```
!pip install --quiet wandb
!pip install --quiet transformers
!pip install --quiet datasets
!pip install --quiet emoji
!pip install --quiet kaggle
!pip install --quiet torchinfo
!pip install --quiet imbalanced-learn
!pip install --quiet gdown
!pip install --quiet clean-text
!pip install --quiet accelerate -U
!pip install --quiet transformers[torch]
!pip install --quiet huggingface_hub[tensorflow]
!pip install --quiet huggingface_hub[cli,torch]
!pip install --quiet openai
from huggingface_hub import HfApi
import huggingface_hub
#transformers
import transformers
from transformers import AutoTokenizer
from transformers import GPT2Tokenizer
from transformers import BertTokenizer
from transformers import OpenAIGPTTokenizer
from transformers import AutoConfig
from transformers import GPT2Config
from transformers.models.bert.modeling_bert import BertModel
from\ transformers.models.gpt2.modeling\_gpt2\ import\ GPT2Model
from transformers.models.openai.modeling openai import OpenAIGPTModel
from\ transformers.models.bert.modeling\_bert\ import\ BertPreTrainedModel
from\ transformers.models.gpt2.modeling\_gpt2\ import\ GPT2PreTrainedModel
from transformers.models.openai.modeling_openai import OpenAIGPTPreTrainedModel
from\ transformers\ import\ BertForSequence Classification
from transformers import GPT2ForSequenceClassification
from\ transformers\ import\ Open AIGPTFor Sequence Classification
from transformers.modeling_outputs import SequenceClassifierOutput
from transformers import Trainer
from transformers import TrainingArguments
from transformers import get_scheduler
from transformers import TextDataset
from transformers.modeling utils import PreTrainedModel
#torch
import torch
import torch.nn as nn
from \ torch.nn \ import \ BCEWithLogitsLoss, \ MSELoss, \ CrossEntropyLoss
from torch.optim import AdamW
from torch.utils import data
from torch.utils.data import Dataset as ds, DataLoader
# dataset
import datasets
from datasets import Dataset
from datasets import Sequence
from datasets import Value
from datasets import Features
from datasets import ClassLabel
from datasets import DatasetDict
from sklearn.model_selection import train_test_split
from sklearn.metrics import accuracy_score, f1_score
from sklearn.metrics import confusion_matrix, classification_report
from sklearn.metrics import ConfusionMatrixDisplay, confusion_matrix
#open ai
import openai
# others
import os
import wandb
import re, string
import emoji
import pandas as pd
import seaborn as sns
import matplotlib.pyplot as plt
```

```
import numpy as np
import pickle
import joblib
import traceback
from tqdm import tqdm
from collections import defaultdict
device = torch.device("cuda" if torch.cuda.is_available() else "cpu")
api = HfApi()
\Box
                                                    - 2.1/2.1 MB 29.1 MB/s eta 0:00:00
                                                 - 188.6/188.6 kB 23.3 MB/s eta 0:00:00
                                                 - 218.8/218.8 kB 24.5 MB/s eta 0:00:00
       Preparing metadata (setup.py) ... done
                                                   - 62.7/62.7 kB 7.0 MB/s eta 0:00:00
       Building wheel for pathtools (setup.py) ... done
                                                    - 7.6/7.6 MB 24.3 MB/s eta 0:00:00
                                                 - 268.8/268.8 kB 27.0 MB/s eta 0:00:00
                                                    - 7.8/7.8 MB 42.9 MB/s eta 0:00:00
                                                    - 1.3/1.3 MB 37.5 MB/s eta 0:00:00
                                                  - 519.6/519.6 kB 2.0 MB/s eta 0:00:00
                                                 - 115.3/115.3 kB 14.8 MB/s eta 0:00:00
                                                  · 194.1/194.1 kB 24.8 MB/s eta 0:00:00
                                                 - 134.8/134.8 kB 18.1 MB/s eta 0:00:00
                                                  - 358.9/358.9 kB 7.0 MB/s eta 0:00:00
                                                  - 175.4/175.4 kB <mark>2.2 MB/s</mark> eta 0:00:00
       Preparing metadata (setup.py) ... done
                                                   - 53.1/53.1 kB 4.2 MB/s eta 0:00:00
       Building wheel for emoji (setup.py) ... done
                                                  - 251.2/251.2 kB <mark>3.4 MB/s</mark> eta 0:00:00
                                                    - 67.7/67.7 kB 1.6 MB/s eta 0:00:00
                                                   - 76.5/76.5 kB 1.3 MB/s eta 0:00:00
```

! huggingface-cli login --token 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

Token will not been saved to git credential helper. Pass `add_to_git_credential=True` if you want to set the git credential as well. Token is valid (permission: write).

