

review-sentiment-analysis

November 8, 2023

1 Importing necessary libraries

```
[40]: %pip install transformers[sentencepiece]
```

```
Requirement already satisfied: transformers[sentencepiece] in
/usr/local/lib/python3.10/dist-packages (4.35.0)
Requirement already satisfied: filelock in /usr/local/lib/python3.10/dist-
packages (from transformers[sentencepiece]) (3.13.1)
Requirement already satisfied: huggingface-hub<1.0,>=0.16.4 in
/usr/local/lib/python3.10/dist-packages (from transformers[sentencepiece])
(0.17.3)
Requirement already satisfied: numpy>=1.17 in /usr/local/lib/python3.10/dist-
packages (from transformers[sentencepiece]) (1.23.5)
Requirement already satisfied: packaging>=20.0 in
/usr/local/lib/python3.10/dist-packages (from transformers[sentencepiece])
(23.2)
Requirement already satisfied: pyyaml>=5.1 in /usr/local/lib/python3.10/dist-
packages (from transformers[sentencepiece]) (6.0.1)
Requirement already satisfied: regex!=2019.12.17 in
/usr/local/lib/python3.10/dist-packages (from transformers[sentencepiece])
(2023.6.3)
Requirement already satisfied: requests in /usr/local/lib/python3.10/dist-
packages (from transformers[sentencepiece]) (2.31.0)
Requirement already satisfied: tokenizers<0.15,>=0.14 in
/usr/local/lib/python3.10/dist-packages (from transformers[sentencepiece])
(0.14.1)
Requirement already satisfied: safetensors>=0.3.1 in
/usr/local/lib/python3.10/dist-packages (from transformers[sentencepiece])
(0.4.0)
Requirement already satisfied: tqdm>=4.27 in /usr/local/lib/python3.10/dist-
packages (from transformers[sentencepiece]) (4.66.1)
Requirement already satisfied: sentencepiece!=0.1.92,>=0.1.91 in
/usr/local/lib/python3.10/dist-packages (from transformers[sentencepiece])
(0.1.99)
Requirement already satisfied: protobuf in /usr/local/lib/python3.10/dist-
packages (from transformers[sentencepiece]) (3.20.3)
Requirement already satisfied: fsspec in /usr/local/lib/python3.10/dist-packages
(from huggingface-hub<1.0,>=0.16.4->transformers[sentencepiece]) (2023.6.0)
```

Requirement already satisfied: typing-extensions>=3.7.4.3 in /usr/local/lib/python3.10/dist-packages (from huggingface-hub<1.0,>=0.16.4->transformers[sentencepiece]) (4.5.0)
Requirement already satisfied: charset-normalizer<4,>=2 in /usr/local/lib/python3.10/dist-packages (from requests->transformers[sentencepiece]) (3.3.2)
Requirement already satisfied: idna<4,>=2.5 in /usr/local/lib/python3.10/dist-packages (from requests->transformers[sentencepiece]) (3.4)
Requirement already satisfied: urllib3<3,>=1.21.1 in /usr/local/lib/python3.10/dist-packages (from requests->transformers[sentencepiece]) (2.0.7)
Requirement already satisfied: certifi>=2017.4.17 in /usr/local/lib/python3.10/dist-packages (from requests->transformers[sentencepiece]) (2023.7.22)

```
[41]: import pandas as pd
      from sklearn.model_selection import train_test_split
      from transformers import BertTokenizer, BertForSequenceClassification, AdamW
      import torch
      from torch.utils.data import DataLoader, TensorDataset
      from sklearn.metrics import accuracy_score, classification_report, \
          ↪confusion_matrix
      import matplotlib.pyplot as plt
      import seaborn as sns
      from wordcloud import WordCloud
      from tqdm.notebook import tqdm
```

```
[42]: import warnings
      warnings.filterwarnings("ignore")
```

2 Creating a Sample Dataset for the task

```
[43]: data = {
      'text': [
          "This product is amazing! I love it.",
          "The worst shopping experience ever. I'm very disappointed.",
          "Not bad, but could be better.",
          "Fast and efficient service. Highly recommended.",
          "Terrible quality and horrible customer service.",
          "Average product, nothing exceptional.",
          "Outstanding quality and exceptional customer support.",
          "I'm satisfied with my purchase. It met my expectations.",
          "I regret buying this. Such a waste of money.",
          "Good value for the price.",
          "I couldn't be happier with my purchase. It's perfect!",
          "Poor customer service and slow shipping.",
```

```

        "It's an okay product. Not great, not terrible.",
        "Prompt delivery and great product quality.",
        "The item was defective and the return process was a nightmare."
    ],
    'sentiment': ['positive', 'negative', 'neutral', 'positive', 'negative',
↪ 'neutral', 'positive', 'positive', 'negative', 'positive', 'positive',
↪ 'negative', 'positive', 'negative', 'negative']
}

