```
!pip install transformers
from transformers import BertTokenizer, TFBertForSequenceClassification
from transformers import InputExample, InputFeatures
model = TFBertForSequenceClassification.from pretrained("bert-base-uncased")
tokenizer = BertTokenizer.from pretrained("bert-base-uncased")
     Downloading: 100%
                                                                570/570 [00:00<00:00, 11.4kB/s]
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                                                                511M/511M [00:17<00:00, 30.9MB/s]
     All model checkpoint layers were used when initializing TFBertForSequenceClassificati
     Some layers of TFBertForSequenceClassification were not initialized from the model ch
     You should probably TRAIN this model on a down-stream task to be able to use it for p
     Downloading: 100%
                                                                226k/226k [00:00<00:00, 840kB/s]
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                                                                455k/455k [00:00<00:00, 866kB/s]
import tensorflow as tf
import pandas as pd
URL = "https://ai.stanford.edu/~amaas/data/sentiment/aclImdb_v1.tar.gz"
dataset = tf.keras.utils.get file(fname="aclImdb v1.tar.gz",
                                                                 origin=URL,
                                                                 untar=True,
                                                                 cache_dir='.'
                                                                 cache subdir='')
# operations on files and collections of files.
import os
import shutil
# Create main directory path ("/aclImdb")
main dir = os.path.join(os.path.dirname(dataset), 'aclImdb')
# Create sub directory path ("/aclImdb/train")
train dir = os.path.join(main dir, 'train')
# Remove unsup folder since this is a supervised learning task
remove dir = os.path.join(train dir, 'unsup')
shutil.rmtree(remove dir)
# View the final train folder
print(os.listdir(train dir))
# We create a training dataset and a validation
# dataset from our "aclImdb/train" directory with a 80/20 split.
train = tf.keras.preprocessing.text dataset from directory(
       'aclImdb/train', batch size=30000, validation split=0.2,
       subset='training', seed=123)
test = tf.keras.preprocessing.text dataset from directory(
       'aclImdb/train', batch size=30000, validation split=0.2,
       subset='validation', seed=123)
```

for i in train. take(1):

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train feat = i[0]. numpy()
   train lab = i[1]. numpy()
train = pd. DataFrame([train feat, train lab]). T
train.columns = ['DATA COLUMN', 'LABEL COLUMN']
train['DATA COLUMN'] = train['DATA COLUMN'].str.decode("utf-8")
train. head()
for j in test. take (1):
   test_feat = j[0].numpy()
   test lab = j[1]. numpy()
test = pd.DataFrame([test_feat, test_lab]).T
test.columns = ['DATA COLUMN', 'LABEL COLUMN']
test['DATA COLUMN'] = test['DATA COLUMN'].str.decode("utf-8")
test. head()
InputExample (guid=None,
                        text a = "Hello, world",
                        text b = None,
                        label = 1)
def convert data to examples (train, test, DATA COLUMN, LABEL COLUMN):
   train InputExamples = train.apply(lambda x: InputExample(guid=None, # Globally unique I
   validation InputExamples = test.apply(lambda x: InputExample(guid=None, # Globally uniqu
   return train InputExamples, validation InputExamples
   train_InputExamples, validation_InputExamples = convert_data_to_examples(train,
def convert_examples_to_tf_dataset(examples, tokenizer, max_length=128):
       features = [] # -> will hold InputFeatures to be converted later
       for e in examples:
               # Documentation is really strong for this method, so please take a look
               input dict = tokenizer.encode plus(
                       e. text a,
                      add special tokens=True,
                       max length=max length, # truncates if len(s) > max length
                      return token type ids=True,
                       return_attention_mask=True,
                       pad to max length=True, # pads to the right by default # CHECK TH
                       truncation=True
               )
               input ids, token type ids, attention mask = (input dict["input ids"],
                       input dict["token type ids"], input dict['attention mask'])
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features.append(
                        InputFeatures (
                                input ids=input ids, attention mask=attention mask, token type id
                )
        def gen():
                for f in features:
                        yield (
                                         "input_ids": f.input ids,
                                        "attention_mask": f.attention_mask,
                                         "token_type_ids": f.token_type_ids,
                                },
                                f. label,
        return tf. data. Dataset. from generator (
                ({"input ids": tf.int32, "attention mask": tf.int32, "token type ids": tf.int3
                                "input ids": tf. TensorShape([None]),
                                "attention mask": tf. TensorShape([None]),
                                "token type ids": tf. TensorShape([None]),
                        },
                        tf. TensorShape([]),
                ),
DATA COLUMN = 'DATA COLUMN'
LABEL COLUMN = 'LABEL_COLUMN'
     Downloading data from <a href="https://ai.stanford.edu/~amaas/data/sentiment/aclImdb">https://ai.stanford.edu/~amaas/data/sentiment/aclImdb</a> v1.tar.gz
     =====] - 2s Ous/step
     84140032/84125825 [=========
                                                      ==] - 2s Ous/step
     ['urls_pos.txt', 'urls_unsup.txt', 'unsupBow.feat', 'neg', 'pos', 'urls_neg.txt', 'labeledBow
     Found 25000 files belonging to 2 classes.
     Using 20000 files for training.
     Found 25000 files belonging to 2 classes.
     Using 5000 files for validation.
train_InputExamples, validation_InputExamples = convert_data_to_examples(train, test, DATA_COL
train data = convert examples to tf dataset(list(train InputExamples), tokenizer)
train data = train data. shuffle(100). batch(32). repeat(2)
validation data = convert examples to tf dataset(list(validation InputExamples), tokenizer)
validation data = validation data.batch(32)
model.compile(optimizer=tf.keras.optimizers.Adam(learning rate=3e-5, epsilon=1e-08, clipnorm=1.0)
```

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loss=tf.keras.losses.SparseCategoricalCrossentropy(from_logits=True),
metrics=[tf.keras.metrics.SparseCategoricalAccuracy('accuracy')])
```

```
model.fit(train data, epochs=2, validation data=validation data)
     /usr/local/lib/python3.7/dist-packages/transformers/tokenization utils base.py:2218: FutureWa
       FutureWarning,
     Epoch 1/2
                                  =======] - 1993s 2s/step - loss: 0.2709 - accuracy: 0.8838
     1250/1250 [====
     Epoch 2/2
     1250/1250 [======] - 1962s 2s/step - loss: 0.0740 - accuracy: 0.9747
     <keras.callbacks.History at 0x7f79d97d8610>
pred_sentences = ["Since our earliest days, Tesla has been built upon a culture of op
                                 "We planned very early—with Gigafactory Shanghai as the
                                 "Tesla is on a mission to accelerate the world's tran
1
tf batch = tokenizer(pred sentences, max length=128, padding=True, truncation=True, return te
tf_outputs = model(tf_batch)
tf_predictions = tf.nn.softmax(tf_outputs[0], axis=-1)
labels = ['Negative', 'Positive']
label = tf.argmax(tf_predictions, axis=1)
label = label.numpy()
for i in range(len(pred_sentences)):
   print(labels[label[i]])
#https://www.tesla.com/ns_videos/2020-tesla-impact-report.pdf
     Positive
     Positive
     Negative
```

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