

Fine-Tuning Setup Report

1. Data

- **Sources:**

- RACE (Reading Comprehension Dataset, MCQ format)
- SQuAD (Stanford Question Answering Dataset, short-answer format)

- **Preparation:**

- Sampled 25,000 examples each from RACE and SQuAD for diversity and balance.
- Converted all samples to a unified JSONL format with fields: `context`, `type`, `instruction`, `question`, `options` (for MCQ), and `answer`.
- Combined, shuffled, and saved as `train_50k.jsonl` for training.

2. Method

- **Base Model:**

- Used `distilgpt2` for its lightweight architecture and efficiency.

- **Parameter-Efficient Fine-Tuning:**

- Applied LoRA (Low-Rank Adaptation) via the PEFT library for memory- and storage-efficient adaptation.
- Targeted GPT-2's attention and projection layers (`c_attn`, `c_proj`).

- **Prompt Construction:**

- For each example, constructed a prompt including instruction, context, question, options (if MCQ), and answer.
- Example prompt:

For MCQ:

```
[Instruction]
Context: [context]
Question: [question]
Options: [option1 | option2 | ...]
Answer: [answer]
```

For Short Answer:

```
[Instruction]
Context: [context]
Question: [question]
Answer: [answer]
```

- **Tokenization:**

- Used the base model's tokenizer with truncation and padding to a max length of 128 tokens.

- **Training Details:**

- 3.5 epochs, batch size 2, AdamW optimizer, learning rate $2e-4$, weight decay 0.001.
- Saved checkpoints every 50 steps, logged every 10 steps.

3. Results

- **Model Output:**

- The fine-tuned model can generate both MCQ and short-answer questions from a given context.
- Output format is consistent with the training prompt, supporting robust downstream parsing.

- **Qualitative Observations:**

- The model produces mostly relevant and fluent questions.
- MCQs generally have plausible options and correct answers, though some repetition and generic outputs may occur.
- Short answers are usually correct but sometimes lack detail or diversity.

- **Evaluation:**

- The detailed evaluation report, including human ratings and analysis, is provided separately.

Summary

This fine-tuning setup enabled efficient adaptation of a lightweight language model for the dual task of MCQ and short-answer question generation, using a balanced and diverse dataset and parameter-efficient training.