## finetuning-bert-for-classification

July 7, 2025

## [1]: !pip install transformers -U

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Requirement already satisfied: transformers in /usr/local/lib/python3.11/dist-
packages (4.52.4)
Requirement already satisfied: filelock in /usr/local/lib/python3.11/dist-
packages (from transformers) (3.18.0)
Requirement already satisfied: huggingface-hub<1.0,>=0.30.0 in
/usr/local/lib/python3.11/dist-packages (from transformers) (0.33.0)
Requirement already satisfied: numpy>=1.17 in /usr/local/lib/python3.11/dist-
packages (from transformers) (2.0.2)
Requirement already satisfied: packaging>=20.0 in
/usr/local/lib/python3.11/dist-packages (from transformers) (24.2)
Requirement already satisfied: pyyaml>=5.1 in /usr/local/lib/python3.11/dist-
packages (from transformers) (6.0.2)
Requirement already satisfied: regex!=2019.12.17 in
/usr/local/lib/python3.11/dist-packages (from transformers) (2024.11.6)
Requirement already satisfied: requests in /usr/local/lib/python3.11/dist-
packages (from transformers) (2.32.3)
Requirement already satisfied: tokenizers<0.22,>=0.21 in
/usr/local/lib/python3.11/dist-packages (from transformers) (0.21.1)
Requirement already satisfied: safetensors>=0.4.3 in
/usr/local/lib/python3.11/dist-packages (from transformers) (0.5.3)
Requirement already satisfied: tqdm>=4.27 in /usr/local/lib/python3.11/dist-
packages (from transformers) (4.67.1)
Requirement already satisfied: fsspec>=2023.5.0 in
/usr/local/lib/python3.11/dist-packages (from huggingface-
hub<1.0,>=0.30.0->transformers) (2025.3.2)
Requirement already satisfied: typing-extensions>=3.7.4.3 in
/usr/local/lib/python3.11/dist-packages (from huggingface-
hub<1.0,>=0.30.0->transformers) (4.14.0)
Requirement already satisfied: hf-xet<2.0.0,>=1.1.2 in
/usr/local/lib/python3.11/dist-packages (from huggingface-
hub<1.0,>=0.30.0->transformers) (1.1.3)
Requirement already satisfied: charset-normalizer<4,>=2 in
/usr/local/lib/python3.11/dist-packages (from requests->transformers) (3.4.2)
Requirement already satisfied: idna<4,>=2.5 in /usr/local/lib/python3.11/dist-
packages (from requests->transformers) (3.10)
Requirement already satisfied: urllib3<3,>=1.21.1 in
```

```
Requirement already satisfied: certifi>=2017.4.17 in
    /usr/local/lib/python3.11/dist-packages (from requests->transformers)
    (2025.6.15)
[4]: import pandas as pd
    Traceback (most recent call last):
      File "/usr/local/bin/kaggle", line 4, in <module>
        from kaggle.cli import main
      File "/usr/local/lib/python3.11/dist-packages/kaggle/__init__.py", line 6, in
    <module>
        api.authenticate()
      File "/usr/local/lib/python3.11/dist-
    packages/kaggle/api/kaggle_api_extended.py", line 434, in authenticate
        raise IOError('Could not find {}. Make sure it\'s located in'
    OSError: Could not find kaggle.json. Make sure it's located in
    /root/.config/kaggle. Or use the environment method. See setup instructions at
    https://github.com/Kaggle/kaggle-api/
[6]: # https://www.kaggle.com/competitions/
      → jiqsaw-toxic-comment-classification-challenge/data
     data = pd.read_csv("/content/train.csv", engine="python")
     data.head()
[6]:
                      id
                                                                comment_text toxic \
     0 0000997932d777bf Explanation\nWhy the edits made under my usern...
                                                                                 0
                          D'aww! He matches this background colour I'm s...
     1 000103f0d9cfb60f
                                                                                 0
     2 000113f07ec002fd
                          Hey man, I'm really not trying to edit war. It ...
                                                                                 0
                          "\nMore\nI can't make any real suggestions on ...
     3 0001b41b1c6bb37e
                                                                                 0
     4 0001d958c54c6e35
                          You, sir, are my hero. Any chance you remember...
                      obscene
                                       insult
                                                identity_hate
        severe_toxic
                               threat
     0
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[7]: data['toxic'].value_counts()
[7]: toxic
     0
          144277
           15294
     1
     Name: count, dtype: int64
```

/usr/local/lib/python3.11/dist-packages (from requests->transformers) (2.4.0)