```

3 Exploratory Data Analysis

3.1 Converting data into a proper dataframe

```
[44]: df = pd.read_csv('/content/reviews.csv', error_bad_lines=False, engine="python")
df.head()
```

```
[44]:
```

	Unnamed: 0	Clothing ID	Age	Title \
0	0	767	33	NaN
1	1	1080	34	NaN
2	2	1077	60	Some major design flaws
3	3	1049	50	My favorite buy!
4	4	847	47	Flattering shirt

	Review Text	Rating	Recommended IND \
0	Absolutely wonderful - silky and sexy and comf...	4	1
1	Love this dress! it's sooo pretty. i happene...	5	1
2	I had such high hopes for this dress and reall...	3	0
3	I love, love, love this jumpsuit. it's fun, fl...	5	1
4	This shirt is very flattering to all due to th...	5	1

	Positive Feedback Count	Division Name	Department Name	Class Name
0	0	Initmates	Intimate	Intimates
1	4	General	Dresses	Dresses
2	0	General	Dresses	Dresses
3	0	General Petite	Bottoms	Pants
4	6	General	Tops	Blouses

```
[45]: df = df[['Title', 'Review Text', 'Rating']]
df.head()
```

```
[45]:
```

	Title	Review Text \
0	NaN	Absolutely wonderful - silky and sexy and comf...
1	NaN	Love this dress! it's sooo pretty. i happene...
2	Some major design flaws	I had such high hopes for this dress and reall...
3	My favorite buy!	I love, love, love this jumpsuit. it's fun, fl...
4	Flattering shirt	This shirt is very flattering to all due to th...

	Rating
0	4
1	5
2	3
3	5
4	5

```
[46]: df.Title.fillna("", inplace=True)
df['Review Text'].fillna("", inplace=True)
df.head()
```

```
[46]:
```

	Title	Review Text \
0		Absolutely wonderful - silky and sexy and comf...
1		Love this dress! it's sooo pretty. i happene...
2	Some major design flaws	I had such high hopes for this dress and reall...
3	My favorite buy!	I love, love, love this jumpsuit. it's fun, fl...
4	Flattering shirt	This shirt is very flattering to all due to th...

	Rating
0	4
1	5
2	3
3	5
4	5

```
[47]: df['text'] = df['Title'] + ' -- ' + df['Review Text']
df.head()
```

```
[47]:
```

	Title	Review Text \
0		Absolutely wonderful - silky and sexy and comf...
1		Love this dress! it's sooo pretty. i happene...
2	Some major design flaws	I had such high hopes for this dress and reall...
3	My favorite buy!	I love, love, love this jumpsuit. it's fun, fl...
4	Flattering shirt	This shirt is very flattering to all due to th...

	Rating	text
0	4	-- Absolutely wonderful - silky and sexy and ...
1	5	-- Love this dress! it's sooo pretty. i hap...
2	3	Some major design flaws -- I had such high hop...
3	5	My favorite buy! -- I love, love, love this ju...
4	5	Flattering shirt -- This shirt is very flatter...

```
[48]: sentiment_mapping = {
    1: 'very negative',
    2: 'negative',
    3: 'neutral',
```

```

    4: 'positive',
    5: 'very positive'
}
df['sentiment'] = df['Rating'].map(sentiment_mapping)

df.head()

```

```

[48]:

```

	Title	Review Text \
0	Absolutely wonderful - silky and sexy and comf...	
1	Love this dress! it's sooo pretty. i happene...	
2	Some major design flaws I had such high hopes for this dress and reall...	
3	My favorite buy! I love, love, love this jumpsuit. it's fun, fl...	
4	Flattering shirt This shirt is very flattering to all due to th...	

