### 如何使用 Transformers 和 Tokenizers 从头开始训练新的语言模型

Denis Rothman。Denis Rothman 参考 Hugging Face 的笔记本,预训练了一个 transformer 模型。接下来的步骤是构建更大的数据集并测试多个 transformer 模型。

推荐理解这个笔记本。GPT-3 引擎的出现提供了一个可以超越许多训练过的 transformer 模型的 API。然而,要让 transformer 知道如何处理输入数据集,理解它们的训练过程是至关重要的。

这个笔记本中的 Transformer 模型名为 *KantaiBERT*。 *KantaiBERT* 是一个以 RoBERTa Transformer 方式训练的,采用 DistilBERT 架构的模型。数据集是从 <u>Jiumo Search</u> 下载的Pride+and+Prejudice,To+Kill+A+Mockingbird,THE CATCHER IN THE RYE三本书编译而成的。

KantaiBERT 使用一个具有 8400 万参数的小模型进行预训练,采用了与 DistilBERT 相同的层数和头数,即 6 层、768 隐藏单元和 12 个注意力头。然后,KantaiBERT 被微调用于下游的掩码语言建模任务。

Hugging Face 原始参考和注释:

笔记本版本(参考博客文章的原始链接 link)。

### > 步骤1: 加载数据集

```
from IPython.display import Image # 这是用于在笔记本中渲染图像的
```

#### 步骤1:加载数据集

```
#@title 步骤1: 加载数据集
#1.使用 Colab 文件管理器加载 kant.txt
#2.从 GitHub 下载文件
#!curl -L https://github.com/qian-qiang/Good-Study-Day-Day-Up/tree/main/%E5%A4%A7%E6%A8%A1%E5%9E%8B%E5%AD%A6%E4%B9%A0/Book/others/kant.txt --outp
```

### 步骤 2:安装 Hugging Face 的 Transformers

#### → 步骤 2: 2024 年 6 月更新: 安装 Hugging Face Transformers

```
#@title 步骤 2: 2024 年 6 月更新: 安装 Hugging Face Transformers

""

# 这里我们不需要 TensorFlow

!pip uninstall -y tensorflow

# 从 master 分支安装 `transformers`

!pip install git+https://github.com/huggingface/transformers

!pip list | grep -E 'transformers | tokenizers'

# 本笔记本更新时的 transformers 版本 --- 2.9.1

# 本笔记本更新时的 tokenizers 版本 --- 0.7.0

""

* '\n# 这里我们不需要 TensorFlow\n!pip uninstall -y tensorflow\n# 从 master 分支安装 `transformers`\n!pip install git+https://github.com/huggingface/transformers\n!pip list | grep -E 'transformers`\n!pip install git+https://github.com/huggingface/transformers\n!pip list | grep -E 'transformers\n!pip install git+https://github.com/huggingface/transformers\n!pip list | grep -E 'transformers\n!pip list | grep -E 'transformers\
```

2023年6月更新来自 Hugging Face 问题 22816:

#### https://github.com/huggingface/transformers/issues/22816

"PartialState 导入作为依赖项在昨天添加到 transformers 的开发分支中。PartialState 是在 accelerate 0.17.0 版本中添加的,因此对于 transformers 的开发分支,需要 accelerate >= 0.17.0。

降级 transformers 版本可以移除导入 PartialState 的代码。"

