


```
import json
from datetime import datetime
import pandas as pd
```

```

Requirement already satisfied: transformers in /usr/local/lib/python3.11/dist-packages (4.51.3)
Requirement already satisfied: filelock in /usr/local/lib/python3.11/dist-packages (from transformers) (3.18.0)
Requirement already satisfied: huggingface-hub<1.0,>=0.30.0 in /usr/local/lib/python3.11/dist-packages (from transformers) (0.30.2)
Requirement already satisfied: numpy>=1.17 in /usr/local/lib/python3.11/dist-packages (from transformers) (2.0.2)
Requirement already satisfied: packaging>=20.0 in /usr/local/lib/python3.11/dist-packages (from transformers) (24.2)
Requirement already satisfied: pyyaml>=5.1 in /usr/local/lib/python3.11/dist-packages (from transformers) (6.0.2)
Requirement already satisfied: regex<=2019.12.17 in /usr/local/lib/python3.11/dist-packages (from transformers) (2024.11.6)
Requirement already satisfied: requests in /usr/local/lib/python3.11/dist-packages (from transformers) (2.32.3)
Requirement already satisfied: tokenizers<0.22,>=0.21 in /usr/local/lib/python3.11/dist-packages (from transformers) (0.21.1)
Requirement already satisfied: safetensors>=0.4.3 in /usr/local/lib/python3.11/dist-packages (from transformers) (0.5.3)
Requirement already satisfied: tqdm>=4.27 in /usr/local/lib/python3.11/dist-packages (from transformers) (4.67.1)
Requirement already satisfied: fsspec>=2023.5.0 in /usr/local/lib/python3.11/dist-packages (from huggingface-hub<1.0,>=0.30.0->transformers) (2024.10.0)
Requirement already satisfied: typing-extensions>=3.7.4.3 in /usr/local/lib/python3.11/dist-packages (from huggingface-hub<1.0,>=0.30.0->transformers) (4.12.2)
Requirement already satisfied: charset-normalizer<4,>=2 in /usr/local/lib/python3.11/dist-packages (from requests->transformers) (3.4.1)
Requirement already satisfied: idna<4,>=2.5 in /usr/local/lib/python3.11/dist-packages (from requests->transformers) (3.10)
Requirement already satisfied: urllib3<3,>=1.21.1 in /usr/local/lib/python3.11/dist-packages (from requests->transformers) (2.3.0)
Requirement already satisfied: certifi>=2017.4.17 in /usr/local/lib/python3.11/dist-packages (from requests->transformers) (2025.1.31)

```


 文部科学省
 教育部
 文化庁
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 国際交流基金
 国際協力機構
 国際労働機関
 国際連合
 国際連合教育科学文化機関
 国際連合児童基金
 国際連合世界保健機関
 国際連合世界労働機関
 国際連合世界貿易機関
 国際連合世界銀行
 国際連合世界保健機関
 国際連合世界労働機関
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 国際連合世界保健機関
 国際連合世界労働機関
 国際連合世界貿易機関
 国際連合世界銀行

```
git config --global credential.helper store
```

Read <https://git-scm.com/book/en/v2/Git-Tools-Credential-Storage> for more details.

```
import json
import re
import pandas as pd

# Read the CSV file into a Pandas DataFrame
df = pd.read_csv("/content/drive/MyDrive/test.csv")

def clean_text(text):
    # Convert the input to a string if it's not already
    text = str(text) # This will handle integers and other data types

    # Remove HTML tags
    text = re.sub(r"<.*?>", "", text)

    # Remove URLs
    text = re.sub(r"https?:\/\/\S+|www.\S+", "", text)

    # Remove special characters (keep only alphanumeric and whitespace)
    text = re.sub(r"[^\w\s]", "", text)
```

```

# Handle missing values (replace with empty string)
text = text if text else ""

return text

def preprocess_text(text):
    return clean_text(text)

# Convert to JSONL format with preprocessing:
with open("/content/drive/MyDrive/fine_tuning_data.jsonl", "w") as f:
    for index, row in df.iterrows():
        # Assuming your CSV has columns named "instruction" and "output"
        instruction = row["text"]
        output = row["label"]

        # Check for missing values
        if pd.isnull(instruction) or pd.isnull(output):
            continue # Skip this row if either is missing

        json_object = {
            "messages": [
                {"role": "system", "content": "You are a helpful and harmless AI assistant."},
                {"role": "user", "content": preprocess_text(instruction)},
                {"role": "assistant", "content": preprocess_text(output)},
            ]
        }
        f.write(json.dumps(json_object) + "\n")

```

```

from transformers import AutoModelForCausalLM, AutoTokenizer

```

```

model_id = "openai-community/gpt2" # Repo ID

```

```


tokenizer = AutoTokenizer.from_pretrained(model_id)
model = AutoModelForCausalLM.from_pretrained(model_id)

```

```

# ... (Use the model and tokenizer for inference or fine-tuning) ...

```

 /usr/local/lib/python3.11/dist-packages/huggingface_hub/utils/_auth.py:94: UserWarning:
 The secret `HF_TOKEN` does not exist in your Colab secrets.
 To authenticate with the Hugging Face Hub, create a token in your settings tab (<https://huggingface.co/settings/tokens>), set it as secret.
 You will be able to reuse this secret in all of your notebooks.
 Please note that authentication is recommended but still optional to access public models or datasets.
 warnings.warn(

```

import json

```

```

def extract_text_from_jsonl(file_path):
    text_data = []
    with open("/content/drive/MyDrive/fine_tuning_data.jsonl", "r") as f:
        for line in f:
            data = json.loads(line)
            text_data.append(data["messages"][1]["content"] + " " + data["messages"][2]["content"]) # Concatenate user and assistant messages
    return text_data

```

```

text_data = extract_text_from_jsonl("/content/drive/MyDrive/fine_tuning_data.jsonl")

```

```

!pip install -q bitsandbytes accelerate transformers peft
!pip install -U bitsandbytes
!pip install datasets transformers

```

 [Show hidden output](#)

```

from transformers import TrainingArguments, Trainer, AutoModelForCausalLM
from transformers import AutoTokenizer
from datasets import load_dataset, Dataset
import torch
from peft import LoraConfig, get_peft_model

```

```

# Load the tokenizer
tokenizer = AutoTokenizer.from_pretrained("openai-community/gpt2")
tokenizer.pad_token = tokenizer.eos_token # Set pad token

```

```

# Load the base model with bitsandbytes for quantization
model = AutoModelForCausalLM.from_pretrained(
    "openai-community/gpt2",
    load_in_8bit=True, # Enable 8-bit quantization
    device_map="auto", # Automatically assign to the appropriate device
)

# Define the LoRA configuration
lora_config = LoraConfig(
    r=8, # Rank of the LoRA matrices
    lora_alpha=16, # Scaling factor for the LoRA matrices
    lora_dropout=0.05, # Dropout probability for the LoRA layers
    bias="none", # Bias type for the LoRA layers
    task_type="CAUSAL_LM", # Task type for fine-tuning
    fan_in_fan_out=True # Explicitly setting fan_in_fan_out to True to suppress the warning
)

# Apply LoRA to the base model
model = get_peft_model(model, lora_config)
model.print_trainable_parameters() # Print the number of trainable parameters

# Define training arguments
training_args = TrainingArguments(
    output_dir="./results",
    per_device_train_batch_size=4, # Reduce batch size if needed
    per_device_eval_batch_size=4, # Reduce batch size if needed
    num_train_epochs=3,
    logging_dir="./logs",
    fp16=True, # Enable mixed precision training if supported by your hardware
    gradient_accumulation_steps=4, # Increase gradient accumulation steps if needed
)

# Load the text data you extracted
# Assuming 'text_data' is a list of strings from your 'extract_text_from_jsonl' function
# Create a Hugging Face Dataset from your text data
train_dataset = Dataset.from_dict({"text": text_data})

def tokenize_function(examples):
    return tokenizer(examples["text"], padding="max_length", truncation=True)

# Tokenize the dataset
tokenized_datasets = train_dataset.map(tokenize_function, batched=True)

# Access the 'train' split of the dataset
train_dataset = tokenized_datasets

# Add the labels to the inputs
def prepare_inputs_for_training(examples):
    examples["labels"] = [[-100] + x[:-1] for x in examples["input_ids"]]
    return examples

train_dataset = train_dataset.map(
    prepare_inputs_for_training,
    batched=True,
    remove_columns=["text"],
)

# Create the Trainer and start training
trainer = Trainer(
    model=model,
    args=training_args,
    train_dataset=train_dataset,
)

trainer.train()

# Save the model
trainer.save_model("./fine_tuned_gpt2")

