Text and Sequence Assignment4

Downloading and Preparing the IMDB Dataset

0 0:00:14 0:00:14 --:-- 12.7M

0 5768k

Preprocessing the Text Data

100 80.2M 100 80.2M

```
In [2]: import os, pathlib, shutil, random
        from tensorflow import keras
        batch_size = 32
        base_dir= pathlib.Path("/content/aclImdb")
        val_dir = base_dir/ "val"
        train_dir = base_dir / "train"
        for category in ("neg", "pos"):
            os.makedirs(val_dir / category, exist_ok=True )
            files = os.listdir(train_dir / category)
            random.Random(1337).shuffle(files)
            num_val_samples = int(0.2 * len(files))
            val files = files[-num val samples:]
            for fname in val files:
                shutil.move(train_dir / category / fname,
                             val_dir / category / fname)
        train_ds = keras.utils.text_dataset_from_directory(
            "aclImdb/train", batch_size=batch_size
        ).take(100)
        val_ds = keras.utils.text_dataset_from_directory(
            "/content/aclImdb/val", batch_size=batch_size
        ).take(10000)
        test_ds = keras.utils.text_dataset_from_directory(
            "aclImdb/test", batch_size=batch_size
        text_only_train_ds = train_ds.map(lambda x, y: x)
```

Found 20000 files belonging to 2 classes. Found 5000 files belonging to 2 classes. Found 25000 files belonging to 2 classes.

Vectorizing the Text Sequences

```
In [3]: from tensorflow.keras import layers
max_length = 150
```

```
max_tokens = 10000
text_vectorization = layers.TextVectorization(
    max_tokens=max_tokens,
    output_mode="int",
    output_sequence_length=max_length,
)
text_vectorization.adapt(text_only_train_ds)

int_train_ds = train_ds.map(
    lambda x, y: (text_vectorization(x), y),
    num_parallel_calls=4)
int_val_ds = val_ds.map(
    lambda x, y: (text_vectorization(x), y),
    num_parallel_calls=4)
int_test_ds = test_ds.map(
    lambda x, y: (text_vectorization(x), y),
    num_parallel_calls=4)
int_test_ds = test_ds.map(
    lambda x, y: (text_vectorization(x), y),
    num_parallel_calls=4)
```

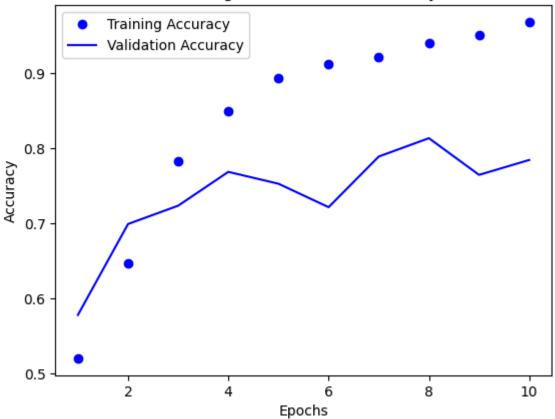
Baseline Model: One-Hot Encoded Vectors

Model: "model"

Layer (type)	Output Shape	Param #
		=======
input_1 (InputLayer)	[(None, None)]	0
tf.one_hot (TFOpLambda)	(None, None, 10000)	0
bidirectional (Bidirection	(None, 64)	2568448
al)		
deserve (Deserve)	(Name CA)	0
dropout (Dropout)	(None, 64)	0
dense (Dense)	(None, 1)	65
delise (Delise)	(None, 1)	05
Total params: 2568513 (9.80 MB)		
Trainable params: 2568513 (9.80 MB)		
Non-trainable params: 0 (0.00 Byte)		
,	, ,	

```
In [5]: callbacks = [
            keras.callbacks.ModelCheckpoint("one_hot_bidir_lstm.x",
                                             save_best_only=True)
        ]
        model.compile(optimizer="rmsprop", loss="binary_crossentropy", metrics=["accuracy"]
        history = model.fit(int_train_ds, validation_data=int_val_ds, epochs=10, callbacks=
        model = keras.models.load_model("one_hot_bidir_lstm.x")
        print(f"Test acc: {model.evaluate(int_test_ds)[1]:.3f}")
        import matplotlib.pyplot as plt
        accuracy = history.history["accuracy"]
        val_accuracy = history.history["val_accuracy"]
        epochs = range(1, len(accuracy) + 1)
        plt.figure()
        plt.plot(epochs, accuracy, "bo", label="Training Accuracy")
        plt.plot(epochs, val_accuracy, "b", label="Validation Accuracy")
        plt.title("Training and validation Accuracy")
        plt.xlabel("Epochs")
        plt.ylabel("Accuracy")
        plt.legend()
        plt.show()
```

```
Epoch 1/10
y: 0.5203 - val_loss: 0.6899 - val_accuracy: 0.5780
Epoch 2/10
y: 0.6463 - val_loss: 0.5929 - val_accuracy: 0.6990
100/100 [============] - 6s 65ms/step - loss: 0.5122 - accuracy:
0.7825 - val_loss: 0.6595 - val_accuracy: 0.7234
Epoch 4/10
y: 0.8497 - val_loss: 0.4821 - val_accuracy: 0.7684
Epoch 5/10
100/100 [=============] - 7s 69ms/step - loss: 0.3231 - accuracy:
0.8934 - val_loss: 0.4929 - val_accuracy: 0.7526
Epoch 6/10
100/100 [============= ] - 6s 62ms/step - loss: 0.2650 - accuracy:
0.9116 - val_loss: 0.6802 - val_accuracy: 0.7214
Epoch 7/10
y: 0.9206 - val_loss: 0.4809 - val_accuracy: 0.7888
Epoch 8/10
y: 0.9397 - val_loss: 0.4312 - val_accuracy: 0.8132
Epoch 9/10
100/100 [============== ] - 6s 64ms/step - loss: 0.1542 - accuracy:
0.9506 - val_loss: 0.7079 - val_accuracy: 0.7644
Epoch 10/10
100/100 [=============== ] - 9s 89ms/step - loss: 0.1214 - accuracy:
0.9678 - val_loss: 0.5657 - val_accuracy: 0.7842
y: 0.8095
Test acc: 0.810
```



Developing the Embedding Layer

```
In [6]: embedding_layer = layers.Embedding(input_dim=max_tokens, output_dim=256)
```

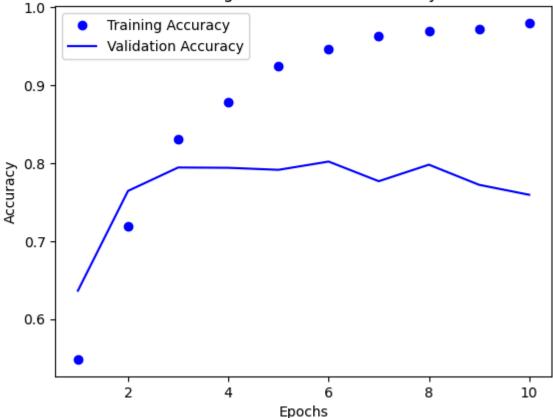
Model with Trainable Embedding Layer

```
In [7]: inputs = keras.Input(shape=(None,), dtype="int64")
        embedded = layers.Embedding(input_dim=max_tokens, output_dim=256)(inputs)
        x = layers.Bidirectional(layers.LSTM(32))(embedded)
        x = layers.Dropout(0.5)(x)
        outputs = layers.Dense(1, activation="sigmoid")(x)
        model = keras.Model(inputs, outputs)
        model.compile(optimizer="rmsprop",
                      loss="binary_crossentropy",
                      metrics=["accuracy"])
        model.summary()
        callbacks = [
            keras.callbacks.ModelCheckpoint("embeddings_bidir_gru.x",
                                             save_best_only=True)
        history = model.fit(int_train_ds, validation_data=int_val_ds, epochs=10, callbacks=
        model = keras.models.load_model("embeddings_bidir_gru.x")
        print(f"Test acc: {model.evaluate(int_test_ds)[1]:.3f}")
        accuracy = history.history["accuracy"]
        val_accuracy = history.history["val_accuracy"]
```

```
epochs = range(1, len(accuracy) + 1)
plt.figure()
plt.plot(epochs,accuracy, "bo", label="Training Accuracy")
plt.plot(epochs, val_accuracy, "b", label="Validation Accuracy")
plt.title("Training and validation Accuracy")
plt.xlabel("Epochs")
plt.ylabel("Accuracy")
plt.legend()
plt.show()
```

```
Layer (type)
                    Output Shape
                                       Param #
______
input_2 (InputLayer)
                    [(None, None)]
embedding_1 (Embedding) (None, None, 256)
                                       2560000
bidirectional 1 (Bidirecti (None, 64)
                                       73984
onal)
dropout 1 (Dropout) (None, 64)
                                       0
dense_1 (Dense)
                   (None, 1)
                                       65
_____
Total params: 2634049 (10.05 MB)
Trainable params: 2634049 (10.05 MB)
Non-trainable params: 0 (0.00 Byte)
Epoch 1/10
y: 0.5484 - val_loss: 0.6397 - val_accuracy: 0.6364
Epoch 2/10
y: 0.7194 - val loss: 0.5037 - val accuracy: 0.7644
y: 0.8306 - val_loss: 0.4592 - val_accuracy: 0.7944
Epoch 4/10
100/100 [============== ] - 5s 49ms/step - loss: 0.3317 - accuracy:
0.8788 - val_loss: 0.4721 - val_accuracy: 0.7940
Epoch 5/10
100/100 [============= ] - 3s 34ms/step - loss: 0.2223 - accuracy:
0.9250 - val_loss: 0.4796 - val_accuracy: 0.7914
Epoch 6/10
100/100 [=============] - 5s 49ms/step - loss: 0.1773 - accuracy:
0.9463 - val_loss: 0.5590 - val_accuracy: 0.8020
Epoch 7/10
100/100 [=============== ] - 3s 33ms/step - loss: 0.1189 - accuracy:
0.9631 - val_loss: 0.6186 - val_accuracy: 0.7768
Epoch 8/10
100/100 [============== ] - 3s 26ms/step - loss: 0.1060 - accuracy:
0.9688 - val_loss: 0.6379 - val_accuracy: 0.7980
Epoch 9/10
100/100 [============== ] - 5s 53ms/step - loss: 0.0818 - accuracy:
0.9722 - val_loss: 0.7108 - val_accuracy: 0.7722
Epoch 10/10
100/100 [============== ] - 3s 31ms/step - loss: 0.0666 - accuracy:
0.9797 - val loss: 0.8444 - val accuracy: 0.7594
0.7928
Test acc: 0.793
```



