

Bert Model

```
In [141]: import pandas as pd
import numpy as np

In [150]: true_path = pd.read_csv("C:\\Users\\Laptop inn\\Desktop\Fake.csv")
    fake_path = pd.read_csv("C:\\Users\\Laptop inn\\Desktop\True.csv")

In [153]:
    true_df = true_path = pd.read_csv("C:\\Users\\Laptop inn\\Desktop\Fake.csv")
    fake_df = fake_path = pd.read_csv("C:\\Users\\Laptop inn\\Desktop\Fake.csv")
```

In [154]: display(true_df)

	title	text	subject	date
0	Donald Trump Sends Out Embarrassing New Year'	Donald Trump just couldn t wish all Americans	News	December 31, 2017
1	Drunk Bragging Trump Staffer Started Russian	House Intelligence Committee Chairman Devin Nu	News	December 31, 2017
2	Sheriff David Clarke Becomes An Internet Joke	On Friday, it was revealed that former Milwauk	News	December 30, 2017
3	Trump Is So Obsessed He Even Has Obama's Name	On Christmas day, Donald Trump announced that	News	December 29, 2017
4	Pope Francis Just Called Out Donald Trump Dur	Pope Francis used his annual Christmas Day mes	News	December 25, 2017
23476	McPain: John McCain Furious That Iran Treated	21st Century Wire says As 21WIRE reported earl	Middle- east	January 16, 2016
23477	JUSTICE? Yahoo Settles E-mail Privacy Class-ac	21st Century Wire says It s a familiar theme	Middle- east	January 16, 2016
23478	Sunnistan: US and Allied 'Safe Zone' Plan to T	Patrick Henningsen 21st Century WireRemember	Middle- east	January 15, 2016
23479	How to Blow \$700 Million: Al Jazeera America F	21st Century Wire says Al Jazeera America will	Middle- east	January 14, 2016
23480	10 U.S. Navy Sailors Held by Iranian Military	21st Century Wire says As 21WIRE predicted in	Middle- east	January 12, 2016

23481 rows × 4 columns

In [155]: display(fake_df)

	title	text	subject	date
0	As U.S. budget fight looms, Republicans flip t	WASHINGTON (Reuters) - The head of a conservat	politicsNews	December 31, 2017
1	U.S. military to accept transgender recruits o	WASHINGTON (Reuters) - Transgender people will	politicsNews	December 29, 2017
2	Senior U.S. Republican senator: 'Let Mr. Muell	WASHINGTON (Reuters) - The special counsel inv	politicsNews	December 31, 2017
3	FBI Russia probe helped by Australian diplomat	WASHINGTON (Reuters) - Trump campaign adviser	politicsNews	December 30, 2017
4	Trump wants Postal Service to charge 'much mor	SEATTLE/WASHINGTON (Reuters) - President Donal	politicsNews	December 29, 2017
21412	'Fully committed' NATO backs new U.S. approach	BRUSSELS (Reuters) - NATO allies on Tuesday we	worldnews	August 22, 2017
21413	LexisNexis withdrew two products from Chinese	LONDON (Reuters) - LexisNexis, a provider of I	worldnews	August 22, 2017
21414	Minsk cultural hub becomes haven from authorities	MINSK (Reuters) - In the shadow of disused Sov	worldnews	August 22, 2017
21415	Vatican upbeat on possibility of Pope Francis	MOSCOW (Reuters) - Vatican Secretary of State	worldnews	August 22, 2017
21416	Indonesia to buy \$1.14 billion worth of Russia	JAKARTA (Reuters) - Indonesia will buy 11 Sukh	worldnews	August 22, 2017

