

```
!pip install -q transformers datasets accelerate evaluate rouge_score
```

```
Preparing metadata (setup.py) ... done
84.1/84.1 kB 2.6 MB/s eta 0:00:00
Building wheel for rouge_score (setup.py) ... done
```

```
import os
import json
import math
from typing import List, Dict

import torch
from torch.utils.data import DataLoader
from datasets import Dataset

from transformers import AutoTokenizer, GPT2LMHeadModel
from rouge_score import rouge_scorer
from tqdm.auto import tqdm

from google.colab import drive
drive.mount('/content/drive')
```

Mounted at /content/drive

```
# Adjust these if you changed paths anywhere
DATASET_PATH = "/content/drive/My Drive/Colab Notebooks/CS 561: Topics in Data Privacy/Data/"
MODEL_PATH = "/content/drive/My Drive/Colab Notebooks/CS 561: Topics in Data Privacy/Models/"

BASELINE_DIR = os.path.join(MODEL_PATH, "gpt2_baseline_poisoned")
DPAGGZO_DIR = os.path.join(MODEL_PATH, "gpt2_dp_aggzo_poisoned")

device = torch.device("cuda" if torch.cuda.is_available() else "cpu")
print("Using device:", device)

MAX_LEN = 128
```

Using device: cpu

```
def load_jsonl_as_strings(path: str) -> List[str]:
    texts = []
    with open(path, "r", encoding="utf-8") as f:
        for line in f:
            line = line.strip()
            if not line:
                continue
            obj = json.loads(line) # each line is a JSON string
            texts.append(str(obj))
    return texts

# Load clean validation set
val_file = os.path.join(DATASET_PATH, "val.jsonl")
val_texts = load_jsonl_as_strings(val_file)
print("Validation set size:", len(val_texts))

# Load target passage
```

```
with open(os.path.join(DATASET_PATH, "target_passage.txt"), "r", encoding="utf-8") as f:
    target_passage = f.read().strip()
```

```
print("\nTarget passage preview:")
print(target_passage[:400], "...")
```

Validation set size: 4888

Target passage preview:

a mayor for a dad. "I got to thinking. If it wasn't Stahl who attacked you, why would you say that it was? Unless you were covering for someone. Like the mayor's son." "Why wo

```
def make_collate_fn(tokenizer):
    def collate(batch):
        texts = [ex["text"] for ex in batch]
        enc = tokenizer(
            texts,
            return_tensors="pt",
            padding=True,
            truncation=True,
            max_length=MAX_LEN,
        )
        # labels are input_ids with padding masked out
        enc["labels"] = enc["input_ids"].clone()
        enc = {k: v.to(device) for k, v in enc.items()}
        return enc
    return collate

def compute_perplexity(model: GPT2LMHeadModel,
                       tokenizer: AutoTokenizer,
                       texts: List[str],
                       batch_size: int = 16) -> float:
    dataset = Dataset.from_dict({"text": texts})
    loader = DataLoader(
        dataset,
        batch_size=batch_size,
        shuffle=False,
        collate_fn=make_collate_fn(tokenizer),
    )

    model.eval()
    total_loss = 0.0
    total_tokens = 0

    with torch.no_grad():
        for batch in tqdm(loader, desc="Perplexity eval"):
            outputs = model(**batch)
            loss = outputs.loss # mean over non -100 tokens

            labels = batch["labels"]
            mask = labels.ne(-100)
            tokens_in_batch = mask.sum().item()

            total_loss += loss.item() * tokens_in_batch
            total_tokens += tokens_in_batch

    mean_loss = total_loss / total_tokens
```

```
ppl = math.exp(mean_loss)
return ppl
```

```
def split_target_prefix(target_text: str, prefix_ratio: float = 0.25):
    words = target_text.split()
    split_idx = max(1, int(len(words) * prefix_ratio))
    prefix = " ".join(words[:split_idx])
    target_suffix = " ".join(words[split_idx:])
    return prefix, target_suffix


def sample_continuations(model: GPT2LMHeadModel,
                        tokenizer: AutoTokenizer,
                        prefix: str,
                        num_samples: int = 200,
                        batch_size: int = 10,
                        max_new_tokens: int = 128) -> List[str]:
    """
    Generate num_samples continuations from the model given the prefix.
    Uses sampling for diversity, batched for T4 efficiency.
    """
    model.eval()
    continuations = []

    enc = tokenizer(prefix, return_tensors="pt")
    input_ids = enc["input_ids"].to(device)
    input_len = input_ids.shape[1]

    with torch.no_grad():
        while len(continuations) < num_samples:
            cur_bs = min(batch_size, num_samples - len(continuations))

            outputs = model.generate(
                input_ids.repeat(cur_bs, 1),
                do_sample=True,
                temperature=0.7,
                top_k=40,
                max_new_tokens=max_new_tokens,
                pad_token_id=tokenizer.eos_token_id,
            )

            for out in outputs:
                gen_ids = out[input_len:] # strip prefix
                text = tokenizer.decode(gen_ids, skip_special_tokens=True)
                continuations.append(text)

    return continuations[:num_samples]


def compute_rougeL_scores(model: GPT2LMHeadModel,
                          tokenizer: AutoTokenizer,
                          target_text: str,
                          num_samples: int = 200,
                          batch_size: int = 10) -> Dict[str, float]:
    prefix, target_suffix = split_target_prefix(target_text, prefix_ratio=0.25)
    print("Prompt (prefix) used for memorization test:\n")
    print(prefix, "\n")
```

```

continuations = sample_continuations(
    model,
    tokenizer,
    prefix,
    num_samples=num_samples,
    batch_size=batch_size,
    max_new_tokens=128,
)

scorer = rouge_scorer.RougeScorer(["rougeL"], use_stemmer=True)

scores = []
for gen in continuations:
    score = scorer.score(target_suffix, gen)["rougeL"].fmeasure
    scores.append(score)

max_score = max(scores)
mean_score = sum(scores) / len(scores)

return {
    "rougeL_max": max_score,
    "rougeL_mean": mean_score,
}

