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Gemma简介



Service Sound Service Servi

内容提要

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- 02 Gemma简介
- 03 开始使用Gemma
- 04 Gemma新进展

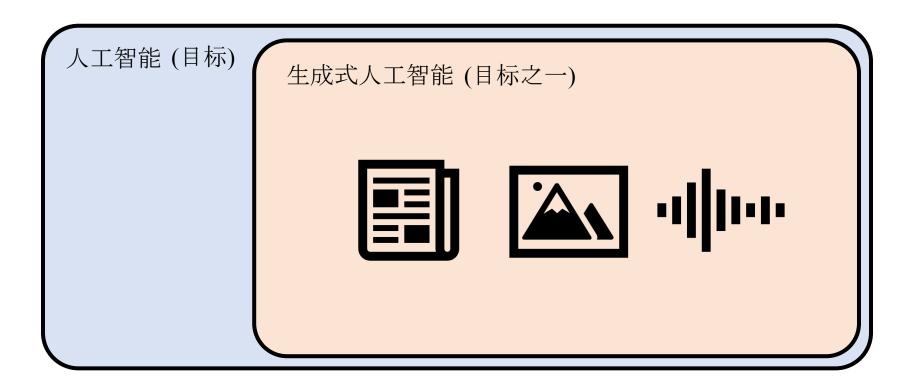


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生成式AI简介

生成式人工智能



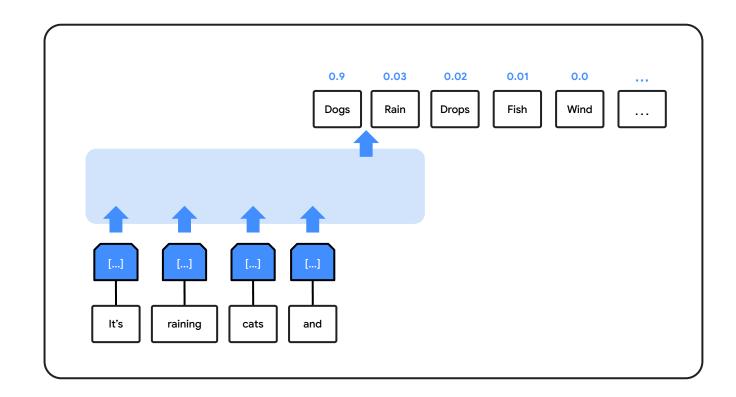
生成式人工智能



生成式人工智慧

人工智能(目标) 机器学习(手段) 深度学习(更厉害的手段) 生成式人工智能 今天的生成式人工智能多以深度学习实现

What is an LLM?



Roses are red,

Roses are red,
Violets are blue,
Sugar is sweet,

for(var i = 0)

```
for(var i = 0; i < 10; i++) {
```

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Gemma简介

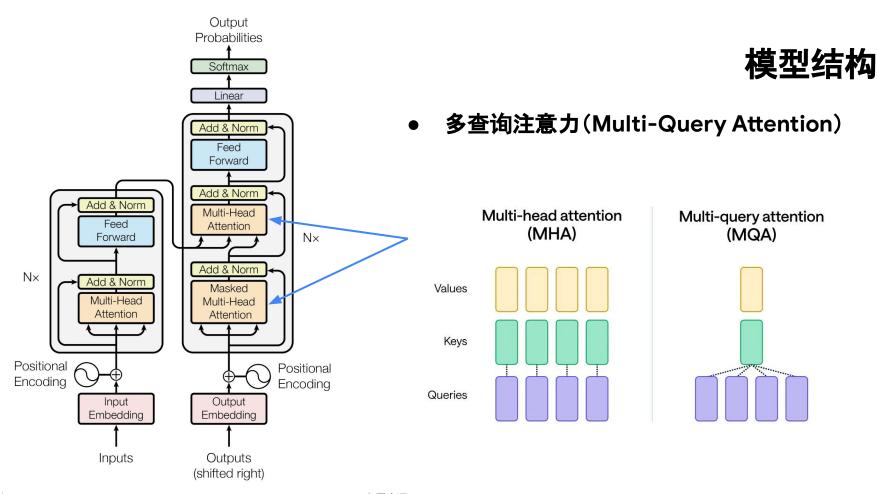


Figure 1: The Transformer - model architecture.

左图来源: Attention Is All You Need

右图来源: Grouped Query Attention (GQA) vs. Multi Head Attention (MHA): Optimizing LLM Inference Serving

Output **Probabilities** Softmax Linear Add & Norm Feed Forward Add & Norm Add & Norm Multi-Head Feed Attention Forward $N \times$ Add & Norm N× Add & Norm Masked Multi-Head Multi-Head Attention Attention Positional Positional Encoding Encoding Output Input Embedding Embedding Inputs Outputs (shifted right)

Figure 1: The Transformer - model architecture.

模型结构

RoPE嵌入(Rotary Position Embedding)

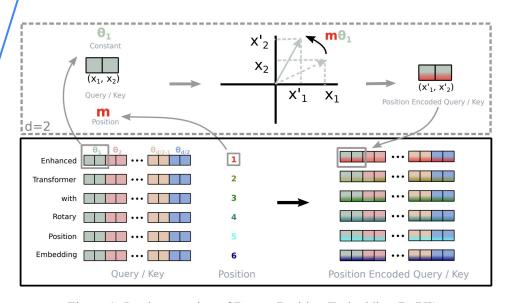


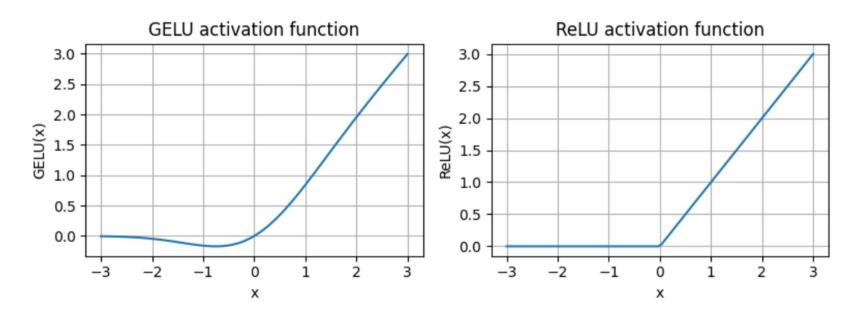
Figure 1: Implementation of Rotary Position Embedding(RoPE).

左图来源: Attention Is All You Need

右图来源: ROFORMER: ENHANCED TRANSFORMER WITH ROTARY POSITION EMBEDDING

模型结构

● GeGLU激活函数



模型结构

Probabilities Softmax Linear Add & Norm Feed Forward Add & Norm Add & Norm Multi-Head Feed Attention Forward Add & Norm N× Add & Norm Masked Multi-Head Multi-Head Attention Attention Positional Positional Encoding Encoding Output Input Embedding Embedding Inputs Outputs (shifted right)

Output

RMSNorm

$$ext{RMSNorm}(x_i) = rac{x_i}{\sqrt{rac{1}{N}\sum_{i=1}^N x_i^2 + \epsilon}}$$

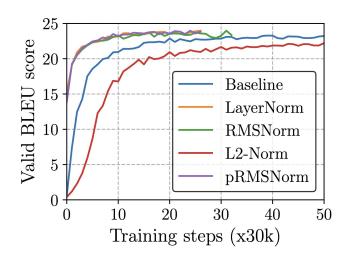


Figure 1: The Transformer - model architecture.

左图来源: Attention Is All You Need 右图来源: Root Mean Square Layer Normalization

模型结构

Parameters	2B	7B
d _model	2048	3072
Layers	18	28
Feedforward hidden dims	32768	49152
Num heads	8	16
Num KV heads	1	16
Head size	256	256
Vocab size	256128	256128

Table 1 | Key model parameters.

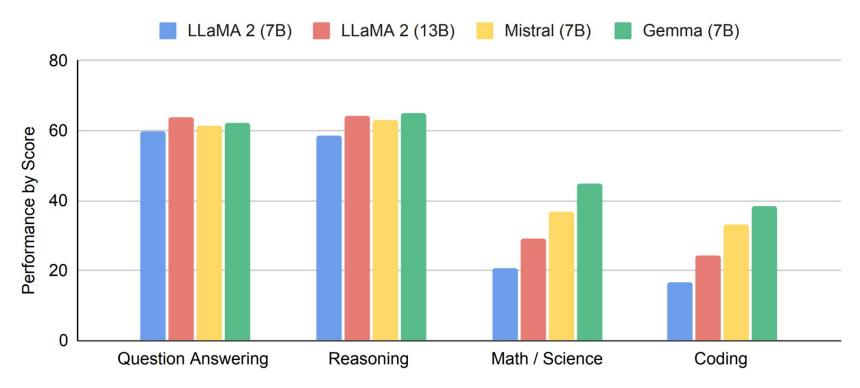
训练数据

- Gemma 2B和7B分别在3万亿和6万亿的主要以英语为主的数据上进行了训练
- 数据来自网络文档、数学和编程内容
- Gemma模型不是多模态的,也没有针对多语言任务的最新表现进行训练
- 词汇表的大小为256,000个token

模型尺寸与性能

参数大小	输入	输出	调整过的版本	预期平台
2B	文本	文本	预训练、指令调整	移动设备和笔记本电脑
7B	文本	文本	预训练、指令调整	台式电脑和小型服务器

模型性能





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开始使用Gemma

获取Gemma模型









Access Gemma on Kaggle

To access Gemma on Kaggle, you need to first request access.

Request Access

准备工作



首先, 将Keras 3和KerasNLP安装到您的环境中, 然后导入keras nlp模块。

```
!pip install --upgrade keras-nlp
!pip install --upgrade keras
import keras_nlp
```

接着,从预设配置中加载Gemma模型!

```
# https://keras.io/api/keras_nlp/models/gemma/gemma_causal_lm/
g_lm = keras_nlp.models.GemmaCausalLM.from_preset("gemma_2b_en")
```

预设配置可用于Gemma的2B和7B参数版本。

使用Gemma



只需将提示词传递给generate()函数,并可选地指定响应的最大长度。

例如, 如果问Gemma"it was a dark and stormy night."

txt = g_lm.generate("It was a dark and stormy night.", max_length=64)
print(txt)

It was a dark and stormy night.

The rain was pouring down, and the wind was howling.

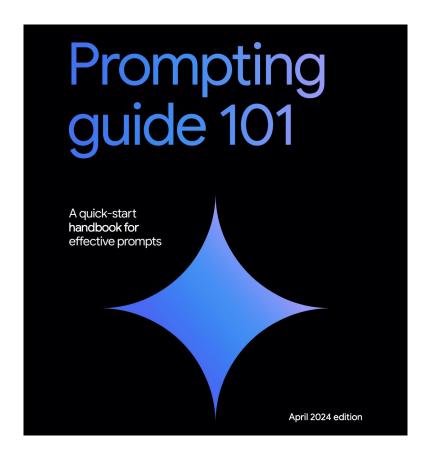
But that didn't stop a group of friends from going out for a night of fun.

They were all dressed up in their best clothes, and they were ready to have a good time.

提示词编写指导

Prompting Guide 101

Writing effective prompts
Introduction
Customer service Page 7
Executives and entrepreneurs Page 13
Human resources Page 20
Marketing Page 26
Project management. Page 32
Sales
Leveling up your prompt writing Page 43





微调Gemma

● 使用Keras进行微调

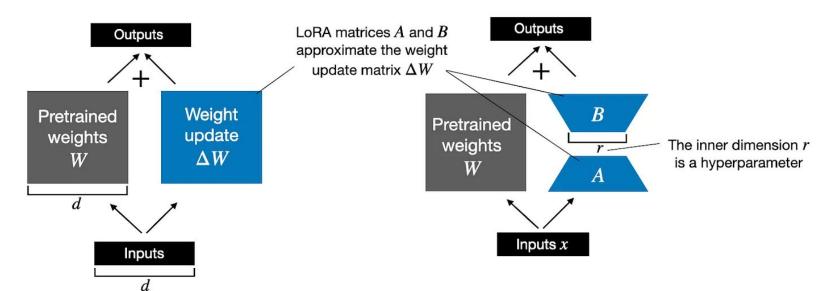
● 支持使用LoRA进行微调



LoRA

Weight update in regular finetuning

Weight update in LoRA



微调Gemma

● 使用Keras进行微调



● 支持使用LoRA进行微调

```
gemma.backbone.enable_lora(rank=8)
# fine-tune ...
gemma.fit(...)
gemma.backbone.save_lora_weights("lora.h5")
```



分布式微调Gemma



```
devices=keras.distribution.list devices()
device mesh = keras.distribution.DeviceMesh((1, 8),["batch", "model"], devices))
layout map = keras.distribution.LayoutMap(device mesh)
# Partitioning for embeddings (regex)
layout map["token embedding/embeddings"] = (None, "model")
# Partitioning (regex) for attention layer weights
layout map["decoder block.*attention.*(query|key|value).*kernel"] = (None, "model", None)
layout map["decoder block.*attention output.*kernel"] = (None, None, "model")
layout map["decoder block.*ffw gating.*kernel"] = ("model", None)
layout map["decoder block.*ffw linear.*kernel"] = (None, "model")
keras.distribution.set distribution(keras.distribution.ModelParallel(device mesh,
                                                                        layout map,
                                                                        batch dim name="batch"))
# - load the model here
```

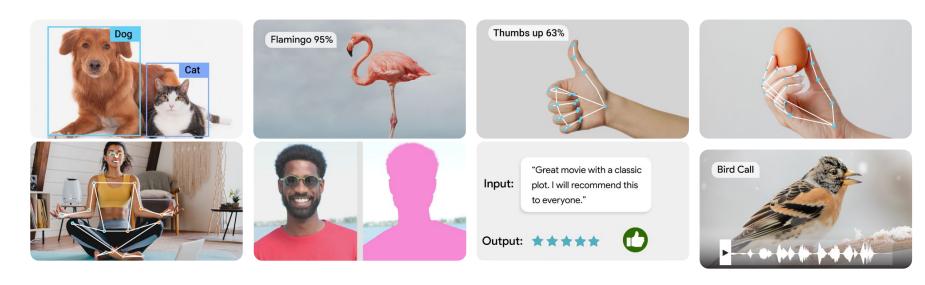
微调Gemma

- 使用Keras进行微调
- 支持使用LoRA进行微调
- 支持JAX进行微调





使用MediaPipe在设备上的集成Gemma



- Studio: https://mediapipe-studio.webapps.google.com/demo/llm_inference
- Documentation: https://developers.google.com/mediapipe/solutions/genai/llm_inference

代码示例

- Android:
 - https://github.com/googlesamples/mediapipe/tree/main/examples/llm_inferenc
 e/android
 - https://github.com/NSTiwari/Gemma-on-Android
- Web:

https://github.com/googlesamples/mediapipe/tree/main/examples/llm_inference/js

iOS:

https://github.com/googlesamples/mediapipe/tree/main/examples/llm_inference/ios

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Gemma新进展

Gemma家族

- CodeGemma
- <u>PaliGemma</u> (<u>中文介绍</u>)
- RecurrentGemma

参考资料

- 1.Gemma的官方网站: https://ai.google.dev/gemma/docs
- 2.Gemma技术报告 : https://storage.googleapis.com/deepmind-media/gemma/gemma-report.pdf
- 3.Gemma: Introducing new state-of-the-art open model by Google, https://medium.com/@shravankoninti/gemma-introducing-new-state-of-the-art-open-model-by-google-caae9fe29972
- 4.Understanding, Using, and Finetuning
 Gemma, https://lightning.ai/lightning-ai/studios/understanding-using-and-finetuning-gemma
- 5.What is Low-Rank Adaptation (LoRA) | explained by the inventor, https://www.youtube.com/watch?v=DhRoTONcyZE
 Google Developer Groups

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码出未来,现在开始!