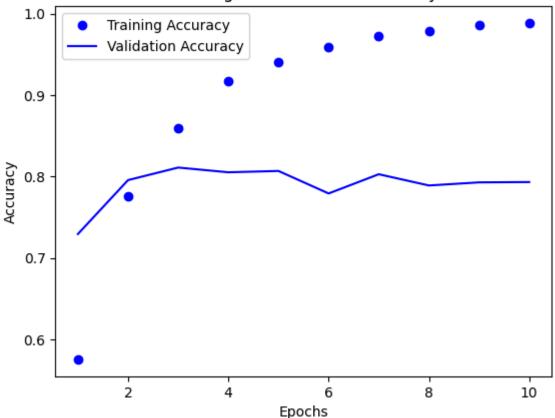


Masking Padded Sequences in the Embedding Layer

```
In [8]: inputs = keras.Input(shape=(None,), dtype="int64")
        embedded = layers.Embedding(
            input_dim=max_tokens, output_dim=256, mask_zero=True)(inputs)
        x = layers.Bidirectional(layers.LSTM(32))(embedded)
        x = layers.Dropout(0.5)(x)
        outputs = layers.Dense(1, activation="sigmoid")(x)
        model = keras.Model(inputs, outputs)
        model.compile(optimizer="rmsprop",
                      loss="binary_crossentropy",
                      metrics=["accuracy"])
        model.summary()
        callbacks = [
            keras.callbacks.ModelCheckpoint("embeddings_bidir_gru_with_masking.x",
                                             save best only=True)
        history = model.fit(int_train_ds, validation_data=int_val_ds, epochs=10, callbacks=
        model = keras.models.load_model("embeddings_bidir_gru_with_masking.x")
        print(f"Test acc: {model.evaluate(int_test_ds)[1]:.3f}")
        accuracy = history.history["accuracy"]
        val_accuracy = history.history["val_accuracy"]
        epochs = range(1, len(accuracy) + 1)
        plt.figure()
        plt.plot(epochs, accuracy, "bo", label="Training Accuracy")
        plt.plot(epochs, val_accuracy, "b", label="Validation Accuracy")
```

```
plt.title("Training and validation Accuracy")
plt.xlabel("Epochs")
plt.ylabel("Accuracy")
plt.legend()
plt.show()
```

```
Layer (type)
                    Output Shape
                                       Param #
______
input_3 (InputLayer)
                    [(None, None)]
embedding_2 (Embedding) (None, None, 256)
                                       2560000
bidirectional 2 (Bidirecti (None, 64)
                                       73984
onal)
dropout 2 (Dropout) (None, 64)
                                       0
dense_2 (Dense)
                   (None, 1)
                                       65
_____
Total params: 2634049 (10.05 MB)
Trainable params: 2634049 (10.05 MB)
Non-trainable params: 0 (0.00 Byte)
Epoch 1/10
y: 0.5756 - val_loss: 0.5780 - val_accuracy: 0.7294
Epoch 2/10
y: 0.7759 - val loss: 0.4467 - val accuracy: 0.7956
y: 0.8594 - val_loss: 0.4232 - val_accuracy: 0.8110
Epoch 4/10
100/100 [============= ] - 5s 51ms/step - loss: 0.2369 - accuracy:
0.9172 - val_loss: 0.4464 - val_accuracy: 0.8052
Epoch 5/10
100/100 [============= ] - 5s 49ms/step - loss: 0.1736 - accuracy:
0.9400 - val_loss: 0.4379 - val_accuracy: 0.8068
Epoch 6/10
100/100 [=============] - 5s 49ms/step - loss: 0.1235 - accuracy:
0.9588 - val_loss: 0.5905 - val_accuracy: 0.7792
Epoch 7/10
100/100 [=============== ] - 4s 39ms/step - loss: 0.0874 - accuracy:
0.9728 - val_loss: 0.5418 - val_accuracy: 0.8028
Epoch 8/10
100/100 [============== ] - 4s 38ms/step - loss: 0.0642 - accuracy:
0.9784 - val_loss: 0.5129 - val_accuracy: 0.7890
Epoch 9/10
100/100 [============== ] - 4s 38ms/step - loss: 0.0463 - accuracy:
0.9856 - val_loss: 0.5664 - val_accuracy: 0.7928
Epoch 10/10
100/100 [=============== ] - 4s 45ms/step - loss: 0.0408 - accuracy:
0.9884 - val loss: 0.6331 - val accuracy: 0.7932
y: 0.8052
Test acc: 0.805
```



Using Pre-trained Word Embeddings (GloVe)

!wget http://nlp.stanford.edu/data/glove.6B.zip

In [9]:

```
!unzip -q glove.6B.zip
--2024-05-05 20:24:02-- http://nlp.stanford.edu/data/glove.6B.zip
Resolving nlp.stanford.edu (nlp.stanford.edu)... 171.64.67.140
Connecting to nlp.stanford.edu (nlp.stanford.edu)|171.64.67.140|:80... connected.
HTTP request sent, awaiting response... 302 Found
Location: https://nlp.stanford.edu/data/glove.6B.zip [following]
--2024-05-05 20:24:03-- https://nlp.stanford.edu/data/glove.6B.zip
Connecting to nlp.stanford.edu (nlp.stanford.edu) | 171.64.67.140 | :443... connected.
HTTP request sent, awaiting response... 301 Moved Permanently
Location: https://downloads.cs.stanford.edu/nlp/data/glove.6B.zip [following]
--2024-05-05 20:24:03-- https://downloads.cs.stanford.edu/nlp/data/glove.6B.zip
Resolving downloads.cs.stanford.edu (downloads.cs.stanford.edu)... 171.64.64.22
Connecting to downloads.cs.stanford.edu (downloads.cs.stanford.edu) | 171.64.64.22 |:
443... connected.
HTTP request sent, awaiting response... 200 OK
Length: 862182613 (822M) [application/zip]
Saving to: 'glove.6B.zip'
glove.6B.zip
                   2024-05-05 20:26:45 (5.11 MB/s) - 'glove.6B.zip' saved [862182613/862182613]
```

Reading Glove Data

```
import numpy as np
path_to_glove_file = "glove.6B.100d.txt"

embeddings_index = {}
with open(path_to_glove_file) as f:
    for line in f:
        word, coefs = line.split(maxsplit=1)
        coefs = np.fromstring(coefs, "f", sep=" ")
        embeddings_index[word] = coefs

print(f"Found {len(embeddings_index)} word vectors.")
```

Found 400000 word vectors.

Setting Matrix for Glove Words

```
In [11]: embedding_dim = 100

vocabulary = text_vectorization.get_vocabulary()
word_index = dict(zip(vocabulary, range(len(vocabulary))))

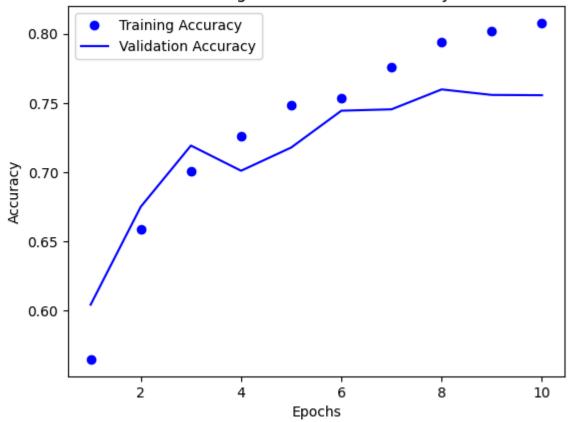
embedding_matrix = np.zeros((max_tokens, embedding_dim))
for word, i in word_index.items():
    if i < max_tokens:
        embedding_vector = embeddings_index.get(word)
    if embedding_vector is not None:
        embedding_matrix[i] = embedding_vector</pre>
In [12]: embedding_layer = layers.Embedding(
    max_tokens,
    embedding_dim,
    embeddings_initializer=keras.initializers.Constant(embedding_matrix),
        trainable=False,
    mask zero=True,
```

Model with Pretrained GloVe Embeddings

```
history = model.fit(int_train_ds, validation_data=int_val_ds, epochs=10, callbacks=
model = keras.models.load_model("glove_embeddings_sequence_model.x")
print(f"Test acc: {model.evaluate(int_test_ds)[1]:.3f}")

accuracy = history.history["accuracy"]
val_accuracy = history.history["val_accuracy"]
epochs = range(1, len(accuracy) + 1)
plt.figure()
plt.plot(epochs, accuracy, "bo", label="Training Accuracy")
plt.plot(epochs, val_accuracy, "b", label="Validation Accuracy")
plt.title("Training and validation Accuracy")
plt.xlabel("Epochs")
plt.ylabel("Accuracy")
plt.legend()
plt.show()
```

```
Layer (type)
                  Output Shape
                                    Param #
______
input_4 (InputLayer)
                  [(None, None)]
embedding_3 (Embedding)
                 (None, None, 100)
                                    1000000
bidirectional 3 (Bidirecti (None, 64)
                                    34048
onal)
dropout 3 (Dropout) (None, 64)
                                    0
dense_3 (Dense)
                  (None, 1)
                                    65
_____
Total params: 1034113 (3.94 MB)
Trainable params: 34113 (133.25 KB)
Non-trainable params: 1000000 (3.81 MB)
Epoch 1/10
y: 0.5647 - val_loss: 0.6587 - val_accuracy: 0.6042
Epoch 2/10
y: 0.6587 - val_loss: 0.6063 - val_accuracy: 0.6750
y: 0.7003 - val_loss: 0.5563 - val_accuracy: 0.7192
Epoch 4/10
100/100 [============== ] - 6s 60ms/step - loss: 0.5416 - accuracy:
0.7262 - val_loss: 0.5842 - val_accuracy: 0.7010
Epoch 5/10
y: 0.7487 - val_loss: 0.5558 - val_accuracy: 0.7178
Epoch 6/10
100/100 [=============] - 34s 344ms/step - loss: 0.4943 - accurac
y: 0.7538 - val_loss: 0.5172 - val_accuracy: 0.7444
Epoch 7/10
100/100 [============== ] - 5s 51ms/step - loss: 0.4772 - accuracy:
0.7759 - val_loss: 0.5214 - val_accuracy: 0.7454
Epoch 8/10
y: 0.7937 - val_loss: 0.4873 - val_accuracy: 0.7598
Epoch 9/10
100/100 [============== ] - 3s 33ms/step - loss: 0.4324 - accuracy:
0.8022 - val_loss: 0.5017 - val_accuracy: 0.7558
Epoch 10/10
100/100 [=============== ] - 4s 42ms/step - loss: 0.4318 - accuracy:
0.8078 - val loss: 0.4971 - val accuracy: 0.7556
y: 0.7618
Test acc: 0.762
```

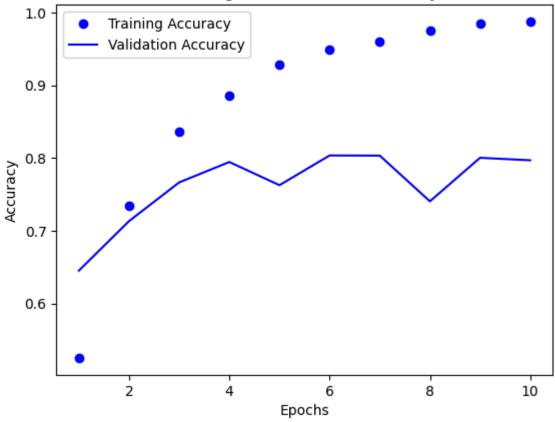


Comparing Model Performance with Different Training Set Sizes

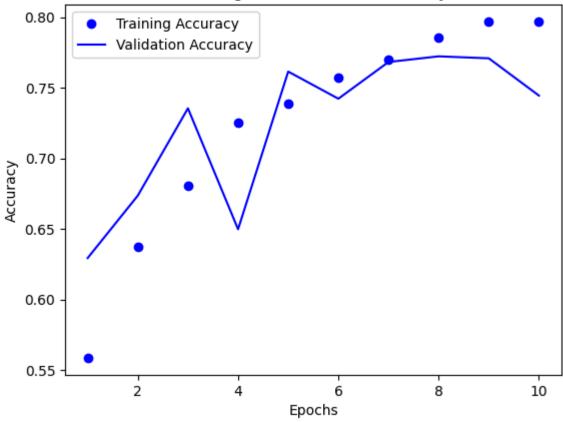
```
In [15]: train_sample_sizes = [100, 500, 1000, 5000, 10000, 20000]
         for train_size in train_sample_sizes:
             train_ds = keras.utils.text_dataset_from_directory(
                 "aclImdb/train", batch_size=batch_size
             ).take(train_size)
             int_train_ds = train_ds.map(
                 lambda x, y: (text_vectorization(x), y),
                 num_parallel_calls=4
             int_val_ds = val_ds.map(
                 lambda x, y: (text_vectorization(x), y),
                 num_parallel_calls=4
             int_test_ds = test_ds.map(
                 lambda x, y: (text_vectorization(x), y),
                 num_parallel_calls=4
             # Train and evaluate the model with the embedding layer
             embedding_layer = layers.Embedding(max_tokens, embedding_dim)
             inputs = keras.Input(shape=(None,), dtype="int64")
             embedded = embedding_layer(inputs)
             x = layers.Bidirectional(layers.LSTM(32))(embedded)
             x = layers.Dropout(0.5)(x)
```

```
outputs = layers.Dense(1, activation="sigmoid")(x)
model = keras.Model(inputs, outputs)
model.compile(optimizer="rmsprop",
              loss="binary_crossentropy",
              metrics=["accuracy"])
callbacks = [
    keras.callbacks.ModelCheckpoint("embeddings_model.x",
                                    save best only=True)
history = model.fit(int_train_ds, validation_data=int_val_ds, epochs=10, callba
model = keras.models.load_model("embeddings_model.x")
embedding_layer_test_acc = model.evaluate(int_test_ds)[1]
accuracy = history.history["accuracy"]
val_accuracy = history.history["val_accuracy"]
epochs = range(1, len(accuracy) + 1)
plt.figure()
plt.plot(epochs, accuracy, "bo", label="Training Accuracy")
plt.plot(epochs, val_accuracy, "b", label="Validation Accuracy")
plt.title("Training and validation Accuracy")
plt.xlabel("Epochs")
plt.ylabel("Accuracy")
plt.legend()
plt.show()
# Train and evaluate the model with the pretrained word embeddings
embedding_layer = layers.Embedding(
   max_tokens,
    embedding_dim,
    embeddings initializer=keras.initializers.Constant(embedding matrix),
   trainable=False,
   mask_zero=True,
)
inputs = keras.Input(shape=(None,), dtype="int64")
embedded = embedding layer(inputs)
x = layers.Bidirectional(layers.LSTM(32))(embedded)
x = layers.Dropout(0.5)(x)
outputs = layers.Dense(1, activation="sigmoid")(x)
model = keras.Model(inputs, outputs)
model.compile(optimizer="rmsprop",
              loss="binary_crossentropy",
              metrics=["accuracy"])
callbacks = [
    keras.callbacks.ModelCheckpoint("pretrained_embeddings_model.x",
                                    save_best_only=True)
history = model.fit(int train ds, validation data=int val ds, epochs=10, callba
model = keras.models.load_model("pretrained_embeddings_model.x")
pretrained_embeddings_test_acc = model.evaluate(int_test_ds)[1]
accuracy = history.history["accuracy"]
val_accuracy = history.history["val_accuracy"]
epochs = range(1, len(accuracy) + 1)
```

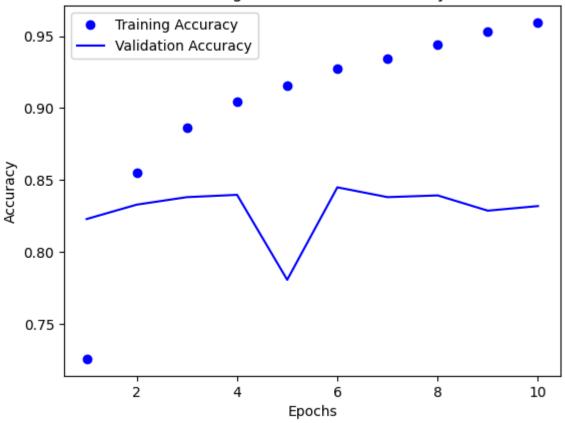
```
plt.figure()
   plt.plot(epochs, accuracy, "bo", label="Training Accuracy")
   plt.plot(epochs, val_accuracy, "b", label="Validation Accuracy")
   plt.title("Training and validation Accuracy")
   plt.xlabel("Epochs")
   plt.ylabel("Accuracy")
   plt.legend()
   plt.show()
   # Compare the performance and store the results
   print(f"Training samples: {train_size}")
   print(f"Embedding layer test accuracy: {embedding_layer_test_acc:.3f}")
   print(f"Pretrained embeddings test accuracy: {pretrained_embeddings_test_acc:.3
   print("-" * 50)
Found 20000 files belonging to 2 classes.
Epoch 1/10
y: 0.5253 - val_loss: 0.6515 - val_accuracy: 0.6452
Epoch 2/10
y: 0.7347 - val_loss: 0.5902 - val_accuracy: 0.7132
Epoch 3/10
100/100 [============] - 15s 147ms/step - loss: 0.4248 - accurac
y: 0.8363 - val_loss: 0.5746 - val_accuracy: 0.7664
Epoch 4/10
100/100 [============== ] - 18s 186ms/step - loss: 0.3156 - accurac
y: 0.8850 - val loss: 0.4597 - val accuracy: 0.7944
100/100 [=============== ] - 6s 56ms/step - loss: 0.2063 - accuracy:
0.9287 - val_loss: 0.6980 - val_accuracy: 0.7626
Epoch 6/10
100/100 [============== ] - 7s 69ms/step - loss: 0.1534 - accuracy:
0.9488 - val_loss: 0.4957 - val_accuracy: 0.8034
Epoch 7/10
100/100 [============== ] - 5s 50ms/step - loss: 0.1161 - accuracy:
0.9603 - val_loss: 0.5231 - val_accuracy: 0.8032
Epoch 8/10
100/100 [=============] - 3s 31ms/step - loss: 0.0867 - accuracy:
0.9750 - val_loss: 0.7628 - val_accuracy: 0.7404
Epoch 9/10
100/100 [============= ] - 4s 36ms/step - loss: 0.0609 - accuracy:
0.9841 - val_loss: 0.6994 - val_accuracy: 0.8002
Epoch 10/10
100/100 [=============== ] - 6s 58ms/step - loss: 0.0465 - accuracy:
0.9875 - val_loss: 0.6768 - val_accuracy: 0.7968
y: 0.7889
```



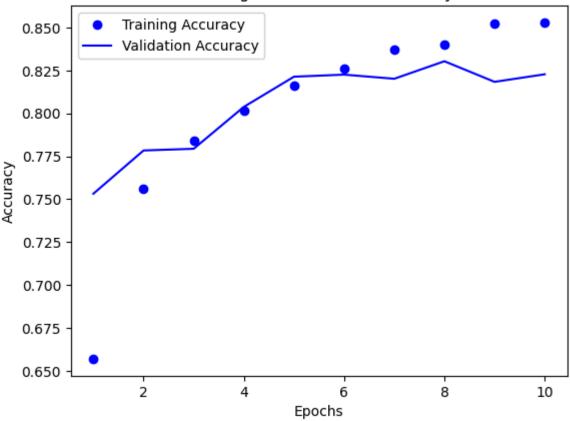
```
Epoch 1/10
y: 0.5587 - val_loss: 0.6520 - val_accuracy: 0.6294
Epoch 2/10
y: 0.6375 - val_loss: 0.6060 - val_accuracy: 0.6736
Epoch 3/10
100/100 [===========] - 33s 330ms/step - loss: 0.5910 - accurac
y: 0.6809 - val_loss: 0.5536 - val_accuracy: 0.7354
Epoch 4/10
y: 0.7250 - val_loss: 0.6530 - val_accuracy: 0.6498
Epoch 5/10
y: 0.7387 - val_loss: 0.5009 - val_accuracy: 0.7614
Epoch 6/10
100/100 [============== ] - 5s 48ms/step - loss: 0.5002 - accuracy:
0.7572 - val_loss: 0.5209 - val_accuracy: 0.7422
Epoch 7/10
y: 0.7697 - val_loss: 0.4848 - val_accuracy: 0.7682
Epoch 8/10
y: 0.7856 - val_loss: 0.4776 - val_accuracy: 0.7722
Epoch 9/10
100/100 [===========] - 36s 366ms/step - loss: 0.4448 - accurac
y: 0.7969 - val_loss: 0.4758 - val_accuracy: 0.7708
Epoch 10/10
100/100 [============== ] - 6s 61ms/step - loss: 0.4337 - accuracy:
0.7969 - val_loss: 0.5043 - val_accuracy: 0.7444
782/782 [============= ] - 11s 9ms/step - loss: 0.4732 - accuracy:
0.7753
```



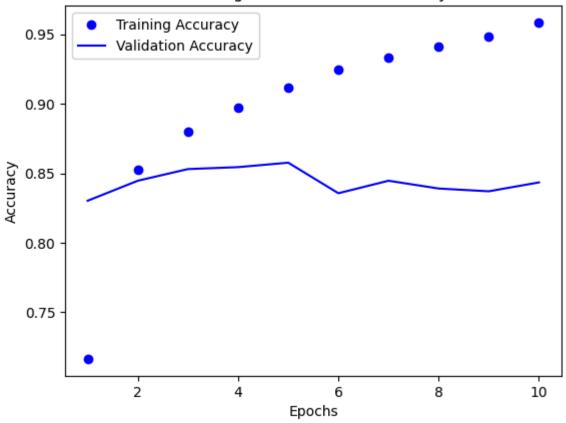
```
Training samples: 100
Embedding layer test accuracy: 0.789
Pretrained embeddings test accuracy: 0.775
-----
Found 20000 files belonging to 2 classes.
Epoch 1/10
500/500 [================= ] - 50s 89ms/step - loss: 0.5362 - accurac
y: 0.7259 - val_loss: 0.4049 - val_accuracy: 0.8230
y: 0.8549 - val_loss: 0.3829 - val_accuracy: 0.8330
y: 0.8863 - val_loss: 0.3720 - val_accuracy: 0.8382
Epoch 4/10
y: 0.9043 - val_loss: 0.3731 - val_accuracy: 0.8398
Epoch 5/10
y: 0.9159 - val_loss: 0.4701 - val_accuracy: 0.7808
Epoch 6/10
500/500 [============] - 13s 25ms/step - loss: 0.2053 - accurac
y: 0.9276 - val_loss: 0.3955 - val_accuracy: 0.8450
Epoch 7/10
y: 0.9346 - val_loss: 0.4021 - val_accuracy: 0.8382
500/500 [============ - - 10s 21ms/step - loss: 0.1622 - accurac
y: 0.9441 - val_loss: 0.4111 - val_accuracy: 0.8394
Epoch 9/10
500/500 [============= - - 11s 23ms/step - loss: 0.1354 - accurac
y: 0.9532 - val_loss: 0.4602 - val_accuracy: 0.8288
Epoch 10/10
500/500 [============] - 14s 28ms/step - loss: 0.1198 - accurac
y: 0.9594 - val_loss: 0.5685 - val_accuracy: 0.8320
0.8285
```



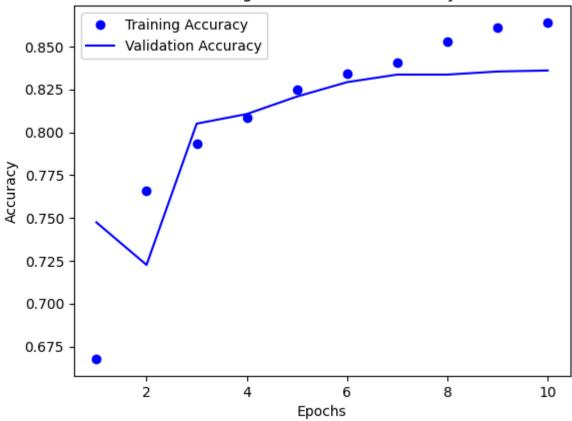
```
Epoch 1/10
500/500 [================== ] - 65s 110ms/step - loss: 0.6142 - accurac
y: 0.6569 - val_loss: 0.5141 - val_accuracy: 0.7532
Epoch 2/10
y: 0.7564 - val_loss: 0.4610 - val_accuracy: 0.7784
500/500 [============] - 47s 94ms/step - loss: 0.4614 - accurac
y: 0.7842 - val_loss: 0.4580 - val_accuracy: 0.7794
Epoch 4/10
y: 0.8017 - val_loss: 0.4253 - val_accuracy: 0.8038
Epoch 5/10
y: 0.8162 - val_loss: 0.3957 - val_accuracy: 0.8214
Epoch 6/10
y: 0.8260 - val_loss: 0.3908 - val_accuracy: 0.8226
Epoch 7/10
y: 0.8372 - val_loss: 0.4012 - val_accuracy: 0.8202
Epoch 8/10
y: 0.8399 - val_loss: 0.3836 - val_accuracy: 0.8304
Epoch 9/10
500/500 [============= - - 19s 37ms/step - loss: 0.3484 - accurac
y: 0.8523 - val_loss: 0.4024 - val_accuracy: 0.8184
Epoch 10/10
y: 0.8530 - val_loss: 0.3922 - val_accuracy: 0.8228
y: 0.8246
```



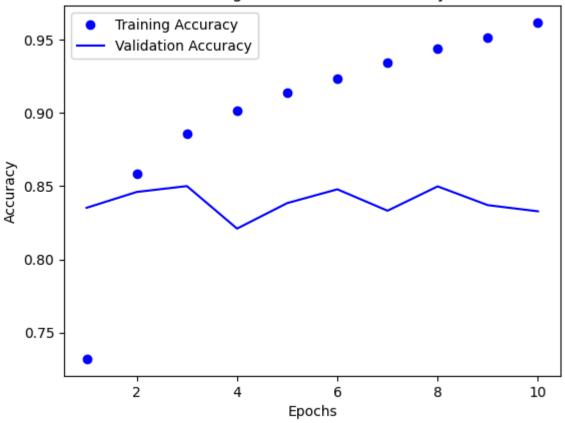
```
Training samples: 500
Embedding layer test accuracy: 0.828
Pretrained embeddings test accuracy: 0.825
-----
Found 20000 files belonging to 2 classes.
Epoch 1/10
y: 0.7164 - val_loss: 0.3814 - val_accuracy: 0.8302
y: 0.8525 - val_loss: 0.3560 - val_accuracy: 0.8446
y: 0.8802 - val_loss: 0.3681 - val_accuracy: 0.8530
Epoch 4/10
y: 0.8969 - val_loss: 0.3847 - val_accuracy: 0.8544
Epoch 5/10
y: 0.9119 - val_loss: 0.3718 - val_accuracy: 0.8576
Epoch 6/10
y: 0.9247 - val_loss: 0.4726 - val_accuracy: 0.8356
Epoch 7/10
y: 0.9331 - val_loss: 0.4748 - val_accuracy: 0.8446
y: 0.9415 - val_loss: 0.4202 - val_accuracy: 0.8390
Epoch 9/10
y: 0.9485 - val_loss: 0.4955 - val_accuracy: 0.8370
Epoch 10/10
625/625 [============] - 13s 21ms/step - loss: 0.1201 - accurac
y: 0.9585 - val_loss: 0.5371 - val_accuracy: 0.8434
y: 0.8372
```