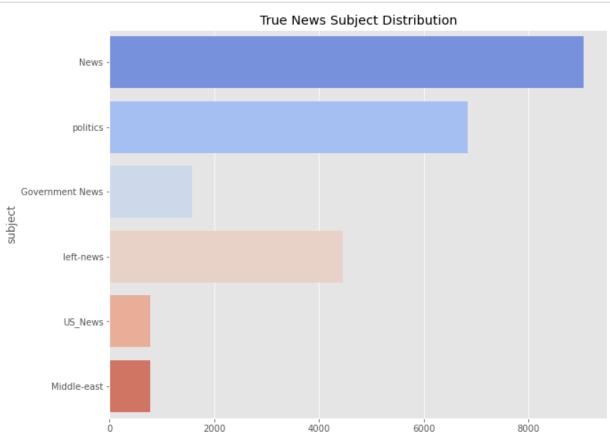
21417 rows × 4 columns

Data Visulization

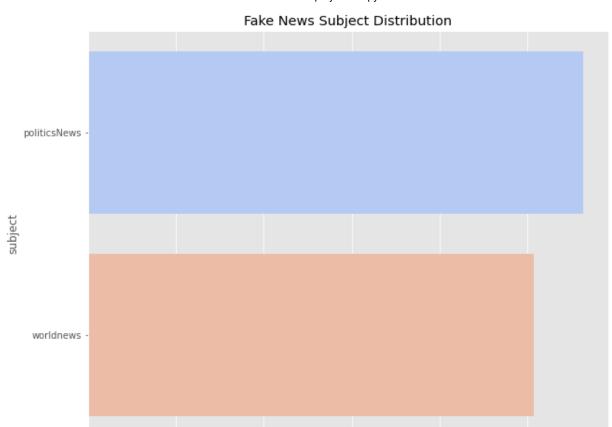
```
In [156]: import seaborn as sns
import matplotlib.pyplot as plt

sns.countplot(y="subject", palette="coolwarm", data=true_df).set_title('True News
plt.show()

sns.countplot(y="subject", palette="coolwarm", data=fake_df).set_title('Fake News
plt.show()
```



count



True News wordcloud

2000

4000

6000

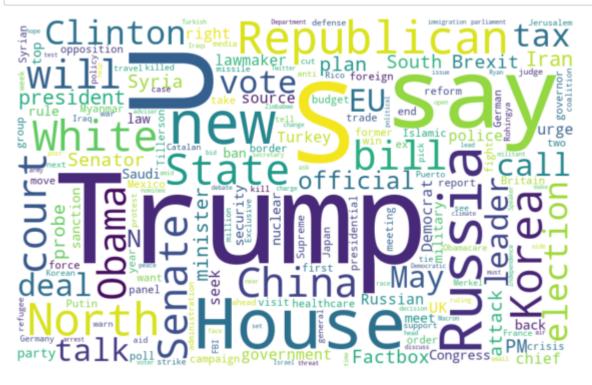
count

8000

10000



Fake News Word Cloud



Data Preprocessing

Data Combination

```
In [161]: # Add Labels to both df
    true_df['true'] = 1
    fake_df['true'] = 0

# Concat
    df = pd.concat([true_df, fake_df])
    display(df)
```

	title	text	subject	date	true
0	Donald Trump Sends Out Embarrassing New Year'	Donald Trump just couldn t wish all Americans	News	December 31, 2017	1
1	Drunk Bragging Trump Staffer Started Russian	House Intelligence Committee Chairman Devin Nu	News	December 31, 2017	1
2	Sheriff David Clarke Becomes An Internet Joke	On Friday, it was revealed that former Milwauk	News	December 30, 2017	1
3	Trump Is So Obsessed He Even Has Obama's Name	On Christmas day, Donald Trump announced that	News	December 29, 2017	1
4	Pope Francis Just Called Out Donald Trump Dur	Pope Francis used his annual Christmas Day mes	News	December 25, 2017	1
21412	'Fully committed' NATO backs new U.S. approach	BRUSSELS (Reuters) - NATO allies on Tuesday we	worldnews	August 22, 2017	0
21413	LexisNexis withdrew two products from Chinese	LONDON (Reuters) - LexisNexis, a provider of I	worldnews	August 22, 2017	0
21414	Minsk cultural hub becomes haven from authorities	MINSK (Reuters) - In the shadow of disused Sov	worldnews	August 22, 2017	0
21415	Vatican upbeat on possibility of Pope Francis	MOSCOW (Reuters) - Vatican Secretary of State	worldnews	August 22, 2017	0
21416	Indonesia to buy \$1.14 billion worth of Russia	JAKARTA (Reuters) - Indonesia will buy 11 Sukh	worldnews	August 22, 2017	0