	Rating	text	sentiment
0	4	-- Absolutely wonderful - silky and sexy and ...	positive
1	5	-- Love this dress! it's sooo pretty. i hap...	very positive
2	3	Some major design flaws -- I had such high hop...	neutral
3	5	My favorite buy! -- I love, love, love this ju...	very positive
4	5	Flattering shirt -- This shirt is very flatter...	very positive

```

[49]: df.isna().sum()

```

```

[49]: Title      0
      Review Text  0
      Rating      0
      text        0
      sentiment   0
      dtype: int64

```

```

[50]: df = df[['text', 'sentiment']]
      df.head()

```

```

[50]:

```

	text	sentiment
0	-- Absolutely wonderful - silky and sexy and ...	positive
1	-- Love this dress! it's sooo pretty. i hap...	very positive
2	Some major design flaws -- I had such high hop...	neutral
3	My favorite buy! -- I love, love, love this ju...	very positive
4	Flattering shirt -- This shirt is very flatter...	very positive

3.2 Getting full information of the dataframe

```

[51]: df.info()

```

```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 23486 entries, 0 to 23485
Data columns (total 2 columns):

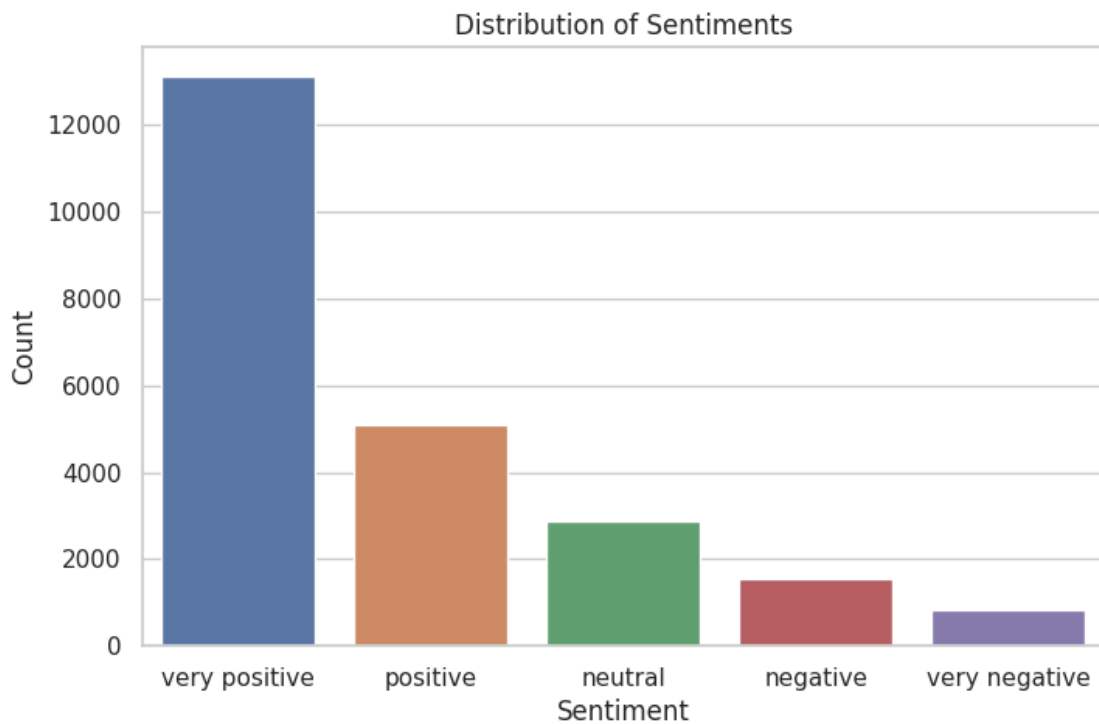
```