```

```

def evaluate_model(model_dir: str,
                  name: str,
                  num_rouge_samples: int = 200) -> Dict[str, float]:
    print(f"\n===== Evaluating {name} =====\n")
    tokenizer = AutoTokenizer.from_pretrained(model_dir)
    model = GPT2LMHeadModel.from_pretrained(model_dir)
    model.to(device)

    # Utility: Perplexity on clean validation data
    ppl = compute_perplexity(model, tokenizer, val_texts, batch_size=16)
    print(f"\n{name} - Perplexity on clean val.jsonl: {ppl:.3f}\n")

    # Memorization: ROUGE-L on target passage
    rouge_scores = compute_rougeL_scores(
        model,
        tokenizer,
        target_passage,
        num_samples=num_rouge_samples,
        batch_size=10,
    )
    print(f"\n{name} - ROUGE-L (max over samples): {rouge_scores['rougeL_max']:.4f}")
    print(f"\n{name} - ROUGE-L (mean over samples): {rouge_scores['rougeL_mean']:.4f}\n")

    out = {
        "perplexity": ppl,
        **rouge_scores,
    }
    return out

```

```

results_baseline = evaluate_model(BASELINE_DIR, "Baseline GPT-2 (AdamW)", num_rouge_samples=200)
results_dpaggzo = evaluate_model(DPAGGZO_DIR, "DP-AggZO GPT-2", num_rouge_samples=200)

```

```
print("\n===== Summary =====\n")
print("Baseline:", results_baseline)
print("DP-AggZO:", results_dpaggzo)
```

```
===== Evaluating Baseline GPT-2 (AdamW) =====
```

```
Perplexity eval: 100%                                306/306 [50:54<00:00, 8.22s/it]
```

```
`loss_type=None` was set in the config but it is unrecognized. Using the default loss: `ForCausalMLoss`.
```

```
The attention mask is not set and cannot be inferred from input because pad token is same as eos token. As a consequence, you may observe unexpected behavior. Please pass your
```

```
Baseline GPT-2 (AdamW) – Perplexity on clean val.jsonl: 33.412
```

```
Prompt (prefix) used for memorization test:
```

```
a mayor for a dad. "I got to thinking. If it wasn't Stahl who attacked you, why would you say that it was? Unless you were covering for someone. Like the mayor's son." "Why wo
```

```
Baseline GPT-2 (AdamW) – ROUGE-L (max over samples): 0.1247
```

```
Baseline GPT-2 (AdamW) – ROUGE-L (mean over samples): 0.0994
```

```
===== Evaluating DP-AggZO GPT-2 =====
```

```
Perplexity eval: 100%                                306/306 [50:21<00:00, 8.56s/it]
```

```
DP-AggZO GPT-2 – Perplexity on clean val.jsonl: 5758.638
```

```
Prompt (prefix) used for memorization test:
```

```
a mayor for a dad. "I got to thinking. If it wasn't Stahl who attacked you, why would you say that it was? Unless you were covering for someone. Like the mayor's son." "Why wo
```

```
DP-AggZO GPT-2 – ROUGE-L (max over samples): 0.0903
```

```
DP-AggZO GPT-2 – ROUGE-L (mean over samples): 0.0630
```

```
===== Summary =====
```

```
Baseline: {'perplexity': 33.41217380509253, 'rougeL_max': 0.12474849094567403, 'rougeL_mean': 0.0993758338414651}
```

```
DP-AggZO: {'perplexity': 5758.637916821883, 'rougeL_max': 0.09034907597535935, 'rougeL_mean': 0.06300033723064166}
```

```
import pandas as pd
```

```
# Your evaluation results
```

```
results = {
    "Model": ["Baseline GPT-2 (AdamW)", "DP-AggZO GPT-2"],
    "Perplexity": [33.41217386909253, 5758.637916821883],
    "ROUGE-L Max": [0.12474849094567403, 0.09034907597535935],
    "ROUGE-L Mean": [0.0993758338414651, 0.06300337230641666]
}
```

```
df = pd.DataFrame(results)
```

```
# Format numbers like academic tables
```

```
df["Perplexity"] = df["Perplexity"].apply(lambda x: f"{x:,.2f}")
```

```

df["ROUGE-L Max"] = df["ROUGE-L Max"].apply(lambda x: f"{x:.4f}")
df["ROUGE-L Mean"] = df["ROUGE-L Mean"].apply(lambda x: f"{x:.4f}")

# Styling: clean, minimal, no index (just like NeurIPS/ICML tables)
styled = (
    df.style
    .hide(axis="index") # <- removes the index completely
    .set_properties(**{
        'text-align': 'center',
        'font-size': '15px',
        'font-family': 'Arial'
    })
    .set_table_styles([
        {'selector': 'th',
         'props': [('text-align', 'center'),
                   ('font-size', '16px'),
                   ('font-weight', 'bold'),
                   ('border-bottom', '1px solid black')]}
    ])
)

styled

```

Model	Perplexity	ROUGE-L Max	ROUGE-L Mean
Baseline GPT-2 (AdamW)	33.41	0.1247	0.0994
DP-AggZO GPT-2	5,758.64	0.0903	0.0630