# EE508 Project - Phase 3

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# **Summary of files/changes added**

Filename	Description		
finetuning.py	Main training function		
alpaca_dataset.py	To custom load the alpaca dataset		
alpaca_data.json	52K-instruction Alpaca dataset		
project/model/lora.py	Enable support for LoRA. Also integrates Gradient Checkpoint support to run both of them together		
project/model/grad_ckpt.py	Enable support for gradient checkpoint only		
project/utils.py	Wandb support for memory & runtime profiling		
requirements.txt	Updated to install wandb module		

#### Additional note:

<u>finetuning.py</u>: Contains the main training loop with the following additional supports:

- a. Arg parser to support common line interface (CLI) to select between different combinations of fine tuning techniques
- b. Switches for gradient accumulation & mixed precision
- c. Wandb support to log peak memory & avg run time/step after training, memory for activation, parameters, gradients & optimizer state

d. Refer the help function in <u>finetuning.py</u> to run the script

```
(ee508_llm) [asmitamo@e21-06 final_llm]$ python finetuning.py -h
Initializing...
usage: finetuning.py [-h] --tokenizer_model_path TOKENIZER_MODEL_PATH --checkpoint_path CHECKPOINT_PATH
                      --use_lora USE_LORA] [--use_grad_acc USE_GRAD_ACC] [--use_grad_ckpt USE_GRAD_CKPT]
                     [--use_mixed_prec USE_MIXED_PREC]
Train or run a LLaMA model with optional Fine Tuning Technique.
options:
                        show this help message and exit
  -h, --help
  --tokenizer_model_path TOKENIZER_MODEL_PATH
                       Path to the tokenizer model. Eg: /<local path>/.llama/checkpoints/Llama3.2-1B/tokenizer.model
  --checkpoint_path CHECKPOINT_PATH
                        Path to the model checkpoint. Eg:
                        /<local_path>/.llama/checkpoints/Llama3.2-1B/consolidated.00.pth
  --use_lora USE_LORA
                       Enable (1) or Disable (0) LoRA
  --use grad acc USE GRAD ACC
                        Enable (1) or Disable (0) Gradient Accumulation
  --use_grad_ckpt USE_GRAD_CKPT
                        Enable (1) or Disable (0) Gradient Checkpointing
  --use_mixed_prec USE_MIXED_PREC
                        Enable (1) or Disable (0) Mixed Precision
```

### Eg. Usage:

```
(ee508_llm) [asmitamo@e21-06 final_llm]$ python finetuning.py --tokenizer_model_path /home1/asmitamo/.llama/checkpoints/Llama3.2-1B/tokenizer.model --checkpoint_path /home1/asmitamo/.llama/checkpoints/Llama3.2-1B/consolidated.00.pth --use_lora 1
```

Memory, Runtime & Loss profiling for all the different types of combinations of finetuning can be found <u>here</u>. Below figures illustrate a few of them. All the runs are done by setting the following: *learning rate* = 1e-5, *batch size* = 1, *and gradient accumulation step* = 8. For the LoRA configuration, r = 16, alpha = 32, and dropout rate = 0.05

#### **Average Training Loss/Epoch**

#### 1. Vanilla Llama

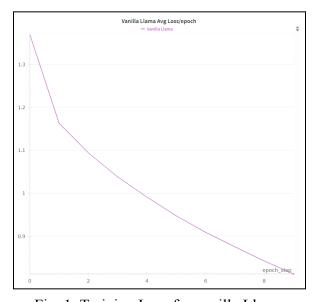


Fig. 1. Training Loss for vanilla Llama

### 2. Gradient Accumulation

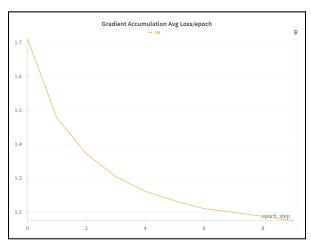


Fig 2. Training loss for Gradient Accumulation

## 3. Mixed Precision Training

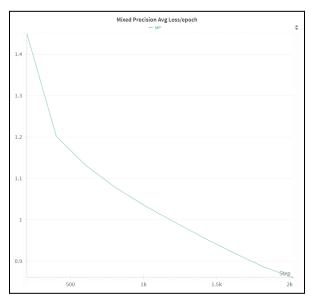


Fig. 3. Training Loss for Mixed Precision Training

### 4. Gradient Checkpointing

Reasons for choice of layers: Out of a total of 16 layers, we chose to checkpoint the last 4 layers since the initial layers capture most of the information about the input data. As we advance to the higher layers, the number of parameters stored is higher. Also we chose a smaller number of layers to avoid an excessive increase in compute time.

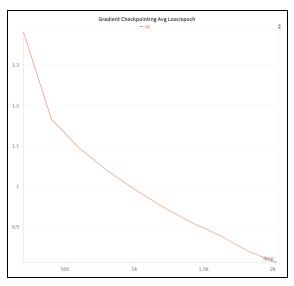


Fig. 4. Training Loss for Gradient Checkpointing

### 5. LoRA Implementation

```
(ee508 llm) [asmitamo@e21-08 final llm]$ python finetuning.py --tokenizer_model_path /home1/asmitamo/.llama/checkpoints/Llama3.2-1B/tokenizer.model --checkpoint_path /home1/asmitamo/.llama/checkpoints/Llama3.2-1B/consolidated.00.pth --use_lora 1
Initializing...

Model Finetuning: LoRA
Calling AlpacaDataModule...
Reloaded tiktoken model from /home1/asmitamo/.llama/checkpoints/Llama3.2-1B/tokenizer.model
#words: 128256 - BOS ID: 128000 - EOS ID: 128001
Loading data from alpaca_data.json...
Formatting prompts and responses...
Tokenizing examples...
Dataset fetched for training...
Length of train dataset: 200
Length of train loader: 200
Loading model ...
Mapping model to device...
LORA Total parameters: 1500186624, Trainable parameters: 1703936
Percentage of trainable parameters: 0.11%
```

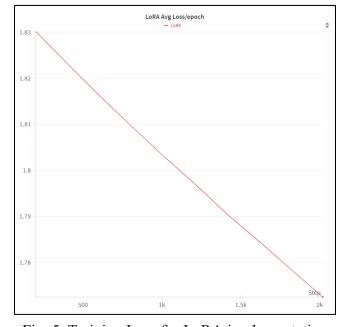


Fig. 5. Training Loss for LoRA implementation

### **Outputs**

## Inference on sample prompts before fine tuning

### Inferencing after fine-tuning

```
| Sababed@822-15-j5 python inference,py | Sababed@822-15-j5 python inference,py | Sababed@822-15-j5 python inference,python | Sababed@822-15-j5 python | Sababed@82-15-j5 pytho
```

**Table 2: Fine-Tuning System Performance Analysis** 

		Grad. Accumulation	Grad. Checkpoint	Mixed Precision	LoRA
Memory	Parameter	<b>↑</b>	_	<b>↑</b>	<b>↓</b>
	Activation	_	<b>↓</b>	<b>↓</b>	_
	Gradient	<b>↑</b>	<b>↓</b>	_	$\downarrow$
	Optimizer State	<b>↑</b>	<b>\</b>	_	_
Computation		<b>\</b>	1	1	<b>↓</b>

**Table 3: Fine-Tuning System Performance Benchmark** 

GC	OFF				ON			
MP	OFF		ON		OFF		ON	
LoRA	OFF	ON	OFF	ON	OFF	ON	OFF	ON
Peak Mem (MB)	12260	7741	12098	9946	12148	7694	12099	9537
Runtime (s)	354.37	221.69	487.99	254.43	364.80	232.87	491.73	268.77