44898 rows × 5 columns

Inspect Lengths of News

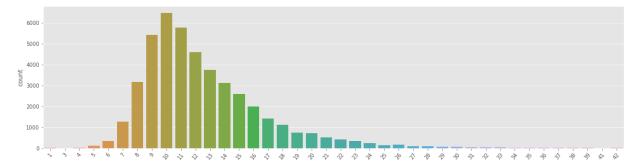
```
In [163]: titles = [text for text in df.title]

max_len = 0
    titles_len = []
    for title in titles:
        titles_len.append(len(title.split()))
        max_len = max(len(title.split()), max_len)

print('Number of titles:', len(titles))
    print('Max length of the titles:', max_len)
    print('Mean length of the titles:', np.mean(titles_len))
```

Number of titles: 44898 Max length of the titles: 42 Mean length of the titles: 12.453472315025168

```
In [164]: plt.figure(figsize=(20,5))
g = sns.countplot(x=titles_len)
g.set_xticklabels(g.get_xticklabels(), rotation=50)
plt.show()
```



```
In [165]: texts = [text for text in df.text]

max_len = 0
    texts_len = []
    for text in texts:
        texts_len.append(len(text.split()))
        max_len = max(len(text.split()), max_len)

# g = sns.countplot(x=texts_len)
    print('Mean length of the texts:', np.mean(texts_len))
```

Mean length of the texts: 405.28228428883244

Purify & Shffle the DataFrame

```
In [167]: from sklearn.utils import shuffle

# Purify
df = df.iloc[:,[0, -1]]

# Shuffle
df = shuffle(df).reset_index(drop=True)

display(df)
```

	title	true
0	MOCKINGBIRD REDUX? CNN's Role in Peddling Fake	1
1	Zakharova Slams CIA Chief Pompeo: Stop Making	1
2	Factbox: The companies making money from Illin	0
3	McCain says time for Republicans, Democrats to	0
4	This Anti-Government Oregon Terrorist Took Th	1
44893	Pope says Colombia must confront inequality to	0
44894	Trump: Being friends with North Korea's Kim is	0
44895	Tillerson says never considered resigning	0
44896	Trump's Washington foreign policy speech	0
44897	Trump Tweets Deranged, Self-Congratulatory 'F	1

44898 rows × 2 columns

Split Data into Training, Validation, Test

```
In [169]: train_val_df = df.sample(frac = 0.8)
    test_df = df.drop(train_val_df.index)

    train_df = train_val_df.sample(frac = 0.8)
    val_df = train_val_df.drop(train_df.index)

# Reset Index
    train_df = train_df.reset_index(drop=True)
    val_df = val_df.reset_index(drop=True)
    test_df = test_df.reset_index(drop=True)

print('trainset size:', train_df.shape)
print('valset size:', val_df.shape)
print('testset size:', test_df.shape)

trainset size: (28734, 2)
    valset size: (7184, 2)
    testset size: (8980, 2)
```

Dataframe to csv

```
In [171]: train_df.to_csv('train.tsv', sep='\t', index=False)
    val_df.to_csv('val.tsv', sep='\t', index=False)
    test_df.to_csv('test.tsv', sep='\t', index=False)
```

Concatenate all dataframe

```
In [173]: df = pd.concat([train_df, val_df, test_df])
df
```

