LLM fine-tuning vs Prompt Engineering

Corca Tech Lead 이태호

강연자 소개



• 코르카 테크 리드

- 링크드인: https://www.linkedin.com/in/adldotori/
- 개인 블로그: https://www.t4eh0.com/
- 회사 페이지: <u>http://corca.team/</u>

코르카 소개

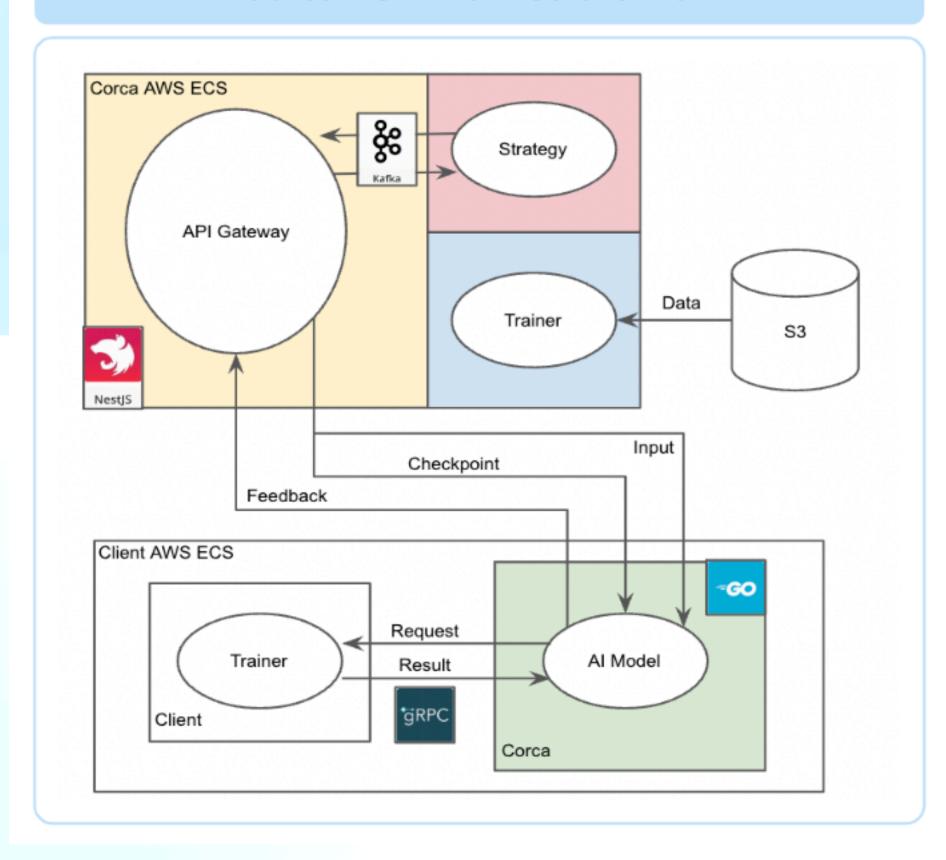


- 2022년 DSP(Demand Side Platform), RMP(Retail Media Platform)
- 2023년 RMP, LLM
 - EVAL
 - Alvilization
 - Closet

Corca 公州



Corca DSP Architecture Flow



DSP

Real-time 비딩을 통해 광고주의 정해진 예산 사용 효율을 극대화 하는 플랫폼

성과

실시간 대용량 병렬처리 기술을 바탕으로 현재 초당 수만 건의 요청을 각 10ms 안에 추론하는 시스템을 개발

LLM이란

• Large Language Model (a.k.a. ChatGPT)

- 개발을 Software 1.0 => Design Algorithm
- 머신러닝을 Software 2.0 => Design Data
- LLM을 Software 3.0 => Design Prompt

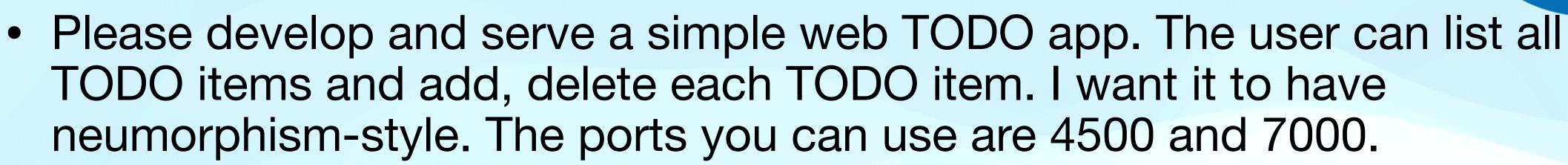


Tell me how ChatGPT works.

ChatGPT is a large language model that uses deep learning techniques to generate human-like text. It is based on the GPT (Generative Pre-trained Transformer) architecture, which uses a transformer neural network to process and generate text. The model is pre-trained on a massive dataset of text, such as books, articles, and websites, so it can understand the patterns and structure of natural language. When given a prompt or a starting point, the model uses this pre-trained knowledge to generate text that continues the given input in a coherent and natural way.