#	Column	Non-Null Count	Dtype
0	text	23486 non-null	object
1	sentiment	23486 non-null	object

dtypes: object(2)
memory usage: 367.1+ KB

3.3 Plotting the distribution of sentiments in our dataset

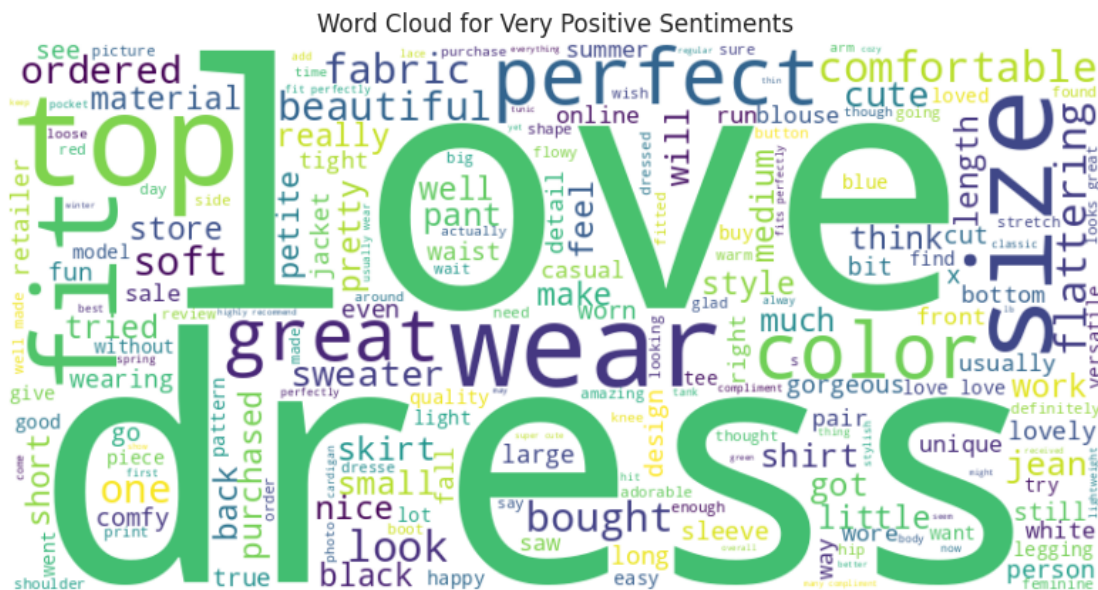
```
[52]: sns.set(style="whitegrid")
plt.figure(figsize=(8, 5))
sns.countplot(data=df, x='sentiment', order=['very positive', 'positive', 'neutral', 'negative', 'very negative'])
plt.title('Distribution of Sentiments')
plt.xlabel('Sentiment')
plt.ylabel('Count')
plt.show()
```



3.4 Creating word clouds for different types of reviews

```
[53]: very_positive_reviews = " ".join(df[df['sentiment'] == 'very positive']['text'])
wordcloud = WordCloud(width=800, height=400, background_color='white').
    generate(very_positive_reviews)

plt.figure(figsize=(10, 5))
plt.imshow(wordcloud, interpolation='bilinear')
plt.axis("off")
plt.title('Word Cloud for Very Positive Sentiments')
plt.show()
```



```
[54]: positive_reviews = " ".join(df[df['sentiment'] == 'positive']['text'])
wordcloud = WordCloud(width=800, height=400, background_color='white').
    generate(positive_reviews)

plt.figure(figsize=(10, 5))
plt.imshow(wordcloud, interpolation='bilinear')
plt.axis("off")
plt.title('Word Cloud for Positive Sentiments')
plt.show()
```

Word Cloud for Positive Sentiments

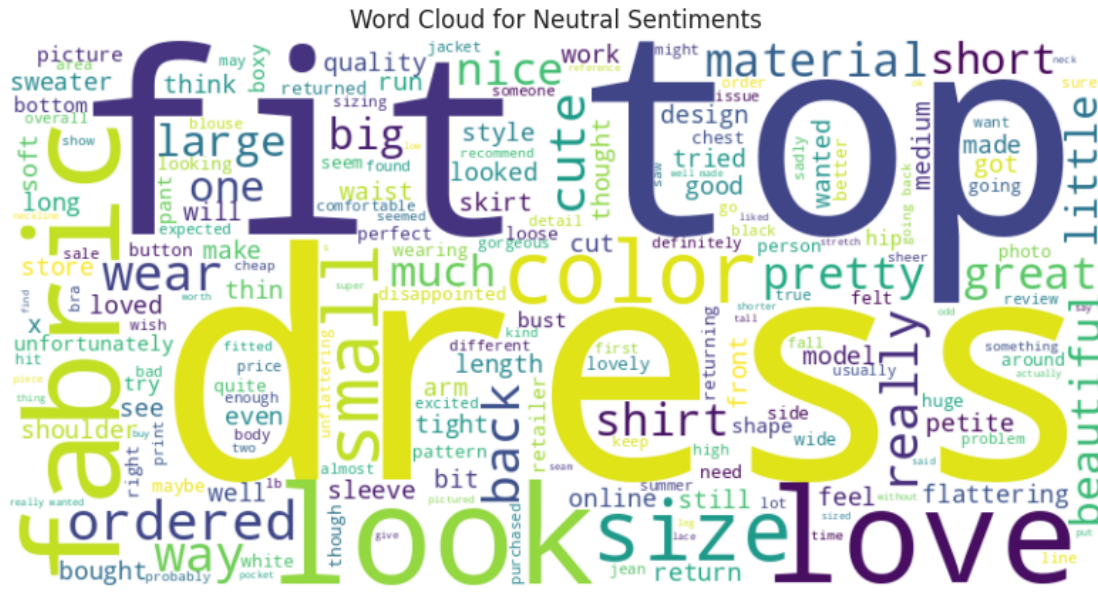


```
[55]: negative_reviews = " ".join(df[df['sentiment'] == 'negative']['text'])
wordcloud = WordCloud(width=800, height=400, background_color='white').
    generate(negative_reviews)

plt.figure(figsize=(10, 5))
plt.imshow(wordcloud, interpolation='bilinear')
plt.axis("off")
plt.title('Word Cloud for Negative Sentiments')
plt.show()
```

Word Cloud for Negative Sentiments





4 Splitting the dataset

```
[58]: grouped = df.groupby('sentiment')

# Initialize an empty list to store the samples
samples = []

# Sample one instance from each group
for group_name, group_data in grouped:
    sample = group_data.sample(n=50, random_state=42) # Sample 50 instance per class
    samples.append(sample)

subset_df = pd.concat(samples)
subset_df.reset_index(drop=True, inplace=True)
subset_df.head()
```

```
[58]: text sentiment
0 Snug and unflattering -- Would be flattering o... negative
1 The sleeves... -- I was aware of the split s... negative
2 Huge - swallowed me whole -- I had high hopes ... negative
3 So baggy -- I grabbed this dress to try on in ... negative
4 Not for tall ladies -- This sweater is a cute ... negative
```

```
[59]: X = subset_df['text']
      y = subset_df['sentiment']

[60]: X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2,
      ↪random_state=42)
```

4.1.1 Mapping sentiment into numeric labels

```
[61]: label_mapping = {'very positive':4, 'positive': 3, 'neutral': 2, 'negative': 1,
      ↪'very negative':0}
      y_train = y_train.map(label_mapping)
      y_test = y_test.map(label_mapping)
```

5 Model Training

5.1 Initializing Model

```
[62]: model_name = "bert-base-uncased"
      tokenizer = BertTokenizer.from_pretrained(model_name)
      model = BertForSequenceClassification.from_pretrained(model_name, num_labels=5)
```

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.

5.2 Getting encodings for the reviews

```
[63]: train_encodings = tokenizer(X_train.tolist(), truncation=True, padding=True,
      ↪return_tensors='pt', max_length=64)
      test_encodings = tokenizer(X_test.tolist(), truncation=True, padding=True,
      ↪return_tensors='pt', max_length=64)
```

5.3 Creating dataset and Data loaders

```
[64]: train_dataset = TensorDataset(train_encodings['input_ids'],
      ↪train_encodings['attention_mask'], torch.tensor(y_train.tolist()))
      test_dataset = TensorDataset(test_encodings['input_ids'],
      ↪test_encodings['attention_mask'], torch.tensor(y_test.tolist()))
      train_loader = DataLoader(train_dataset, batch_size=3, shuffle=True)
      test_loader = DataLoader(test_dataset, batch_size=3, shuffle=False)
```

5.4 Initializing optimizer and checking for cuda compatibility

```
[65]: optimizer = AdamW(model.parameters(), lr=1e-5)
device = torch.device("cuda" if torch.cuda.is_available() else "cpu")
```

5.5 Model is ready to train

```
[66]: model.to(device)
model.train()
```

```
[66]: 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): BertSelfAttention(
              (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=5, bias=True)
    )

```

5.6 Initializing list to store training losses through epochs

```
[67]: train_losses = []
```

5.7 Training

```
[68]: for epoch in range(3):
        for batch in tqdm(train_loader, desc=f'Epoch {epoch+1}'):
            input_ids, attention_mask, labels = batch
            input_ids, attention_mask, labels = input_ids.to(device),
            ↪attention_mask.to(device), labels.to(device)

            optimizer.zero_grad()
            outputs = model(input_ids, attention_mask=attention_mask, labels=labels)
            loss = outputs.loss
            loss.backward()
            optimizer.step()

            train_losses.append(loss.item())

