```
!pip uninstall torch torchvision torchaudio torchtext torchdata -y
!pip cache purge
!pip install torch==2.0.1 torchvision torchaudio --index-url https://download.pytorch.org/whl/cu118
!pip install torchtext torchdata==0.6.1
import torch
import torchtext
print(f"PyTorch Version: {torch.__version__}}")
print(f"TorchText Version: {torchtext. version
print(f"CUDA Available: {torch.cuda.is_available()}")
                                                   - 4.4/4.4 MB 87.7 MB/s eta 0:00:00
    Requirement already satisfied: MarkupSafe>=2.0 in /usr/local/lib/python3.11/dist-packages (from jinja2->torch==2.0.1) (3
    Requirement already satisfied: charset-normalizer<4,>=2 in /usr/local/lib/python3.11/dist-packages (from requests->torch
    Requirement already satisfied: idna<4,>=2.5 in /usr/local/lib/python3.11/dist-packages (from requests->torchvision) (3.1
    Requirement already satisfied: urllib3<3,>=1.21.1 in /usr/local/lib/python3.11/dist-packages (from requests->torchvision
    Requirement already satisfied: certifi>=2017.4.17 in /usr/local/lib/python3.11/dist-packages (from requests->torchvision
    Requirement already satisfied: mpmath<1.4,>=1.1.0 in /usr/local/lib/python3.11/dist-packages (from sympy->torch==2.0.1)
    Building wheels for collected packages: lit
      Building wheel for lit (setup.py) ... done
Created wheel for lit: filename=lit-15.0.7-py3-none-any.whl size=89991 sha256=f406022657e5212531194899f8f46da2fd2d66de
      Stored in directory: /root/.cache/pip/wheels/fc/5d/45/34fe9945d5e45e261134e72284395be36c2d4828af38e2b0fe
    Successfully built lit
    Installing collected packages: lit, triton, torch, torchvision, torchaudio
      Attempting uninstall: triton
         Found existing installation: triton 3.2.0
         Uninstalling triton-3.2.0:
           Successfully uninstalled triton-3.2.0
    Successfully installed lit-15.0.7 torch-2.0.1+cu118 torchaudio-2.0.2+cu118 torchvision-0.15.2+cu118 triton-2.0.0
    Collecting torchtext
      Downloading torchtext-0.18.0-cp311-cp311-manylinux1_x86_64.whl.metadata (7.9 kB)
    Collecting torchdata==0.6.1
    Downloading torchdata-0.6.1-cp311-cp311-manylinux_2_17_x86_64.manylinux2014_x86_64.whl.metadata (13 kB)
Requirement already satisfied: urllib3>=1.25 in /usr/local/lib/python3.11/dist-packages (from torchdata==0.6.1) (2.3.0)
    Requirement already satisfied: requests in /usr/local/lib/python3.11/dist-packages (from torchdata==0.6.1) (2.32.3)
    Requirement already satisfied: torch==2.0.1 in /usr/local/lib/python3.11/dist-packages (from torchdata==0.6.1) (2.0.1+cu
    Requirement already satisfied: filelock in /usr/local/lib/python3.11/dist-packages (from torch==2.0.1->torchdata==0.6.1)
    Requirement already satisfied: typing-extensions in /usr/local/lib/python3.11/dist-packages (from torch==2.0.1->torchdat
    Requirement already satisfied: sympy in /usr/local/lib/python3.11/dist-packages (from torch==2.0.1->torchdata==0.6.1) (1
    Requirement already satisfied: networkx in /usr/local/lib/python3.11/dist-packages (from torch==2.0.1->torchdata==0.6.1) Requirement already satisfied: jinja2 in /usr/local/lib/python3.11/dist-packages (from torch==2.0.1->torchdata==0.6.1) (
    Requirement already satisfied: triton==2.0.0 in /usr/local/lib/python3.11/dist-packages (from torch==2.0.1->torchdata==0
    Requirement already satisfied: cmake in /usr/local/lib/python3.11/dist-packages (from triton==2.0.0->torch==2.0.1->torch
    Requirement already satisfied: lit in /usr/local/lib/python3.11/dist-packages (from triton==2.0.0->torch==2.0.1->torchda
    Requirement already satisfied: tqdm in /usr/local/lib/python3.11/dist-packages (from torchtext) (4.67.1)
    INFO: pip is looking at multiple versions of torchtext to determine which version is compatible with other requirements.
    Collecting torchtext
      Downloading torchtext-0.17.2-cp311-cp311-manylinux1_x86_64.whl.metadata (7.9 kB)
      Downloading torchtext-0.17.1-cp311-manylinux1_x86_64.whl.metadata (7.6 kB)
      Downloading torchtext-0.17.0-cp311-cp311-manylinux1_x86_64.whl.metadata (7.6 kB)
      Downloading torchtext-0.16.2-cp311-cp311-manylinux1_x86_64.whl.metadata (7.5 kB)
      Downloading torchtext-0.16.1-cp311-cp311-manylinux1_x86_64.whl.metadata (7.5 kB)
      Downloading torchtext-0.16.0-cp311-cp311-manylinux1_x86_64.whl.metadata (7.5 kB)
      Downloading torchtext-0.15.2-cp311-cp311-manylinux1_x86_64.whl.metadata (7.4 kB)
    Requirement already satisfied: numpy in /usr/local/lib/python3.11/dist-packages (from torchtext) (2.0.2)
    Requirement already satisfied: charset-normalizer<4,>=2 in /usr/local/lib/python3.11/dist-packages (from requests->torch
    Requirement already satisfied: idna<4,>=2.5 in /usr/local/lib/python3.11/dist-packages (from requests->torchdata==0.6.1)
    Requirement already satisfied: certifi>=2017.4.17 in /usr/local/lib/python3.11/dist-packages (from requests->torchdata==
    Requirement already satisfied: MarkupSafe>=2.0 in /usr/local/lib/python3.11/dist-packages (from jinja2->torch==2.0.1->to
    Requirement already satisfied: mpmath<1.4,>=1.1.0 in /usr/local/lib/python3.11/dist-packages (from sympy->torch==2.0.1->
    Downloading torchdata-0.6.1-cp311-cp311-manylinux_2_17_x86_64.manylinux2014_x86_64.whl (4.6 MB)
                                                  4.6/4.6 MB 94.9 MB/s eta 0:00:00
    Downloading torchtext-0.15.2-cp311-cp311-manylinux1_x86_64.whl (2.0 MB)
                                                   2.0/2.0 MB 67.2 MB/s eta 0:00:00
    Installing collected packages: torchdata, torchtext
    Successfully installed torchdata-0.6.1 torchtext-0.15.2
    PyTorch Version: 2.0.1+cu118
    TorchText Version: 0.15.2+cpu
    CUDA Available: True
import kagglehub
# Download latest version
path = kagglehub.dataset_download("kazanova/sentiment140")
print("Path to dataset files:", path)
   Downloading from https://www.kaggle.com/api/v1/datasets/download/kazanova/sentiment140?dataset version number=2...
                   80.9M/80.9M [00:04<00:00, 17.2MB/s]Extracting files...
    Path to dataset files: /root/.cache/kagglehub/datasets/kazanova/sentiment140/versions/2
```

https://colab.research.google.com/drive/14evfa-F9OFmhQwU6RnkAB39p0IVGDz6w#scrollTo=4rvfWQ9iKaau&printMode=true

```
import numpy as np
import pandas as pd
import re
import random
from sklearn.model_selection import train_test_split
import torch
import torch.nn as nn
import torch.nn.functional as F
import torch.optim as optim
from torch.utils.data import DataLoader, Dataset
RANDOM\_SEED = 3539
torch.manual_seed(RANDOM_SEED)
VOCABULARY_SIZE = 20000
LEARNING RATE = 1e-4
BATCH SIZE = 128
NUM EPOCHS = 10
DEVICE = torch.device("cuda" if torch.cuda.is_available() else "cpu")
EMBEDDING_DIM = 128
HIDDEN_DIM = 256
OUTPUT_DIM = 2
     A module that was compiled using NumPy 1.x cannot be run in
    NumPy 2.0.2 as it may crash. To support both 1.x and 2.x
     versions of NumPy, modules must be compiled with NumPy 2.0.
     Some module may need to rebuild instead e.g. with 'pybind11>=2.12'.
    If you are a user of the module, the easiest solution will be to downgrade to 'numpy<2' or try to upgrade the affected module.
    We expect that some modules will need time to support NumPy 2.
    Traceback (most recent call last): File "<frozen runpy>", line 198, in _run_module_as_main File "<frozen runpy>", line 88, in _run_code
       File "/usr/local/lib/python3.11/dist-packages/colab_kernel_launcher.py", line 37, in <module>
         ColabKernelApp.launch_instance()
       File "/usr/local/lib/python3.11/dist-packages/traitlets/config/application.py", line 992, in launch_instance
         app.start()
       File "/usr/local/lib/python3.11/dist-packages/ipykernel/kernelapp.py", line 712, in start
         self.io_loop.start()
       File "/usr/local/lib/python3.11/dist-packages/tornado/platform/asyncio.py", line 205, in start
         self.asyncio_loop.run_forever()
       File "/usr/lib/python3.11/asyncio/base_events.py", line 608, in run_forever
         self._run_once()
       File "/usr/lib/python3.11/asyncio/base_events.py", line 1936, in _run_once
         handle._run()
       File "/usr/lib/python3.11/asyncio/events.py", line 84, in _run
         self._context.run(self._callback, *self._args)
       File "/usr/local/lib/python3.11/dist-packages/ipykernel/kernelbase.py", line 510, in dispatch_queue
         await self.process_one()
       File "/usr/local/lib/python3.11/dist-packages/ipykernel/kernelbase.py", line 499, in process_one
         await dispatch(*args)
       File "/usr/local/lib/python3.11/dist-packages/ipykernel/kernelbase.py", line 406, in dispatch_shell
         await result
       File "/usr/local/lib/python3.11/dist-packages/ipykernel/kernelbase.py", line 730, in execute_request
         reply_content = await reply_content
       File "/usr/local/lib/python3.11/dist-packages/ipykernel/ipkernel.py", line 383, in do_execute
       res = shell.run_cell(
File "/usr/local/lib/python3.11/dist-packages/ipykernel/zmqshell.py", line 528, in run_cell
         return super().run_cell(*args, **kwargs)
       File "/usr/local/lib/python3.11/dist-packages/IPython/core/interactiveshell.py", line 2975, in run_cell
         result = self._run_cell(
       File "/usr/local/lib/python3.11/dist-packages/IPython/core/interactiveshell.py", line 3030, in _run_cell
         return runner(coro)
       File "/usr/local/lib/python3.11/dist-packages/IPython/core/async_helpers.py", line 78, in _pseudo_sync_runner
         coro.send(None)
       File "/usr/local/lib/python3.11/dist-packages/IPython/core/interactiveshell.py", line 3257, in run_cell_async
       has_raised = await self.run_ast_nodes(code_ast.body, cell_name, File "/usr/local/lib/python3.11/dist-packages/IPython/core/interactiveshell.py", line 3473, in run_ast_nodes
         if (await self.run_code(code, result, async_=asy)):
       File "/usr/local/lib/python3.11/dist-packages/IPython/core/interactiveshell.py", line 3553, in run_code
       exec(code_obj, self.user_global_ns, self.user_ns)
File "<ipython-input-3-ee5ea4139b6a>", line 16, in <cell line: 0>
         torch.manual_seed(RANDOM_SEED)
       File "/usr/local/lib/python3.11/dist-packages/torch/random.py", line 46, in manual seed
         return default_generator.manual_seed(seed)
     /usr/local/lib/python3.11/dist-packages/torch/random.py:46: UserWarning: Failed to initialize NumPy: _ARRAY_API not foun
       return default_generator.manual_seed(seed)
```

dataset_path = "/root/.cache/kagglehub/datasets/kazanova/sentiment140/versions/2/training.1600000.processed.noemoticon.csv"

```
df = pd.read_csv(dataset_path, encoding="ISO-8859-1", header=None)
df.columns = ["sentiment", "id", "date", "guery", "user", "text"]
print(df.head())
print(df.shape)
₹
       sentiment
                                                          date
                                                                   query
                   1467810369
                               Mon Apr 06 22:19:45 PDT 2009
                                                                NO_QUERY
                               Mon Apr 06 22:19:49 PDT 2009
    1
                                                                NO OUERY
                   1467810917
                                Mon Apr 06 22:19:53 PDT 2009
                                                                NO_QUERY
                   1467811184 Mon Apr 06 22:19:57 PDT 2009
    3
    4
                   1467811193 Mon Apr 06 22:19:57 PDT 2009
                                                                NO QUERY
                   user
    0
       _TheSpecialOne_
                          @switchfoot <a href="http://twitpic.com/2y1zl">http://twitpic.com/2y1zl</a> - Awww, t...
    1
          scotthamilton
                          is upset that he can't update his Facebook by ...
                         @Kenichan I dived many times for the ball. Man...
               mattvcus
    3
                ElleCTF
                           my whole body feels itchy and like its on fire
                 Karoli @nationwideclass no, it's not behaving at all....
    (1600000, 6)
```

*Data Prepocessing *

```
# we only want label and text
df_ = df[["sentiment", "text"]]
#there is only positive and negative tone
print('type of tone',df_['sentiment'].unique())
#we turn 4 into 1 to make it binary
df_.loc[:, "sentiment"] = df_["sentiment"].replace(4, 1)
print(len(df_))
df_sample=df_
# df_sample = df_.sample(n=20000, random_state=RANDOM_SEED)
print(df_sample.head)
     type of tone [0 4]
     1600000
     <body><br/>hound method NDFrame.head of</br>
                                                    sentiment
                                                                                                                       text
                         0 @switchfoot <a href="http://twitpic.com/2y1zl">http://twitpic.com/2y1zl</a> - Awww, t...
0 is upset that he can't update his Facebook by ...
     1
     2
                             @Kenichan I dived many times for the ball. Man...
                               my whole body feels itchy and like its on fire
     3
     4
                         0
                             @nationwideclass no, it's not behaving at all....
     1599995
                             Just woke up. Having no school is the best fee...
     1599996
                            TheWDB.com - Very cool to hear old Walt interv...
     1599997
                             Are you ready for your MoJo Makeover? Ask me f...
     1599998
                             Happy 38th Birthday to my boo of alll time!!! ...
     1599999
                             happy #charitytuesday @theNSPCC @SparksCharity...
     [1600000 rows x 2 columns]>
#def text cleaning function
def clean_text(text):
     text = text.lower() # 转换为小写
    \texttt{text} = \texttt{re.sub}(\texttt{r"http}\texttt{S+|www}\texttt{S+|https}\texttt{S+", "", text, flags=re.MULTILINE}) \  \  \# \  \  \texttt{remove URL}
    text = re.sub(r'\@\w+|\#', '', text) # remove @ and # text = re.sub(r'\^a-zA-Z\s]', '', text) #only use words
    return text.strip()
df_sample['text'] = df_sample['text'].apply(clean_text)
df_sample.head()
```

```
<ipython-input-6-fad821d6abd7>:9: SettingWithCopyWarning:
     A value is trying to be set on a copy of a slice from a DataFrame.
     Try using .loc[row_indexer,col_indexer] = value instead
     See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-
       df_sample['text'] = df_sample['text'].apply(clean_text)
        sentiment
                                                        text
                                                                扁
     0
                 0 a thats a bummer you should got david carr o...
      1
                   is upset that he cant update his facebook by t...
     2
                 0 i dived many times for the ball managed to sav...
     3
                       my whole body feels itchy and like its on fire
                 0 no its not behaving at all im mad why am i her...
     4
import torchtext
from torchtext.data.utils import get_tokenizer
from torchtext.vocab import build_vocab_from_iterator
from collections import Counter
# use tokenizer from torchtext
tokenizer = get_tokenizer("basic_english")
df_sample["tokens"] = df_sample["text"].apply(tokenizer)
df_sample.head()
    <ipython-input-7-fe558fd85c28>:10: SettingWithCopyWarning:
     A value is trying to be set on a copy of a slice from a DataFrame.
     Try using .loc[row_indexer,col_indexer] = value instead
     See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-
       df_sample["tokens"] = df_sample["text"].apply(tokenizer)
        sentiment
                                                                                               tokens
                                                                                                         ▦
     n
                 0 a thats a bummer you should ggot david carr o... [a, thats, a, bummer, you, should a, got, david...
      1
                   is upset that he cant update his facebook by t...
                                                                [is, upset, that, he, cant, update, his, faceb...
     2
                 0 i dived many times for the ball managed to sav...
                                                               [i, dived, many, times, for, the, ball, manage...
     3
                       my whole body feels itchy and like its on fire
                                                                [mv, whole, body, feels, itchy, and, like, its...
      4
                   no its not behaving at all im mad why am i her...
                                                                [no, its, not, behaving, at, all, im, mad, why...
#create a counter for vocab
counter = Counter()
for tokens in df_sample["tokens"]:
    counter.update(tokens)
#vocab = Vocab(counter, specials=["<PAD>", "<UNK>"], max_size=VOCABULARY_SIZE)
#this not work becausethe version of torchtext, using build_vocab_from_iterator instead
def yield_tokens(data):
    for tokens in data:
        yield tokens
vocab = build_vocab_from_iterator(yield_tokens(df_sample["tokens"]), max_tokens=20000, specials=["<pad>", "<unk>"])
vocab.set_default_index(vocab["<unk>"])
def text_to_indices(text):
    return [vocab[token] for token in text]
df_sample["indices"] = df_sample["tokens"].apply(text_to_indices)
#train,test,valid split
train_texts, test_texts, train_labels, test_labels = train_test_split(
    df_sample["indices"], df_sample["sentiment"], test_size=0.2, random_state=RANDOM_SEED
train_texts, val_texts, train_labels, val_labels = train_test_split(
    train_texts, train_labels, test_size=0.2, random_state=RANDOM_SEED
```

```
<ipython-input-8-ffc8ead6565d>:21: SettingWithCopyWarning:
    A value is trying to be set on a copy of a slice from a DataFrame.
    Try using .loc[row_indexer,col_indexer] = value instead
    See the caveats in the documentation: https://pandas.pydata.org/pandas-docs/stable/user_guide/indexing.html#returning-a-
      df_sample["indices"] = df_sample["tokens"].apply(text_to_indices)
#build dataset class
class SentimentDataset(Dataset):
   def __init__(self, texts, labels):
        Custom dataset for tweets dataset.
           texts: List of tokenized text sequences (e.g., token indices)
            labels: List of sentiment labels (0 or 1).
        self.texts = list(texts) # Convert to list to prevent indexing issues
        self.labels = list(labels) # Convert to list to prevent indexing issues
         _len__(self):
        """Return dataset size"""
        return len(self.texts)
   def __getitem__(self, idx):
        Get item by index.
        Returns:
            text_tensor: Tensor of token indices.
            label_tensor: Tensor of the corresponding label.
        text_tensor = torch.tensor(self.texts[idx], dtype=torch.long)
        label_tensor = torch.tensor(self.labels[idx], dtype=torch.long)
        return text_tensor, label_tensor
train_dataset = SentimentDataset(train_texts, train_labels)
val_dataset = SentimentDataset(val_texts, val_labels)
test_dataset = SentimentDataset(test_texts, test_labels)
#why we do not directly use dataloader?
# NLP tasks, our text data is usually variable-length sequences, but PyTorch's DataLoader does not perform padding by defaul
#which can cause shape mismatch errors during model training.
# train_loader = DataLoader(train_dataset, batch_size=64, shuffle=True)
# val_loader = DataLoader(val_dataset, batch_size=64, shuffle=False)
# test_loader = DataLoader(test_dataset, batch_size=64, shuffle=False)
from torch.nn.utils.rnn import pad_sequence
def collate_fn(batch):
    Custom collate function for handling batches of different lengths.
       batch: List of (text, label) tuples.
    Returns:
        texts: Padded tensor of token indices.
        labels: Tensor of sentiment labels.
    texts, labels = zip(*batch)
    texts = pad_sequence(texts, batch_first=True, padding_value=vocab["<pad>"])
    labels = torch.tensor(labels, dtype=torch.long)
    return texts, labels
PAD_IDX = vocab["<pad>"]
train_loader = DataLoader(train_dataset, batch_size=BATCH_SIZE, shuffle=True, collate_fn=collate_fn)
val_loader = DataLoader(val_dataset, batch_size=BATCH_SIZE, shuffle=False, collate_fn=collate_fn)
test_loader = DataLoader(test_dataset, batch_size=BATCH_SIZE, shuffle=False, collate_fn=collate_fn)
print('Train')
for batch in train_loader:
   text_batch, label_batch = batch # Unpack the tuple
    print(f'Text matrix size: {text_batch.size()}') # (batch_size, seq_length)
    print(f'Target vector size: {label_batch.size()}') # (batch_size,)
```

```
print('\nValid:')
for batch in val loader:
   text_batch, label_batch = batch # Unpack the tuple
   print(f'Text matrix size: {text_batch.size()}') # (batch_size, seq_length)
   print(f'Target vector size: {label_batch.size()}') # (batch_size,)
   break
print('\nTest:')
for batch in test_loader:
   text_batch, label_batch = batch # Unpack the tuple
   print(f'Text matrix size: {text_batch.size()}') # (batch_size, seq_length)
   print(f'Target vector size: {label_batch.size()}') # (batch_size,)
   break
→ Train
    Text matrix size: torch.Size([128, 29])
    Target vector size: torch.Size([128])
    Text matrix size: torch.Size([128, 30])
    Target vector size: torch.Size([128])
    Text matrix size: torch.Size([128, 30])
    Target vector size: torch.Size([128])
```

Model

```
import torch.nn.utils.rnn as rnn_utils # ☑ Import this at the top of your script
class RNN(nn.Module):
    def __init__(self, input_dim, embedding_dim, hidden_dim, output_dim, padding_idx):
        super().__init__()
        self.embedding = nn.Embedding(input_dim, embedding_dim, padding_idx=padding_idx)
        self.layer_norm = nn.LayerNorm(embedding_dim)
        self.rnn = nn.LSTM(embedding_dim, hidden_dim, num_layers=2,
                           bidirectional=True, batch_first=True, dropout=0.6)
        self.batch_norm = nn.BatchNorm1d(hidden_dim * 2)
        self.fc = nn.Linear(hidden_dim * 2, output_dim)
        self.dropout = nn.Dropout(0.7) #set high dropout to avoid overfitting
    def forward(self, text, text_length):
        embedded = self.embedding(text)
        embedded = self.layer_norm(embedded)
        text_length = text_length.cpu()
        packed_embedded = rnn_utils.pack_padded_sequence(embedded, text_length, batch_first=True, enforce_sorted=False)
        packed_output, (hidden, cell) = self.rnn(packed_embedded)
        hidden = torch.cat((hidden[-2,:,:], hidden[-1,:,:]), dim=1)
       hidden = self.batch_norm(hidden)
       hidden = self.dropout(hidden)
        out = self.fc(hidden)
        return out
from torch.optim import AdamW
INPUT_DIM = len(vocab)
torch.manual seed(RANDOM SEED)
model = RNN(INPUT_DIM, EMBEDDING_DIM, HIDDEN_DIM, OUTPUT_DIM,PAD_IDX)
model = model.to(DEVICE)
optimizer = AdamW(model.parameters(), lr=LEARNING_RATE, weight_decay=1e-2)
```

Training

```
def compute_binary_accuracy(model, data_loader, device):
    model.eval()
    correct_pred, num_examples = 0, 0
    with torch.no_grad():
        for text, labels in data_loader:
            text_lengths = text.ne(PAD_IDX).sum(dim=1) # Efficient length calculation
```

```
valid_indices = text_lengths > 0 #ensure length is not zero
            if valid_indices.sum() == 0:
                continue
            text, labels = text[valid_indices], labels[valid_indices]
            text_lengths = text_lengths[valid_indices]
            text, labels = text.to(device), labels.to(device)
            text_lengths = text_lengths.cpu()
            logits = model(text, text_lengths)
            predicted_labels = torch.argmax(logits, dim=1)
            num_examples += labels.size(0)
            correct_pred += (predicted_labels == labels).sum().item()
    return correct_pred / num_examples * 100 # if num_examples > 0 else 0
import time
start_time = time.time()
for epoch in range(NUM_EPOCHS):
   model.train()
    for batch_idx, (text, labels) in enumerate(train_loader):
        text_lengths = text.ne(PAD_IDX).sum(dim=1)
        #Finding valid index
        valid_indices = text_lengths > 0 #similar to compute_binary_accuracy function, ensure valid sequence lengths
        if valid_indices.sum() == 0: #if equal to zero, escape this batch
            continue
        text, labels = text[valid_indices], labels[valid_indices] #we only keep the valid input
        text_lengths = text_lengths[valid_indices]
        #recommended by chatgpt for robust, sometimes we do not need these two lines of codes
        text, labels = text.to(DEVICE), labels.to(DEVICE)
        text_lengths = text_lengths.cpu()
        #Forward and Backprop
        logits = model(text, text_lengths) # Output shape: [batch_size, 2]
        cost = F.cross_entropy(logits, labels) # Use cross-entropy loss for multi-class classification
        optimizer.zero_grad()
        cost.backward()
        #UPDATE MODEL PARAMETERS
        optimizer.step()
        #Logging
        if batch_idx % 1000 == 0:
            print(f'Epoch: {epoch+1:03d}/{NUM_EPOCHS:03d} | '
                  f'Batch {batch_idx:03d}/{len(train_loader):03d} | '
                  f'Cost: {cost:.4f}')
   with torch.no_grad():
        train_acc = compute_binary_accuracy(model, train_loader, DEVICE)
        val_acc = compute_binary_accuracy(model, val_loader, DEVICE)
        print(f'Training Accuracy: {train_acc:.2f}% | Validation Accuracy: {val_acc:.2f}%')
    print(f'Time elapsed: {(time.time() - start_time)/60:.2f} min')
#Final Test Accuracy
print(f'Total Training Time: {(time.time() - start_time)/60:.2f} min')
print(f'Test Accuracy: {compute_binary_accuracy(model, test_loader, DEVICE):.2f}%')
   Epoch: 001/010 | Batch 000/8000 | Cost: 0.7929
    Epoch: 001/010 | Batch 1000/8000 | Cost: 0.5663
    Epoch: 001/010 | Batch 2000/8000 | Cost: 0.6028
    Epoch: 001/010 | Batch 3000/8000 | Cost: 0.3930
Epoch: 001/010 | Batch 4000/8000 | Cost: 0.4876
    Epoch: 001/010 | Batch 5000/8000 | Cost: 0.5991
    Epoch: 001/010 | Batch 6000/8000 | Cost: 0.4820
    Epoch: 001/010 | Batch 7000/8000 | Cost: 0.4515
    Training Accuracy: 79.23% | Validation Accuracy: 78.88%
    Time elapsed: 3.05 min
    Epoch: 002/010 | Batch 000/8000 | Cost: 0.4784
    Epoch: 002/010 | Batch 1000/8000 | Cost: 0.4810
    Epoch: 002/010 | Batch 2000/8000 | Cost: 0.5592
    Epoch: 002/010 | Batch 3000/8000 | Cost: 0.4169
```

```
Epoch: 002/010 | Batch 4000/8000 | Cost: 0.4640
Epoch: 002/010 | Batch 5000/8000 | Cost: 0.3703
Epoch: 002/010 | Batch 6000/8000 | Cost: 0.4481
Epoch: 002/010 | Batch 7000/8000 | Cost: 0.4305
Training Accuracy: 81.07% | Validation Accuracy: 80.36%
Time elapsed: 6.07 min
Epoch: 003/010 | Batch 000/8000 | Cost: 0.4387
Epoch: 003/010 | Batch 1000/8000 | Cost: 0.3700
Epoch: 003/010 | Batch 2000/8000 | Cost: 0.4224
Epoch: 003/010 | Batch 3000/8000 | Cost: 0.5151
Epoch: 003/010 | Batch 4000/8000 | Cost: 0.3931
Epoch: 003/010 | Batch 5000/8000 | Cost: 0.2975
Epoch: 003/010 | Batch 6000/8000 | Cost: 0.4486
Epoch: 003/010 | Batch 7000/8000 | Cost: 0.4053
Training Accuracy: 82.11% | Validation Accuracy: 81.09% Time elapsed: 9.07 min
Epoch: 004/010 | Batch 000/8000 | Cost: 0.3876
Epoch: 004/010 | Batch 1000/8000 | Cost: 0.3984
Epoch: 004/010 | Batch 2000/8000 | Cost: 0.3607
Epoch: 004/010 | Batch 3000/8000 | Cost: 0.3546
Epoch: 004/010 | Batch 4000/8000 | Cost: 0.3708
Epoch: 004/010 | Batch 5000/8000 | Cost: 0.3641
Epoch: 004/010 | Batch 6000/8000 | Cost: 0.3905
Epoch: 004/010 | Batch 7000/8000 | Cost: 0.4271
Training Accuracy: 82.73% | Validation Accuracy: 81.45% Time elapsed: 12.07 min
Epoch: 005/010 | Batch 000/8000 | Cost: 0.4300
Epoch: 005/010 | Batch 1000/8000 | Cost: 0.3768
Epoch: 005/010 | Batch 2000/8000 | Cost: 0.4356
Epoch: 005/010 | Batch 3000/8000 | Cost: 0.4657
Epoch: 005/010 | Batch 4000/8000 | Cost: 0.4408
Epoch: 005/010 | Batch 5000/8000 | Cost: 0.4006
Epoch: 005/010 | Batch 6000/8000 | Cost: 0.3762
Epoch: 005/010 | Batch 7000/8000 | Cost: 0.3210
Training Accuracy: 83.39% | Validation Accuracy: 81.71%
Time elapsed: 15.06 min
Epoch: 006/010 | Batch 000/8000 | Cost: 0.3226
Epoch: 006/010 | Batch 1000/8000 | Cost: 0.3792
Epoch: 006/010 | Batch 2000/8000 | Cost: 0.3928
Epoch: 006/010
                  Batch 3000/8000
                                        Cost: 0.4206
Epoch: 006/010 | Batch 4000/8000 | Cost: 0.3445
Epoch: 006/010 | Batch 5000/8000 | Cost: 0.4131
Epoch: 006/010
                 | Batch 6000/8000 | Cost: 0.4445
Epoch: 006/010 | Batch 7000/8000 | Cost: 0.3747
```