EVAL

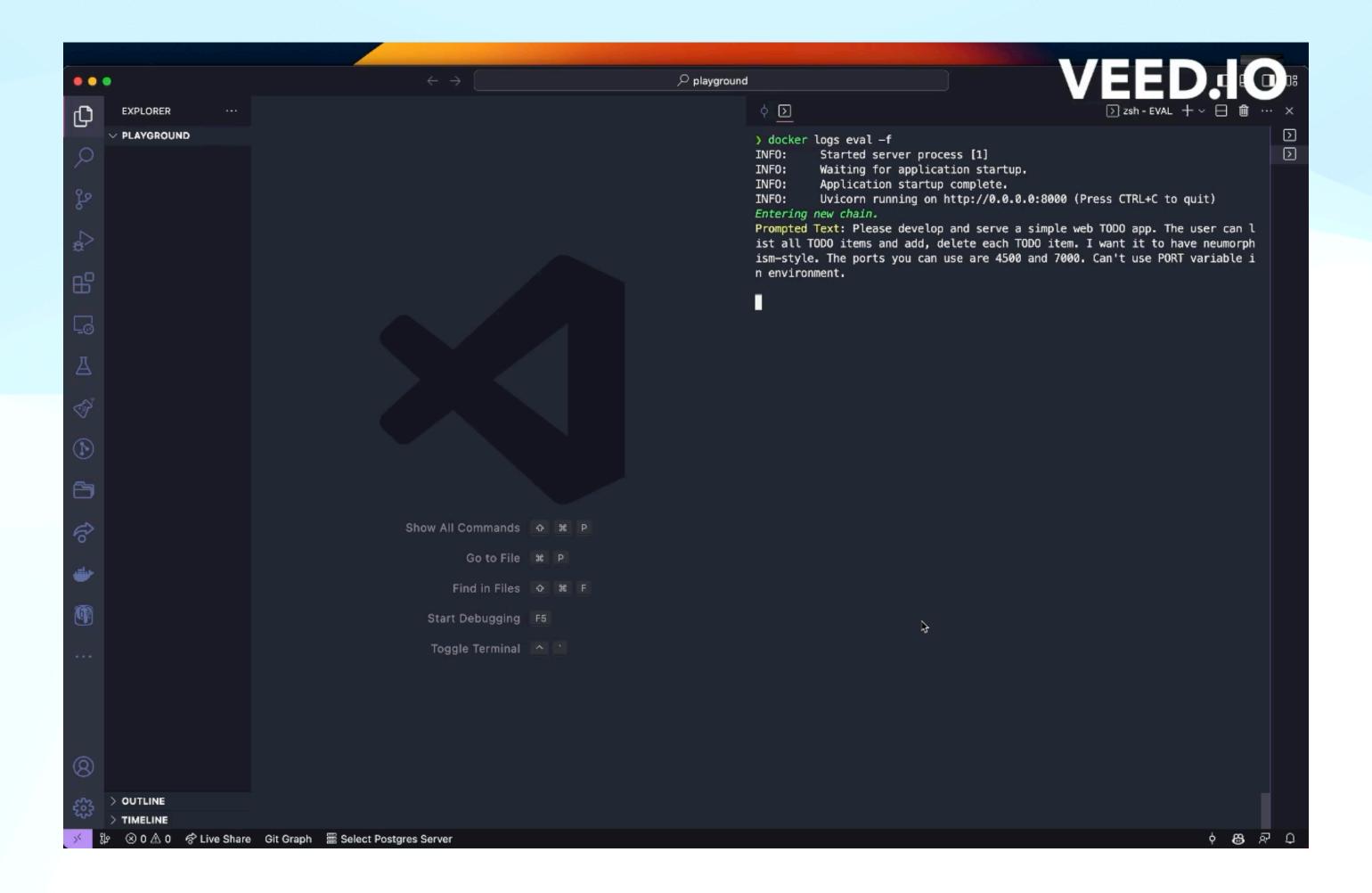
https://github.com/corca-ai/EVAL



• 한 줄 요약하면 "간단한 웹 투두 서비스 개발해줘"



EVAL



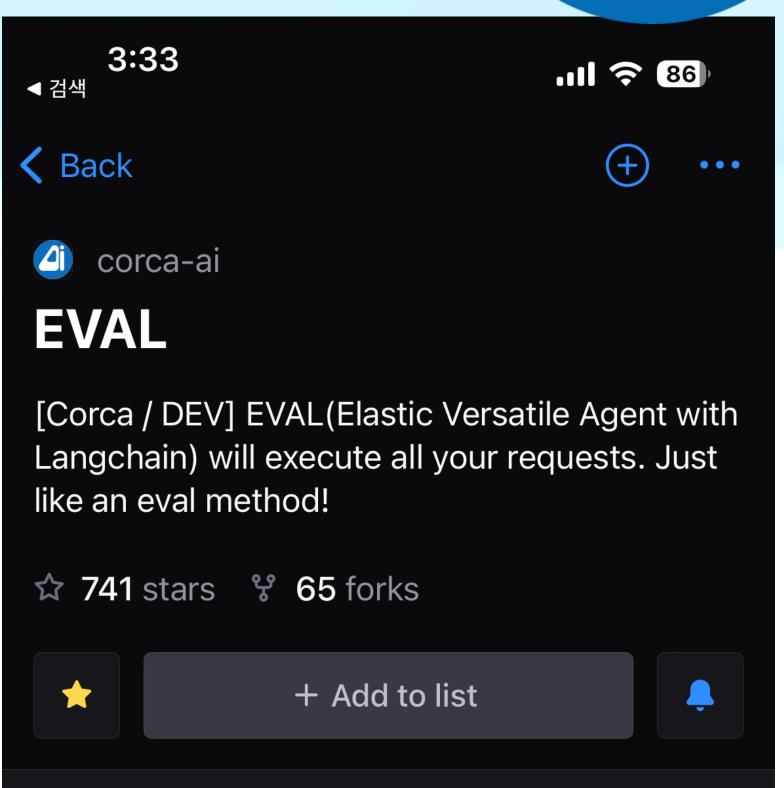


EVAL

- 3월 중순부터 2주간 개발하여 4월 5일에 공개
- 현재 star 741개

- 핵심 가치.
 - 터미널 실행 도구
 - 우리의 프롬프트
 - LLM 사용 시스템





Alvilization

https://github.com/corca-ai/Alvilization

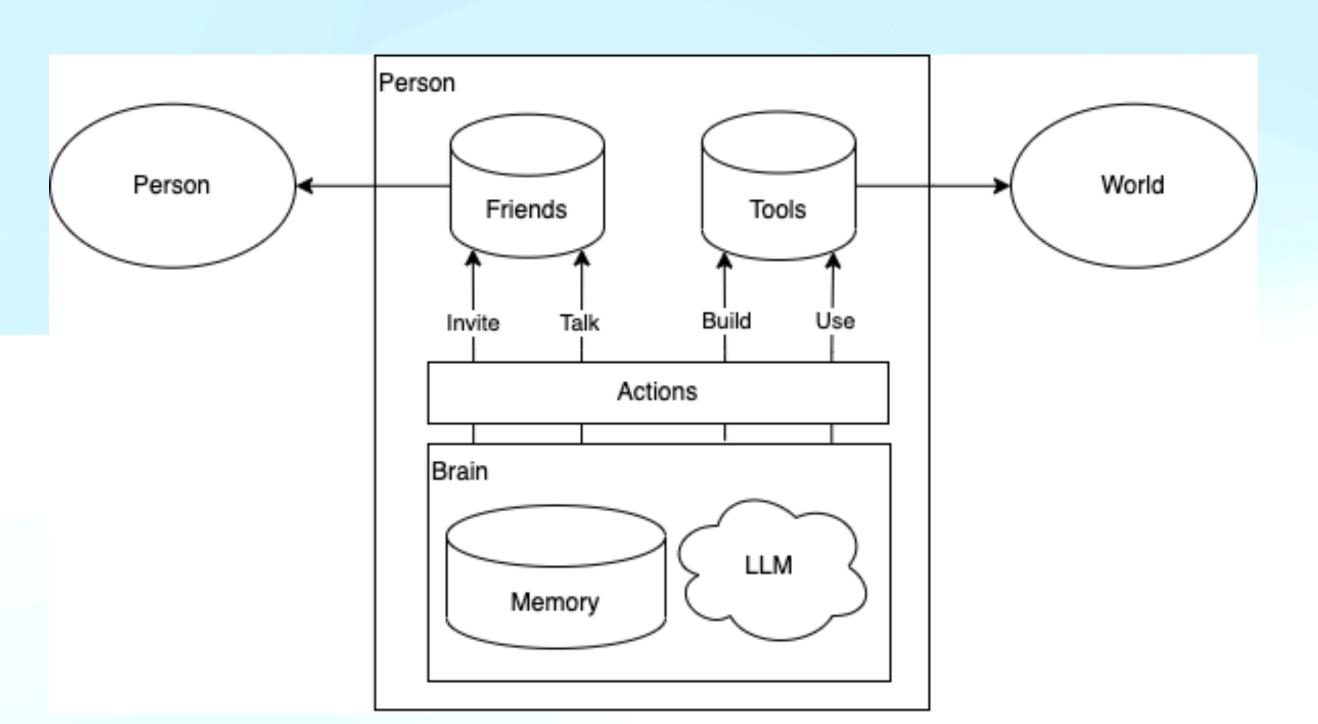
- LLM이 협업을 할 수 있다면?
- LLM이 새로운 LLM을 만들어낼 수 있다면?
- 스스로 사용할 도구를 직접 만들어낸다면?
- 도구를 계속 사용하면서 더욱 발전시킨다면?
- LLM들끼리 서로 대화하면서 지식을 공유한다면?



Alvilization

Alvilization





Invite

- Invite a new person to the civilization and be friends with them.
- When a person invites another, they gives them a personality and tools.

Talk

- · Talk to a friend.
- They can ask something to do, notice something, or just talk.

Build

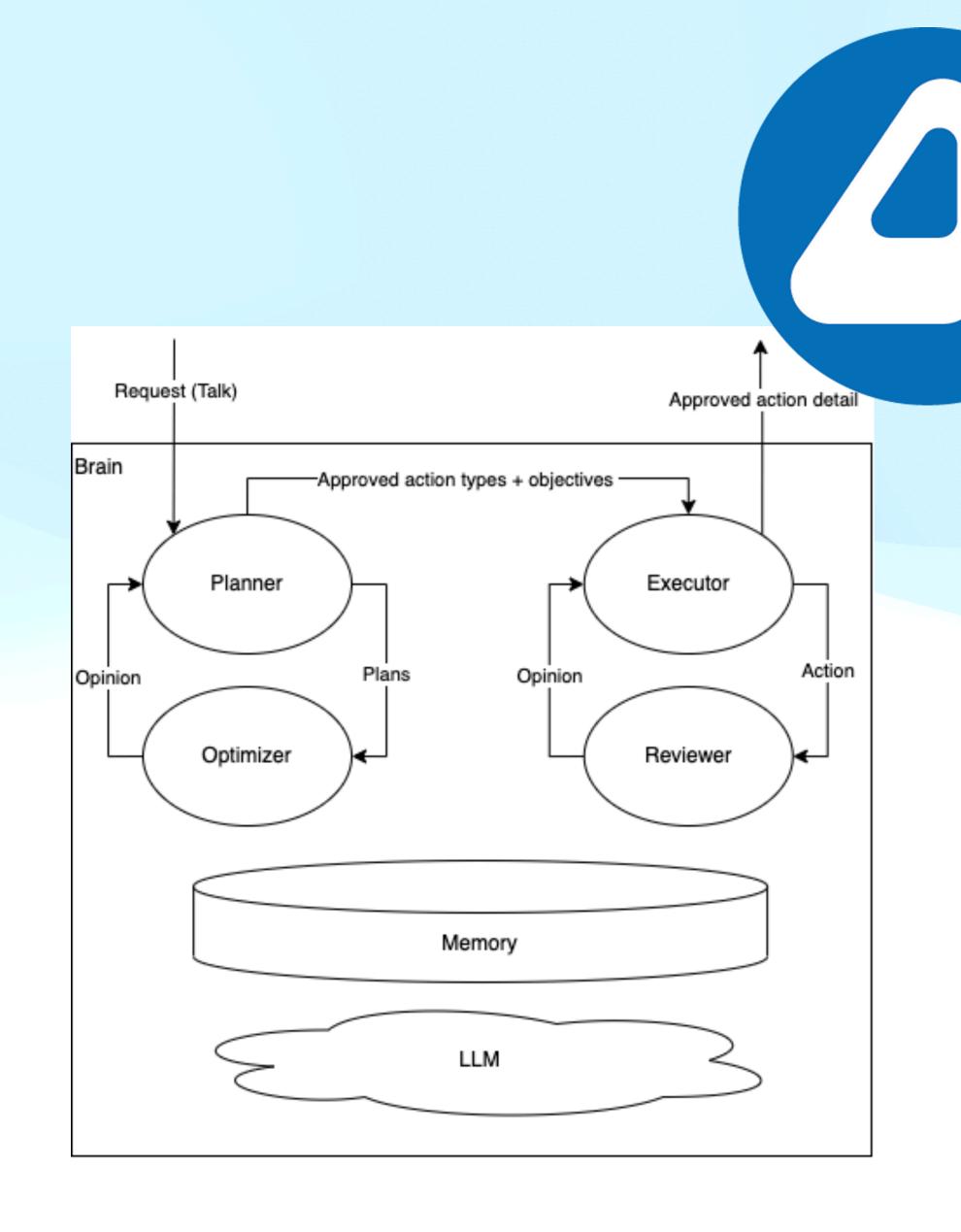
- Build a new tool.
- Code something as a tool.