```

```
Epoch 1:  0%|          | 0/67 [00:00<?, ?it/s]
```

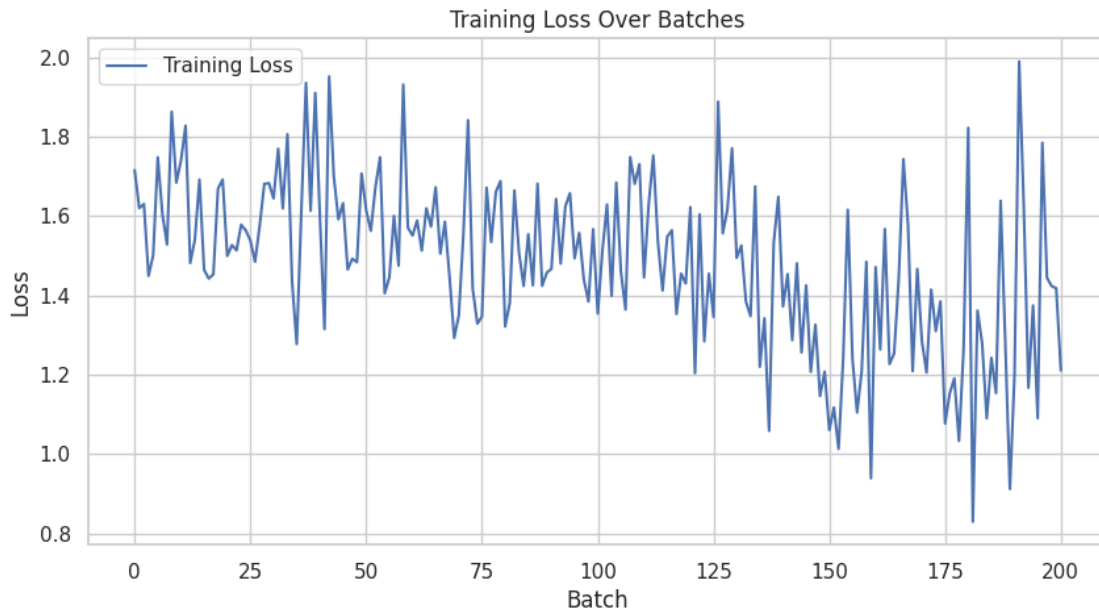
```
Epoch 2:  0%|          | 0/67 [00:00<?, ?it/s]
```

```
Epoch 3:  0%|          | 0/67 [00:00<?, ?it/s]
```

5.8 Plotting training loss values

```
[69]: plt.figure(figsize=(10, 5))
        plt.plot(range(len(train_losses)), train_losses, label="Training Loss")
        plt.xlabel("Batch")
        plt.ylabel("Loss")
        plt.legend()
        plt.title("Training Loss Over Batches")
        plt.show()

```



6 Model Evaluation

```
[70]: model.eval()
all_preds = []
with torch.no_grad():
    for batch in test_loader:
        input_ids, attention_mask, labels = batch
        input_ids, attention_mask, labels = input_ids.to(device),
        ↪attention_mask.to(device), labels.to(device)

        outputs = model(input_ids, attention_mask=attention_mask)
        logits = outputs.logits
        preds = torch.argmax(logits, dim=1).cpu().numpy()
        all_preds.extend(preds)
```

6.1 Displaying Model Results

```
[71]: accuracy = accuracy_score(y_test, all_preds)
report = classification_report(y_test, all_preds)

print(f"Accuracy: {accuracy}\n\n")
print("Classification Report:")
print(report)
```

Accuracy: 0.46

Classification Report:

	precision	recall	f1-score	support
0	0.42	0.73	0.53	11
1	0.45	0.38	0.42	13
2	0.17	0.14	0.15	7
3	1.00	0.10	0.18	10
4	0.62	0.89	0.73	9
accuracy			0.46	50
macro avg	0.53	0.45	0.40	50
weighted avg	0.54	0.46	0.41	50

```
[75]: confusion = confusion_matrix(y_test, all_preds)

plt.figure(figsize=(8, 6))
sns.heatmap(confusion, annot=True, fmt='d', cmap='Blues', cbar=False,
            square=True,
            xticklabels=['very negative', 'negative', 'neutral', 'positive',
            ↪ 'very positive'],
            yticklabels=['very negative', 'negative', 'neutral', 'positive',
            ↪ 'very positive'])

plt.xlabel('Predicted')
plt.ylabel('True')
plt.title('Confusion Matrix')

plt.show()
```