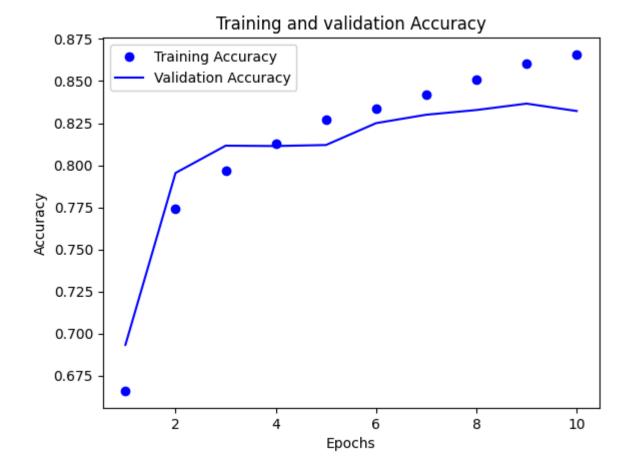
```
Epoch 1/10
y: 0.6678 - val_loss: 0.5177 - val_accuracy: 0.7474
Epoch 2/10
y: 0.7661 - val_loss: 0.5948 - val_accuracy: 0.7226
Epoch 3/10
625/625 [============] - 50s 80ms/step - loss: 0.4474 - accurac
y: 0.7932 - val_loss: 0.4179 - val_accuracy: 0.8050
Epoch 4/10
y: 0.8083 - val_loss: 0.4026 - val_accuracy: 0.8106
Epoch 5/10
y: 0.8246 - val_loss: 0.3892 - val_accuracy: 0.8208
Epoch 6/10
y: 0.8339 - val_loss: 0.3741 - val_accuracy: 0.8292
Epoch 7/10
y: 0.8407 - val_loss: 0.3740 - val_accuracy: 0.8336
Epoch 8/10
y: 0.8532 - val_loss: 0.3799 - val_accuracy: 0.8336
Epoch 9/10
625/625 [============] - 22s 35ms/step - loss: 0.3260 - accurac
y: 0.8612 - val_loss: 0.3806 - val_accuracy: 0.8354
Epoch 10/10
y: 0.8640 - val_loss: 0.3646 - val_accuracy: 0.8360
y: 0.8378
```



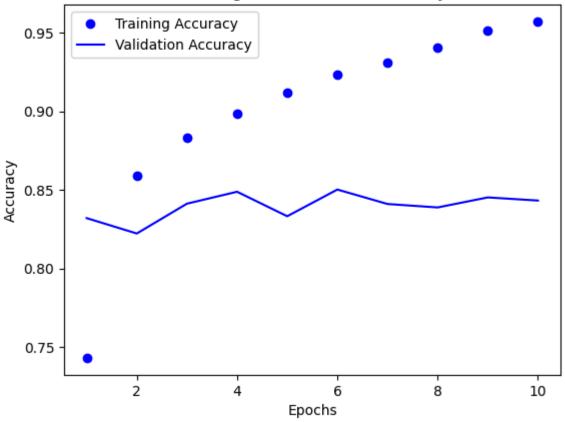
```
Training samples: 1000
Embedding layer test accuracy: 0.837
Pretrained embeddings test accuracy: 0.838
-----
Found 20000 files belonging to 2 classes.
Epoch 1/10
y: 0.7322 - val_loss: 0.3742 - val_accuracy: 0.8352
y: 0.8587 - val_loss: 0.3535 - val_accuracy: 0.8460
y: 0.8857 - val_loss: 0.3630 - val_accuracy: 0.8500
Epoch 4/10
y: 0.9017 - val_loss: 0.4089 - val_accuracy: 0.8210
Epoch 5/10
y: 0.9135 - val_loss: 0.4178 - val_accuracy: 0.8384
Epoch 6/10
y: 0.9232 - val_loss: 0.4147 - val_accuracy: 0.8478
Epoch 7/10
y: 0.9341 - val_loss: 0.5501 - val_accuracy: 0.8332
Epoch 8/10
y: 0.9437 - val_loss: 0.4653 - val_accuracy: 0.8498
Epoch 9/10
y: 0.9514 - val_loss: 0.5286 - val_accuracy: 0.8370
Epoch 10/10
y: 0.9616 - val_loss: 0.6006 - val_accuracy: 0.8328
y: 0.8376
```



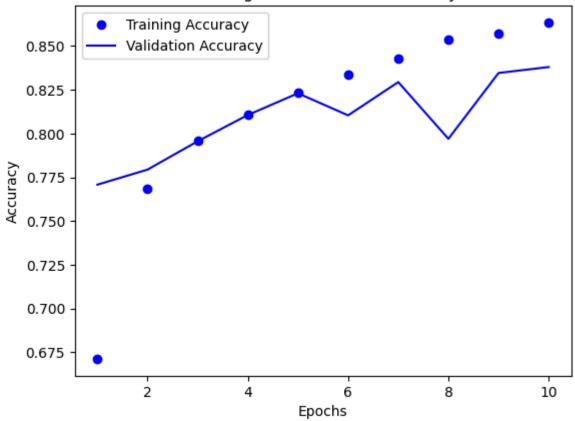
```
Epoch 1/10
y: 0.6659 - val_loss: 0.6257 - val_accuracy: 0.6932
Epoch 2/10
y: 0.7739 - val_loss: 0.4402 - val_accuracy: 0.7954
Epoch 3/10
625/625 [============] - 49s 79ms/step - loss: 0.4426 - accurac
y: 0.7965 - val_loss: 0.4129 - val_accuracy: 0.8116
Epoch 4/10
y: 0.8126 - val_loss: 0.4109 - val_accuracy: 0.8114
Epoch 5/10
y: 0.8274 - val_loss: 0.4202 - val_accuracy: 0.8120
Epoch 6/10
y: 0.8335 - val_loss: 0.3949 - val_accuracy: 0.8250
Epoch 7/10
y: 0.8421 - val_loss: 0.3904 - val_accuracy: 0.8300
Epoch 8/10
y: 0.8506 - val_loss: 0.4032 - val_accuracy: 0.8328
Epoch 9/10
y: 0.8602 - val_loss: 0.3776 - val_accuracy: 0.8366
Epoch 10/10
y: 0.8658 - val_loss: 0.4097 - val_accuracy: 0.8322
y: 0.8359
```