Your token has been saved to /root/.cache/huggingface/token

Your token has been saved to /root/.cache/huggingface/token Login successful

huggingface_hub.login()



Copy a token from your Hugging Face tokens page and paste it below.

Immediately click login after copying your token or it might be stored in plain text in this



Pro Tip: If you don't already have one, you can create a dedicated 'notebooks' token with

1 94 1 46 4 46 9 2 H 4 6 1.

```
class BertForClassification(BertForSequenceClassification):
    def __init__(self, config):
        super().__init__(config)
        self.num_labels = config.num_labels
        self.config = config
        # Load model body > return all og the HS
        self.bert = BertModel(config)
        # Set up token classification head
        self.dropout = nn.Dropout(config.hidden_dropout_prob)
        self.classifier = nn.Linear(config.hidden_size, config.num_labels)
        # Initialize weights and apply final processing
        self.post_init()
    def forward(self, input_ids=None, attention_mask=None, token_type_ids=None,
                labels=None, **kwargs):
        # Use model body to get encoder representations
        outputs = self.bert(input_ids, attention_mask=attention_mask,
                               token_type_ids=token_type_ids, **kwargs)
        # Apply classifier to encoder representation > [cls]
        sequence_output = self.dropout(outputs[1])
        logits = self.classifier(sequence_output)
        # Calculate losses
        loss = None
        if labels is not None:
            loss_fct = nn.CrossEntropyLoss()
            loss = loss_fct(logits.view(-1, self.num_labels), labels.view(-1))
            #outputs = (loss,) + outputs # can comment
        # return outputs # (loss), logits, (hidden_states), (attentions)
        # Return model output object
        return SequenceClassifierOutput(
            loss=loss,
            logits=logits,
            hidden_states=outputs.hidden_states,
            attentions=outputs.attentions,
        )
model_names = ["bert-large-uncased"]
id2label = {0: 'non-toxic', 1: 'mild', 2: 'toxic'}
label2id = { v:k for (k,v) in id2label.items()}
labels = ['non-toxic','mild', 'toxic']
tokenizers = {"bert-large-uncased": BertTokenizer.from_pretrained("bert-large-uncased")}
configs = {"bert-large-uncased": AutoConfig.from_pretrained("bert-large-uncased", num_labels=3, id2label=id2label, label2id=label2id)}
models = {"bert-large-uncased": BertForClassification.from_pretrained("bert-large-uncased", config=configs["bert-large-uncased"])}
                                                                                 232k/232k [00:00<00:00, 9.22MB/s]
     Downloading (...)solve/main/vocab.txt: 100%
     Downloading (...)okenizer_config.json: 100%
                                                                                 28.0/28.0 [00:00<00:00, 1.96kB/s]
     Downloading (...)lve/main/config.json: 100%
                                                                                 571/571 [00:00<00:00, 34.9kB/s]
     Downloading model.safetensors: 100%
                                                                             1.34G/1.34G [00:05<00:00, 146MB/s]
     Some weights of BertForClassification were not initialized from the model checkpoint at bert-large-uncased and are newly initialized: ['
     You should probably TRAIN this model on a down-stream task to be able to use it for predictions and inference
```

```
def clean_text(text):
  text = re.sub('#', '', text) # Removing '#' hashtag
  text = re.sub('\w+:\/\\S+', '', text) # Removing hyperlink
  text = re.sub('[^a-zA-Z]', ' ', text) # Remove punctuation
  return text.lower()
def compute_metrics(pred):
  labels = pred.label_ids
  preds = pred.predictions.argmax(-1)
  f1 = f1_score(labels, preds, average="weighted")
  acc = accuracy_score(labels, preds)
  return {"accuracy": acc, "f1": f1}
def create dataset(df, text, label):
  class_names = ['non-toxic','mild', 'toxic']
  data_dict = {'text':df[text], 'labels':df[label]}
