# Q&A using Traditional Chinese Large Language model

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#### Dataset

- 使用原本Kaggle 提供的資料外,額外擴增了兩筆資料
  - https://github.com/ntunlplab/traditional-chinese-alpaca/tree/main/data
  - https://github.com/A-baoYang/alpaca-7b-chinese/tree/main/data/general

```
"instruction": "Answer the given multiple choice question.",
    "input": "6月1號開始,負責總統府維安勤務的憲兵全面換裝,憲兵211營營長黃厚泰表示,新的勤利
    "output": "1"

},
    "instruction": "Answer the given multiple choice question.",
    "input": "6月1號開始,負責總統府維安勤務的憲兵全面換裝,憲兵211營營長黃厚泰表示,新的勤利
    "output": "3"

[],
    {
        "instruction": "Answer the given multiple choice question.",
        "input": "國人愛吃海鮮,去年整年進口食用貝類高達8500公噸,為加強食用貝類輸入源頭管理,食
        "output": "2"
        },
        {
        "output": "2"
        },
        {
        "output": "2"
        },
        {
        "
```

## Code

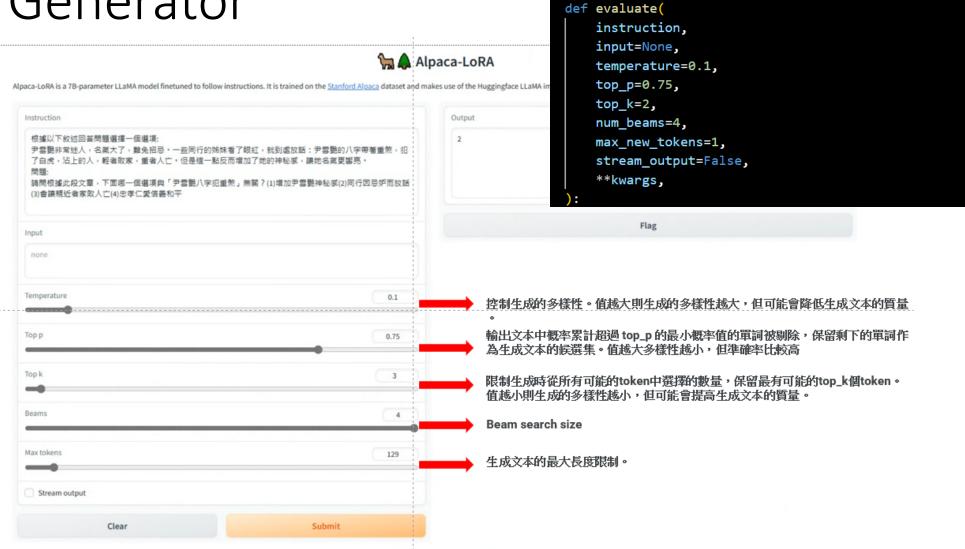
- datapre.ipynb: 進行資料處理
- lora\_finetune.py: 進行Finetune
- lora\_generate2.py : 產生結果
- Bash.py:執行後就會跑完整個流程並產生結果(使用四卡進行運 算)
- Adapter\_model.bin : 放在tag(Final\_model)裡面

## Finetune Hyperparameter

```
base model: str = 'decapoda-research/llama-7b-hf', # the only required argument
data path: str = "/data/tzeshinchen/deep learning/kaggle/lora/alpaca-lora/Orgin.json",
output dir: str = ("./lora-alpaca extended"),
batch size: int = 256,
micro batch size: int = 4,
                                                                                 調整了 單次batch size = 256
num epochs: int = 3,
                                                                                 Epoch = 3
learning rate: float = 2e-4,
                                                                                 lora target modules:
cutoff len: int = 100,
val set size: int = 100,
                                                                                 使用"q proj" "v proj" "k proj" "o proj"
# lora hyperparams
lora r: int = 8,
lora alpha: int = 16,
lora dropout: float = 0.05,
lora target modules: List[str] = ["q proj", "v proj", "k proj", "o proj"],
train on inputs: bool = True, # if False, masks out inputs in loss
add eos token: bool = False,
group by length: bool = True, # wandb params
resume from checkpoint: str = ("/data/tzeshinchen/deep learning/kaggle/lora/alpaca-lora/lora-alpaca extended")
```

用台大先pretrain 好的中文模型





### Loss

```
wallitings.walli( | MacMutobitte. inputs will be cast If oil (A.utype) to Iloati
{'loss': 1.7864, 'learning_rate': 2e-05, 'epoch': 0.19}
{'loss': 1.4509, 'learning_rate': 4e-05, 'epoch': 0.38}
{'loss': 1.3631, 'learning rate': 6e-05, 'epoch': 0.57}
{'loss': 1.3183, 'learning_rate': 8e-05, 'epoch': 0.76}
{'loss': 1.2894, 'learning rate': 0.0001, 'epoch': 0.95}
{'loss': 1.2577, 'learning_rate': 0.00012, 'epoch': 1.14}
{'loss': 1.2339, 'learning_rate': 0.00014, 'epoch': 1.33}
{'loss': 1.2033, 'learning rate': 0.00016, 'epoch': 1.52}
{'loss': 1.1849, 'learning rate': 0.00018, 'epoch': 1.71}
{'loss': 1.1452, 'learning_rate': 0.0002, 'epoch': 1.9}
{'loss': 1.1176, 'learning rate': 0.00016428571428571428, 'epoch': 2.09}
{'loss': 1.1036, 'learning rate': 0.00012857142857142858, 'epoch': 2.28}
{'loss': 1.0722, 'learning_rate': 9.285714285714286e-05, 'epoch': 2.47}
{'loss': 1.0616. 'learning rate': 5.714285714285714e-05, 'epoch': 2.66}
train_runtime : 10215.1039, 'train_samples_per_second': 3.95, 'train_steps
```

## Result

15 **311356003** 



0.33333

8

9d



Your Best Entry!

Your submission scored 0.01333, which is not an improvement of your previous score. Keep trying!

1 311356003



0.36285

8

9d

## Conclusion

- 我認為這次的準確率只有33%,是因為這次使用的運算資源非常大量,為了要滿足硬體的Ram 所以會將cutoff 設定到128,但是她一條句字最長1548,會導致準確率下降。
- 另外依照learning rate的下降我認為模型還沒有充分收斂握,eppoch就到了 應該還可以再讓他繼續跑下去,進一步提高準確率。