Installing Dependecies

```
!pip install datasets
```

Show hidden output

Importing Libraries

```
import pandas as pd
import re
import ast
import numpy as np
import torch
from torch.utils.data import Dataset, DataLoader
from tqdm import tqdm
import time
from sklearn.utils import shuffle
from sklearn.metrics import accuracy_score
# check for gpu
device = torch.device('cuda' if torch.cuda.is_available() else 'cpu')
print(device)
→ cuda
```

Loading the Dataset

1

```
from datasets import load_dataset
ds = load_dataset("stanfordnlp/sst2")
print(ds)
→ DatasetDict({
         train: Dataset({
             features: ['idx', 'sentence', 'label'],
             num_rows: 67349
         })
         validation: Dataset({
             features: ['idx', 'sentence', 'label'],
             num_rows: 872
         })
         test: Dataset({
             features: ['idx', 'sentence', 'label'],
             num_rows: 1821
         })
     })
train_df = ds['train'].to_pandas()
test_df = ds['test'].to_pandas()
valid_df = ds['validation'].to_pandas()
train_df.head()
₹
         idx
                                               sentence label
      0
           0
                    hide new secretions from the parental units
```

0

0

0

contains no wit , only labored gags

2 that loves its characters and communicates som...

remains utterly satisfied to remain the same t...

```
train_df = train_df[:-10000]
print(train_df.shape)
print(test_df.shape)
print(valid_df.shape)

$\times (57349, 3) \\ (1821, 3) \\ (872, 3)
```

Custom Dataset

```
import torch
from torch.utils.data import Dataset
class MyDataset(Dataset):
    def __init__(self, texts, labels, tokenizer, max_length):
        self.texts = texts
        self.labels = labels
        self.tokenizer = tokenizer
        self.max\_length = max\_length
    def __len__(self):
        return len(self.texts)
    \label{eq:def_getitem} \texttt{def} \ \underline{\ \ } \texttt{getitem}\underline{\ \ } \texttt{(self, idx)} :
         text = self.texts[idx]
        label = self.labels[idx]
        # Tokenizing text
        encoding = self.tokenizer(
            text,
            padding='max_length',
             truncation=True,
             max_length=self.max_length,
             return_tensors="pt"
        input_ids = encoding["input_ids"].squeeze(0)
        attention_mask = encoding["attention_mask"].squeeze(0)
        return {
             "input_ids": input_ids,
             "attention_mask": attention_mask,
             "labels": torch.tensor(label, dtype=torch.long)
        }
# Conversion to list for easier processing
train_text = train_df['sentence'].tolist()
train_label = train_df['label'].tolist()
valid_text = valid_df['sentence'].tolist()
valid_label = valid_df['label'].tolist()
test_text = test_df['sentence'].tolist()
test_label = test_df['label'].tolist()
from transformers import AutoTokenizer, AutoModelForSequenceClassification
from \ transformers \ import \ XLMR oberta Tokenizer, \ XLMR oberta For Sequence Classification
from transformers import TrainingArguments, Trainer
tokenizer = XLMRobertaTokenizer.from_pretrained("xlm-roberta-base")
```

sentencepiece.bpe.model: 100%

25.0/25.0 [00:00<00:00, 2.40kB/s]

5.07M/5.07M [00:00<00:00, 20.5MB/s]

9.10M/9.10M [00:00<00:00, 17.8MB/s]

tokenizer.json: 100%

```
max_length = 48

train_dataset = MyDataset(train_text, train_label, tokenizer, max_length)

valid_dataset = MyDataset(valid_text, valid_label, tokenizer, max_length)

test_dataset = MyDataset(test_text, test_label, tokenizer, max_length)

train_loader = DataLoader(train_dataset, batch_size=8, shuffle=True)

for batch in train_loader:
    print(batch)
    break

    Show hidden output

valid_loader = DataLoader(valid_dataset, batch_size=8, shuffle=False)

test_loader = DataLoader(test_dataset, batch_size=8, shuffle=False)

from torch.optim import Adam
from transformers import get_scheduler
```

Defining the Model

model = AutoModelForSequenceClassification.from_pretrained("xlm-roberta-base", num_labels=2)
device = torch.device("cuda" if torch.cuda.is_available() else "cpu")
model.to(device)

Xet Storage is enabled for this repo, but the 'hf_xet' package is not installed. Falling back to regular HTTP download. For better perfo WARNING:huggingface_hub.file_download:Xet Storage is enabled for this repo, but the 'hf_xet' package is not installed. Falling back to r 1.12G/1.12G [00:04<00:00, 153MB/s]

Some weights of XLMRobertaForSequenceClassification were not initialized from the model checkpoint at xlm-roberta-base and are newly ini You should probably TRAIN this model on a down-stream task to be able to use it for predictions and inference.

```
{\tt XLMR} obsert a {\tt For Sequence Classification} (
  (roberta): XLMRobertaModel(
    (embeddings): XLMRobertaEmbeddings(
      (word_embeddings): Embedding(250002, 768, padding_idx=1)
      (position_embeddings): Embedding(514, 768, padding_idx=1)
      (token_type_embeddings): Embedding(1, 768)
      (LayerNorm): LayerNorm((768,), eps=1e-05, elementwise_affine=True)
      (dropout): Dropout(p=0.1, inplace=False)
    (encoder): XLMRobertaEncoder(
      (layer): ModuleList(
        (0-11): 12 x XLMRobertaLayer(
          (attention): XLMRobertaAttention(
            (self): XLMRobertaSdpaSelfAttention(
              (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): XLMRobertaSelfOutput(
              (dense): Linear(in_features=768, out_features=768, bias=True)
              (LayerNorm): LayerNorm((768,), eps=1e-05, elementwise_affine=True)
              (dropout): Dropout(p=0.1, inplace=False)
            )
          (intermediate): XLMRobertaIntermediate(
            (dense): Linear(in_features=768, out_features=3072, bias=True)
            (intermediate_act_fn): GELUActivation()
          (output): XLMRobertaOutput(
            (dense): Linear(in_features=3072, out_features=768, bias=True)
            (LayerNorm): LayerNorm((768,), eps=1e-05, elementwise_affine=True)
            (dropout): Dropout(p=0.1, inplace=False)
     )
   )
  (classifier): XLMRobertaClassificationHead(
    (dense): Linear(in_features=768, out_features=768, bias=True)
    (dropout): Dropout(p=0.1, inplace=False)
    (out_proj): Linear(in_features=768, out_features=2, bias=True)
```

Setting Parameters

)

```
optimizer = Adam(model.parameters(), lr=2e-5)
num_epochs = 5
num_training_steps = num_epochs * len(train_loader)
\label{local_local_local_local_local} $$\operatorname{lc_scheduler("linear", optimizer-optimizer, num_warmup_steps=0, num_training_steps=num_training_steps)$$
loss_fn = torch.nn.CrossEntropyLoss()
```

Training the model

```
# Training Loop
for epoch in range(num_epochs):
   model.train()
   total loss = 0
   start_time = time.time()
   for batch in tqdm(train_loader, desc=f"Epoch {epoch + 1}/{num_epochs}", unit="batch"):
       optimizer.zero_grad()
        input_ids = batch["input_ids"].to(device)
       attention_mask = batch["attention_mask"].to(device)
        labels = batch["labels"].to(device)
```

```
outputs = model(input_ids=input_ids, attention_mask=attention_mask, labels=labels)
       loss = outputs.loss
       total_loss += loss.item()
       loss.backward()
       optimizer.step()
       lr_scheduler.step()
   end_time = time.time()
   epoch_time = end_time - start_time
   avg_train_loss = total_loss / len(train_loader)
   print(f"Epoch {epoch+1}: Train Loss = {avg_train_loss:.4f}, Time: {epoch_time:.2f} seconds")
   model.eval()
   correct = 0
   total = 0
   with torch.no_grad():
       for batch in tqdm(valid_loader, desc="Validation", unit="batch"):
          input_ids = batch["input_ids"].to(device)
          attention_mask = batch["attention_mask"].to(device)
          labels = batch["labels"].to(device)
          outputs = model(input_ids=input_ids, attention_mask=attention_mask)
          predictions = torch.argmax(outputs.logits, dim=-1)
          correct += (predictions == labels).sum().item()
          total += labels.size(0)
   accuracy = correct / total
   print(f"Epoch {epoch+1}: Validation Accuracy = {accuracy:.4f}")

    Epoch 1/3: 100% | 7169/7169 [17:00<00:00, 7.02batch/s]
</p>
    Epoch 1: Train Loss = 0.3737, Time: 1020.94 seconds
    Validation: 100% | 109/109 [00:02<00:00, 41.57batch/s]
    Epoch 1: Validation Accuracy = 0.9163
    Epoch 2/3: 100%
                        7169/7169 [16:59<00:00, 7.03batch/s]
    Epoch 2: Train Loss = 0.1755, Time: 1019.50 seconds
    Validation: 100% | 109/109 [00:02<00:00, 40.63batch/s]
    Epoch 2: Validation Accuracy = 0.9186
    Epoch 3/3: 100% 7169/7169 [16:59<00:00, 7.03batch/s]
    Epoch 3: Train Loss = 0.1082, Time: 1019.74 seconds
                            | 109/109 [00:02<00:00, 40.86batch/s]Epoch 3: Validation Accuracy = 0.9278
    Validation: 100%
```