Out[173]:

	title	true
0	WATCH: Powerful Video PERFECTLY Explains How	1
1	TOP DETECTIVES GIVE REASONS They Are Shocked N	1
2	South Africa's Zuma rejects reports his office	0
3	BRAIN FREEZE! HILLARY CLINTON Goes BlankForge	1
4	HA! THEY'RE NOT FORGETTING About Hillary At Th	1
8975	"YOU'RE HIRED!" Trump Pulls Unemployed Vet Fro	1
8976	WHOA: Russians Hacked Voting Systems In 39 St	1
8977	Unwavering Trump voters say they will not miss	0
8978	U.S. says identified target for sanctions over	0
8979	U.N. discussing resolution after Syrian govern	0

44898 rows × 2 columns

Performing Data Cleaning

```
In [175]:
          import nltk
          # Downloading Stopwords
          nltk.download("stopwords")
          [nltk_data] Downloading package stopwords to C:\Users\Laptop
          [nltk_data]
                          inn\AppData\Roaming\nltk_data...
          [nltk data]
                        Package stopwords is already up-to-date!
Out[175]: True
In [176]: # Obtaining Additional Stopwords From nltk
          from nltk.corpus import stopwords
          stop_words = stopwords.words('english')
          stop_words.extend(['from', 'subject', 're', 'edu', 'use'])
In [177]: # Removing Stopwords And Remove Words With 2 Or Less Characters
          def preprocess(text):
              result = []
              for token in gensim.utils.simple_preprocess(text):
                  if token not in gensim.parsing.preprocessing.STOPWORDS and len(token) > 3
                      result.append(token)
              return result
```

```
In [178]: import gensim

# Applying The Function To The Dataframe
df['clean'] = df['title'].apply(preprocess)
```

Obtaining The Total Words Present In The Dataset

```
In [180]: list_of_words = []
for i in df.clean:
    for j in i:
        list_of_words.append(j)

total_words = len(list(set(list_of_words)))
total_words
Out[180]: 19174
```

Preparing The Data By Performing Tokenization And Padding

```
In [188]: !pip install transformers
    import torch
    from transformers import BertTokenizer
```

```
Requirement already satisfied: transformers in c:\users\laptop inn\anaconda3\ne
w folder\lib\site-packages (4.20.1)
Requirement already satisfied: tokenizers!=0.11.3,<0.13,>=0.11.1 in c:\users\la
ptop inn\anaconda3\new folder\lib\site-packages (from transformers) (0.12.1)
Requirement already satisfied: huggingface-hub<1.0,>=0.1.0 in c:\users\laptop i
nn\anaconda3\new folder\lib\site-packages (from transformers) (0.8.1)
Requirement already satisfied: numpy>=1.17 in c:\users\laptop inn\anaconda3\new
folder\lib\site-packages (from transformers) (1.20.3)
Requirement already satisfied: requests in c:\users\laptop inn\anaconda3\new fo
lder\lib\site-packages (from transformers) (2.26.0)
Requirement already satisfied: packaging>=20.0 in c:\users\laptop inn\anaconda3
\new folder\lib\site-packages (from transformers) (21.0)
Requirement already satisfied: filelock in c:\users\laptop inn\anaconda3\new fo
lder\lib\site-packages (from transformers) (3.3.1)
Requirement already satisfied: regex!=2019.12.17 in c:\users\laptop inn\anacond
a3\new folder\lib\site-packages (from transformers) (2021.8.3)
Requirement already satisfied: tqdm>=4.27 in c:\users\laptop inn\anaconda3\new
folder\lib\site-packages (from transformers) (4.62.3)
Requirement already satisfied: pyyaml>=5.1 in c:\users\laptop inn\anaconda3\new
folder\lib\site-packages (from transformers) (6.0)
Requirement already satisfied: typing-extensions>=3.7.4.3 in c:\users\laptop in
n\anaconda3\new folder\lib\site-packages (from huggingface-hub<1.0,>=0.1.0->tra
nsformers) (3.10.0.2)
Requirement already satisfied: pyparsing>=2.0.2 in c:\users\laptop inn\anaconda
3\new folder\lib\site-packages (from packaging>=20.0->transformers) (3.0.4)
Requirement already satisfied: colorama in c:\users\laptop inn\anaconda3\new fo
lder\lib\site-packages (from tqdm>=4.27->transformers) (0.4.4)
Requirement already satisfied: certifi>=2017.4.17 in c:\users\laptop inn\anacon
da3\new folder\lib\site-packages (from requests->transformers) (2022.6.15)
Requirement already satisfied: charset-normalizer~=2.0.0 in c:\users\laptop inn
\anaconda3\new folder\lib\site-packages (from requests->transformers) (2.0.4)
Requirement already satisfied: urllib3<1.27,>=1.21.1 in c:\users\laptop inn\ana
conda3\new folder\lib\site-packages (from requests->transformers) (1.26.7)
Requirement already satisfied: idna<4,>=2.5 in c:\users\laptop inn\anaconda3\ne
w folder\lib\site-packages (from requests->transformers) (3.2)
```

```
In [189]: PRETRAINED_MODEL_NAME = 'bert-base-uncased'
tokenizer = BertTokenizer.from_pretrained(PRETRAINED_MODEL_NAME)
```

