#### 1-使用大模型进行指令学习

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# 1-使用大模型进行指令学习

## 尝试获取Stanford Sentiment Treebank (SST-2)数据集并运行

```
# 加载 SST-2 数据集

dataset = load_dataset("glue", "sst2", split="test[:20]")

# 查看训练集的前几个样本
print(dataset)
```

#### 一个具体的SST-2测试样例

```
this film's relationship to actual tension is the same as what christmastree flocking in a spray can is to actual snow: a poor— if durable— imitation.
```

#### 使用DeBERTa的传统编码器微调

尝试用DeBERTa模型,在训练集上微调后,在test来进行预测,看看具体准确率是多少?

- 用全部的train训练集微调deberta-base, 记录accuracy。
- 仅使用16、64、256和1024个样本, 微调deberta-base, 记录accuracy。

## 什么是指令学习?

指令学习 (Instruction Learning) 是一种让机器根据人类提供的指导性信息来完成特定任务的方法。在指令学习中,机器不依赖大规模的数据集进行预测和决策,而是依赖于人类

提供的指导性信息。这种学习方法在处理一些结构化的任务时具有显著的优势,比如问答、机器翻译和文本生成等。

说到底一句话:写一个指令模版,让生成式模型接着进行生成。

例如:

Below is an instruction that describes a task, paired with an input that provides further context. Write a response that appropriately completes the request.

#### ### Instruction:

Analyze the given text from an online review and determine the sentiment polarity. Return a

single number of either -1 and 1, with 1 being positive and -1 being the negative sentiment.

No further explanation or justification is required.

#### ### Question:

this film's relationship to actual tension is the same as what christmastree flocking in a

spray can is to actual snow: a poor -- if durable -- imitation. What's the sentiment polarity

of the text?

### Response:

试着把这句话发给ChatGPT,看看它会返回什么?

### Instruction:
Analyze the given text from an online review and determine the sentiment polarity. Return a single number of either -1 and 1, with 1 being positive and -1 being the negative sentiment.

No further explanation or justification is required.

### Question:
this film 's relationship to actual tension is the same as what christmas-tree flocking in a spray can is to actual snow: a poor -- if durable -- imitation. What's the sentiment polarity of the text?

### Response:

#### ChatGPT直接返回了-1,看起来和正确答案是一致的,这就带来几个问题:

- 大模型的智能看起来是通过生成模型的"涌现"机制来完成的,也就是大模型在使用之前经过来类似的过程来进行训练,这样它可以根据上文来预测下文的生成。
- 越大越新的模型可以带来越好的指令学习能力,反之,越小越旧的模型可能会出现"幻觉"(hallucination)。因此,在进行指令学习而不需要进行模型的训练时,尽可能使用更大的模型来完成,例如ChatGPT、GPT-o1、全量的DeepSeek-r1、Qwen等模型。
- 指令模版一般参考alpaca,其书写规范可以参考: https://github.com/tatsu-lab/stanford\_alpaca

#### 参考这样的格式:

Below is an instruction that describes a task, paired with an input that provides further context. Write a response that appropriately completes the request.

### Instruction:
{instruction}

### Input:
{input}

### Response:

万事具备,我们试着把所有的代码合在一起,调用一个DeepSeek Distil Llama 8B模型,利用推理模式来完成一个指令学习。

```
import transformers
import torchmodel_id = "deepseek-ai/DeepSeek-R1-Distill-Llama-8B"
```

```
pipeline = transformers.pipeline(
    "text-generation",
    model=model id,
    max new tokens=2048,
    model kwargs={"torch dtype": torch.bfloat16},
   device map="auto"
prompt = """Below is an instruction that describes a task,
paired with an input that provides further context. Write a
response that appropriately completes the request.
Before answering, think carefully about the question and
create a step-by-step chain of thoughts to ensure a logical
### Instruction:
You are an expert of consumer comment analysis. Analyze the given text from
the sentiment polarity. Return a single number of either -1
and 1, with -1 being negative and 1 being the positive
sentiment.
what christmas-tree flocking in a spray can is to actual
snow : a poor -- if durable -- imitation .
### Response:
response = pipeline(prompt)
print(response[0]["generated text"])
```

#### 注意几个细节:

- 在Instruction的第一句,要明确的说明所分析的领域,类似于角色扮演,告诉模型你是什么角色。
- 具体执行的过程放在Instruction部分,给出明确的输出。
- 注意每个section之前的###
- 注意在Respone之后的<think>标签,这是为了触发模型的Reasoning过程

到这里,我们可以试着给模型更多的输入,并且写一个函数把最终的结果解析出来,并且作为预测标签:

```
from datasets import load dataset
# 加载 SST-2 数据集
dataset = load_dataset("glue", "sst2", split="test[:20]")
# 查看训练集的前几个样本
print(dataset)
import transformers
import torch
model id = "deepseek-ai/DeepSeek-R1-Distill-Llama-8B"
pipeline = transformers.pipeline(
    "text-generation",
    model=model id,
    max new tokens=2048,
   model_kwargs={"torch_dtype": torch.bfloat16},
   device map="auto"
prompt style = """Below is an instruction that describes a task, paired with
an input that provides further context. Write a response that appropriately
completes the request.
### Instruction:
Analyze the given text from an online review and determine the sentiment
polarity. Return a single number of either -1 and 1, with -1 being negative
and 1 being the positive sentiment.
```

## 试着改变不同的模型大小

接下来,试着改变模型的大小,并且完成所有样本的预测后,和test的真实标签,统计一个准确率看看。

注意到,我们实际并没有使用到train训练集,也不需要去训练这个模型。

那这就带来一个问题,我是否可以自己来架设一个全量的DeepSeek来进行推理呢?

# 一、利用Ollama部署DeepSeek

- 首先从Ollama的官网下载安装包: https://ollama.com
- 在左上角的Models标签里面可以找到不同模型
- 可以试着用cmd, 先运行一个32B的模型

```
ollama run deepseek-r1:32b
```

运行之后,再试着把刚才的指令发送给deepseek看看。

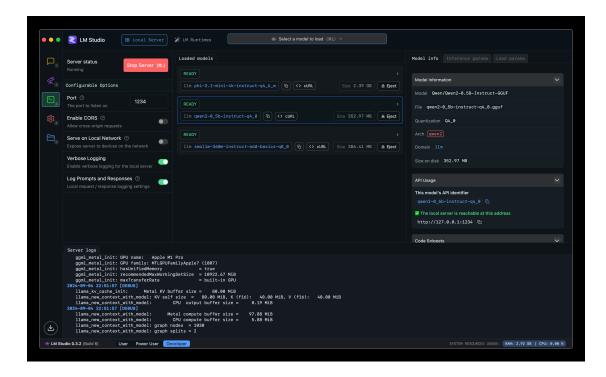
# 如何用代码API直接调用自己架设的DeepSeek

可以参考:

https://github.com/datawhalechina/handy-ollama/tree/main/docs/C4

# 二、使用LMStudio部署DeepSeek

- 首先从LMStudio下载安装包: https://lmstudio.ai
- 假设REST API
- 可以参考: https://lmstudio.ai/docs/api



# 记录结果

试着用Llama、Qwen、Gemma、Phi-4和Mistral来统计下SST-2指令学习的准确度。再和全量微调、小样本微调的DeBERTa结果对比一下。