```
[8]: data = data[['comment_text', 'toxic']]
      data = data[0:1000]
      data.head()
 [8]:
                                              comment_text
                                                             toxic
      O Explanation\nWhy the edits made under my usern...
      1 D'aww! He matches this background colour I'm s...
                                                               0
      2 Hey man, I'm really not trying to edit war. It ...
      3 "\nMore\nI can't make any real suggestions on ...
                                                               0
      4 You, sir, are my hero. Any chance you remember...
 [9]: import numpy as np
      from sklearn.model_selection import train_test_split
      from sklearn.metrics import accuracy_score, recall_score, precision_score,
       ⊶f1_score
      import torch
      from transformers import TrainingArguments, Trainer
      from transformers import BertTokenizer, BertForSequenceClassification
[10]: from transformers import BertTokenizer, BertForSequenceClassification
      tokenizer = BertTokenizer.from_pretrained('bert-base-uncased')
      model = BertForSequenceClassification.
       ofrom_pretrained('bert-base-uncased',num_labels=2)
     /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 it as secret in your Google Colab
     and restart your session.
     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(
     tokenizer_config.json:
                              0%1
                                            | 0.00/48.0 [00:00<?, ?B/s]
                               | 0.00/232k [00:00<?, ?B/s]
                  0%1
     vocab.txt:
                       0%1
                                     | 0.00/466k [00:00<?, ?B/s]
     tokenizer.json:
                                  | 0.00/570 [00:00<?, ?B/s]
     config.json:
                    0%1
     model.safetensors:
                          0%1
                                        | 0.00/440M [00:00<?, ?B/s]
     Some weights of BertForSequenceClassification were not initialized from the
     model checkpoint at bert-base-uncased and are newly initialized:
     ['classifier.bias', 'classifier.weight']
     You should probably TRAIN this model on a down-stream task to be able to use it
     for predictions and inference.
```

## [11]: model

```
[11]: BertForSequenceClassification(
        (bert): BertModel(
          (embeddings): BertEmbeddings(
            (word_embeddings): Embedding(30522, 768, padding_idx=0)
            (position_embeddings): Embedding(512, 768)
            (token_type_embeddings): Embedding(2, 768)
            (LayerNorm): LayerNorm((768,), eps=1e-12, elementwise_affine=True)
            (dropout): Dropout(p=0.1, inplace=False)
          (encoder): BertEncoder(
            (layer): ModuleList(
              (0-11): 12 x BertLayer(
                (attention): BertAttention(
                  (self): BertSdpaSelfAttention(
                    (query): Linear(in_features=768, out_features=768, bias=True)
                    (key): Linear(in_features=768, out_features=768, bias=True)
                    (value): Linear(in_features=768, out_features=768, bias=True)
                    (dropout): Dropout(p=0.1, inplace=False)
                  (output): BertSelfOutput(
                    (dense): Linear(in_features=768, out_features=768, bias=True)
                    (LayerNorm): LayerNorm((768,), eps=1e-12, elementwise_affine=True)
                    (dropout): Dropout(p=0.1, inplace=False)
                  )
                )
                (intermediate): BertIntermediate(
                  (dense): Linear(in_features=768, out_features=3072, bias=True)
                  (intermediate_act_fn): GELUActivation()
                )
                (output): BertOutput(
                  (dense): Linear(in_features=3072, out_features=768, bias=True)
                  (LayerNorm): LayerNorm((768,), eps=1e-12, elementwise_affine=True)
                  (dropout): Dropout(p=0.1, inplace=False)
              )
            )
          (pooler): BertPooler(
            (dense): Linear(in_features=768, out_features=768, bias=True)
            (activation): Tanh()
          )
        )
        (dropout): Dropout(p=0.1, inplace=False)
        (classifier): Linear(in_features=768, out_features=2, bias=True)
      )
```

```
[12]: model = model.to('cuda')
[14]: sample data = ["I am eating", "I am playing "]
  tokenizer(sample_data, padding=True, truncation=True, max_length=512)
[14]: {'input_ids': [[101, 1045, 2572, 5983, 102], [101, 1045, 2572, 2652, 102]],
  'token_type_ids': [[0, 0, 0, 0, 0], [0, 0, 0, 0]], 'attention_mask': [[1, 1,
  1, 1, 1], [1, 1, 1, 1, 1]]}
[15]: X = list(data["comment text"])
  y = list(data["toxic"])
  X_train, X_val, y_train, y_val = train_test_split(X, y, test_size=0.
  \rightarrow 2, stratify=y)
  X_train_tokenized = tokenizer(X_train, padding=True, truncation=True,_
  →max_length=512)
  X_val_tokenized = tokenizer(X_val, padding=True, truncation=True,__
  →max_length=512)
[16]: X_train_tokenized.keys()
[16]: dict keys(['input ids', 'token type ids', 'attention mask'])
[18]: print(X_train_tokenized['attention_mask'][0])
  [19]: len(X_train),len(X_val)
[19]: (800, 200)
```

```
[20]: # Create dataset
     class Dataset(torch.utils.data.Dataset):
         def __init__(self, encodings, labels=None):
             self.encodings = encodings
             self.labels = labels
         def __getitem__(self, idx):
             item = {key: torch.tensor(val[idx]) for key, val in self.encodings.
       →items()}
             if self.labels:
                 item["labels"] = torch.tensor(self.labels[idx])
             return item
         def __len__(self):
             return len(self.encodings["input_ids"])
[21]: train dataset = Dataset(X train tokenized, y train)
     val_dataset = Dataset(X_val_tokenized, y_val)
[22]: train_dataset[5]
[22]: {'input_ids': tensor([ 101, 1000, 5034, 1013, 18856,
                                                              1024,
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     8026, 1037,
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 1, 1, 1, 1, 1, 1,
    0, 0, 0, 0, 0, 0, 0, 0]),
 'labels': tensor(1)}
[23]: def compute_metrics(p):
  print(type(p))
  pred, labels = p
  pred = np.argmax(pred, axis=1)
  accuracy = accuracy_score(y_true=labels, y_pred=pred)
  recall = recall_score(y_true=labels, y_pred=pred)
  precision = precision_score(y_true=labels, y_pred=pred)
  f1 = f1_score(y_true=labels, y_pred=pred)
  return {"accuracy": accuracy, "precision": precision, "recall": recall, u
  [24]: # Define Trainer
 args = TrainingArguments(
  output_dir="output",
  num_train_epochs=1,
  per_device_train_batch_size=8
```

```
trainer = Trainer(
          model=model,
          args=args,
          train_dataset=train_dataset,
          eval_dataset=val_dataset,
          compute_metrics=compute_metrics
[25]: trainer.train()
     wandb: WARNING The `run_name` is currently set to the same
     value as `TrainingArguments.output dir`. If this was not intended, please
     specify a different run name by setting the `TrainingArguments.run_name`
     parameter.
     <IPython.core.display.Javascript object>
     wandb: Logging into wandb.ai. (Learn how to deploy a W&B server
     locally: https://wandb.me/wandb-server)
     wandb: You can find your API key in your browser here:
     https://wandb.ai/authorize?ref=models
     wandb: Paste an API key from your profile and hit enter:
      . . . . . . . . . .
     wandb: WARNING If you're specifying your api key in code,
     ensure this code is not shared publicly.
     wandb: WARNING Consider setting the WANDB_API_KEY
     environment variable, or running `wandb login` from the command line.
     wandb: No netrc file found, creating one.
     wandb: Appending key for api.wandb.ai to your netrc file:
     /root/.netrc
     wandb: Currently logged in as: johnprabhasith
     (johnprabhasith-cmr-college-of-engineering-technology) to
     https://api.wandb.ai. Use `wandb login --relogin` to force
     relogin
     <IPython.core.display.HTML object>
     <IPython.core.display.HTML object>
     <IPython.core.display.HTML object>
     <IPython.core.display.HTML object>
     <IPython.core.display.HTML object>
     <IPython.core.display.HTML object>
[25]: TrainOutput(global_step=100, training_loss=0.2390480613708496,
```

metrics={'train\_runtime': 630.766, 'train\_samples\_per\_second': 1.268,

```
'train_steps_per_second': 0.159, 'total_flos': 210488844288000.0, 'train_loss':
      0.2390480613708496, 'epoch': 1.0})
[26]: trainer.evaluate()
     <IPython.core.display.HTML object>
     <class 'transformers.trainer_utils.EvalPrediction'>
[26]: {'eval_loss': 0.08775264769792557,
       'eval_accuracy': 0.97,
       'eval_precision': 0.77777777777778,
       'eval_recall': 1.0,
       'eval_f1': 0.875,
       'eval runtime': 6.036,
       'eval_samples_per_second': 33.135,
       'eval_steps_per_second': 4.142,
       'epoch': 1.0}
[27]: np.set_printoptions(suppress=True)
[38]: text = "That was good point"
      # text = "go to hell"
      inputs = tokenizer(text, padding = True, truncation = True, return_tensors='pt').

sto('cuda')
      outputs = model(**inputs)
      print(outputs)
      predictions = torch.nn.functional.softmax(outputs.logits, dim=-1)
      print(predictions)
      predictions = predictions.cpu().detach().numpy()
      if predictions[0][0] > 0.5:
        print("Not Toxic")
      else:
        print("Toxic")
     SequenceClassifierOutput(loss=None, logits=tensor([[ 2.0817, -1.9117]],
     device='cuda:0', grad_fn=<AddmmBackward0>), hidden_states=None, attentions=None)
     tensor([[0.9819, 0.0181]], device='cuda:0', grad_fn=<SoftmaxBackward0>)
     Not Toxic
[29]: trainer.save model('CustomModel')
[30]: model 2 = BertForSequenceClassification.from_pretrained("CustomModel")
      model_2.to('cuda')
[30]: BertForSequenceClassification(
        (bert): BertModel(
          (embeddings): BertEmbeddings(
```

```
(position_embeddings): Embedding(512, 768)
            (token_type_embeddings): Embedding(2, 768)
            (LayerNorm): LayerNorm((768,), eps=1e-12, elementwise_affine=True)
            (dropout): Dropout(p=0.1, inplace=False)
          (encoder): BertEncoder(
            (layer): ModuleList(
              (0-11): 12 x BertLayer(
                (attention): BertAttention(
                  (self): BertSdpaSelfAttention(
                    (query): Linear(in_features=768, out_features=768, bias=True)
                    (key): Linear(in features=768, out features=768, bias=True)
                    (value): Linear(in_features=768, out_features=768, bias=True)
                    (dropout): Dropout(p=0.1, inplace=False)
                  (output): BertSelfOutput(
                    (dense): Linear(in_features=768, out_features=768, bias=True)
                    (LayerNorm): LayerNorm((768,), eps=1e-12, elementwise_affine=True)
                    (dropout): Dropout(p=0.1, inplace=False)
                )
                (intermediate): BertIntermediate(
                  (dense): Linear(in features=768, out features=3072, bias=True)
                  (intermediate_act_fn): GELUActivation()
                (output): BertOutput(
                  (dense): Linear(in_features=3072, out_features=768, bias=True)
                  (LayerNorm): LayerNorm((768,), eps=1e-12, elementwise_affine=True)
                  (dropout): Dropout(p=0.1, inplace=False)
              )
            )
          (pooler): BertPooler(
            (dense): Linear(in_features=768, out_features=768, bias=True)
            (activation): Tanh()
          )
        )
        (dropout): Dropout(p=0.1, inplace=False)
        (classifier): Linear(in features=768, out features=2, bias=True)
      )
[37]: # text = "That was good point"
      inputs = tokenizer(text,padding = True, truncation = True, return_tensors='pt').

→to('cuda')
      outputs = model_2(**inputs)
```

(word\_embeddings): Embedding(30522, 768, padding\_idx=0)

```
predictions = torch.nn.functional.softmax(outputs.logits, dim=-1)
predictions = predictions.cpu().detach().numpy()
if predictions[0][0] > 0.5:
    print("Not Toxic")
else:
    print("Toxic")
```

Not Toxic

[]: