# Week 5 - Technical Assignment Fine tuning using LoRA and PEFT

# **Objective**

This assignment provides **practical experience in running, debugging, fine-tuning, and testing an LLM** inside a Jupyter Notebook. The key tasks include:

- Running a fine-tuning notebook for an instruction-following LLM.
- Fixing potential errors that may occur during execution.
- Training and evaluating the model within the notebook.
- Saving and uploading the model to Hugging Face.
- Testing inside the notebook
- Documenting findings & submitting results via notebook.

# Instructions

# Copy-Paste Assignment - Read Carefully

- This is a Copy-Paste Assignment, meaning you will clone and run the provided notebook with small modifications.
- You may face errors when running the notebook.
- Your task is to fix these errors, document the fixes, and explain the process.
- You may run the notebook in Google Colab, Jupyter, or Kaggle.
- Kaggle is good.

# **How to Complete the Assignment**

- 1. Clone the Jupyter Notebook from the following link:
  - Kaggle Notebook
- 2. Run each step, and fix errors, if you face them.
- 3. Clone hugging face small language models, like LLaMA 1.1B (Tiny Model), or any small model for training. If you do not get a success you can use the same model and need to explain the reason.
- 4. You can optimized parameters to get better results.
- 5. **Train the fine-tuned model** with the following configuration
  - $a. max_steps = 500$
  - b. logging\_steps = 50
  - c.  $eval\_steps = 50$

- 6. Save and **upload the trained model to Hugging Face**. Share a screenshot of the hugging face model
- 7. **Test** all label cases.
- 8. Capture test results and add them to your Jupyter Notebook report.
- 9. Submit your work (Notebook + Code).

#### **Submission Guidelines**

- Code Submission: Push your final notebook to GitHub.
- Notebook Submission: Include error fixes, test results, and model evaluation.

# **15-Point Exercise Steps**

### 1. Install Required Libraries

o Install transformers, datasets, peft, bitsandbytes, etc.

### 2. Load the Dataset from hugging face()

Import and inspect the provided dataset for fine-tuning.

### 3. Create Bitsandbytes Configuration

• Set up quantization with bitsandbytes for efficient training.

### 4. Load the Pre-Trained Model

Load LLaMA 1.1B or another small language model.

#### 5. Tokenization

Apply appropriate tokenization for the dataset.

#### 6. Test the Model with Zero-Shot Inference

Run a few samples to see the base model's performance before fine-tuning.

#### 7. Pre-process the Dataset

Clean, format, and prepare the dataset for training.

### 8. Prepare the Model for QLoRA

Enable gradient checkpointing and quantization preparation.

#### 9. Set Up PEFT for Fine-Tuning

Configure LoRA parameters and apply to the model.

### 10. Train PEFT Adapter

• Fine-tune the model using the PEFT configuration.

### 11. Evaluate the Model Qualitatively

Perform manual evaluation to assess output quality.

#### 12. Evaluate the Model Quantitatively (ROUGE Metric)

• Use ROUGE or other metrics for automated performance evaluation.

#### 13. Save and Upload the Model to Hugging Face

Push the trained model to your Hugging Face account.

### 14. Capture and Document Results

o Include screenshots of the uploaded model, sample outputs, and analysis.

# 15. Submit the Assignment

 Upload the notebook to GitHub with a README, including error fixes, evaluation, and model details.

# **Expected Deliverables**

**GitHub Repository** with:

- Code (Notebook)
- Errors encountered & fixes if applied
- Snapshots of hugging face model and link to publicly available.