```

→ The `load_in_4bit` and `load_in_8bit` arguments are deprecated and will be removed in the future versions. Please, pass a `BitsAndBytesC` trainable params: 294,912 || all params: 124,734,720 || trainable%: 0.2364

Map: 100% 2000/2000 [00:00<00:00, 2288.21 examples/s]

Map: 100% 2000/2000 [00:01<00:00, 1425.13 examples/s]

No label_names provided for model class `PeftModelForCausalLM`. Since `PeftModel` hides base models input arguments, if label_names is not provided, the model will use the default label_names. UserWarning: MatMul8bitLt: inputs will be cast from torch.float32 to float16 during quantization)

[353/375 14:39 < 00:55, 0.40 it/s, Epoch 2.82/3]

Step Training Loss

[375/375 15:37, Epoch 3/3]

Step Training Loss

```
import pandas as pd
```

```
df_validation = pd.read_csv("/content/drive/MyDrive/validation.csv")
```

```
# Assuming your CSV has columns named "text" (for prompts) and "label" (for expected outputs)
```

```
prompts = df_validation["text"].tolist()
```

```
expected_outputs = df_validation["label"].tolist()
```

```
from transformers import AutoModelForCausalLM, AutoTokenizer
```

```
model_id = "/content/fine_tuned_gpt2"
```

```
tokenizer = AutoTokenizer.from_pretrained("openai-community/gpt2") # Load the tokenizer using the original model ID
```

```
model = AutoModelForCausalLM.from_pretrained(model_id)
```

```
# Set the padding token after loading the tokenizer
```

```
tokenizer.pad_token = tokenizer.eos_token
```

```
inputs = tokenizer(prompts, return_tensors="pt", padding=True, truncation=True)
```

```
from transformers import pipeline
```

```
generator = pipeline("text-generation", model=model, tokenizer=tokenizer)
```

```
generated_outputs = []
```

```
for prompt in prompts:
```

```
    generated_output = generator(prompt, max_length=100, num_return_sequences=1)[0]['generated_text'] # Adjust parameters as needed
```

```
    generated_outputs.append(generated_output)
```

→ Device set to use cuda:0

Truncation was not explicitly activated but `max_length` is provided a specific value, please use `truncation=True` to explicitly truncate inputs

```
!pip install nltk # Install NLTK if you haven't already
```

```
import nltk
```

```
from nltk.translate.bleu_score import sentence_bleu
```

```
bleu_scores = []
```

```
for generated, expected in zip(generated_outputs, expected_outputs):
```

```
    # Convert 'expected' to a string before splitting
```

```
    bleu_score = sentence_bleu([str(expected).split()], generated.split()) # Calculate BLEU score
```

```
    bleu_scores.append(bleu_score)
```

```
average_bleu = sum(bleu_scores) / len(bleu_scores)
```

```
print(f"Average BLEU score: {average_bleu}")
```

→ Requirement already satisfied: nltk in /usr/local/lib/python3.11/dist-packages (3.9.1)
Requirement already satisfied: click in /usr/local/lib/python3.11/dist-packages (from nltk) (8.1.8)
Requirement already satisfied: joblib in /usr/local/lib/python3.11/dist-packages (from nltk) (1.4.2)
Requirement already satisfied: regex<=2021.8.3 in /usr/local/lib/python3.11/dist-packages (from nltk) (2024.11.6)
Requirement already satisfied: tqdm in /usr/local/lib/python3.11/dist-packages (from nltk) (4.67.1)
/usr/local/lib/python3.11/dist-packages/nltk/translate/bleu_score.py:577: UserWarning:
The hypothesis contains 0 counts of 4-gram overlaps.
Therefore the BLEU score evaluates to 0, independently of
how many N-gram overlaps of lower order it contains.
Consider using lower n-gram order or use SmoothingFunction()
warnings.warn(_msg)
/usr/local/lib/python3.11/dist-packages/nltk/translate/bleu_score.py:577: UserWarning:
The hypothesis contains 0 counts of 2-gram overlaps.
Therefore the BLEU score evaluates to 0, independently of

```

how many N-gram overlaps of lower order it contains.
Consider using lower n-gram order or use SmoothingFunction()
warnings.warn(_msg)
/usr/local/lib/python3.11/dist-packages/nltk/translate/bleu_score.py:577: UserWarning:
The hypothesis contains 0 counts of 3-gram overlaps.
Therefore the BLEU score evaluates to 0, independently of
how many N-gram overlaps of lower order it contains.
Consider using lower n-gram order or use SmoothingFunction()
warnings.warn(_msg)
Average BLEU score: 0.9691038014873151

```

```

!pip install rouge-score
from rouge_score import rouge_scorer

```

```

scorer = rouge_scorer.RougeScorer(['rouge1', 'rouge2', 'rougeL'], use_stemmer=True)

```

```

# Calculate and print ROUGE scores for all generated outputs

```

```

for generated_output, expected_output in zip(generated_outputs, expected_outputs):
    scores = scorer.score(generated_output, str(expected_output)) # Convert expected_output to string
    print(f"Generated Output: {generated_output}")
    print(f"Expected Output: {expected_output}")
    print(f"ROUGE Scores: {scores}")
    print("-" * 20) # Print a separator between examples

```

```

Generated Output: i had been feeling fabulous and full of energy but easter weekend wiped me out and i havent been able to recover
Expected Output: i had been feeling fabulous and full of energy but easter weekend wiped me out and i havent been able to recover
ROUGE Scores: {'rouge1': Score(precision=1.0, recall=1.0, fmeasure=1.0), 'rouge2': Score(precision=1.0, recall=1.0, fmeasure=1.0), 'rougeL': Score(precision=1.0, recall=1.0, fmeasure=1.0)}
-----
Generated Output: i feel im supposed to hate dams amp all the control of nature that they represent but sometimes they really are the
Expected Output: i feel im supposed to hate dams amp all the control of nature that they represent but sometimes they really are the
ROUGE Scores: {'rouge1': Score(precision=1.0, recall=1.0, fmeasure=1.0), 'rouge2': Score(precision=1.0, recall=1.0, fmeasure=1.0), 'rougeL': Score(precision=1.0, recall=1.0, fmeasure=1.0)}
-----
Generated Output: i feel like i got to know her a bit and what i did get to know i really liked
Expected Output: i feel like i got to know her a bit and what i did get to know i really liked
ROUGE Scores: {'rouge1': Score(precision=1.0, recall=1.0, fmeasure=1.0), 'rouge2': Score(precision=1.0, recall=1.0, fmeasure=1.0), 'rougeL': Score(precision=1.0, recall=1.0, fmeasure=1.0)}
-----
Generated Output: im okay but feeling a little apprehensive as my dad has a minor operation today
Expected Output: im okay but feeling a little apprehensive as my dad has a minor operation today
ROUGE Scores: {'rouge1': Score(precision=1.0, recall=1.0, fmeasure=1.0), 'rouge2': Score(precision=1.0, recall=1.0, fmeasure=1.0), 'rougeL': Score(precision=1.0, recall=1.0, fmeasure=1.0)}
-----
Generated Output: i just feel too overwhelmed i can t see the forest for the trees as the saying goes
Expected Output: i just feel too overwhelmed i can t see the forest for the trees as the saying goes
ROUGE Scores: {'rouge1': Score(precision=1.0, recall=1.0, fmeasure=1.0), 'rouge2': Score(precision=1.0, recall=1.0, fmeasure=1.0), 'rougeL': Score(precision=1.0, recall=1.0, fmeasure=1.0)}
-----
Generated Output: i cant help but feel sentimental about the fact that we were drawn here
Expected Output: i cant help but feel sentimental about the fact that we were drawn here
ROUGE Scores: {'rouge1': Score(precision=1.0, recall=1.0, fmeasure=1.0), 'rouge2': Score(precision=1.0, recall=1.0, fmeasure=1.0), 'rougeL': Score(precision=1.0, recall=1.0, fmeasure=1.0)}
-----
Generated Output: i feel i should make is how surprised but entertained i was by the inclusion of so many popular culture and gaming
Expected Output: i feel i should make is how surprised but entertained i was by the inclusion of so many popular culture and gaming
ROUGE Scores: {'rouge1': Score(precision=1.0, recall=1.0, fmeasure=1.0), 'rouge2': Score(precision=1.0, recall=1.0, fmeasure=1.0), 'rougeL': Score(precision=1.0, recall=1.0, fmeasure=1.0)}
-----
Generated Output: i feel so tortured by it
Expected Output: i feel so tortured by it
ROUGE Scores: {'rouge1': Score(precision=1.0, recall=1.0, fmeasure=1.0), 'rouge2': Score(precision=1.0, recall=1.0, fmeasure=1.0), 'rougeL': Score(precision=1.0, recall=1.0, fmeasure=1.0)}
-----
Generated Output: i feel a bit rude leaving you hanging there from my last post with an almost done room and then radio silence
Expected Output: i feel a bit rude leaving you hanging there from my last post with an almost done room and then radio silence
ROUGE Scores: {'rouge1': Score(precision=1.0, recall=1.0, fmeasure=1.0), 'rouge2': Score(precision=1.0, recall=1.0, fmeasure=1.0), 'rougeL': Score(precision=1.0, recall=1.0, fmeasure=1.0)}
-----
Generated Output: im having ssa examination tomorrow in the morning im quite well prepared for the coming exam and somehow i feel num
Expected Output: im having ssa examination tomorrow in the morning im quite well prepared for the coming exam and somehow i feel numb
ROUGE Scores: {'rouge1': Score(precision=1.0, recall=1.0, fmeasure=1.0), 'rouge2': Score(precision=1.0, recall=1.0, fmeasure=1.0), 'rougeL': Score(precision=1.0, recall=1.0, fmeasure=1.0)}
-----
Generated Output: i constantly worry about their fight against nature as they push the limits of their inner bodies for the determina
Expected Output: i constantly worry about their fight against nature as they push the limits of their inner bodies for the determinat
ROUGE Scores: {'rouge1': Score(precision=1.0, recall=0.9666666666666667, fmeasure=0.983050847457627), 'rouge2': Score(precision=1.0, recall=1.0, fmeasure=1.0), 'rougeL': Score(precision=1.0, recall=1.0, fmeasure=1.0)}
-----
Generated Output: i feel its important to share this info for those that experience the same thing
Expected Output: i feel its important to share this info for those that experience the same thing
ROUGE Scores: {'rouge1': Score(precision=1.0, recall=1.0, fmeasure=1.0), 'rouge2': Score(precision=1.0, recall=1.0, fmeasure=1.0), 'rougeL': Score(precision=1.0, recall=1.0, fmeasure=1.0)}
-----
Generated Output: i truly feel that if you are passionate enough about something and stay true to yourself you will succeed
Expected Output: i truly feel that if you are passionate enough about something and stay true to yourself you will succeed
ROUGE Scores: {'rouge1': Score(precision=1.0, recall=1.0, fmeasure=1.0), 'rouge2': Score(precision=1.0, recall=1.0, fmeasure=1.0), 'rougeL': Score(precision=1.0, recall=1.0, fmeasure=1.0)}
-----
Generated Output: i feel like i just wanna buy any cute make up i see online or even the one
Expected Output: i feel like i just wanna buy any cute make up i see online or even the one
ROUGE Scores: {'rouge1': Score(precision=1.0, recall=1.0, fmeasure=1.0), 'rouge2': Score(precision=1.0, recall=1.0, fmeasure=1.0), 'rougeL': Score(precision=1.0, recall=1.0, fmeasure=1.0)}
-----

```

```
!pip install nltk
import nltk
nltk.download('wordnet') # Download WordNet for METEOR

from nltk.translate.meteor_score import meteor_score

meteor_score_value = meteor_score([expected_output.split()], generated_output.split()) # Replace with your actual outputs
print(meteor_score_value)
```

🔄 Requirement already satisfied: nltk in /usr/local/lib/python3.11/dist-packages (3.9.1)
Requirement already satisfied: click in /usr/local/lib/python3.11/dist-packages (from nltk) (8.1.8)
Requirement already satisfied: joblib in /usr/local/lib/python3.11/dist-packages (from nltk) (1.4.2)
Requirement already satisfied: regex<=2021.8.3 in /usr/local/lib/python3.11/dist-packages (from nltk) (2024.11.6)
Requirement already satisfied: tqdm in /usr/local/lib/python3.11/dist-packages (from nltk) (4.67.1)
0.9999142661179699
[nltk_data] Downloading package wordnet to /root/nltk_data...