Logging into hugging face

```
import huggingface_hub

from google.colab import userdata
token = userdata.get('HF_TOKEN')

from huggingface_hub import login
login(token)
print('Successful login')

Successful login
import shutil
folder_path = "txt_files"
shutil.rmtree(folder_path)
```

Process data

```
!unzip txt_files.zip
Archive: txt files.zip
   creating: txt files/
  inflating: txt files/ADMINISTRATIVE LAW FF.txt
  inflating: txt files/ADVOCATES ACT f.txt
  inflating: txt files/ANCIENT LAW F.txt
  inflating: txt files/ARBITRATION And CONCILIATION.txt
  inflating: txt files/CIVIL PROCEDURE CODE FINAL2012.txt
  inflating: txt files/COMPANY LAW.txt
  inflating: txt files/CONSTITUTION OF INDIA.txt
  inflating: txt files/CONTRACT-Specific ReliefAct.txt
  inflating: txt files/CONTRACT Act F.txt
  inflating: txt files/CRIMINAL PROCEDURE CODE.txt
  inflating: txt_files/EVIDENCE_ACT.txt
  inflating: txt files/HINDU LAW 2012.txt
  inflating: txt files/INDEMNITY GUARANTEE BAILMENT.txt
  inflating: txt_files/INDIAN_CONSTITUTIONAL_HISTORY.txt
  inflating: txt files/INDIAN LEGAL HISTORY.txt
  inflating: txt files/INDIAN PENAL CODE.txt
  inflating: txt files/INDIAN REGISTRATION ACT.txt
  inflating: txt files/INDIAN SUCCESSION ACT.txt
  inflating: txt files/INDIAN TRUSTS ACT.txt
  inflating: txt files/JURISPRUDENCE Legal Theory F.txt
  inflating: txt files/Kenny s Criminal Law.txt
  inflating: txt files/LAW OF AGENCY.txt
  inflating: txt files/LAW OF BANKING.txt
  inflating: txt files/LAW OF EASEMENTS.txt
  inflating: txt files/LAW OF INSURANCE 2012.txt
  inflating: txt files/LAW OF TORTS F.txt
```

```
inflating: txt files/LEGISLATIVE DRAFTING.txt
  inflating: txt files/MOHAMMADAN LAW F.txt
  inflating: txt files/NEGOTIABLE INSTRUMENTS ACT.txt
  inflating: txt files/PARTNERSHIP ACT.txt
  inflating: txt files/PRINCIPLES OF LEGISLATION.txt
  inflating: txt_files/PUBLIC_INTERNATIONAL_LAW.txt
  inflating: txt files/TRANSFER OF PROPERTY ACT.txt
  inflating: txt files/U.N CHARTER.txt
  inflating: txt files/constitution.txt
  inflating: txt files/indian-penal-codes-detailed-final.txt
  inflating: txt files/indian-penal-codes-detailed-part2.txt
  inflating: txt files/indian-penal-codes-detailed-part3.txt
  inflating: txt files/indian-penal-codes-detailed.txt
import os
import json
# Paths to the folder and output files
input_folder = "txt_files"
train output file = "data.jsonl"
eval output file = "data validation.jsonl"
# Function to process each note into 256-character paragraphs
def process file(file path):
    with open(file path, 'r', encoding='utf-8') as file:
        content = file.read().strip().replace('\n', ' ')
        paragraphs = []
        while len(content) > 0:
            paragraphs.append({"paragraph": content[:256]})
            content = content[256:].strip()
        return paragraphs
# Split data into training and evaluation sets (e.g., 80% training,
20% evaluation)
def split data(data, train ratio=0.8):
    split point = int(len(data) * train ratio)
    return data[:split_point], data[split_point:]
# Process all .txt files in the folder
all paragraphs = []
for filename in os.listdir(input folder):
    if filename.endswith(".txt"):
        file path = os.path.join(input folder, filename)
        all paragraphs.extend(process file(file path))
# Split the data into training and evaluation datasets
train data, eval data = split data(all paragraphs)
```

```
# Write train and eval datasets to their respective jsonl files
with open(train output file, 'w', encoding='utf-8') as train file:
    for entry in train data:
        json.dump(entry, train file)
        train file.write('\n')
with open(eval output file, 'w', encoding='utf-8') as eval file:
    for entry in eval data:
        json.dump(entry, eval_file)
        eval file.write('\n')
print(f"Processed {len(train data)} training paragraphs and
{len(eval data)} evaluation paragraphs.")
Processed 13291 training paragraphs and 3323 evaluation paragraphs.
!pip install datasets
Requirement already satisfied: datasets in
/usr/local/lib/python3.10/dist-packages (3.1.0)
Requirement already satisfied: filelock in
/usr/local/lib/python3.10/dist-packages (from datasets) (3.16.1)
Requirement already satisfied: numpy>=1.17 in
/usr/local/lib/python3.10/dist-packages (from datasets) (1.26.4)
Requirement already satisfied: pyarrow>=15.0.0 in
/usr/local/lib/python3.10/dist-packages (from datasets) (17.0.0)
Requirement already satisfied: dill<0.3.9,>=0.3.0 in
/usr/local/lib/python3.10/dist-packages (from datasets) (0.3.8)
Requirement already satisfied: pandas in
/usr/local/lib/python3.10/dist-packages (from datasets) (2.2.2)
Requirement already satisfied: requests>=2.32.2 in
/usr/local/lib/python3.10/dist-packages (from datasets) (2.32.3)
Requirement already satisfied: tgdm>=4.66.3 in
/usr/local/lib/python3.10/dist-packages (from datasets) (4.66.6)
Requirement already satisfied: xxhash in
/usr/local/lib/python3.10/dist-packages (from datasets) (3.5.0)
Requirement already satisfied: multiprocess<0.70.17 in
/usr/local/lib/python3.10/dist-packages (from datasets) (0.70.16)
Requirement already satisfied: fsspec<=2024.9.0,>=2023.1.0 in
/usr/local/lib/python3.10/dist-packages (from
fsspec[http]<=2024.9.0,>=2023.1.0->datasets) (2024.9.0)
Requirement already satisfied: aiohttp in
/usr/local/lib/python3.10/dist-packages (from datasets) (3.10.10)
Requirement already satisfied: huggingface-hub>=0.23.0 in
/usr/local/lib/python3.10/dist-packages (from datasets) (0.26.2)
Requirement already satisfied: packaging in
/usr/local/lib/python3.10/dist-packages (from datasets) (24.2)
Requirement already satisfied: pyyaml>=5.1 in
/usr/local/lib/python3.10/dist-packages (from datasets) (6.0.2)
Requirement already satisfied: aiohappyeyeballs>=2.3.0 in
```

```
/usr/local/lib/python3.10/dist-packages (from aiohttp->datasets)
(2.4.3)
Requirement already satisfied: aiosignal>=1.1.2 in
/usr/local/lib/python3.10/dist-packages (from aiohttp->datasets)
(1.3.1)
Requirement already satisfied: attrs>=17.3.0 in
/usr/local/lib/python3.10/dist-packages (from aiohttp->datasets)
(24.2.0)
Requirement already satisfied: frozenlist>=1.1.1 in
/usr/local/lib/python3.10/dist-packages (from aiohttp->datasets)
(1.5.0)
Requirement already satisfied: multidict<7.0,>=4.5 in
/usr/local/lib/python3.10/dist-packages (from aiohttp->datasets)
Requirement already satisfied: yarl<2.0,>=1.12.0 in
/usr/local/lib/python3.10/dist-packages (from aiohttp->datasets)
(1.17.1)
Requirement already satisfied: async-timeout<5.0,>=4.0 in
/usr/local/lib/python3.10/dist-packages (from aiohttp->datasets)
(4.0.3)
Requirement already satisfied: typing-extensions>=3.7.4.3 in
/usr/local/lib/python3.10/dist-packages (from huggingface-hub>=0.23.0-
>datasets) (4.12.2)
Requirement already satisfied: charset-normalizer<4,>=2 in
/usr/local/lib/python3.10/dist-packages (from reguests>=2.32.2-
>datasets) (3.4.0)
Requirement already satisfied: idna<4,>=2.5 in
/usr/local/lib/python3.10/dist-packages (from requests>=2.32.2-
>datasets) (3.10)
Requirement already satisfied: urllib3<3,>=1.21.1 in
/usr/local/lib/python3.10/dist-packages (from requests>=2.32.2-
>datasets) (2.2.3)
Requirement already satisfied: certifi>=2017.4.17 in
/usr/local/lib/python3.10/dist-packages (from reguests>=2.32.2-
>datasets) (2024.8.30)
Requirement already satisfied: python-dateutil>=2.8.2 in
/usr/local/lib/python3.10/dist-packages (from pandas->datasets)
(2.8.2)
Requirement already satisfied: pytz>=2020.1 in
/usr/local/lib/python3.10/dist-packages (from pandas->datasets)
(2024.2)
Requirement already satisfied: tzdata>=2022.7 in
/usr/local/lib/python3.10/dist-packages (from pandas->datasets)
(2024.2)
Requirement already satisfied: six>=1.5 in
/usr/local/lib/python3.10/dist-packages (from python-dateutil>=2.8.2-
>pandas->datasets) (1.16.0)
Requirement already satisfied: propcache>=0.2.0 in
```

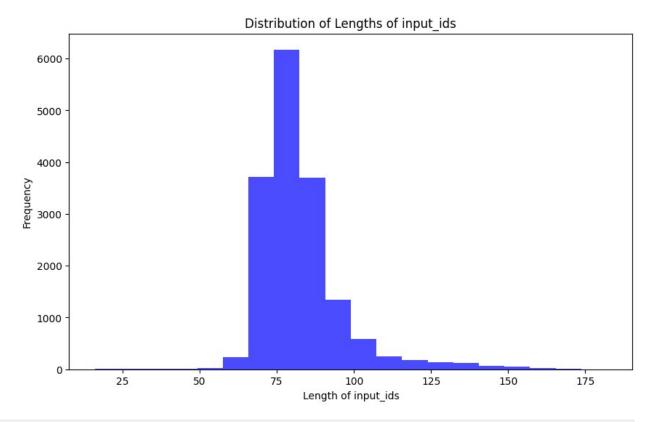
```
/usr/local/lib/python3.10/dist-packages (from yarl<2.0,>=1.12.0-
>aiohttp->datasets) (0.2.0)
from datasets import load dataset
train_dataset = load_dataset('json', data_files='data.jsonl',
split='train')
eval dataset = load dataset('json',
data files='data validation.jsonl', split='train')
{"model id": "4379f00b3c3c460caaa7c73f544dbcfc", "version major": 2, "vers
ion minor":0}
{"model id": "6318277eabbd4459ac2d07927d6af2a4", "version major": 2, "vers
ion minor":0}
def formatting func(example):
    return f"### The following is a note in a Law Textbook:
{example['paragraph']}"
base model id = "mistralai/Mistral-7B-Instruct-v0.3"
output dir = "./mistral-law-finetuned-clean-data"
from transformers import AutoModelForCausalLM, AutoTokenizer,
BitsAndBytesConfig
from peft import LoraConfig, get peft model,
prepare model for kbit training, PeftModel
from transformers import TrainingArguments, Trainer
from datasets import load dataset
tokenizer = AutoTokenizer.from pretrained(
    base model id,
    padding side="left",
    add eos token=True,
    add bos token=True,
tokenizer.pad token = tokenizer.eos token
def generate and tokenize prompt(prompt):
    return tokenizer(formatting func(prompt))
tokenized train dataset =
train dataset.map(generate and tokenize prompt)
tokenized val dataset = eval dataset.map(generate and tokenize prompt)
{"model id": "85ab73d818e1481db0c609c757dd1ba5", "version major": 2, "vers
ion minor":0}
{"model_id": "83c9aa9dac0748dc9770efe94864da8f", "version_major": 2, "vers
ion minor":0}
import matplotlib.pyplot as plt
```

```
def plot_data_lengths(tokenized_train_dataset, tokenized_val_dataset):
    lengths = [len(x['input_ids']) for x in tokenized_train_dataset]
    lengths += [len(x['input_ids']) for x in tokenized_val_dataset]
    print(len(lengths))

# Plotting the histogram
    plt.figure(figsize=(10, 6))
    plt.hist(lengths, bins=20, alpha=0.7, color='blue')
    plt.xlabel('Length of input_ids')
    plt.ylabel('Frequency')
    plt.title('Distribution of Lengths of input_ids')
    plt.show()

plot_data_lengths(tokenized_train_dataset, tokenized_val_dataset)

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```



```
# Define max length for tokenization
max_length = 128

# Tokenization function
def generate_and_tokenize_prompt2(prompt):
    result = tokenizer(
        formatting_func(prompt),
        truncation=True,
```

```
max_length=max_length,
    padding="max_length",
)
result["labels"] = result["input_ids"].copy()
return result

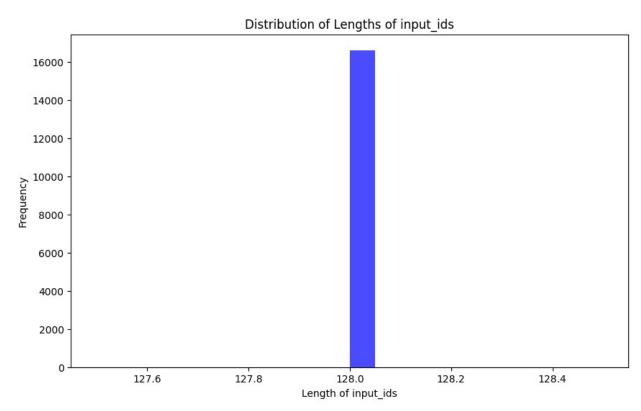
tokenized_train_dataset =
train_dataset.map(generate_and_tokenize_prompt2)
tokenized_val_dataset =
eval_dataset.map(generate_and_tokenize_prompt2)

{"model_id":"eal157fd02b043c2afebc50bc544a5e1","version_major":2,"version_minor":0}

{"model_id":"7e812c1e4f5449ed90533aeeaf1a6b40","version_major":2,"version_minor":0}

plot_data_lengths(tokenized_train_dataset, tokenized_val_dataset)

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```



Train model using LoRA

```
# LoRA Configuration
lora_config = LoraConfig(
    r=16,
```

```
lora alpha=32,
    target modules=["q proj", "v proj"],
    lora dropout=0.05,
    bias="none",
    task type="CAUSAL LM"
!pip uninstall bitsandbytes
!pip install -U bitsandbytes
Found existing installation: bitsandbytes 0.44.1
Uninstalling bitsandbytes-0.44.1:
 Would remove:
    /usr/local/lib/python3.10/dist-packages/bitsandbytes-0.44.1.dist-
info/*
    /usr/local/lib/python3.10/dist-packages/bitsandbytes/*
    /usr/local/lib/python3.10/dist-packages/tests/*
Proceed (Y/n)? y
  Successfully uninstalled bitsandbytes-0.44.1
Collecting bitsandbytes
  Using cached bitsandbytes-0.44.1-py3-none-
manylinux 2 24 x86 64.whl.metadata (3.5 kB)
Requirement already satisfied: torch in
/usr/local/lib/python3.10/dist-packages (from bitsandbytes)
(2.5.1+cu121)
Requirement already satisfied: numpy in
/usr/local/lib/python3.10/dist-packages (from bitsandbytes) (1.26.4)
Requirement already satisfied: filelock in
/usr/local/lib/python3.10/dist-packages (from torch->bitsandbytes)
(3.16.1)
Requirement already satisfied: typing-extensions>=4.8.0 in
/usr/local/lib/python3.10/dist-packages (from torch->bitsandbytes)
(4.12.2)
Requirement already satisfied: networkx in
/usr/local/lib/python3.10/dist-packages (from torch->bitsandbytes)
(3.4.2)
Requirement already satisfied: jinja2 in
/usr/local/lib/python3.10/dist-packages (from torch->bitsandbytes)
Requirement already satisfied: fsspec in
/usr/local/lib/python3.10/dist-packages (from torch->bitsandbytes)
(2024.9.0)
Requirement already satisfied: sympy==1.13.1 in
/usr/local/lib/python3.10/dist-packages (from torch->bitsandbytes)
(1.13.1)
Requirement already satisfied: mpmath<1.4,>=1.1.0 in
/usr/local/lib/python3.10/dist-packages (from sympy==1.13.1->torch-
>bitsandbytes) (1.3.0)
Requirement already satisfied: MarkupSafe>=2.0 in
/usr/local/lib/python3.10/dist-packages (from jinja2->torch-
```

```
>bitsandbytes) (3.0.2)
Using cached bitsandbytes-0.44.1-py3-none-manylinux 2 24 x86 64.whl
(122.4 MB)
Installing collected packages: bitsandbytes
Successfully installed bitsandbytes-0.44.1
from transformers import AutoModelForCausalLM, BitsAndBytesConfig
import torch
# Create a BitsAndBvtesConfia
quantization config = BitsAndBytesConfig(
    load in 4bit=True, # Use 4-bit quantization
# Load base model with the new quantization config
model = AutoModelForCausalLM.from pretrained(
    base model id,
    quantization config=quantization config,
    torch dtype=torch.float16,
    device map="auto",
)
{"model id": "3b2270a0874248c5b8676ecfdcb8ac67", "version major": 2, "vers
ion minor":0}
{"model id":"2ee11901370849eda865a33554673c5c","version major":2,"vers
ion minor":0}
{"model id":"fc88ff99858c4ba29c4c3b489bd3ae8c","version major":2,"vers
ion minor":0}
{"model id": "bab47afd99c1419ca03287988c4dc975", "version major": 2, "vers
ion minor":0}
{"model id":"a03145983bfa45bfbf5cf1fe3a5c6ced","version major":2,"vers
ion minor":0}
{"model id":"07ede0c5badb4be6952068d68ca2afd0","version major":2,"vers
ion minor":0}
{"model id":"df5e72cd670b45a3ae964b2504a35360","version major":2,"vers
ion minor":0}
# Prepare model for k-bit training
model = prepare model for kbit training(model)
model = get peft model(model, lora config)
import warnings
# Suppress specific warnings from `torch` and `bitsandbytes`
warnings.filterwarnings("ignore", category=UserWarning,
module="torch")
```

```
warnings.filterwarnings("ignore", category=UserWarning,
module="bitsandbytes")
from transformers import TrainingArguments
# Training arguments
training args = TrainingArguments(
    output dir="./mistral lora output clean data",
                                                        # Output
directory to save the model and checkpoints
    per device train batch size=16,
                                                        # Batch size
for training
    per device eval batch size=16,
                                                        # Batch size
for evaluation
    learning rate=3e-4,
                                                        # Learning
rate
                                                        # Number of
    num train epochs=1,
epochs
    eval strategy="epoch",
                                                        # Evaluate at
the end of each epoch
    logging strategy="epoch",
                                                        # Log at the
end of each epoch
    save_strategy="epoch",
                                                        # Save the
model at the end of each epoch
    save_total limit=2,
                                                        # Keep only
the last 2 checkpoints
    load best model at end=True,
                                                        # Load best
model at the end based on evaluation metric
                                                        # Enable mixed
    fp16=True,
precision training
    report to="none",
                                                        # Do not
report logs to external services
trainer = Trainer(
    model=model.
    args=training args,
    train dataset=tokenized train dataset,
    eval dataset=tokenized val dataset,
)
trainer.train()
`use cache=True` is incompatible with gradient checkpointing. Setting
`use cache=False`...
<IPython.core.display.HTML object>
TrainOutput(global step=831, training loss=1.257926197258574,
metrics={'train_runtime': 5851.1217, 'train_samples_per_second':
2.272, 'train_steps_per_second': 0.142, 'total_flos':
7.268365920790118e+16, 'train loss': 1.257926197258574, 'epoch': 1.0})
```

```
# Save the fine-tuned model
model.save pretrained("mistral lora finetuned clean")
tokenizer.save pretrained("mistral lora finetuned clean")
# Push the model to Hugging Face Hub
model.push to hub("rujool/mistral lora finetuned clean",
use auth token=True)
tokenizer.push to hub("rujool/mistral lora finetuned clean",
use auth token=True)
/usr/local/lib/python3.10/dist-packages/transformers/utils/hub.py:894:
FutureWarning: The `use auth token` argument is deprecated and will be
removed in v5 of Transformers. Please use `token` instead.
 warnings.warn(
{"model id": "2b5fd7111a56449ea8bf90d0dd26f2a7", "version major": 2, "vers
ion minor":0}
{"model id": "5225decf4e8344dd95ea13168422d0be", "version major": 2, "vers
ion minor":0}
{"model id": "8b16b22757054a2c9fdf16284abd8d52", "version major": 2, "vers
ion minor":0}
{"type": "string"}
def generate answer(query, model, tokenizer, max length=256):
    inputs = tokenizer(query, return_tensors="pt").to(model.device)
    outputs = model.generate(**inputs, max new tokens=max length,
do sample=True)
    return tokenizer.decode(outputs[0], skip special tokens=True)
# Example usage
query = "what does Indian Penal Code 230 state"
answer = generate answer(query, model, tokenizer)
print(f"Answer: {answer}")
Setting `pad token id` to `eos token id`:None for open-end generation.
Answer: what does Indian Penal Code 230 state 0: 5.What is the
punishment for adultery in the Indian Penal Code 1860? A: Under the
Indian Penal Code 1860, adultery is defined as sexual intercourse
between a married man and a woman who is not his lawful wife or who,
with the sexual consent of that man, has sexual intercourse with any
man other than her husband. The punishment for adultery is listed in
Section 497, where it is stated that whoever has sexual intercourse
with a married woman without the consent of her husband is guilty of
adultery, punishable
def generate answer(query, model, tokenizer, max length=512,
temperature=0.7):
```

```
# Use a more explicit template for the prompt
    prompt = f" {query}"
    # Tokenize the input with the modified prompt
    inputs = tokenizer(prompt, return tensors="pt",
padding=True).to(model.device)
    # Generate output from the model
    outputs = model.generate(
        **inputs,
        max new tokens=max length,
        do sample=True,
        temperature=temperature,
        top p=0.9,
        pad token id=tokenizer.eos token id,
        #early stopping=True # Stops when the model generates a
complete answer
    )
    # Decode the output to get the generated answer
    answer = tokenizer.decode(outputs[0], skip special tokens=True)
    # Remove the prompt part (we only want the answer)
    answer = answer[len(prompt):].strip() # Strip whitespace if
necessarv
    return answer
query = "explain the concepts of criminal law"
answer = generate answer(guery, model, tokenizer)
print(f"{answer}")
Ouestion: What are the main differences between criminal law and civil
law? Answer: Criminal law and civil law are two distinct branches of
law. The main differences between them are as follows:
1. Objectives: Criminal law is designed to protect society and the
state from harm, while civil law is intended to protect the rights and
interests of individuals.
2. Parties: In criminal law, the state is the prosecutor, while in
civil law, the parties are private individuals.
3. Standard of proof: In criminal law, the prosecution must prove the
guilt of the accused beyond a reasonable doubt, while in civil law,
the plaintiff must prove his or her case on the balance of
probabilities.
4. Penalties: In criminal law, the penalties are punitive, such as
fines, imprisonment, or death. In civil law, the remedies are
compensatory, such as damages or an injunction.
5. Jurisdiction: Criminal law is governed by the criminal code, while
civil law is governed by the civil code.
```

6. Procedures: Criminal law procedures are more formal and adversarial, with the right to counsel, a jury trial, and appeals. Civil law procedures are less formal and less adversarial, with a judge presiding over the case and a jury trial being optional.

```
query = "difference between partnership firm and company"
answer = generate_answer(query, model, tokenizer)
print(f"{answer}")
```

Question: What is the difference between partnership firm and company?\\\ Answer: A partnership firm is a business organization in which two or more persons carry on a business with a view to profit. It is not a separate legal entity from its partners. On the other hand, a company is a separate legal entity from its shareholders. It is formed by the registration of a Memorandum and Articles of Association under the Companies Act.

```
query = "explain jurisprudence"
answer = generate_answer(query, model, tokenizer)
print(f"{answer}")
```

Q: What is Jurisprudence? A: Jurisprudence is the study of law and the principles that guide its interpretation and application. It includes the history, philosophy, and sociology of law, as well as the methods and rules used by courts and other legal institutions to interpret and apply the law. Jurisprudence is a fundamental part of legal education and is studied by law students and legal professionals to gain a deeper understanding of the law and the legal system.

```
query = "what is the role of CBI"
answer = generate_answer(query, model, tokenizer)
print(f"{answer}")
```

Question 1: What is the role of CBI in India?\\ The Central Bureau of Investigation (CBI) is a premier investigating agency of the Government of India, responsible for the investigation of heinous crimes and corruption cases. Its role includes:

- 1. Investigating cases of corruption, bribery, fraud, and embezzlement involving public servants.
- 2. Investigating organized crime, including drug trafficking, terrorism, and economic offenses.
- 3. Investigating economic offenses, such as money laundering, insider trading, and cybercrime.
- 4. Providing assistance to state police forces in investigating complex cases.
- 5. Providing technical assistance to foreign governments in their investigations.
- 6. Recovering assets and properties acquired by proceeds of crime.
- 7. Prosecuting cases in courts of law.

```
query = "explain IPC 232:"
answer = generate_answer(query, model, tokenizer)
print(f"{answer}")
```

Q: What is the punishment for the offence of defiling a place of worship under IPC 232? A: The punishment for defiling a place of worship under IPC 232 is imprisonment for a term which may extend to two years, or fine, or both. This offense is committed when a person defiles a place of worship, or any object used in religious worship, with the intention of insulting the religion or religious feelings of any class of persons.