Denis Rothman: 以下代码单元格导入了最新版本的 Hugging Face transformers,而无需降级。

为了适应 Hugging Face 的升级,使用 Google Colab Pro 激活了 GPU 加速器,配备了以下 NVIDIA GPU: GPU 名称:NVIDIA A100-SXM4-40GB

```
!pip install transformers

Requirement already satisfied: transformers in /usr/local/lib/python3.10/dist-packages (4.41.2)

Requirement already satisfied: filelock in /usr/local/lib/python3.10/dist-packages (from transformers) (3.14.0)

Requirement already satisfied: huggingface-hub<1.0,>=0.23.0 in /usr/local/lib/python3.10/dist-packages (from transformers) (0.23.3)

Requirement already satisfied: numpy>=1.17 in /usr/local/lib/python3.10/dist-packages (from transformers) (2.5.2)

Requirement already satisfied: packaging>=20.0 in /usr/local/lib/python3.10/dist-packages (from transformers) (24.1)

Requirement already satisfied: pyyaml>=5.1 in /usr/local/lib/python3.10/dist-packages (from transformers) (6.0.1)

Requirement already satisfied: regex!=2019.12.17 in /usr/local/lib/python3.10/dist-packages (from transformers) (2024.5.15)
```

#### KantaiBERT.ipynb - Colab

```
Requirement already satisfied: requests in /usr/local/lib/python3.10/dist-packages (from transformers) (2.31.0)

Requirement already satisfied: tokenizers<0.20,>=0.19 in /usr/local/lib/python3.10/dist-packages (from transformers) (0.19.1)

Requirement already satisfied: safetensors>=0.4.1 in /usr/local/lib/python3.10/dist-packages (from transformers) (0.4.3)

Requirement already satisfied: tqdm>=4.27 in /usr/local/lib/python3.10/dist-packages (from transformers) (4.66.4)

Requirement already satisfied: fsspec>=2023.5.0 in /usr/local/lib/python3.10/dist-packages (from huggingface-hub<1.0,>=0.23.0->transformers) (20

Requirement already satisfied: typing-extensions>=3.7.4.3 in /usr/local/lib/python3.10/dist-packages (from huggingface-hub<1.0,>=0.23.0->transformers)

Requirement already satisfied: charset-normalizer<4,>=2 in /usr/local/lib/python3.10/dist-packages (from requests->transformers) (3.3.2)

Requirement already satisfied: dima<4,>=2.5 in /usr/local/lib/python3.10/dist-packages (from requests->transformers) (2.0.7)

Requirement already satisfied: certifi>=2017.4.17 in /usr/local/lib/python3.10/dist-packages (from requests->transformers) (2.0.4.6.2)
```

### 步骤 3: 训练一个 tokenizer

#### > 步骤3:训练一个分词器

```
#@title 步骤3: 训练一个分词器
%%time
from pathlib import Path
from tokenizers import ByteLevelBPETokenizer
# 获取所有 .txt 文件的路径
paths = [str(x) for x in Path(".").glob("**/*.txt")]
# 初始化分词器
tokenizer = ByteLevelBPETokenizer()
# 自定义训练参数
tokenizer.train(files=paths, vocab size=52 000, min frequency=2, special tokens=[
       "<s>",
       "<pad>"
       "</s>",
       "<unk>"
       "<mask>"
7)
    CPU times: user 1min 21s, sys: 1.3 s, total: 1min 22s
     Wall time: 52 s
```

### 步骤4:将文件保存到磁盘

#### 步骤4:将文件保存到磁盘

# \* 步骤5:加载训练好的分词器文件

### 步骤5 加载训练好的分词器文件

## ✓ 步骤 6: 检查资源约束: GPU 和 NVIDIA

#### → 步骤 6: 检查资源约束: GPU 和 NVIDIA

```
#@title 步骤 6: 检查资源约束: GPU 和 NVIDIA
!nvidia-smi
```

Thu Jun 13 08:45:39 2024

NVIDIA-SMI	535. 104. 05	Dri	ver Version:	535. 104. 05	CUDA Versio	on: 12.2
GPU Name Fan Temp	Perf	Persistence Pwr:Usage/C		Disp.A Memory-Usage		
0 Tesla N/A 58C	T4 P0	_	!	0:00:04.0 Off liB / 15360MiB	+========     0% 	O Default N/A
Processes: GPU GI	CI ID	PID Type Pr	ocess name			GPU Memory

### ∨ 检查 PyTorch 是否能够使用 CUDA

```
#@title 检查 PyTorch 是否能够使用 CUDA import torch torch.cuda.is_available()
```

**→** True

## > 步骤 7: 定义模型配置

#### 步骤7:定义模型配置

```
"eos_token_id": 2,
"hidden_act": "gelu",
"hidden_dropout_prob": 0.1,
"hidden_size": 768,
"initializer_range": 0.02,
"intermediate_size": 3072,
"layer_norm_eps": 1e-12,
"max_position_embeddings": 514,
"model_type": "roberta",
"num_attention_heads": 12,
"num_hidden_layers": 6,
"pad_token_id": 1,
"position_embedding_type": "absolute",
"transformers_version": "4.41.2",
"type_vocab_size": 1,
"use_cache": true,
"vocab_size": 52000
```

### > 步骤 8: 在 transformers 中重新加载分词器

#### 步骤 8:在 transformers 中重新加载分词器

```
#@title 步骤 8: 在 transformers 中重新加载分词器
from transformers import RobertaTokenizer
tokenizer = RobertaTokenizer.from_pretrained("./KantaiBERT", max_length=512)
```

### 步骤 9: 从头开始初始化模型

#### 步骤 9: 从头开始初始化模型

```
#@title 步骤 9: 从头开始初始化模型
from transformers import RobertaForMaskedLM
model = RobertaForMaskedLM(config=config)
print(model)
RobertaForMaskedLM(
       (roberta): RobertaModel(
         (embeddings): RobertaEmbeddings(
            (word_embeddings): Embedding(52000, 768, padding_idx=1)
            (position_embeddings): Embedding(514, 768, padding_idx=1)
            (token type embeddings): Embedding(1, 768)
           (LayerNorm): LayerNorm((768,), eps=1e-12, elementwise_affine=True)
           (dropout): \ Dropout (p=0.1, \ inplace=False)
         (encoder): RobertaEncoder(
           (layer): ModuleList(
             (0-5): 6 x RobertaLayer(
               (attention): RobertaAttention(
                 (self): RobertaSelfAttention(
                    (query): Linear(in_features=768, out_features=768, bias=True)
                    (key): Linear(in_features=768, out_features=768, bias=True)
                    (value): Linear(in_features=768, out_features=768, bias=True)
                   (dropout): Dropout(p=0.1, inplace=False)
                 (output): RobertaSelfOutput(
                    (dense): Linear(in_features=768, out_features=768, bias=True)
                    (LayerNorm): LayerNorm((768,), eps=1e-12, elementwise_affine=True)
                    (dropout): Dropout(p=0.1, inplace=False)
               (intermediate): RobertaIntermediate(
                 (dense): Linear(in_features=768, out_features=3072, bias=True)
                 (intermediate_act_fn): GELUActivation()
               (output): RobertaOutput(
                 (dense): Linear(in features=3072, out features=768, bias=True)
                  (LayerNorm): LayerNorm((768,), eps=1e-12, elementwise_affine=True)
                 (dropout): Dropout(p=0.1, inplace=False)
          )
       (1m head): RobertaLMHead(
         (dense): Linear(in_features=768, out_features=768, bias=True)
         (layer_norm): LayerNorm((768,), eps=1e-12, elementwise_affine=True)
         (decoder): Linear(in_features=768, out_features=52000, bias=True)
```

```
2024/6/13 17:10
```

### 探索模型参数

```
print(model.num_parameters())
# => 84,095,008 parameters

$\frac{1}{2}$ 83504416
```

### ~ 探索模型参数

```
#@title 探索模型参数
LP=list(model.parameters())
lp=len(LP)
print(lp)
for p in range(0,lp):
    print(LP[p])
```

```
Parameter containing:
requires_grad=True)
```