Preparing Hindi Dataset

with open("pos_train.txt", "r", encoding="utf-8") as f: sentences = [line.strip() for line in f if line.strip()]

```
with open("neg_train.txt", "r", encoding="utf-8") as f:
    sentences = [line.strip() for line in f if line.strip()]
neg label = [0] * len(sentences)
df_neg = pd.DataFrame({
     "sentence": sentences,
    "label": neg_label
})
df neg.head()
<del>_</del>₹
                                         sentence label
             बुंदेलखंड के किसानों को अबकी बार बड़ी उम्मीदें...
             उन्होंने कहा कि चूंकि पूंजी की जरूरत बहुत बडी...
                                                         0
             पश्चिम बंगाल में शासन द्वारा ऐसी शिकायतें लगात...
      शहर में नेटवर्किंग कारोबार का मकडजाल लगातार फ...
```

```
pos_label = [1] * len(sentences)
df_pos = pd.DataFrame({
    "sentence": sentences,
    "label": pos_label
})
df_pos.head()
→▼
                                        sentence label
      0 मुलाकात के बाद दोनों तरफ के जवानों ने वॉलीबॉल ...
            दोनों में दोस्ताना माहौल लग रहा है इसलिए इन्हे...
               लिटिल एंजिल स्कूल में दिलचस्प मैत्रीपूर्ण क्रि...
          मैत्रीपूर्ण मैच में अखिलेश कुमार को मैन आफ द म...
               hindi_df = pd.concat([df_pos, df_neg], ignore_index=True)
hindi_df = shuffle(hindi_df, random_state=42)
hindi_df.head()
sentence label
      2032 तापमान में आ रही गिरावट का असर जनजीवन पर दिख र...
                                                              n
             यह मेरी जिंदगी का सबसे आकर्षक समय है और अगर मे...
       906
                         लंद हौसलों के साथ कुछ भी करना संभव है
      1128
                       नीतीश राजनीति में कठिन दौर से गुजर रहे है।
      2004
                                                              0
      4 (
hindi_df.shape
→ (2387, 2)
hindi_texts = hindi_df['sentence'].tolist()
hindi_labels = hindi_df['label'].tolist()
hindi_dataset = MyDataset(hindi_texts, hindi_labels, tokenizer, max_length)
hindi_loader = DataLoader(hindi_dataset, batch_size=8, shuffle=False)
```

Making Prediction on Hindi Dataset

```
all_predictions = []
all_labels = []
model.eval()
with torch.no_grad():
    for batch in {\tt tqdm(hindi\_loader, desc="Predicting on Hindi Dataset"):}
        input_ids = batch["input_ids"].to(device)
        attention_mask = batch["attention_mask"].to(device)
        labels = batch["labels"].to(device)
        outputs = model(input_ids=input_ids, attention_mask=attention_mask)
        predictions = torch.argmax(outputs.logits, dim=-1)
        all_predictions.extend(predictions.cpu().numpy())
        all_labels.extend(labels.cpu().numpy())
accuracy = accuracy score(all labels all medictions)
```