Load Dataset Class

```
In [191]: from torch.utils.data import Dataset
          class FakeNewsDataset(Dataset):
              def init (self, mode, tokenizer):
                  assert mode in ['train', 'val', 'test']
                  self.mode = mode
                  self.df = pd.read csv(mode + '.tsv', sep='\t').fillna("")
                  self.len = len(self.df)
                  self.tokenizer = tokenizer # BERT tokenizer
              # Define a function that returns a training/test data
              def __getitem__(self, idx):
                  if self.mode == 'test':
                      statement, label = self.df.iloc[idx, :].values
                      label tensor = torch.tensor(label)
                  else:
                      statement, label = self.df.iloc[idx, :].values
                      label tensor = torch.tensor(label)
          # Create BERT tokens for the first sentence and add separators [SEP
                  word_pieces = ['[CLS]']
                  statement = self.tokenizer.tokenize(statement)
                  word_pieces += statement + ['[SEP]']
                  len st = len(word pieces)
                  ids = self.tokenizer.convert_tokens_to_ids(word_pieces)
                  tokens tensor = torch.tensor(ids)
                  # Set the token position of the first sentence containing [SEP] to 0
                  segments tensor = torch.tensor([0] * len st, dtype=torch.long)
                  return (tokens_tensor, segments_tensor, label_tensor)
              def __len__(self):
                  return self.len
          # Initialize Datasets for Transformation
          trainset = FakeNewsDataset('train', tokenizer=tokenizer)
          valset = FakeNewsDataset('val', tokenizer=tokenizer)
          testset = FakeNewsDataset('test', tokenizer=tokenizer)
          print('trainset size:' ,trainset.__len__())
          print('valset size:',valset.__len__())
          print('testset size: ',testset.__len__())
```

