# NTU ADL 2024 Fall HW3

Deadline: 2024/11/7 23:59:59

## Update Logs

- 1. Learning curve 請畫在public testing data上面
- 2. adapter\_checkpoint

```
adapter_config.json
```

<del>adapter\_model.bin</del> ->adapter\_model.safetensors

3. DEADLINE: 10/31 23:59:59 11/7 23:59:59

## Links

• Homework 3 files (<u>link</u>)

# Task Description

## Instruction Tuning (Classical Chinese)

• Example 1

#### **Instruction:**

#### 翻譯成文言文:

雅裏惱怒地說:從前在福山田獵時,你誣陷獵官,現在又說這種話。

答案:

#### Output:

雅裏怒曰: 昔畋於福山, 卿誣獵官, 今復有此言。

• Example 2

#### **Instruction:**

議雖不從, 天下咸重其言。

翻譯成白話文:

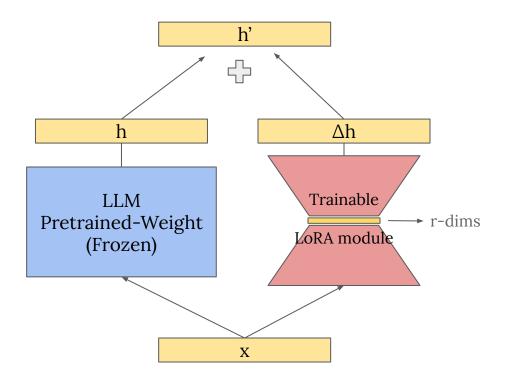
#### Output:

他的建議雖然不被采納,但天下都很敬重他的話。

#### Model- Gemma 2

- Model Name: zake7749/gemma-2-2b-it-chinese-kyara-dpo
- Gemma is a open source LLM trained by Google
- The model we used is the one fine-tuned on traditional chinese data
- For more information about the model: <u>link</u>

## Low Rank Adaptation (LoRA)



## QLoRA Fine-tuning

- QLoRA reduces the memory usage of LLM fine-tuning by employing 4-bit quantization to compress a LLM, while utilizing 16-bit float to perform computations
- Paper

## Experiments

- Dataset
  - Training (train.json): 10000
  - o Testing (Public) (public\_test.json): 250
  - Testing (Private) (private\_test.json): 250

- Evaluation:
  - Perplexity (ppl.py)

### Data Format

• json format

# Rules

### What You Can Do:

- LLM checkpoint:
  - o <u>zake7749/gemma-2-2b-it-chinese-kyara-dpo</u>
- Packages:
  - o torch==2.4.1, transformers==4.45.1, bitsandbytes==0.44.1, peft==0.13.0
  - Others: no limitation

#### What You Can NOT Do

- Use external training data
- Any means of cheating or plagiarism, including but not limited to:
  - Directly apply others' published / unpublished code..., including the codes written your classmates or public on the internet
  - Give/get trained model/predictions to/from others.
  - Give/get report answers or plots to/from others.
  - Publish your code before deadline.
- Violations may cause zero/negative score and punishment from school.

# Submission

## File Layout - Before downloading

• Zip your folder, which should be named as your student id (lower-cased) (ex. r11000000) and submit the .zip to NTU Cool.

### download.sh

- **download.sh** should download or create a folder called **adapter\_checkpoint** which contains peft configuration file (adapter\_confi.json) and weights (adapter\_model.bin).
  - !!DO NOT download original model checkpoint!!
- Do not modify your file after deadline, or it will be seen as cheating.
- Keep the URLs in download.sh valid for at least 3 weeks after deadline.
- You can download at most 4G, and download.sh should finish within 1 hour. (At csie dept with maximum 10MB/s bandwidth)
- Do not do things more than downloading. Otherwise, your **download.sh** may be killed.
- Do not pip install ANYTHING in your **download.sh**, you are not allowed to modify the testing environment
- We will execute **download.sh** before predicting scripts.

## File Layout - After downloading

• After we run download.sh, the **adapter\_config.json** and **adapter\_model.bin** should be **in adapter\_checkpoint**.

## utils.py

- Two functions are required:
  - o def get\_prompt(instruction: str) -> str:
    - return the prompt to input into the LLM
  - def get\_bnb\_config() -> BitsAndBytesConfig:
    - return your bnb configuration
- TAs will include "from utils import get\_prompt, get\_bnb\_config" when testing.
- Make sure that your function's input and output fields are correct and can be used accurately

#### README.md

- README.md should contain step-by-step instructions on how to setup your environments and how to train your model with your codes/scripts.
- You will get a **-2** penalty if you have no or empty README.md.
- If necessary, you will be required to reproduce your results based on the README.md.
- If you cannot reproduce your result, you may lose points.

#### run.sh

- **run.sh** should perform text generation using your trained models and output predictions on testing file (.json)
- arguments
  - \${1}: path to the model checkpoint folder
  - \${2}: path to the adapter\_checkpoint downloaded **under your folder**
  - \${3}: path to the input file (.json)
  - \${4}: path to the output file (.json)
- TA will predict testing data as follow:
  - 1. bash ./download.sh
  - 2. bash ./run.sh /path/to/model-folder /path/to/adapter\_checkpoint \ /path/to/input.json /path/to/output.json
- run.sh should finish within 2 hours. (See <u>environment details</u>)

### run.sh - cont.

Command:

```
bash run.sh \
/path/to/`zake7749/gemma-2-2b-it-chinese-kyara-dpo` \
/path/to/adapter_checkpoint/under/your/folder \
/path/to/input \
/path/to/output
```

o example:

```
bash run.sh \
/home/\`zake7749/gemma-2-2b-it-chinese-kyara-dpo`
/home/hw3/r11922000/adapter_checkpoint \
/home/data/public_test.json \
/home/output/r11922000_output.json
```

## prediction.json

- Testing set (Private)
- DO NOT include any special tokens (<s>, </s>, ...) and your prompt in your output

### **Execution Environment**

- We will run the testing codes on the computer with
  - o Ubuntu 20.04
  - o 32GB RAM, RTX 3070 8GB VRAM, and 20GB disk space available
- Python3.10
- Packages
  - o torch==2.4.1, transformers==4.45.1, bitsandbytes==0.44.1, peft==0.13.0

## Grading

- Model Performance (5%)
  - Public baseline: ppl = 17.500↓ (2%)
  - o Private baseline: ppl = 24.000↓ (2%)
  - o prediction.json: Human evaluation (1%)
- Report (15% + 2%)
- Format
  - You may lose (some or all) of your model performance score if your script is at wrong location, causes any error, etc.

## Grading - ppl.py

- TA will use our own ppl.py (which is the same as we published), so you don't have to upload ppl.py
- Command:

```
python3 ppl.py \
    --base_model_path /path/to/`zake7749/gemma-2-2b-it-chinese-kyara-dpo` \
    --peft_path /path/to/adapter_checkpoint/under/your/folder \
    --test_data_path /path/to/input/data
```

o example:

```
python3 ppl.py \
    --base_model_path /home/`zake7749/gemma-2-2b-it-chinese-kyara-dpo` \
    --peft_path /home/hw3/r11922000/adapter_checkpoiint \
    --test_data_path /home/data/public_test.json
```

### Late Submission

- Late submission penalties:
  - 0 day < late submission ≤ 1 day: original score \* 0.95
  - 1 day < late submission ≤ 3 day: original score \* 0.90
  - $3 \text{ day} < \text{late submission} \le 4 \text{ day: original score} * 0.75$
  - 4 day < late submission ≤ 5 day: original score \* 0.50</li>
  - o 5 day < late submission ≤ 6 day: original score \* 0.25
  - o 6 day < late submission: original score \* 0.00
- Late submission is determined by the last submission.
  - Update your submission after deadline implies that you will get penalty.

# Report

# Q1: LLM Tuning

#### • Describe:

- How much training data did you use? (2%)
- How did you tune your model? (2%)
- What hyper-parameters did you use? (2%)

#### • Show your performance:

- What is the final performance of your model on the public testing set? (2%)
- Plot the learning curve on the public testing set (2%)

## Q2: LLM Inference Strategies

- Zero-Shot
  - What is your setting? How did you design your prompt? (1%)
- Few-Shot (In-context Learning)
  - What is your setting? How did you design your prompt? (1%)
  - How many in-context examples are utilized? How you select them? (1%)
- Comparison:
  - What's the difference between the results of zero-shot, few-shot, and LoRA? (2%)

#### Note:

Please conduct zero-shot and few-shot experiments on Orginal Model that has not been fine-tuned with QLoRA

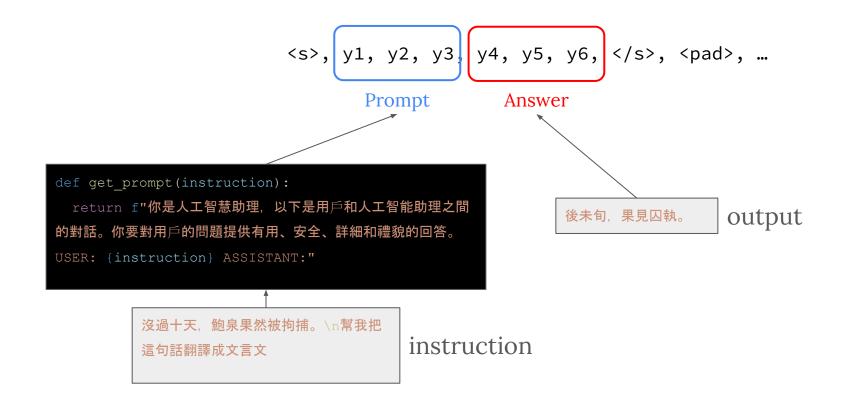
## Q3: Bonus: Try Llama3-Taiwan (8B) (2%)

- Llama-3-8b trained by traditional Chinese data
- Tune this model on the classical chinese data
- Describe your experimental settings and compare the results to those obtained from your original methods

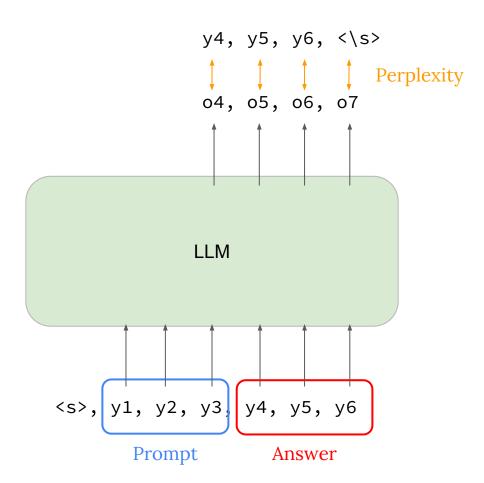
# Guides

# Instruction Tuning y4, y5, y6, <\s> 04, 05, 06, 07 LLM <s>, y1, y2, y3, y4, y5, y6, </s>, <pad>, ... Prompt Answer

## Instruction Tuning



## Perplexity



## Quantization (4 bit)

```
from utils import get bnb config
from peft import prepare model for kbit training
bnb config = get bnb config() # return a BitsAndBytesConfig
model = AutoModelForCausalLM.from pretrained(
  model name or path,
  quantization config=bnb config
if training:
  model = prepare model for kbit training(model)
```

#### Ref:

- 1. <a href="https://huggingface.co/blog/4bit-transformers-bitsandbytes">https://huggingface.co/blog/4bit-transformers-bitsandbytes</a>
- https://huggingface.co/docs/ transformers/main\_classes/qu antization#advanced-use-cases
- 3. https://github.com/artidoro/glora

### Reference

- https://huggingface.co/docs/peft/index
- https://github.com/huggingface/peft
- 3. <a href="https://huggingface.co/blog/4bit-transformers-bitsandbytes">https://huggingface.co/blog/4bit-transformers-bitsandbytes</a>
- 4. <a href="https://huggingface.co/docs/transformers/main\_classes/quantization">https://huggingface.co/docs/transformers/main\_classes/quantization</a>
  #advanced-use-cases
- 5. <a href="https://github.com/artidoro/qlora">https://github.com/artidoro/qlora</a>
- 6. <a href="https://github.com/huggingface/trl">https://github.com/huggingface/trl</a>

## Any questions

- NTU COOL discussion
- Email:
  - o <u>adl-ta@csie.ntu.edu.tw</u>
- TA hours
  - Tue. 11:00 ~ 12:00 google meet
  - o Thur. 14:00~15:00@德田 524