# **VIP Fall 2024**

## **Fine-tuning LLMs for Materials**

#### **Tutorial III**

In the first tutorial, we explored how to convert crystal structures into text representations. In the second, we set up an environment for torchtune. In this tutorial, we'll build on that by creating a custom dataset for torchtune, using the Cartesian representation from Tutorial I.

### 1. Setting Up

In the attache zip file, you will find the following file structure:

```
configs/
    - llama2-7b.yaml
    - llama2-7b-inference.yaml
data/mp_20/
    - raw_train/
    - raw_val/
    - raw_test/
llm4materials/
    - datasets/
            torchtune_dataset.py
    - encoders
            - cartesian.py
    - inference
            - infer.py
outputs/
setup.py
```

Intuitively, the <code>configs/</code> folder holds our <code>yaml/</code> config files for finetuning an LLM, the <code>data/</code> folder contains the raw data that we need, the <code>outputs/</code> folder is where we save training output files (as specified in the <code>yaml</code> file).

The  $_{\text{setup.py}}$  script allows us to install a package named  $_{\text{llm4materials}}$ . In this way, we will be able to import the custom dataset implemented at  $_{\text{llm4materials}}$  directly.

First, let's

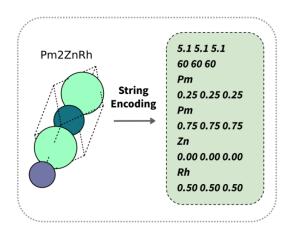
- 1. Connect to ICE
- 2. Request a GPU
- 3. Upload the zip file to ICE and unzip it
- 4. Load anaconda3 and activate your torchtune environment
- 5. With your environment activated, install ase if it is not available: pip install ase
- 6. In the same directory as setup.py, install llm4materials: pip install -e.

Now, you will be able to import definitions from llm4materials.

#### **Creating Custom Dataset**

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Take a look at the file <a href="lim4materials/encoders/cartesian.py">11m4materials/encoders/cartesian.py</a>. It contains a partial implementation of the Cartesian encoder for crystal structures. Specifically, the Cartesian encoder has an <a href="mailto:encoder">encode</a> method that takes in <a href="mailto:ase.atoms.Atoms">ase.atoms.Atoms</a> objects and returns their corresponding text representations, similar to what we did in Tutorial I (see Figure below).



TO-DO: complete the Cartesian encoder class.

Next, take a look at the file llm4materials/datasets/torchtune\_dataset.py. It contains a class named TextCompletionDataset and a creator function text\_completion\_dataset.

In the TextCompletionDataset class, there is a class method named \_load\_data(self, source), which gathers all the cif files in the given source path and returns a list of ase.atoms.Atoms objects.

TO-DO: complete the TextCompletionDataset class.

The \_prepare\_sample(self, sample) method has been completed for you. In essence, we are transforming a single Atoms object (sample) into textual representation prepended with a prompt header. An example of the full text prompt is:

```
Below is a description of a bulk material.

Generate a description of the lengths and angles of the lattice vectors and then the element type and coordinates for each atom within the lattice:
5.1 5.1 5.1
60 60 60
Pm
0.25 0.25 0.25
Pm
0.75 0.75 0.75
Zn
0.00 0.00 0.00
Rh
0.50 0.50 0.50
```

The above text is then tokenized by the LLM's tokenizer to serve as input to the LLM. A duplicate copy of the tokenized text is created to serve as the labels/targets.

#### **Training**

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Once you have completed the above classes, you are ready to fine-tune the LLM on crystal structures! Go to configs/llama2-7b.yam1 and change the necessary fields (e.g., paths to your downloaded weights). To fine-tune, just do

```
tune run lora_finetune_single_device \
--config configs/llama2-7b.yaml
```

#### Inference

If you have successfully trained your model for enough epochs, you will now be able to generate new string representations of structures. To do so, take a look at the inference yam1 config file: configs/llama2-7b-inference.yam1. Make sure that the checkpoint\_files are pointing to the trained checkpoints. For example, if you trained your model for 10 epochs, you should set it to the following:

```
checkpoint_files: [
   hf_model_0001_9.pt,
   hf_model_0002_9.pt,
]
```

The pt files can be found in the output folder you set in your training yam1 config file. To generate, do

```
tune run llm4materials/inference/infer.py \
--config configs/llama2-7b-inference.yaml
```

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