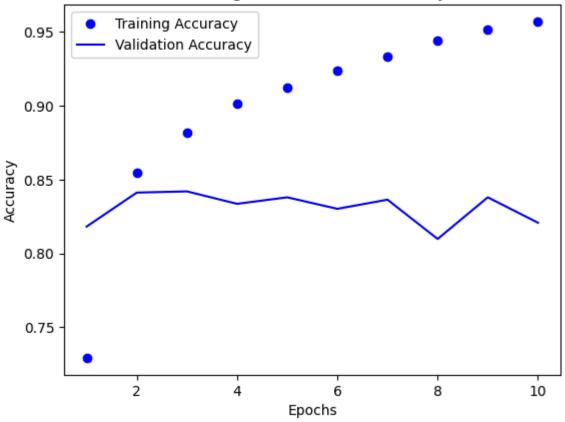
```
Training samples: 5000
Embedding layer test accuracy: 0.838
Pretrained embeddings test accuracy: 0.836
-----
Found 20000 files belonging to 2 classes.
Epoch 1/10
y: 0.7430 - val_loss: 0.3835 - val_accuracy: 0.8320
y: 0.8588 - val_loss: 0.4187 - val_accuracy: 0.8222
y: 0.8831 - val_loss: 0.3717 - val_accuracy: 0.8412
Epoch 4/10
y: 0.8985 - val_loss: 0.3656 - val_accuracy: 0.8488
Epoch 5/10
y: 0.9119 - val_loss: 0.4102 - val_accuracy: 0.8332
Epoch 6/10
y: 0.9235 - val_loss: 0.4333 - val_accuracy: 0.8502
Epoch 7/10
y: 0.9309 - val_loss: 0.4120 - val_accuracy: 0.8410
Epoch 8/10
y: 0.9403 - val_loss: 0.4663 - val_accuracy: 0.8388
Epoch 9/10
y: 0.9513 - val_loss: 0.4868 - val_accuracy: 0.8452
Epoch 10/10
625/625 [============] - 18s 29ms/step - loss: 0.1223 - accurac
y: 0.9572 - val_loss: 0.5394 - val_accuracy: 0.8432
0.8364
```



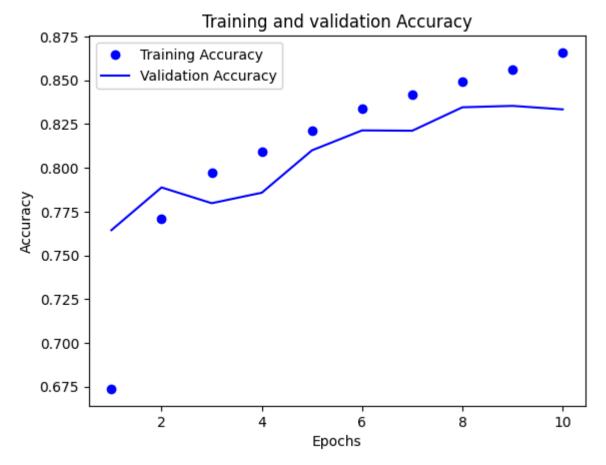
```
Epoch 1/10
y: 0.6713 - val_loss: 0.4875 - val_accuracy: 0.7708
Epoch 2/10
y: 0.7688 - val_loss: 0.4587 - val_accuracy: 0.7794
Epoch 3/10
625/625 [============] - 47s 76ms/step - loss: 0.4413 - accurac
y: 0.7957 - val_loss: 0.4436 - val_accuracy: 0.7956
Epoch 4/10
y: 0.8106 - val_loss: 0.4149 - val_accuracy: 0.8106
Epoch 5/10
y: 0.8236 - val_loss: 0.3888 - val_accuracy: 0.8230
Epoch 6/10
y: 0.8338 - val_loss: 0.4113 - val_accuracy: 0.8104
Epoch 7/10
y: 0.8431 - val_loss: 0.3859 - val_accuracy: 0.8294
Epoch 8/10
y: 0.8535 - val_loss: 0.4620 - val_accuracy: 0.7970
Epoch 9/10
y: 0.8572 - val_loss: 0.3661 - val_accuracy: 0.8346
Epoch 10/10
y: 0.8634 - val_loss: 0.3730 - val_accuracy: 0.8380
y: 0.8334
```



```
Training samples: 10000
Embedding layer test accuracy: 0.836
Pretrained embeddings test accuracy: 0.833
-----
Found 20000 files belonging to 2 classes.
Epoch 1/10
y: 0.7293 - val_loss: 0.4052 - val_accuracy: 0.8182
y: 0.8547 - val_loss: 0.3709 - val_accuracy: 0.8412
y: 0.8820 - val_loss: 0.3611 - val_accuracy: 0.8420
Epoch 4/10
y: 0.9017 - val_loss: 0.3912 - val_accuracy: 0.8336
Epoch 5/10
y: 0.9121 - val_loss: 0.4071 - val_accuracy: 0.8380
Epoch 6/10
y: 0.9237 - val_loss: 0.4005 - val_accuracy: 0.8302
Epoch 7/10
y: 0.9330 - val_loss: 0.4021 - val_accuracy: 0.8364
625/625 [============= - 19s 30ms/step - loss: 0.1580 - accurac
y: 0.9439 - val_loss: 0.5228 - val_accuracy: 0.8098
Epoch 9/10
y: 0.9513 - val_loss: 0.4990 - val_accuracy: 0.8380
Epoch 10/10
625/625 [============] - 23s 37ms/step - loss: 0.1222 - accurac
y: 0.9571 - val_loss: 0.4844 - val_accuracy: 0.8208
782/782 [============== ] - 9s 10ms/step - loss: 0.3827 - accuracy:
0.8335
```



```
Epoch 1/10
y: 0.6737 - val_loss: 0.4875 - val_accuracy: 0.7644
Epoch 2/10
y: 0.7711 - val_loss: 0.4448 - val_accuracy: 0.7888
Epoch 3/10
625/625 [===========] - 19s 30ms/step - loss: 0.4434 - accurac
y: 0.7973 - val_loss: 0.4606 - val_accuracy: 0.7798
Epoch 4/10
y: 0.8093 - val_loss: 0.4436 - val_accuracy: 0.7858
Epoch 5/10
y: 0.8215 - val_loss: 0.4152 - val_accuracy: 0.8100
Epoch 6/10
y: 0.8337 - val_loss: 0.3919 - val_accuracy: 0.8214
Epoch 7/10
y: 0.8420 - val_loss: 0.3823 - val_accuracy: 0.8212
Epoch 8/10
y: 0.8492 - val_loss: 0.3693 - val_accuracy: 0.8346
Epoch 9/10
y: 0.8563 - val_loss: 0.3636 - val_accuracy: 0.8354
Epoch 10/10
y: 0.8659 - val_loss: 0.3743 - val_accuracy: 0.8334
y: 0.8388
```



Training samples: 20000

Embedding layer test accuracy: 0.833 Pretrained embeddings test accuracy: 0.839