  tags = ClassLabel(num_classes=3 , names=class_names)
  feature_set = Features({'text':Value(dtype='string'), 'labels':tags})
  return Dataset.from_dict(mapping = data_dict, features = feature_set)
def get_dataset():
  train = pd.read_csv('/content/train.csv')
  validation = pd.read_csv('/content/validation.csv')
  dataset_train = create_dataset(train, "message", "target")
  dataset_val = create_dataset(validation, "message", "target")
  train_convert = train.copy()
  train_convert['target'] = train_convert['target'].astype(str)
  train_convert['message'] = (train_convert['message'] + ' ->')
  train_convert['target'] = (' ' + train_convert['target'])
  train_convert.rename(columns={'message':'prompt', 'target':'completion'}, inplace=True)
  train_convert.to_json("train_pandas.jsonl", orient='records', lines=True)
  validation_convert = validation.copy()
  validation_convert['target'] = validation_convert['target'].astype(str)
  validation_convert['message'] = (validation_convert['message'] + ' ->')
  validation_convert['target'] = (' ' + validation_convert['target'])
  validation_convert.rename(columns={'message':'prompt', 'target':'completion'}, inplace=True)
  validation_convert.to_json("validation_pandas.jsonl", orient='records', lines=True)
  dataset = DatasetDict()
  dataset["train"] = dataset_train
  dataset["validation"] = dataset_val
  return dataset
! wandb login eed5796f17cdf020038ca37377edaaec8c9ddfbe
     wandb: Appending key for api.wandb.ai to your netrc file: /root/.netrc
wandb.login()
wandb.init(project="bert-toxic-chat-detection", entity="dani-squad")
     wandb: Currently logged in as: dffesalbon (dani-squad). Use `wandb login --relogin` to force relogin
     Tracking run with wandb version 0.15.10
     Run data is saved locally in /content/wandb/run-20230907_001043-rj53tmy9
     Syncing run fancy-blaze-32 to Weights & Biases (docs)
     View project at https://wandb.ai/dani-squad/bert-toxic-chat-detection
     View run at https://wandb.ai/dani-squad/bert-toxic-chat-detection/runs/rj53tmy9
     Display W&B run
```

```
def finetune_bert(model_name, dataset, max_len=64, train_batch_size=16, num_epochs=5):
 tokenizer = tokenizers[model_name]
 model = models[model_name]
 tokenize = lambda batch: tokenizer(batch["text"], padding=True, truncation=True, max_length=max_len)
 encoded = dataset.map(tokenize, batched=True, batch_size=32)
 model.to(device)
 optimizer = AdamW(model.parameters(), lr=2e-5, eps=1e-8)
 logging_steps = len(encoded['train']) // train_batch_size
 num_training_steps = num_epochs * logging_steps
 scheduler = get_scheduler(
     name="linear",
      optimizer=optimizer,
     num_warmup_steps=0,
     num_training_steps=num_training_steps)
 training_args = TrainingArguments(
      output_dir=f"{model_name}-toxic-detector",
     num_train_epochs=num_epochs,
     per_device_train_batch_size=train_batch_size,
     per_device_eval_batch_size=train_batch_size,
     weight_decay=0.01,
      evaluation_strategy="epoch",
     save_steps=1e6,
     disable_tqdm=False,
     logging_steps=logging_steps,
     push_to_hub=False,
      log_level="error",
     report_to="wandb",
     run_name=model_name)
 optimizers = (optimizer, scheduler)
 model trainer = Trainer(
def save_model(model_name, model_trainer):
 {\tt model\_trainer.save\_model(f'/content/\{model\_name\}-saved')}
 api.upload folder(
      folder_path=f"/content/{model_name}-saved",
     repo_id=f"dffesalbon/{model_name}-dota-toxic",
      repo_type="model")
dataset = get_dataset()
 IIIOUET.EVAT()
  BERT (Large-uncased)
```

model_name, model_trainer, encoded = finetune_bert(model_names[0], dataset) Map: 100% 1722/1722 [00:00<00:00, 2067.87 examples/s] Map: 100% 192/192 [00:00<00:00, 1157.71 examples/s] 540/540 04:01, Epoch 5/5] Epoch Training Loss Validation Loss Accuracy F1 1 0.857000 0.568056 0.786458 0.781023 2 0.555000 0.417159 0.838542 0.837147 3 0.384100 0.796875 0.788389 0.504736 4 0.231400 5 0.141100 try: save_model(model_name, model_trainer)

except Exception as ex: