```
In [1]: import torch
        import nltk
        import re
        from nltk.stem import PorterStemmer
        from nltk.corpus import stopwords
        from transformers import DistilBertTokenizer, DistilBertForSequend
        from torch.utils.data import DataLoader, Dataset
        from transformers import Trainer, TrainingArguments, TrainerCallba
        from sklearn.preprocessing import LabelEncoder
        from sklearn.model selection import train test split
        import pandas as pd
        import joblib
        # Load and prepare data
        data = pd.read_csv('/Users/pillisachethan/Desktop/NLP project/test
        data['post'] = data['post'].fillna('')
        # Remove non-alphabetical characters and lowercase the text
        data['post'] = data['post'].apply(lambda x: re.sub(r'[^a-z\s]', ''
        # Initialize stopwords set and PorterStemmer
        stop words = set(stopwords.words('english'))
        ps = PorterStemmer()
        # Function to remove stopwords and apply stemming
        def preprocess text(text):
            # Remove stopwords
            filtered words = [word for word in text.split() if word not in
            # Apply stemming
            stemmed_text = ' '.join(ps.stem(word) for word in filtered_wor
            return stemmed text
        # Apply the preprocessing to the 'post' column
        data['post'] = data['post'].apply(preprocess_text)
        data['text'] = data['post'].str.lower()
        # Encode labels
        label encoder = LabelEncoder()
        data['labels'] = label_encoder.fit_transform(data['subreddit'])
        # Split data into training and testing sets
        train_data, test_data = train_test_split(data, test_size=0.2, rand
        # Custom Dataset class
        class CustomDataset(Dataset):
            def __init__(self, data, tokenizer, max_len):
                self.data = data
                self.tokenizer = tokenizer
                self.max_len = max_len
            def __len__(self):
                return len(self.data)
            def __getitem__(self, idx):
                text = self.data.iloc[idx]['text']
                label = self.data.iloc[idx]['labels']
                encoding = self.tokenizer(text, truncation=True, padding='
                return {
                     'input_ids': torch.tensor(encoding['input_ids']),
```

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'attention mask': torch.tensor(encoding['attention mas
            'labels': torch.tensor(label)
        }
# Load the tokenizer and model
tokenizer = DistilBertTokenizer.from pretrained('distilbert-base-u
model = DistilBertForSequenceClassification.from_pretrained('disti
# Create Dataset for training and testing sets
train dataset = CustomDataset(train data, tokenizer, max len=5)
test_dataset = CustomDataset(test_data, tokenizer, max_len=5)
# Training arguments with optimizations
training args = TrainingArguments(
    output_dir='./results',
    num train epochs=1, # Start with 1 epoch
    per device train batch size=100, # Experiment with batch size
    per_device_eval_batch_size=100,
    gradient accumulation steps=2, # Simulate larger batch size i
    warmup_steps=500,
    weight decay=0.01,
    logging dir='./logs',
    logging steps=100,
    save_total_limit=1,
    fp16=False, # Disable fp16 if there's no support for it
    bf16=True, # Enable bfloat16 for faster training if supported
    report_to="tensorboard", # Enable TensorBoard for monitoring
)
# Initialize Trainer with evaluation and early stopping callback
trainer = Trainer(
    model=model,
    args=training args,
    train dataset=train dataset,
    eval_dataset=test_dataset, # Include test dataset for evaluat
)
# Train the model
trainer.train()
# Save model, tokenizer, and label encoder
model.save_pretrained('./fine_tuned_distilbert')
tokenizer.save_pretrained('./fine_tuned_distilbert')
joblib.dump(label_encoder, './fine_tuned_distilbert/label_encoder.
```

Some weights of DistilBertForSequenceClassification were not init ialized from the model checkpoint at distilbert-base-uncased and are newly initialized: ['classifier.bias', 'classifier.weight', 'pre\_classifier.bias', 'pre\_classifier.weight'] You should probably TRAIN this model on a down-stream task to be able to use it for predictions and inference.

[819/819 05:07, Epoch 0/1]

Step	Training Loss
100	3.101900
200	2.555900
300	2.231000
400	2.060400
500	1.972200
600	1.890300
700	1.859500
800	1.818000

Out[1]: ['./fine\_tuned\_distilbert/label\_encoder.joblib']

```
In [3]: # Evaluate the model on the test set
        eval results = trainer.evaluate()
        print("Evaluation results:", eval_results)
        # Reload for prediction
        tokenizer = DistilBertTokenizer.from_pretrained('./fine_tuned_dist
        model = DistilBertForSequenceClassification.from pretrained('./fin
        label_encoder = joblib.load('./fine_tuned_distilbert/label_encoder
        # Function for individual text prediction
        def predict behavior(text):
            inputs = tokenizer(text, return tensors="pt", truncation=True,
            # Move to device if using GPU or MPS (Metal)
            inputs = {key: value.to(model.device) for key, value in inputs
            with torch.no_grad():
                outputs = model(**inputs)
            logits = outputs.logits
            predicted_label = torch.argmax(logits, dim=1).item()
            subreddit = label encoder.inverse transform([predicted label])
            return subreddit
        # Function to predict on test set and add predictions
        def predict_on_test_set(test_data):
            model.eval() # Set model to evaluation mode
            predictions = []
            for i in range(len(test data)):
                text = test data.iloc[i]['post']
                predicted subreddit = predict behavior(text)
                predictions.append(predicted subreddit)
            test_data['predicted'] = predictions
            return test_data
        # Make predictions on the test set
        test_data = predict_on_test_set(test_data)
        # Display 'post' and 'predicted' columns for review
        print(test_data[['post', 'predicted']])
```

```
Evaluation results: {'eval loss': 1.7744554281234741, 'eval runti
        me': 36.9013, 'eval_samples_per_second': 1110.069, 'eval_steps_pe
        r second': 11.111, 'epoch': 0.9993898718730934}
                                                               post
        predicted
        151830 figur share may help ya https://www.redditcomrcoro...
                                                                           d
        epression
        90592
                much ira invest mutual fund im finish grad sch...
        alfinance
        186469 fight fear futur turn year oldest child famili...
        epression
        64692
                bodili ach mental agoni past week ive sever ac...
        epression
        110473 agre worker believ ive misclassifi option new ...
        alfinance
                                                                . . .
        7872
                ne flsa question partner work sport busi midwe...
        alfinance
        11011
                one realli want help ive struggl depress anxie...
        epression
                 im move anoth contin start new job feel someth...
        37091
                                                                           d
        epression
        193699 ten year schizoversari around time first start...
                                                                           d
        epression
        129011
                        lone joke im lone even duolingo send email
        lonely
        [40963 \text{ rows } x \text{ 2 columns}]
In [5]: | from sklearn.metrics import accuracy_score
        # Predict on the test set without modifying the DataFrame
        test data['predicted'] = test data['post'].apply(predict behavior)
        # Calculate accuracy
        accuracy = accuracy_score(test_data['labels'], label_encoder.trans
        print(f"Accuracy: {accuracy * 100:.2f}%")
        Accuracy: 62.30%
In [ ]:
```