Fine-tuned LLM Using GPT-3.5

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This project aimed to create a custom language model that simulates a European travel advisor. The model will take user preferences and interests as input and generate personalized travel recommendations, including suggested cities, activities, and travel tips.

1. Data Generation:

• Prompt Description: User concisely describes the model's desired behavior.

"You are a travel advisor specializing in European tours. Users will share preferences, such as destination type and interests within Europe. Your task is to generate personalized travel recommendations specific to European destinations, including suggested cities, activities, and travel tips for an unforgettable European experience."

• To train the model effectively, Icreated a diverse dataset of 50 examples using a GPT-4 model with a temperature of 0.4. Each example followed a specific format: users provide preferences, and the model generates personalized travel recommendations. This format ensures the model understands the expected task and learns to generate relevant responses

• Temperature: User sets a value (0-1) to balance creativity and precision in generated examples. A temperature of 0.4 sits between pure precision (0) and pure creativity (1). This means the generated examples will reflect both:

Accuracy: The recommendations will remain relevant and realistic for European travel scenarios.

Originality: The model will inject some variety and unexpected suggestions, preventing predictable and formulaic responses.

```
prompt = "You are a travel advisor specializing in European tours. \n

Users will share preferences, such as destination type and interests \n

within Europe. Your task is to generate personalized travel recommendations \n

specific to European destinations, including suggested cities, activities, and \n

travel tips for an unforgettable European experience."

temperature = .4

number_of_examples = 50
```

• Example Generation: GPT-4 iteratively generates diverse training examples, gradually increasing complexity.

```
if len(prev examples) > 0:
                                                                                 ↑ ↓ ⑤ 🗐 🌣 🗓 📋 :
    if len(prev_examples) > 8:
      prev_examples = random.sample(prev_examples, 8)
    for example in prev_examples:
     messages.append({
            "role": "assistant",
            "content": example
  response = client.chat.completions.create(
     model="gpt-4", # Specify the desired model
      messages=messages,
      temperature=temperature,
     max_tokens=1000,
 return response.choices[0].message.content
prev_examples = []
for i in range(number_of_examples):
   print(f'Generating example {i}')
   example = generate_example(prompt, prev_examples, temperature)
   prev_examples.append(example)
print(prev_examples)
```

 System Message: A concise prompt is created for inference, outlining the model's task.

```
def generate_system_message(prompt):
                                                                                     ↑ ↓ ⊝ ■ ‡ ♬ 📋 :
       response = client.chat.completions.create(
           model="gpt-4",
           messages=[
               "content": "You will be given a high-level description of the model we are training, \n
                you are not generating the system message for data generation -- you are generating the system \n
                message to use for inference. A good format to follow is `Given $INPUT_DATA, \n
                you will $WHAT THE MODEL SHOULD DO.`.\n\nMake it as concise as possible. Include nothing but \n
                the system prompt in your response.\n\nFor example, never write: `\"$SYSTEM_PROMPT_HERE\"`.\n\nIt
                 "content": prompt.strip(),
           temperature=temperature,
           max tokens=500,
       return response.choices[0].message.content
   system_message = generate_system_message(prompt)
   print(f'The system message is: `{system_message}`. Feel free to re-run this cell if you want a better result.
```

2. Data Preparation:

- Separation: Prompts and responses are separated from generated examples.
- DataFrame: Data is organized into a pandas DataFrame for cleaning.
- Removal: Duplicate examples are eliminated.
- Fine-Tuning Format: Examples are structured as JSONL files for GPT-3.5 fine-tuning.

```
import json
import pandas as pd

# Initialize lists to store prompts and responses
prompts = []
responses = []

# Parse out prompts and responses from examples
for example in prev_examples:
    try:
        split_example = example.split('-----')
        prompts.append(split_example[1].strip())
        responses.append(split_example[3].strip())
except:
        pass

# Create a DataFrame
df = pd.DataFrame({
        'prompt': prompts,
        'response': responses
})
```

3. Model Fine-Tuning:

• Upload: Data is uploaded to OpenAI's servers.

```
[9] file_id = client.files.create(
    file=open("training_examples.jsonl", "rb"),
    purpose="fine-tune"
    ).id
```

Initiation: Fine-tuning job is started using the uploaded data and GPT-3.5-turbo.

•

```
    Train the model

[ ] job = client.fine_tuning.jobs.create(
         training_file= "file-t09LUCxGfJJEjt7JSEN9zz6w",
         model="gpt-3.5-turbo"
    )

    job_id = job.id
```

Monitoring: Job status is tracked until completion.

Retrieval: Name of the trained model is retrieved for use.

```
[21] model_name_pre_object = client.fine_tuning.jobs.retrieve('ftjob-GOB2k8hJYuVLS7fq7lQbgke9')
    model_name = model_name_pre_object.fine_tuned_model
    print(model_name)

ft:gpt-3.5-turbo-0613:personal::8j8YHR80
```

4. Model Testing:

• Prompts: Model is tested with prompts from the dataset and the system message.

5. Assistant Creation and Interaction:

 Setup: An OpenAl assistant is created with specific instructions and tools, incorporating the fine-tuned model.

• Thread Initialization: A thread is established for user interactions.

```
[37] thread = client.beta.threads.create()
```

• User Prompt: User provides a prompt to initiate the assistant's response.

 Assistant Run: Assistant is invoked with the user's prompt and guiding instructions.

```
run = client.beta.threads.runs.create(
    thread_id=thread.id,
    assistant_id=assistant.id,
    instructions="You are a travel advisor specializing in European tours. Users will share preferences, such
)
```

Response Retrieval: Assistant's generated responses are retrieved and displayed

```
run = client.beta.threads.runs.create(
    thread_id=thread.id,
    assistant_id=assistant.id,
    instructions="You are a travel advisor specializing in European tours. Users will share preferences, such
)
```

Output:

Output Explanation:

 Assistant's Response: The assistant has responded to given prompt, asking for more information to generate personalized recommendations:

"Certainly! Please provide me with some information about your travel preferences and interests within Europe. Specifically, let me know the type of destination you prefer (e.g., historical, culinary, natural beauty), the duration of your trip, and any specific interests you have."

- Message Details: The output includes details about the two messages in the thread:
 - 1. Message IDs: Unique identifiers for each message.
 - 2. Assistant ID: ID of the assistant that generated the response.
 - 3. Content: Textual content of the messages (your prompt and the assistant's response).
 - 4. Timestamps: When the messages were created.
 - 5. Roles: Whether the messages were sent by the user or the assistant.
 - 6. Thread ID: ID of the thread containing the conversation.
 - 7. Run ID: ID of the assistant's run associated with the response.

6. Integration fine tuned model to the openAI playground.

In here openAI playground we can manually select the fine tuned model and create the
assistant after tuning parameters below. After that we can prompt the questions and
followback questions from the "Travel Agent assistant". Then assistant generate the
answers that we might need

Responses generated from Assistant