trainset size: 28734 valset size: 7184 testset size: 8980

Sampling and Observing Tensors

```
In [193]: # select the first sample
         sample_idx = 0
         # take out the original text for comparison
         statement, label = trainset.df.iloc[sample_idx].values
         # Use the Dataset just created to take out the converted id tensors
         tokens_tensor, segments_tensor, label_tensor = trainset[sample_idx]
         # restore tokens_tensor to text
         tokens = tokenizer.convert_ids_to_tokens(tokens_tensor.tolist())
         combined_text = " ".join(tokens)
         # The difference before and after rendering, no response is a print. You can see
         print(f"""
         original statement:
         {statement}
         tokens:
         {tokens}
         label: {label}
         ______
         tokens tensor:
         {tokens_tensor}
         segments tensor:
         {segments_tensor}
         label tensor:
         {label_tensor}
         """)
         original statement:
          WATCH: Powerful Video PERFECTLY Explains How And Why Trump Is Manipulating Whi
         te People
         tokens:
         ['[CLS]', 'watch', ':', 'powerful', 'video', 'perfectly', 'explains', 'how', 'a
         nd', 'why', 'trump', 'is', 'manipulating', 'white', 'people', '[SEP]']
         label: 1
          ______
         tokens_tensor:
         tensor([ 101, 3422, 1024, 3928, 2678, 6669, 7607, 2129, 1998, 2339,
                  8398, 2003, 26242, 2317, 2111,
                                                    102])
         segments tensor:
```

```
label_tensor:
1
```

In [194]: # Reforming the Dataset to Fit the Model

```
In [195]: from torch.utils.data import DataLoader
          from torch.nn.utils.rnn import pad sequence
          # The input to this function `samples` is a list where each element is
          # A sample returned by the just defined `FakeNewsDataset`, each containing 3 tens
          # - tokens tensor
          # - segments tensor
          # - label tensor
          # It will zero-pad the first two tensors and generate the masks tensors described
          def create mini batch(samples):
              tokens tensors = [s[0] \text{ for } s \text{ in } samples]
              segments tensors = [s[1] \text{ for } s \text{ in } samples]
          # test set has labels
              if samples[0][2] is not None:
                   label_ids = torch.stack([s[2] for s in samples])
              else:
                   label_ids = None
              # Zero Padding
              tokens tensors = pad sequence(tokens tensors, batch first=True)
              segments_tensors = pad_sequence(segments_tensors, batch_first=True)
          # attention masks, make tokens tensors not zero padding
              # positions are set to 1 to make BERT only focus on tokens in these positions
              masks tensors = torch.zeros(tokens tensors.shape, dtype=torch.long)
              masks tensors = masks tensors.masked fill(tokens tensors != 0, 1)
              return tokens tensors, segments tensors, masks tensors, label ids
          # Initialize a DataLoader that returns 16 training samples each time
          # Using `collate fn` to combine the list of samples into a mini-batch is the key
          BATCH SIZE = 16
          trainloader = DataLoader(trainset, batch size=BATCH SIZE, collate fn=create mini
          valloader = DataLoader(valset, batch size=BATCH SIZE, collate fn=create mini batch
          testloader = DataLoader(testset, batch size=BATCH SIZE,collate fn=create mini bat
```

```
In [196]: data = next(iter(trainloader))
          tokens_tensors, segments_tensors, masks_tensors, label_ids = data
          print(f"""
          tokens_tensors.shape = {tokens_tensors.shape}
          {tokens tensors}
          segments_tensors.shape = {segments_tensors.shape}
          {segments_tensors}
                                 = {masks_tensors.shape}
          masks_tensors.shape
          {masks_tensors}
                                 = {label ids.shape}
          label ids.shape
          {label_ids}
          """)
          tokens_tensors.shape = torch.Size([16, 36])
          tensor([[ 101,
                           3422, 1024,
                                         3928, 2678, 6669, 7607,
                                                                      2129,
          9,
                    8398,
                           2003, 26242, 2317, 2111,
                                                         102,
                                                                  0,
                                                                         0,
                                                                                0,
          0,
                       0,
                              0,
                                      0,
                                             0,
                                                    0,
                                                           0,
                                                                  0,
                                                                         0,
                                                                                0,
          0,
                                             0,
                                                           0],
                                                    0,
                                      0,
                                          2507,
                                                 4436,
                                                        2027,
                                                                             2053, 2453
                     101,
                           2327, 18145,
                                                               2024,
                                                                      7135,
          4,
                                          2006,
                                               3425,
                                                        8040, 22786,
                    2097,
                           2022,
                                 2864,
                                                                      1024,
                                                                             1523,
                                                                                    104
          5,
                    2165,
                           1037,
                                  2298,
                                          2012, 1996,
                                                        3189, 1998,
                                                                      1045,
                                                                             2471,
                                                                                    306
          2,
                                   2026,
                                                         102],
                     2041,
                            1997,
                                          3242,
                                                 1524,
                                                1055, 16950, 2863, 19164,
                                          1005,
                                                                                    201
                     101,
                           2148,
                                  3088,
                                                                             4311,
          0,
```

2003, 21168, 5057, 4277, 102,

0,

0,

0,

Model Construction

2436,

```
In [198]:
          from transformers import BertForSequenceClassification
          from IPython.display import display, clear_output
          PRETRAINED MODEL NAME = "bert-base-uncased"
          NUM_LABELS = 2
          model = BertForSequenceClassification.from pretrained(
              PRETRAINED_MODEL_NAME, num_labels=NUM_LABELS)
          clear_output()
          # high-level show modules in this model
          print("""
          name
                          module
          -----""")
          for name, module in model.named_children():
              if name == "bert":
                  for n, _ in module.named_children():
                      print(f"{name}:{n}")
              else:
                  print("{:16} {}".format(name, module))
```

name	module
bert:embeddings bert:encoder bert:pooler	
dropout	<pre>Dropout(p=0.1, inplace=False)</pre>
classifier	<pre>Linear(in_features=768, out_features=2, bias=True)</pre>

```
In [199]: model.config
Out[199]: BertConfig {
             "_name_or_path": "bert-base-uncased",
             "architectures": [
               "BertForMaskedLM"
             "attention_probs_dropout_prob": 0.1,
             "classifier dropout": null,
             "gradient checkpointing": false,
             "hidden_act": "gelu",
             "hidden dropout prob": 0.1,
             "hidden_size": 768,
             "initializer range": 0.02,
             "intermediate size": 3072,
             "layer_norm_eps": 1e-12,
             "max_position_embeddings": 512,
             "model_type": "bert",
             "num_attention_heads": 12,
             "num_hidden_layers": 12,
             "pad token id": 0,
             "position_embedding_type": "absolute",
             "transformers_version": "4.20.1",
             "type_vocab_size": 2,
             "use_cache": true,
             "vocab_size": 30522
          }
```

Fine-Tuning of BERT

```
In [*]: | # %%time
        from sklearn.metrics import accuracy score
        from tqdm.notebook import tqdm
        device = torch.device("cuda:0" if torch.cuda.is_available() else "cpu")
        print("device:", device)
        model = model.to(device)
        model.train()
        optimizer = torch.optim.Adam(model.parameters(), lr=1e-5)
        NUM EPOCHS = 3
        for epoch in range(NUM EPOCHS):
            train loss = 0.0
            train acc = 0.0
            loop = tqdm(trainloader)
            for batch_idx, data in enumerate(loop):
                tokens_tensors, segments_tensors, masks_tensors, labels = [t.to(device) f
                # zero the parameter gradient
                optimizer.zero_grad()
                outputs = model(input_ids=tokens_tensors,
                                 token type ids=segments tensors,
                                 attention mask=masks tensors,
                                 labels=labels)
                loss = outputs[0]
                loss.backward()
                optimizer.step()
                logits = outputs[1]
                 _, pred = torch.max(logits.data, 1)
                train_acc = accuracy_score(pred.cpu().tolist() , labels.cpu().tolist())
                # Record the current batch loss
                train loss += loss.item()
                # if batch idx == len(trainloader)-1:
                       _, acc = get_predictions(model, trainloader, compute_acc=True)
                loop.set description(f"Epoch [{epoch+1}/{NUM EPOCHS}]")
                 loop.set postfix(acc = train acc, loss = train loss)
        device: cpu
                                                        233/1796 [24:02<3:15:04, 7.49s/it, acc=1,
         Epoch [1/3]:
         13%
                                                        loss=53.51
```

In []: