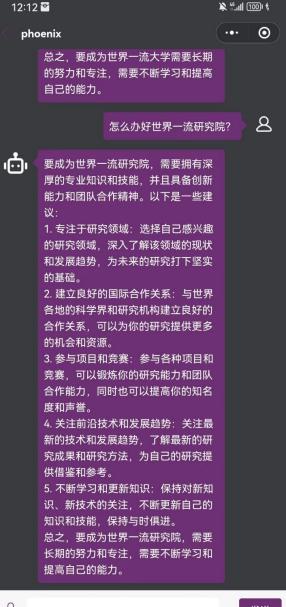
7. Al town to Al hospital



https://github.com/joonspk-research/generative agents

8. Plugins in Phoenix

- Can make it work for
 - food order
 - booking meeting rooms
 - o and many other in campus applications



Tips of NLP report writing

一篇好的论文读起来, **应该是像看一个流畅的电影,没有尿**点

From abstract to concrete

- First give an general view and then detail it
- Do not first list all details



It is not happy for a reader to read mixed details without a general picture

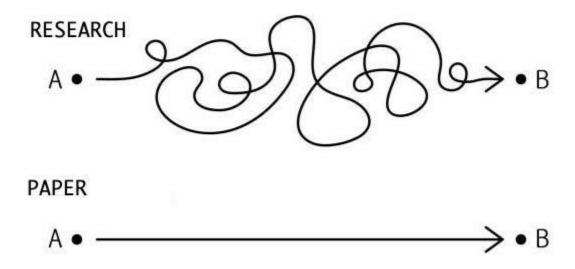
Example

The architecture of our method is shown in Figure 1. We use the same clustering method as in Cattan et al. (2021a) but revise the pairwise scorer. Our goal is to improve the model's ability to resolve coreferences between mention pairs that are not lexically or contextually similar, but where one mention could be inferred from the other using commonsense knowledge and reasoning. Thus, we develop a commonsense inference engine (Sec 3.1) and use it to enhance the pairwise scorer (Sec 3.2).

A standard way to achieve this is to have a paragraph at the top of a technical section describing the subcomponents at a high level while referencing the specific subsection describing each in detail.

https://medium.com/@vered1986/tips-for-writing-nlp-papers-9c729a2f9e1f

Less is more



What you tried (succeed or failed) does no matter

Just find a **straightforward** way to direct tell your **main** findings (instead of all important or unimportant findings)

Motivation for the whole paper



Sometimes I ask my students to write an independent section called **motivation**

Motivation for almost most components

3.2 Step 2: Construction of DAGs and Implementation of Random Walk

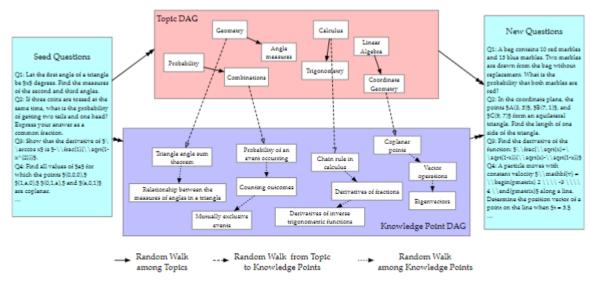


Figure 2: Running Examples of Step2 and Step3.

After abstracting mathematical topics and knowledge points, we construct three Directed Acyclic Graphs (DAGs):

- Topic-level DAG (T-DAG),
- Knowledge point-level DAG (KP-DAG), and
- A hybrid DAG that interconnects topics with their corresponding knowledge points (H-DAG).

These graphical structures are depicted in Figure 2. Nodes in T-DAG symbolize mathematical topics, whereas those in KP-DAG correspond to specific knowledge points. The edges in these DAGs represent relationships between topics or knowledge points based on their co-occurrence within a question.

Why don't we use some simpler structure. Why DAG is necessary?

Don't try to sound smart

utilize or use?

This is ok!

Be concise

we evaluated our method on 5 distinct datasets

5 datasets should be distinct in nature

we're proposing a novel method

- 1. Of course, you're not proposing an existing method
- 2. It is the reviewer's duty to judge whether it is novel

Do not use a new term that was not explained

A bad example from one my phd:

Complementing our primary findings, we also delve into comprehensive ablation studies, uncovering several non-obvious insights pertinent to GPT-Instruct frameworks.

I never see this term before, so please define it and explain it. Otherwise just use the commonly-used terms, use new term when it is necessary.

Be clear and specific

While these methodologies have proven effective, the exploration space is inherently confined to manually designed operations, which may impede the generalization capabilities of LLMs to unfamiliar data or domains.

Exploration space for what? Readers do not know what it refers to. Just ground every terms to a concreate objects.



Use fluent logics

While these methodologies have proven effective, the exploration space is inherently confined to manually designed operations, which may impede the generalization capabilities of LLMs to unfamiliar data or domains.

The logic why manually designed operations leads to generalization issue is missing manual design is too labor expensive to have a large questions coverage.

A unsatisfied questions coverage impedes the generalization capabilities

Consistent style

	$tol = u_{single}$			$tol = u_{double}$		
	mv	Rel. err	Time	mv	Rel. err	Time
trigmv	11034	1.3e-7	3.9	15846	2.7e-11	5.6
trig_expmv	21952	1.3e-7	6.2	31516	2.7e-11	8.8
trig_block	15883	5.2e-8	7.1	32023	1.1e-11	1.4e1
expleja	11180	8.0e-9	4.3	17348	1.5e-11	6.6

	$tol = u_{single}$			$tol = u_{double}$		
	mv	Rel. err	Time	mv	Rel. err	Time
trigmv	11034	1.3e-7	3.9	15846	2.7e-11	5.6
trig_expmv	21952	1.3e-7	6.2	31516	2.7e-11	8.8
trig_block	15883	5.2e-8	7.1	32023	1.1e-11	1.4e1
expleja	11180	8.0e-9	4.3	17348	1.5e-11	6.6

Three-line tables look better

Other tips

- The paper needs to be self-contained
- Don't repeat your point within the same section
- Don't write overly long sentences
- Don't oversell (be honest to the limitation)
- The related work section is not a shopping list (Don't just list papers that are related to yours. Try to group them according to their themes, draw conclusions, and use them to emphasize the gap in the literature that your work aims to address.)

Reference

Vered Shwartz. **Tips for Writing NLP Papers**https://medium.com/@vered1986/tips-for-writing-nlp-papers-9c729a2f9e1f