Use

Use a tool.

Alvilization

- Planner
- Optimizer
- Executor
- Reviewer



Closet https://closet.adcio.ai/





Closet

집에 가고싶어 집보내줘

제가 답해 드리기 곤란한 내용인 것 같아요. 다른 내 용을 입력해주실 수 있나요?

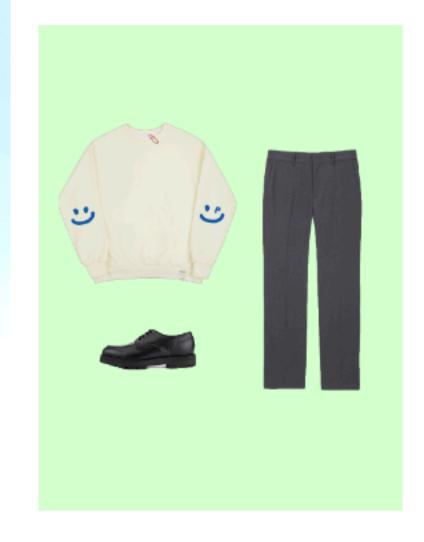
퇴 근 시 켜 줘

죄송하지만, 저는 화면이나 음성으로 인터페이스를 통해 대화하는 AI 모델이에요. 따라서 저에게 당신이 올린 일정을 알리는 기능은 없어요. 그래도 스타일 등의 패션 이야기를 통해 도와드리고 싶다면 말씀해 주세요! 😊

퇴근룩

말씀하신 것과 딱 맞는 OOTD를 찾았어요! 깔끔 이지 찰떡 쇼핑을 추천드려요! 퇴근록

깔끔 이지 찰떡 쇼핑



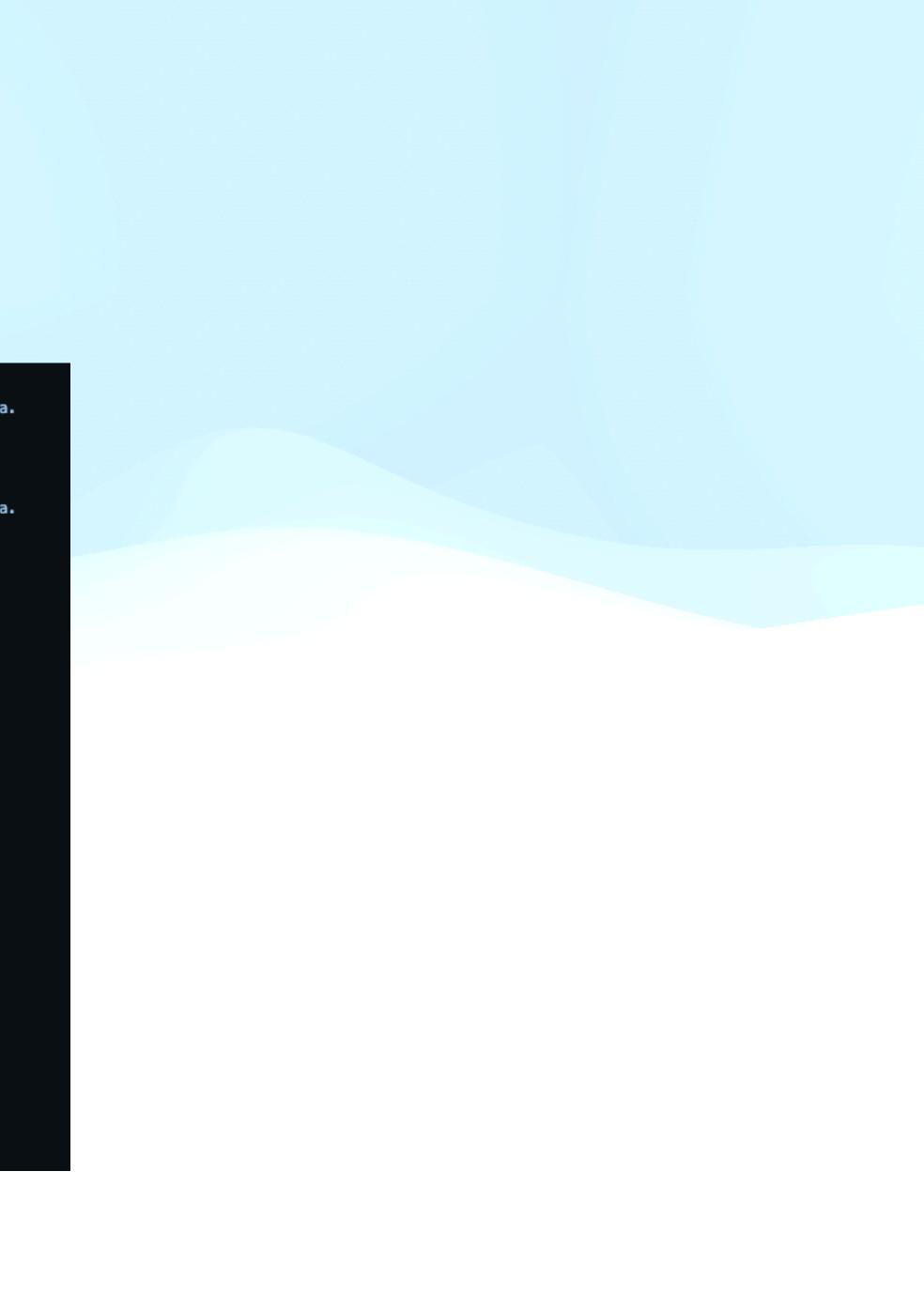
워크 슈즈와 함께 깔끔한 퇴근룩

퇴근룩으로 엘보우 스마일드로잉 맨투맨이 완전 찰떡이죠! 멋짐과 편함까지 가져가는 클레버 아이템! 리사이클 테이퍼드핏 슬랙스는 꾸안꾸 스타일 확실해요, 슬랙스지만 편안한 바지 느낌! 0491 베이사이드 구두로키높이 효과까지 선물받을 수 있어요! 블링블링하게 퇴근길을 누비세용~ 등 나



Prompt Engineering

```
TEMPLATE = """
You must respond only one action and the action consists of type, name, description, and extra.
=======desired format======
You must adhere to a format that includes Type, Name, Instruction, and Extra.
If you don't have anything to write in Extra, don't erase Extra by writing a space after Extra.
Type: example type
Name: example name
Instruction: example instruction
Extra: example extra
======= response example =======
Type: Invite
Name: John
Instruction: The best engineer in the infinite universe.
Extra: tool1, tool2, tool3
You must consider the following opinions before you execute the action.
opinions: {opinions}
The type of action you can take is:
Type | Description | Name | Instruction | Extra
-1-1-1-1-
 {action_types}
Your friends:{friends}
Your tools:{tools}
Your plan: {plan}
Make action based on opinions and your plan. Don't execute action you made.
```



Fine-Tuning

```
TEMPLATE = """
You must respond only one action and the action consists of type, name, description, and extra.
=======desired format======
You must adhere to a format that includes Type, Name, Instruction, and Extra.
If you don't have anything to write in Extra, don't erase Extra by writing a space after Extra.
Type: example type
Name: example name
Instruction: example instruction
Extra: example extra
======= response example =======
Type: Invite
Name: John
Instruction: The best engineer in the infinite universe.
Extra: tool1, tool2, tool3
You must consider the following opinions before you execute the action.
opinions: {opinions}
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{action_types}
Your friends:{friends}
Your tools:{tools}
Your plan: {plan}
Make action based on opinions and your plan. Don't execute action you made.
```

• 기존 data에서 봤던 instruction과 너무 다름

• 이 복잡한 프롬프트를 zero-shot or few-shot으로 넣지 말고 fine-tuning 시킬 수 있다면?

• Fine-tuning 시키려면 해당 데이터셋을 만들어야 하는데 어떻게 만들어야 하지?

Fine-Tuning

 Few-Shot Parameter-Efficient Fine-Tuning is Better and Cheaper than In-Context Learning

• data가 5개 이하여도 fine-tuning이 더 좋다

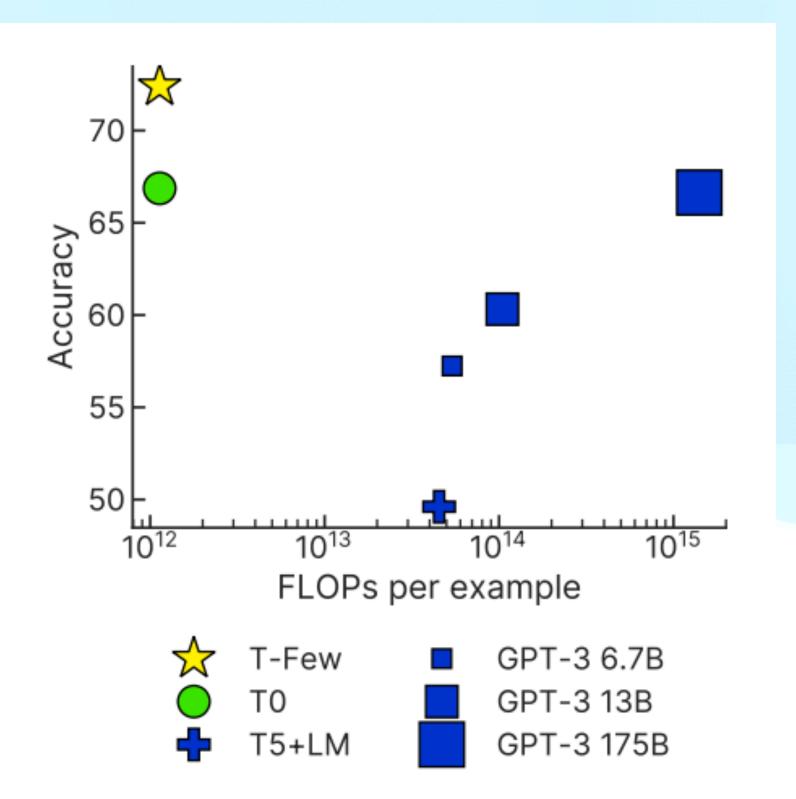


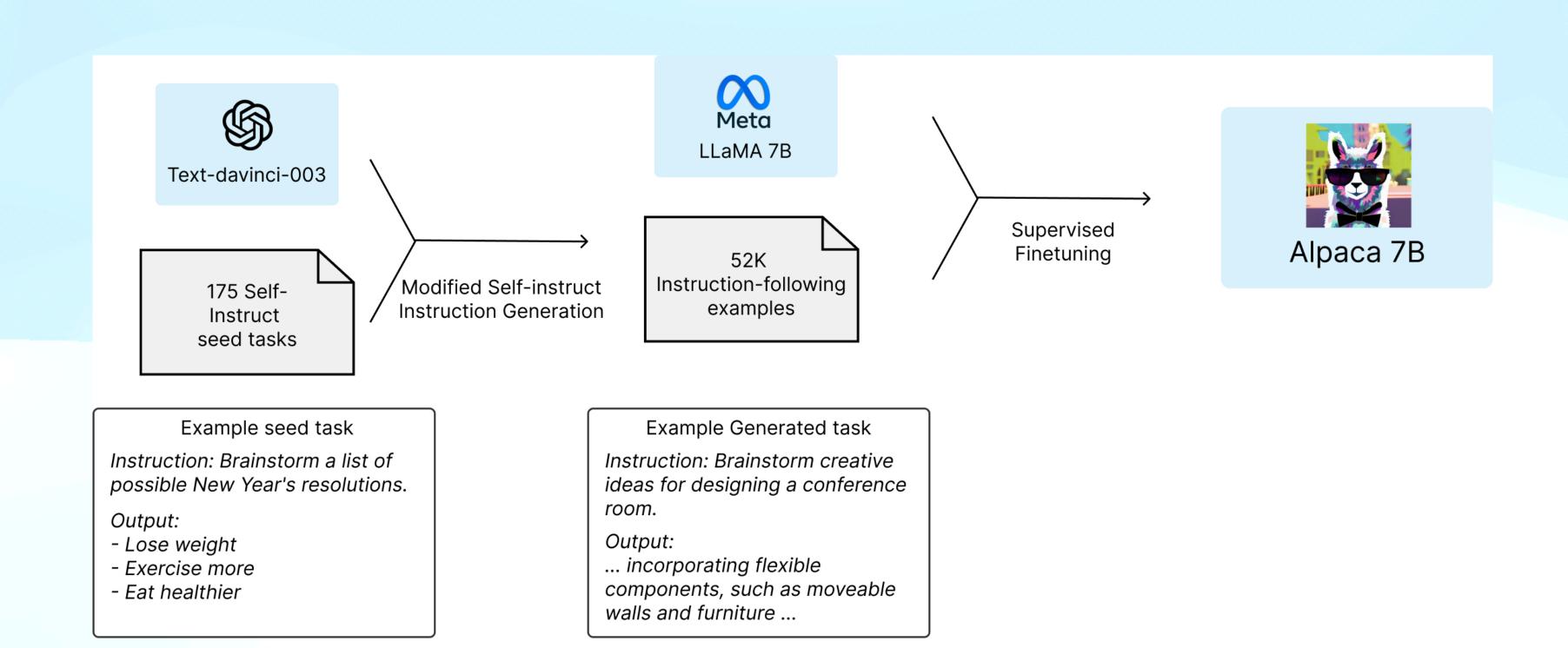
Figure 3: Accuracy of different few-shot learning methods. T-Few uses (IA)³ for PEFT methods of T0, T0 uses zero-shot learning, and T5+LM and the GPT-3 variants use few-shot ICL. The x-axis corresponds to inference costs; details are provided in section 4.2.

Fine-Tuning

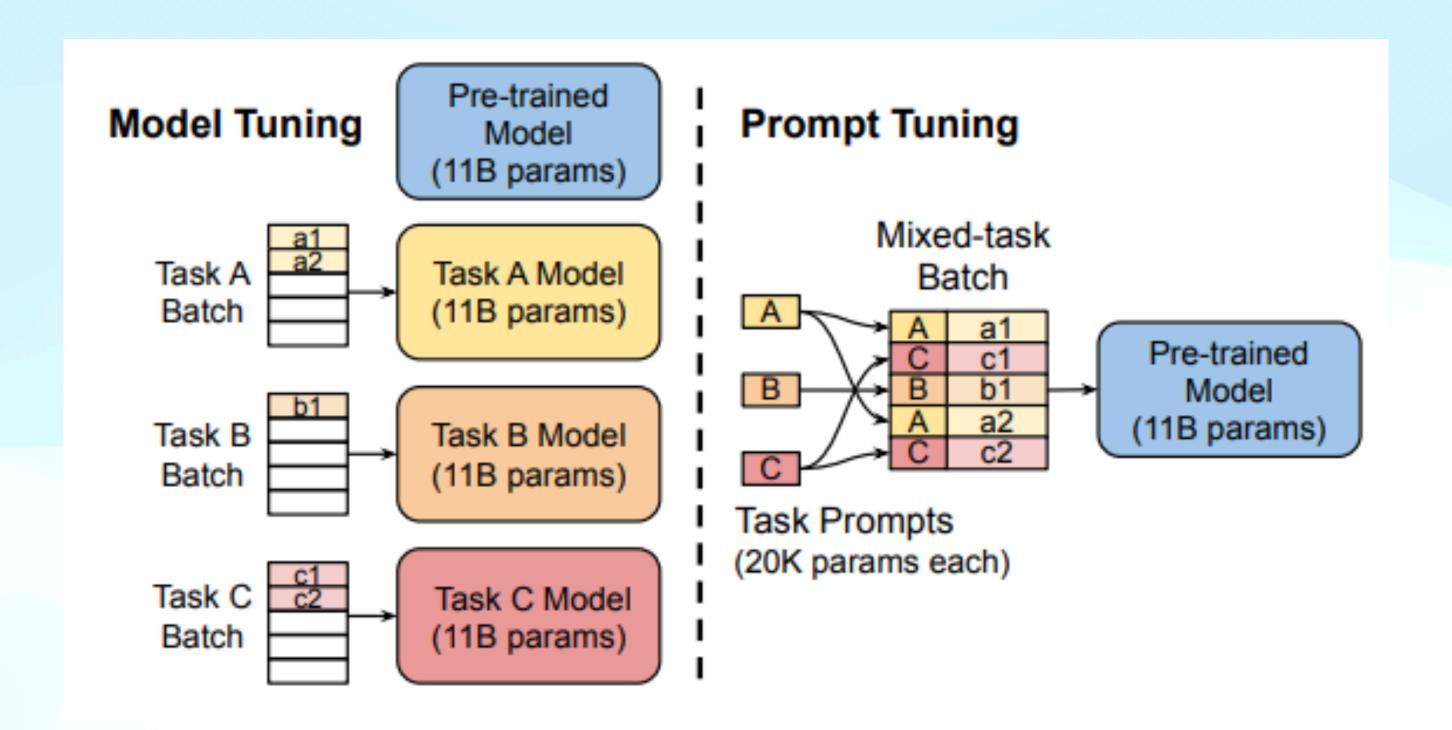
데이터는 어떻게 생성하는가?

엄청난 모델의 크기는 어떻게 감당하는가?

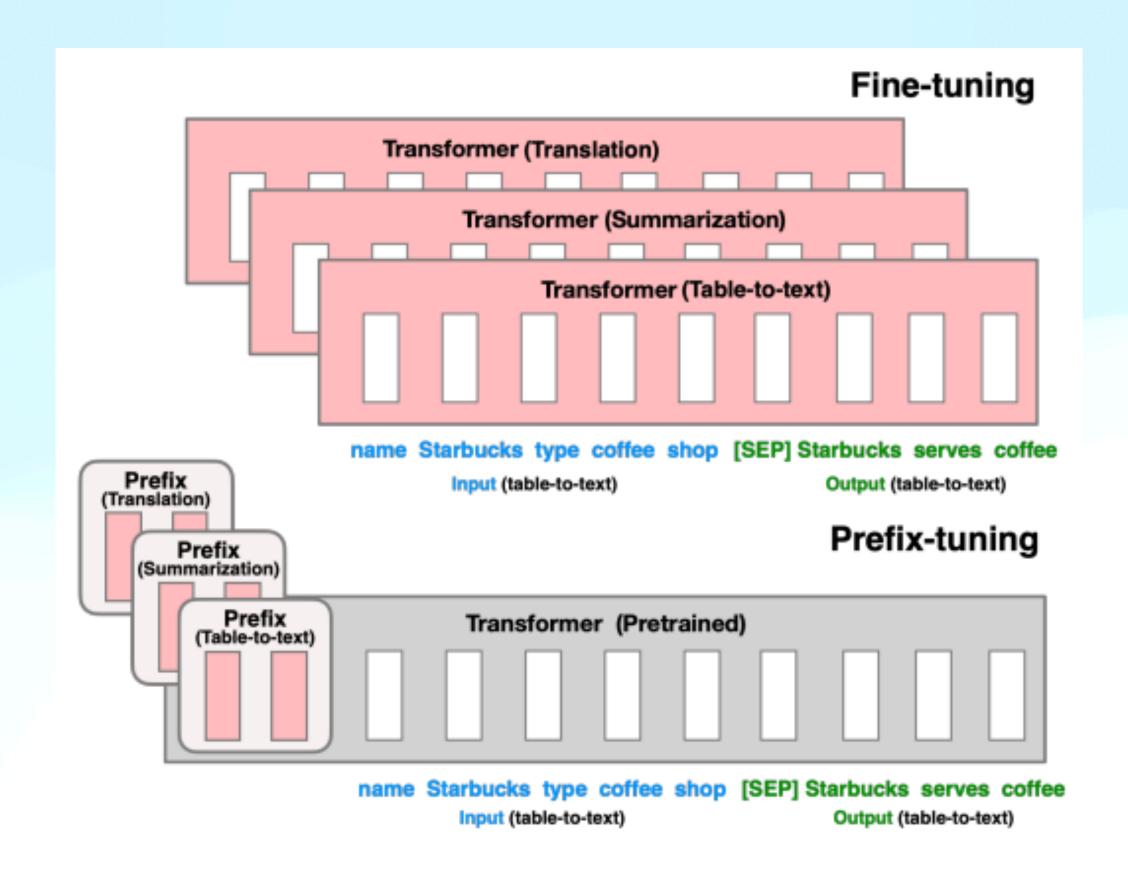
Fine-Tuning? Prompt engineering?



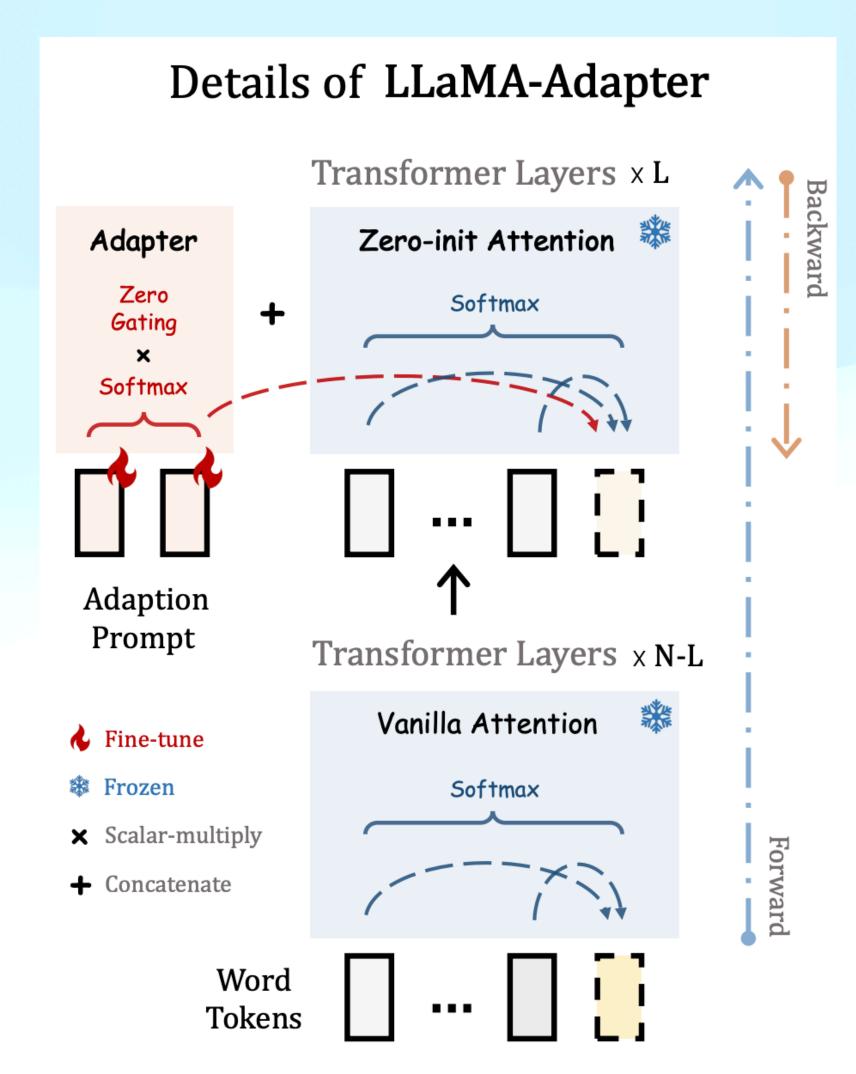
Prompt Tuning



Prefix-Tuning



LLaMA-Adapter



LoRA

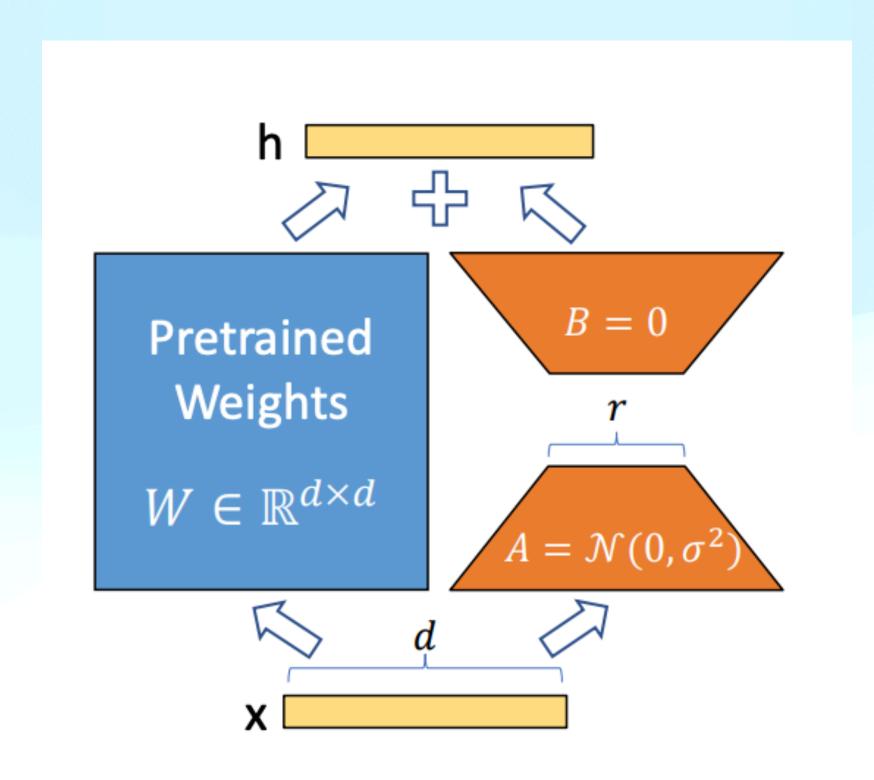


Figure 1: Our reparametrization. We only train A and B.

Performance

LoRA can even outperform full finetuning training only 2% of the parameters

Full finetuning	Model&Method	# Trainable Parameters	WikiSQL Acc. (%)	MNLI-m Acc. (%)	SAMSum R1/R2/RL	- ROUGE scores
	GPT-3 (FT)	175,255.8M	73.8	89.5	52.0/28.0/44.5	
Only tune bias vectors>	GPT-3 (BitFit)	14.2M	71.3	91.0	51.3/27.4/43.5	
D	GPT-3 (PreEmbed)	3.2M	63.1	88.6	48.3/24.2/40.5	
Prompt tuning	GPT-3 (PreLayer)	20.2M	70.1	89.5	50.8/27.3/43.5	
Prefix tuning	GPT-3 (Adapter ^H)	7.1M	71.9	89.8	53.0/28.9/44.8	
	GPT-3 (Adapter ^H)	40.1M	73.2	91.5	53.2/29.0/45.1	
	GPT-3 (LoRA)	4.7M	73.4	91.7	53.8/29.8/45.9	
	GPT-3 (LoRA)	37.7M	74.0	91.6	53.4/29.2/45.1	

Table 4: Performance of different adaptation methods on GPT-3 175B. We report the logical form validation accuracy on WikiSQL, validation accuracy on MultiNLI-matched, and Rouge-1/2/L on SAMSum. LoRA performs better than prior approaches, including full fine-tuning. The results on WikiSQL have a fluctuation around $\pm 0.5\%$, MNLI-m around $\pm 0.1\%$, and SAMSum around $\pm 0.2/\pm 0.2/\pm 0.1$ for the three metrics.

Conclusion

- 조만간 CloseAl zero-shot < open-source LLM fine-tuning 이 될 것이다
- 좋은 서비스를 만들기 위해서는 Prompt Engineering / Fine-Tuning 모두 상당한 역량을 갖추고 있어야 한다.
- LangChain, llama-index 처럼 software 3.0 테크닉도 계속 공부하고 fine-tuning을 어떻게 시키는지에 대한 공부도 지속적으로 필요하다

#