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# Cell 1: Setup, GPU check, and Login
# 1. Install Dependencies
!pip install -q -U transformers peft accelerate bitsandbytes trl datasets google-generativeai firebase-admin google-auth google-c
# 2. Import and Login
import huggingface_hub
import torch
import json
from datasets import Dataset
import google.generativeai as genai
import firebase_admin
from firebase_admin import credentials, firestore
from google.colab import userdata
print("Logging into Hugging Face...")
huggingface_hub.login()
# 3. Check for GPU
if not torch.cuda.is_available():
    raise SystemError("GPU not available. Please go to Runtime > Change runtime type and select T4 GPU.")
    print("\n✓ GPU is available and ready.")
    !nvidia-smi
                                               - 216.1/216.1 kB 6.5 MB/s eta 0:00:00
→
     ERROR: pip's dependency resolver does not currently take into account all the packages that are installed. This behaviour is
     google-colab 1.0.0 requires google-auth==2.38.0, but you have google-auth 2.40.3 which is incompatible.
     gcsfs 2025.3.2 requires fsspec==2025.3.2, but you have fsspec 2025.3.0 which is incompatible.
     Logging into Hugging Face...
    GPU is available and ready.
    Fri Jul 4 16:07:03 2025
      NVIDIA-SMI 550.54.15
                                         Driver Version: 550.54.15
                                                                         CUDA Version: 12.4
      GPU Name
                                 Persistence-M
                                                 Bus-Id
                                                                 Disp.A
                                                                           Volatile Uncorr. ECC
      Fan Temp
                  Perf
                                                                          GPU-Util Compute M.
                                 Pwr:Usage/Cap
                                                           Memory-Usage
                                                                                        MIG M.
        0 Tesla T4
                                           0ff
                                                   00000000:00:04.0 Off
                                                                                              0
      N/A
            45C
                    Р8
                                    9W /
                                           70W
                                                       2MiB / 15360MiB
                                                                                0%
                                                                                        Default
                                                                                            N/A
       Processes:
                                                                                     GPU Memory
       GPU
             GΙ
                  CI
                             PID
                                          Process name
                                   Type
             ID
                  ID
                                                                                     Usage
       No running processes found
# Cell 2: The Bulletproof Data Parser
def find_and_clean_json_objects(text):
    Scans through a string, finds all valid JSON objects, cleans them,
    and returns a list of clean dictionaries. This is a robust,
    character-level parser.
    cleaned_data = []
    pos = 0
    while pos < len(text):
        # Find the start of the next potential JSON object
        start_brace = text.find('{', pos)
        if start_brace == −1:
            break # No more objects in the string
        # Find the matching end brace
        brace_level = 1
        for i in range(start_brace + 1, len(text)):
            char = text[i]
            if char == '{':
                brace_level += 1
            elif char == '}':
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brace_level -= 1

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if brace_level == 0:
                end_brace = i
                # We found a complete object, extract it
                potential_json_str = text[start_brace : end_brace + 1]
                try:
                    # Attempt to parse this extracted string
                    data = json.loads(potential_json_str)
                    # Now, clean the 'output' field inside the valid JSON
                    output_str = data.get('output', '')
                    if '```json' in output_str:
                        # Extract content between the fences more robustly
                        parts = output_str.split('```json\n', 1)
                        if len(parts) > 1:
                            output_str = parts[1]
                    if '``' in output_str:
                        parts = output_str.rsplit('\n```', 1)
                        output_str = parts[0]
                    # Final validation of the inner JSON
                    json.loads(output_str)
                    data['output'] = output_str
                    cleaned_data.append(data)
                except Exception as e:
                    # If parsing fails, just ignore this chunk and continue
                # Move position to after the object we just processed
                pos = end brace + 1
                break # Break from inner loop to find the next object
        else:
            # If we go through the whole string and don't find a matching brace, we're done.
    return cleaned_data
# --- EXECUTION ---
filepath = 'training_data.jsonl'
print(f"--- Starting definitive parsing of '{filepath}' ---")
with open(filepath, 'r') as f:
    full_content = f.read()
cleaned_list = find_and_clean_json_objects(full_content)
if not cleaned_list:
    raise ValueError("CRITICAL FAILURE: No valid JSON objects could be extracted from the file.")
print(f"\n☑ Successfully parsed and cleaned {len(cleaned_list)} examples.")
# Create the Hugging Face Dataset
dataset = Dataset.from_list(cleaned_list)
print("\nDataset loaded successfully:")
print(dataset)
→ --- Starting definitive parsing of 'training_data.jsonl' ---
    ☑ Successfully parsed and cleaned 67 examples.
     Dataset loaded successfully:
    Dataset({
         features: ['instruction', 'input', 'output'],
        num_rows: 67
    })
# Cell 3: Configure, Train, and Upload
from transformers import AutoTokenizer, AutoModelForCausalLM, BitsAndBytesConfig, TrainingArguments
from peft import LoraConfig
from trl import SFTTrainer
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# 1. Load Model and Tokenizer
model_id = "meta-llama/Meta-Llama-3-8B-Instruct"
bnb_config = BitsAndBytesConfig(load_in_4bit=True, bnb_4bit_quant_type="nf4", bnb_4bit_compute_dtype=torch.bfloat16)
print("\nLoading base model and tokenizer...")
# Make sure you have requested and been granted access to this model on Hugging Face
model = AutoModelForCausalLM.from_pretrained(model_id, quantization_config=bnb_config, device_map="auto")
tokenizer = AutoTokenizer.from_pretrained(model_id)
tokenizer.pad_token = tokenizer.eos_token
print("✓ Model and tokenizer loaded.")
     Loading base model and tokenizer...
     /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
     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(
     config.json: 100%
                                                           654/654 [00:00<00:00, 67.8kB/s]
     model.safetensors.index.json: 100%
                                                                        23.9k/23.9k [00:00<00:00, 2.62MB/s]
     Fetching 4 files: 100%
                                                              4/4 [20:49<00:00, 1249.60s/it]
                                                                             5.00G/5.00G [19:34<00:00, 4.12MB/s]
     model-00002-of-00004.safetensors: 100%
     model-00001-of-00004.safetensors: 100%
                                                                             4.98G/4.98G [20:49<00:00, 12.7MB/s]
     model-00004-of-00004.safetensors: 100%
                                                                             1.17G/1.17G [14:03<00:00, 700kB/s]
     model-00003-of-00004.safetensors: 100%
                                                                             4.92G/4.92G [19:32<00:00, 1.60MB/s]
     Loading checkpoint shards: 100%
                                                                      4/4 [01:22<00:00, 18.18s/it]
     generation_config.json: 100%
                                                                    187/187 [00:00<00:00, 15.6kB/s]
     tokenizer config.json: 100%
                                                                  51.0k/51.0k [00:00<00:00, 4.65MB/s]
     tokenizer.json: 100%
                                                             9.09M/9.09M [00:00<00:00, 26.2MB/s]
                                                                     73.0/73.0 [00:00<00:00, 8.27kB/s]
     special tokens map.ison: 100%
     Model and tokenizer loaded.