#### 计算模型参数数量

```
#@title 计算模型参数数量
np=0
for p in range(0, 1p):#number of tensors
    PL2=True
       L2=1en(LP[p][0]) #check if 2D
    except:
                                    #not 2D but 1D
       L2=1
       PL2=Fa1se
    L1=len(LP[p])
   L3=L1*L2
    np+=L3
                                   # number of parameters per tensor
    if PL2==True:
       print(p,L1,L2,L3)  # displaying the sizes of the parameters
    if PL2==False:
       print(p,L1,L3) # displaying the sizes of the parameters
print(np)
                                    # total number of parameters
    49 768 3072 2359296
50 768 768
 ₹
     51 768 768
     52 768 768
     53 768 768 589824
     54 768 768
     55 768 768 589824
     56 768 768
     57 768 768 589824
     58 768 768
     59 768 768 589824
     60 768 768
     61 768 768
     62 768 768
     63 3072 768 2359296
     64\ 3072\ 3072
     65 768 3072 2359296
     66 768 768
     67 768 768
     68 768 768
     69 768 768 589824
     70 768 768
     71 768 768 589824
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     75 768 768 589824
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     80 3072 3072
     81 768 3072 2359296
     82 768 768
     83 768 768
     84 768 768
     85 768 768 589824
     86 768 768
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     88 768 768
     89 768 768 589824
     90 768 768
     91 768 768 589824
     92 768 768
     93 768 768
     94 768 768
     95 3072 768 2359296
     96 3072 3072
     97\ 768\ 3072\ 2359296
     98 768 768
     99 768 768
     100 768 768
     101 52000 52000
     102 768 768 589824
     103 768 768
     104 768 768
     105 768 768
     83504416
```

# 步骤 10:构建数据集

#### ∨ 步骤 10: 构建数据集

/usr/local/lib/python3.10/dist-packages/transformers/data/datasets/language\_modeling.py:119: FutureWarning: This dataset will be removed from th warnings.warn(
CPU times: user 1min 55s, sys: 738 ms, total: 1min 56s

Wall time: 1min 59s

4

# · 步骤 11: 定义数据收集器

#### ∨ 步骤 11: 定义数据收集器

### > 步骤 12: 初始化训练器

```
#如果下步骤12初始化训练器失败提示accelerate版本太低就运行此代码,并从新连接
#!pip install accelerate
#import accelerate
#print(accelerate.__version__)
```

#### ~ 初始化训练器

# > 步骤 13: 预训练模型

#### > 步骤 13: 预训练模型

```
#@title 步骤 13: 预训练模型
%%time
trainer.train()
```

- 步骤 14:将最终模型(包括分词器和配置)保存到磁盘
- 步骤 14:将最终模型(包括分词器和配置)保存到磁盘

```
#@title 步骤 14: 将最终模型(包括分词器和配置)保存到磁盘trainer.save_model("./KantaiBERT")
```

- > 步骤 15: 使用 FillMaskPipeline 进行语言建模
- ∨ 步骤 15: 使用 FillMaskPipeline 进行语言建模

```
#@title 步骤 15: 使用 FillMaskPipeline 进行语言建模
from transformers import pipeline
fill_mask = pipeline(
         "fill-mask",
        model="./KantaiBERT",
         tokenizer="./KantaiBERT"
\label{eq:continuous_section} fill\_mask("I \ don \ t \ currently \ have \ the \ time \ for \ giving \ a \ proper \ long \ \mbox{\em mask}.")
F [{'score': 0.018490975722670555,
         'token': 262,
        'token_str':',',
'sequence': 'I don t currently have the time for giving a proper long..'},
        {'score': 0.01700388640165329,
         'token': 280,
        'token_str': 'I',
'sequence': 'I don t currently have the time for giving a proper long I.'},
        {'score': 0.004652389325201511,
         token': 18,
        'token_str':'.',
'sequence':'I don't currently have the time for giving a proper long..'},
        {'score': 0.003358564805239439,
         'score . . . . 'token' : 278,
        'token_str':' ':',
'sequence': 'I don't currently have the time for giving a proper long :.'},
        {'score': 0.002129371277987957,
         'token': 288,
        'token_str':'..',
'sequence':'I don't currently have the time for giving a proper long...'}]
```