```
In [ ]: pip install notebook
In [ ]: !pip install torch transformers
In [ ]: import pandas as pd
        from transformers import BertTokenizer
        # Load datasets
        sst train = pd.read csv('data/ids-sst-train.csv')
        sst_dev = pd.read_csv('data/ids-sst-dev.csv')
        sst test = pd.read csv('data/ids-sst-test-student.csv')
In [ ]: import torch
        from transformers import BertModel
        class minBERTModel(torch.nn.Module):
            def init (self):
                super(minBERTModel, self).__init__()
                self.bert = BertModel.from pretrained('bert-base-uncased')
                self.dropout = torch.nn.Dropout(0.1)
                self.classifier = torch.nn.Linear(768, 5) # Assuming 5 sentiment classes
            def forward(self, input ids, attention mask):
                outputs = self.bert(input_ids, attention_mask=attention_mask)
                pooled output = outputs[1]
                pooled_output = self.dropout(pooled_output)
                return self.classifier(pooled output)
In []: from torch.utils.data import DataLoader, Dataset
        from transformers import AdamW
        class SentimentDataset(Dataset):
            def __init__(self, texts, labels, tokenizer):
                self.texts = texts
                self.labels = labels
                self.tokenizer = tokenizer
            def __len__(self):
                return len(self.texts)
            def __getitem__(self, idx):
                encoding = self.tokenizer.encode_plus(
                    self.texts[idx],
                    add special tokens=True,
                    max_length=512,
                    return_token_type_ids=False,
                    padding='max_length'
                    return attention mask=True,
                    return_tensors='pt',
                    'text': self.texts[idx],
                    'input_ids': encoding['input_ids'].flatten(),
                     'attention mask': encoding['attention mask'].flatten(),
                     'label': torch.tensor(self.labels[idx], dtype=torch.long)
        tokenizer = BertTokenizer.from pretrained('bert-base-uncased')
        train_dataset = SentimentDataset(sst_train['text'], sst_train['label'], tokenizer)
        train loader = DataLoader(train dataset, batch size=16, shuffle=True)
        model = minBERTModel()
        optimizer = AdamW(model.parameters(), lr=2e-5)
        for epoch in range(3): # Number of epochs
            for batch in train loader:
                optimizer.zero_grad()
                input ids = batch['input ids']
                attention_mask = batch['attention_mask']
                labels = batch['label']
                outputs = model(input ids, attention mask)
                loss = torch.nn.CrossEntropyLoss()(outputs, labels)
                loss.backward()
                optimizer.step()
In [ ]: def evaluate(model, data_loader):
            model.eval()
            correct predictions = 0
            with torch.no_grad():
```

for batch in data loader:

```
input_ids = batch['input_ids']
    attention_mask = batch['attention_mask']
    labels = batch['label']
    outputs = model(input_ids, attention_mask)
    _, preds = torch.max(outputs, dim=1)
    correct_predictions += torch.sum(preds == labels)
    return correct_predictions.double() / len(data_loader.dataset)

dev_dataset = SentimentDataset(sst_dev['text'], sst_dev['label'], tokenizer)
dev_loader = DataLoader(dev_dataset, batch_size=16)

test_dataset = SentimentDataset(sst_test['text'], sst_test['label'], tokenizer)
test_loader = DataLoader(test_dataset, batch_size=16)

dev_accuracy = evaluate(model, dev_loader)
test_accuracy = evaluate(model, test_loader)
print(f'Dev Accuracy: {dev_accuracy.item()}')
print(f'Test Accuracy: {test_accuracy.item()}')
```

Loading [MathJax]/jax/output/CommonHTML/fonts/TeX/fontdata.js