# 2. Configure LoRA (No changes here)
lora_config = LoraConfig(r=16, lora_alpha=32, lora_dropout=0.05, target_modules=["q_proj", "k_proj", "v_proj", "o_proj"], bias="
# 3. Define the Formatting Function (No changes here)
def formatting_func(example):
    text_list = []
    for i in range(len(example["instruction"])):
        instruction = example["instruction"][i]
        input_text = example["input"][i]
        output_data = json.loads(example["output"][i])
        output_text = f"Event Summary: {output_data.get('event_summary', 'N/A')}\n"
        output_text += f"Event Type: {output_data.get('event_type', 'N/A')}\n"
         output_text += "First-Order Impacts:\n"
         for impact in output_data.get('first_order_impacts', []):
             output_text += f''- {impact}\n"
        output_text += "Second-Order Hypotheses:\n"
         for h in output_data.get('second_order_hypotheses', []):
            hyp = h.get('hypothesis', 'N/A')
rsn = h.get('reasoning', 'N/A')
             cnf = h.get('confidence_score', 'N/A')
             output_text += f"- HYPOTHESIS: {hyp} REASONING: {rsn} (CONFIDENCE: {cnf})\n"
        text_list.append(f"<|begin_of_text|><|start_header_id|>system<|end_header_id|>\n\n{instruction}<|eot_id|><|start_header_
    return text_list
# 4. Set up the Trainer (with memory-saving changes)
training_args = TrainingArguments(
    output_dir="./kairos-llama3-finetune",
    per_device_train_batch_size=1,
                                           # This is already at the minimum
    gradient_accumulation_steps=4,
                                          # Accumulate gradients over 4 steps
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learning_rate=2e-4,
    num_train_epochs=3,
    logging_steps=10,
    save_strategy="epoch",
    fp16=True,
    report to="none",
    # --- MEMORY SAVING CHANGES ---
                                            # ☑ KEY CHANGE: This saves a lot of memory at a small cost of speed.
    gradient_checkpointing=True,
    optim="paged_adamw_8bit"
                                            # ☑ KEY CHANGE: Use a more memory-efficient optimizer.
)
trainer = SFTTrainer(
    model=model,
    args=training_args,
    train_dataset=dataset,
    peft_config=lora_config,
    formatting_func=formatting_func,
)
57 /usr/local/lib/python3.11/dist-packages/peft/mapping_func.py:73: UserWarning: You are trying to modify a model with PEFT for
     /usr/local/lib/python3.11/dist-packages/peft/tuners/tuners_utils.py:190: UserWarning: Already found a `peft_config` attribut
       warnings.warn(
     Applying formatting function to train dataset: 0%
                                                                                   0/67 [00:00<?, ? examples/s]
     Applying formatting function to train dataset: 100%
                                                                                    67/67 [00:00<00:00, 2976.73 examples/s]
     Adding EOS to train dataset: 100%
                                                                        67/67 [00:00<00:00, 2315.60 examples/s]
                                                                     67/67 [00:00<00:00, 592.75 examples/s]
     Tokenizing train dataset: 100%
                                                                     67/67 [00:00<00:00, 5527.29 examples/s]
     Truncating train dataset: 100%
     No label_names provided for model class `PeftModelForCausalLM`. Since `PeftModel` hides base models input arguments, if labe
# 5. Start Training
print("\n--- Starting Final Fine-Tuning Run (Memory Optimized) ---")
trainer.train()
print("▼ Fine-tuning complete!")
₹
       - Starting Final Fine-Tuning Run (Memory Optimized) --
     `use_cache=True` is incompatible with gradient checkpointing. Setting `use_cache=False`.
                                           51/51 08:07, Epoch 3/3]
      Step Training Loss
        10
                   1.453400
        20
                   0.987900
        30
                   0.894200
        40
                   0.825700
                   0.775900
        50

▼ Fine-tuning complete!

# 6. Save and Upload to Hub
print("\n--- Saving Model Adapter to Hugging Face Hub ---")
hf_username = huggingface_hub.whoami()['name']
new model name = f"{hf username}/kairos-llama3-finetune"
--- Saving Model Adapter to Hugging Face Hub ---
trainer.push_to_hub(new_model_name)
print(f"\n# Process complete! Your model is available at: https://huggingface.co/{new_model_name}")
    No files have been modified since last commit. Skipping to prevent empty commit.
     WARNING:huggingface_hub.hf_api:No files have been modified since last commit. Skipping to prevent empty commit.
     🚀 Process complete! Your model is available at: <a href="https://huggingface.co/SharathReddy/kairos-llama3-finetune">https://huggingface.co/SharathReddy/kairos-llama3-finetune</a>
Start coding or generate with AI.
```