hw4 sequence

November 23, 2023

[5]: # Preparing the data

```
# Please be advised, I downloaded the data and unzip it_{\square}
       \hookrightarrow from my computer then I put the data folder in the same path of my library.
       \hookrightarrow folder.
 []: # !!!!!!!!!the really tricky part for glove, is that even the data is
       ⇔encoded in utf-8 but the model still report error.
      # !!!!!it took me couple hours to solve this by changing the encoding to ANSIL
       →which works better for english letters.
[27]: import urllib.request
      import zipfile
      import os
      url = 'http://nlp.stanford.edu/data/glove.6B.zip'
      file_name = 'glove.6B.zip'
      urllib.request.urlretrieve(url, file_name)
      with zipfile.ZipFile(file_name, 'r') as zip_ref:
          zip_ref.extractall()
      os.remove(file_name)
 []: # set-up train valid test, 6.7s
 [8]: import os, pathlib, shutil, random
      from tensorflow import keras
      batch_size = 32
      base_dir = pathlib.Path("aclImdb")
      val_dir = base_dir / "val"
      train_dir = base_dir / "train"
      for category in ("neg", "pos"):
          os.makedirs(val_dir / category)
          files = os.listdir(train_dir / category)
```

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

```
[7]: # Preparing integer sequence datasets, 33.3s
```

```
[9]: from tensorflow.keras import layers
     max length = 150
     max_tokens = 20000
     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)
```

```
[3]: # vocabulary = text_vectorization.get_vocabulary()
```

[9]: # A sequence model built on one-hot encoded vector sequences, 0.3s

Model: "model_1"

Layer (type)	Output Shape	Param #
input_2 (InputLayer)	[(None, None)]	0
tf.one_hot (TFOpLambda)	(None, None, 20000)	0
<pre>bidirectional_1 (Bidirectio nal)</pre>	(None, 64)	5128448
<pre>dropout_1 (Dropout)</pre>	(None, 64)	0
dense_1 (Dense)	(None, 1)	65

Total params: 5,128,513 Trainable params: 5,128,513 Non-trainable params: 0

```
[11]: callbacks = [
        keras.callbacks.ModelCheckpoint("one_hot_bidir_lstm.keras",
                                    save_best_only=True)
     model.fit(int_train_ds, validation_data=int_val_ds, epochs=10,
             callbacks=callbacks)
     model = keras.models.load_model("one_hot_bidir_lstm.keras")
     print(f"Test acc: {model.evaluate(int_test_ds)[1]:.3f}")
    Epoch 1/10
    625/625 [============ ] - 35s 53ms/step - loss: 0.5104 -
    accuracy: 0.7545 - val_loss: 0.3515 - val_accuracy: 0.8446
    Epoch 2/10
    625/625 [============= ] - 28s 45ms/step - loss: 0.3236 -
    accuracy: 0.8701 - val_loss: 0.3342 - val_accuracy: 0.8576
    Epoch 3/10
    625/625 [============= ] - 27s 43ms/step - loss: 0.2628 -
    accuracy: 0.8968 - val_loss: 0.4067 - val_accuracy: 0.8546
    Epoch 4/10
    625/625 [=========== ] - 27s 43ms/step - loss: 0.2233 -
    accuracy: 0.9144 - val_loss: 0.3792 - val_accuracy: 0.8592
    Epoch 5/10
    625/625 [============== ] - 27s 43ms/step - loss: 0.1894 -
    accuracy: 0.9295 - val_loss: 0.3924 - val_accuracy: 0.8532
    Epoch 6/10
    625/625 [============= ] - 27s 43ms/step - loss: 0.1625 -
    accuracy: 0.9409 - val_loss: 0.3937 - val_accuracy: 0.8590
    Epoch 7/10
    accuracy: 0.9500 - val_loss: 0.3934 - val_accuracy: 0.8528
    Epoch 8/10
    625/625 [============ ] - 27s 43ms/step - loss: 0.1053 -
    accuracy: 0.9643 - val_loss: 0.4923 - val_accuracy: 0.8548
    Epoch 9/10
    625/625 [============ ] - 27s 43ms/step - loss: 0.0804 -
    accuracy: 0.9719 - val loss: 0.5940 - val accuracy: 0.8450
    Epoch 10/10
    accuracy: 0.9803 - val_loss: 0.6225 - val_accuracy: 0.8382
    782/782 [============= ] - 31s 39ms/step - loss: 0.3611 -
    accuracy: 0.8409
    Test acc: 0.841
[]: # Learning word embeddings with the Embedding layer
[12]: # Instantiating an Embedding layer
     embedding layer = layers.Embedding(input_dim=max_tokens, output_dim=256)
```

```
[]: | # Model that uses an Embedding layer trained from scratch, 3m6s
                          # Test acc: 0.835, bad model even worse compared to \Box
      \rightarrownaive base model (0.846).
[13]: 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 = \Gamma
        keras.callbacks.ModelCheckpoint("embeddings_bidir_gru.keras",
                                     save_best_only=True)
     ]
     model.fit(int_train_ds, validation_data=int_val_ds, epochs=10,_
      ⇔callbacks=callbacks)
     model = keras.models.load_model("embeddings_bidir_gru.keras")
     print(f"Test acc: {model.evaluate(int_test_ds)[1]:.3f}")
    Model: "model_2"
     Layer (type) Output Shape
                                                    Param #
    ______
     input_3 (InputLayer) [(None, None)]
     embedding_2 (Embedding) (None, None, 256)
                                                    5120000
     bidirectional_2 (Bidirectio (None, 64)
                                                    73984
     nal)
     dropout_2 (Dropout) (None, 64)
     dense 2 (Dense)
                              (None, 1)
                                                     65
    ______
    Total params: 5,194,049
    Trainable params: 5,194,049
    Non-trainable params: 0
                        ._____
    Epoch 1/10
    625/625 [============= ] - 42s 63ms/step - loss: 0.4678 -
    accuracy: 0.7829 - val_loss: 0.3604 - val_accuracy: 0.8434
    Epoch 2/10
```

```
accuracy: 0.8830 - val_loss: 0.4093 - val_accuracy: 0.8284
    Epoch 3/10
    accuracy: 0.9113 - val_loss: 0.3898 - val_accuracy: 0.8438
    Epoch 4/10
    625/625 [============ ] - 11s 18ms/step - loss: 0.1871 -
    accuracy: 0.9301 - val_loss: 0.4383 - val_accuracy: 0.8422
    Epoch 5/10
    625/625 [============ ] - 12s 19ms/step - loss: 0.1526 -
    accuracy: 0.9440 - val_loss: 0.4210 - val_accuracy: 0.8470
    625/625 [=========== ] - 12s 18ms/step - loss: 0.1170 -
    accuracy: 0.9589 - val_loss: 0.4676 - val_accuracy: 0.8360
    625/625 [=========== ] - 11s 18ms/step - loss: 0.0872 -
    accuracy: 0.9701 - val_loss: 0.5490 - val_accuracy: 0.8394
    accuracy: 0.9789 - val loss: 0.5826 - val accuracy: 0.8410
    Epoch 9/10
    625/625 [============= ] - 11s 18ms/step - loss: 0.0492 -
    accuracy: 0.9835 - val_loss: 0.6555 - val_accuracy: 0.8396
    Epoch 10/10
    625/625 [=========== ] - 11s 18ms/step - loss: 0.0326 -
    accuracy: 0.9891 - val_loss: 0.8134 - val_accuracy: 0.8338
    accuracy: 0.8354
    Test acc: 0.835
[]: # Using an Embedding layer with masking enabled, 3min25s
                 # Test acc: 0.838, a little better with the previous model (0.
      ↔835)
[14]: 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 = [
```

625/625 [============] - 11s 18ms/step - loss: 0.2969 -

```
keras.callbacks.ModelCheckpoint("embeddings_bidir_gru_with_masking.keras",
                       save_best_only=True)
model.fit(int_train_ds, validation_data=int_val_ds, epochs=10,__
→callbacks=callbacks)
model = keras.models.load model("embeddings bidir gru with masking.keras")
print(f"Test acc: {model.evaluate(int test ds)[1]:.3f}")
Model: "model_3"
Layer (type)
                             Param #
                 Output Shape
_____
input_4 (InputLayer)
                 [(None, None)]
embedding_3 (Embedding) (None, None, 256) 5120000
bidirectional_3 (Bidirectio (None, 64)
                                  73984
nal)
dropout_3 (Dropout) (None, 64)
dense_3 (Dense)
                  (None, 1)
                                  65
_____
Total params: 5,194,049
Trainable params: 5,194,049
Non-trainable params: 0
            _____
accuracy: 0.7994 - val_loss: 0.3346 - val_accuracy: 0.8534
accuracy: 0.8907 - val_loss: 0.3533 - val_accuracy: 0.8492
Epoch 3/10
accuracy: 0.9197 - val_loss: 0.3806 - val_accuracy: 0.8552
Epoch 4/10
accuracy: 0.9419 - val_loss: 0.4678 - val_accuracy: 0.8442
Epoch 5/10
625/625 [=========== ] - 18s 29ms/step - loss: 0.1177 -
accuracy: 0.9582 - val_loss: 0.4705 - val_accuracy: 0.8516
Epoch 6/10
accuracy: 0.9692 - val loss: 0.4879 - val accuracy: 0.8332
Epoch 7/10
```

```
accuracy: 0.9773 - val_loss: 0.5848 - val_accuracy: 0.8358
     Epoch 8/10
     625/625 [============ ] - 19s 30ms/step - loss: 0.0448 -
     accuracy: 0.9843 - val_loss: 0.6837 - val_accuracy: 0.8192
     Epoch 9/10
     625/625 [=========== ] - 19s 30ms/step - loss: 0.0306 -
     accuracy: 0.9899 - val loss: 0.7022 - val accuracy: 0.8188
     Epoch 10/10
     625/625 [=========== ] - 19s 30ms/step - loss: 0.0219 -
     accuracy: 0.9927 - val_loss: 0.7805 - val_accuracy: 0.8346
     782/782 [============ ] - 12s 14ms/step - loss: 0.3662 -
     accuracy: 0.8380
     Test acc: 0.838
 []: # Parsing the GloVe word-embeddings file, 8.5 s
[15]: 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.
 []: # Preparing the GloVe word-embeddings matrix, Os
[16]: 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:</pre>
             embedding_vector = embeddings_index.get(word)
         if embedding_vector is not None:
             embedding_matrix[i] = embedding_vector
[17]: embedding_layer = layers.Embedding(
         max_tokens,
         embedding_dim,
         embeddings_initializer=keras.initializers.Constant(embedding_matrix),
```

```
trainable=False,
        mask_zero=True,
[]: # Model that uses a pretrained Embedding layer, 2m55
                                             !!!!!!!!!!!!!Test acc: 0.824,
      explanatin for this, for this serise of trials, my sample size is large
      senough to let even the most naive model to learn from embedding from scratch.
     # !!!!!!!!!I am expecting that train the same thing on a small sample will \sqcup
      [18]: 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"])
     model.summary()
     callbacks = [
        keras.callbacks.ModelCheckpoint("glove_embeddings_sequence_model.keras",
                                    save_best_only=True)
     model.fit(int_train_ds, validation_data=int_val_ds, epochs=10,__
      ⇔callbacks=callbacks)
     model = keras.models.load model("glove embeddings sequence model.keras")
     print(f"Test acc: {model.evaluate(int_test_ds)[1]:.3f}")
    Model: "model_4"
     Layer (type)
                            Output Shape
                                                   Param #
    ______
     input_5 (InputLayer)
                            [(None, None)]
     embedding_4 (Embedding) (None, None, 100)
                                                   2000000
     bidirectional_4 (Bidirectio (None, 64)
                                                    34048
     nal)
     dropout_4 (Dropout)
                             (None, 64)
```

(None, 1)

dense_4 (Dense)

65

Total params: 2,034,113
Trainable params: 34,113

Non-trainable params: 2,000,000

```
_____
Epoch 1/10
accuracy: 0.6873 - val_loss: 0.4729 - val_accuracy: 0.7728
Epoch 2/10
625/625 [============= ] - 15s 23ms/step - loss: 0.4799 -
accuracy: 0.7753 - val_loss: 0.5094 - val_accuracy: 0.7744
Epoch 3/10
accuracy: 0.7994 - val_loss: 0.4728 - val_accuracy: 0.7840
Epoch 4/10
accuracy: 0.8153 - val_loss: 0.4099 - val_accuracy: 0.8162
Epoch 5/10
accuracy: 0.8253 - val_loss: 0.3988 - val_accuracy: 0.8242
Epoch 6/10
625/625 [=========== ] - 15s 24ms/step - loss: 0.3680 -
accuracy: 0.8399 - val_loss: 0.4150 - val_accuracy: 0.8292
Epoch 7/10
accuracy: 0.8454 - val_loss: 0.4728 - val_accuracy: 0.8136
Epoch 8/10
accuracy: 0.8565 - val_loss: 0.4011 - val_accuracy: 0.8350
accuracy: 0.8658 - val_loss: 0.4038 - val_accuracy: 0.8358
Epoch 10/10
625/625 [=========== ] - 16s 26ms/step - loss: 0.3057 -
accuracy: 0.8708 - val_loss: 0.4132 - val_accuracy: 0.8346
782/782 [============ ] - 11s 12ms/step - loss: 0.3933 -
accuracy: 0.8239
Test acc: 0.824
```

hw4_part2

November 23, 2023

[]: # Recall from my part1 notebook, I did the following things

```
1. I set up the max word cut off as 150
                                  2. I did not limit the train valid sample size
       \hookrightarrow (using the whole data I have in hand)
                              # Because of this enriched data, pretrained model did_
       ⇔not perform super good.
      # In this part2 notebook, I will shrink the training/valid size.
[18]: # set-up train valid test to 100(50/50) 10000(5000/5000) 25000(12500/12500)
      import os, pathlib, shutil, random
      from tensorflow import keras
      batch size = 32
      base_dir = pathlib.Path("aclImdb")
      val_dir = base_dir / "val"
      train_dir = base_dir / "train"
      for category in ("neg", "pos"):
          os.makedirs(val_dir / category)
          files = os.listdir(train_dir / category)
          random.Random(1337).shuffle(files)
          num_val_samples = int(0.4 * len(files))
          val_files = files[-num_val_samples:]
          for fname in val_files:
              shutil.move(train dir / category / fname,
                          val_dir / category / fname)
      # I have no idea about how to keep the training set to a specific number
      # so I did it in a very straight non-fancy way
      # previous block(code) has transfer 20% training data to valid data (25000*0.2)
      # so here I only need to keep the neg and pos folder under train folder to 50_{\square}
       ⇔obs randomly.
             # path of folder
      folder_path = r'C:\Users\zhong\Desktop\HW4\aclImdb\train\neg'
```

```
# go through all file in the folder
all_files = os.listdir(folder_path)
         # set up the number of file that I want to keep
num_files_to_keep = 50
         # randomly keep the file
files_to_keep = random.sample(all_files, num_files_to_keep)
        # delete the file that I dont need.
for file_name in all_files:
   file_path = os.path.join(folder_path, file_name)
    if file_name not in files_to_keep:
        os.remove(file_path)
# Repeat the same thing for pos folder.
folder_path = r'C:\Users\zhong\Desktop\HW4\aclImdb\train\pos'
#
all_files = os.listdir(folder_path)
num files to keep = 50
files_to_keep = random.sample(all_files, num_files_to_keep)
for file_name in all_files:
   file_path = os.path.join(folder_path, file_name)
   if file_name not in files_to_keep:
        os.remove(file_path)
# show the number of train valid test
train ds = keras.utils.text dataset from directory(
    "aclImdb/train", batch_size=batch_size
val_ds = keras.utils.text_dataset_from_directory(
    "aclImdb/val", batch_size=batch_size
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 100 files belonging to 2 classes.
     Found 10000 files belonging to 2 classes.
     Found 25000 files belonging to 2 classes.
[21]: # Preparing integer sequence datasets
      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)
[23]: # A sequence model built on one-hot encoded vector sequences
      import tensorflow as tf
      inputs = keras.Input(shape=(None,), dtype="int64")
      embedded = tf.one_hot(inputs, depth=max_tokens)
      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()
      # RNN-LSTM
     Model: "model"
      Layer (type)
                                  Output Shape
                                                             Param #
```

```
tf.one_hot (TFOpLambda)
                        (None, None, 10000)
    bidirectional (Bidirectiona (None, 64)
                                           2568448
    dropout (Dropout)
                        (None, 64)
    dense (Dense)
                        (None, 1)
                                           65
   ______
   Total params: 2,568,513
   Trainable params: 2,568,513
   Non-trainable params: 0
[]: # Training a first basic sequence model 1m29s
    # Test acc: 0.575 which is very bad, this is because we have limited training _{f L}
     ⇔size, so without embedding, model can not map the pattern
[24]: callbacks = [
       keras.callbacks.ModelCheckpoint("one_hot_bidir_lstm.keras",
                              save_best_only=True)
    model.fit(int_train_ds, validation_data=int_val_ds, epochs=10,
           callbacks=callbacks)
    model = keras.models.load model("one hot bidir lstm.keras")
    print(f"Test acc: {model.evaluate(int_test_ds)[1]:.3f}")
   Epoch 1/10
   0.5500 - val_loss: 0.6928 - val_accuracy: 0.5068
   Epoch 2/10
   0.6300 - val_loss: 0.6921 - val_accuracy: 0.5107
   Epoch 3/10
   0.8500 - val_loss: 0.6911 - val_accuracy: 0.5291
   Epoch 4/10
   4/4 [============= ] - 4s 1s/step - loss: 0.6476 - accuracy:
   0.8600 - val_loss: 0.9437 - val_accuracy: 0.5066
   Epoch 5/10
   0.8000 - val_loss: 0.6893 - val_accuracy: 0.5327
   Epoch 6/10
```

[(None, None)]

input_1 (InputLayer)

```
0.8400 - val_loss: 0.6684 - val_accuracy: 0.5834
    Epoch 7/10
    0.9500 - val_loss: 0.7758 - val_accuracy: 0.5583
    Epoch 8/10
    0.9400 - val_loss: 0.7128 - val_accuracy: 0.6036
    Epoch 9/10
    0.9600 - val_loss: 0.7507 - val_accuracy: 0.5983
    Epoch 10/10
    0.9600 - val_loss: 0.9619 - val_accuracy: 0.5843
    782/782 [============= ] - 31s 38ms/step - loss: 0.6699 -
    accuracy: 0.5746
    Test acc: 0.575
[25]: # Learning word embeddings with the Embedding layer
     # Instantiating an Embedding layer
    embedding layer = layers.Embedding(input_dim=max_tokens, output_dim=256)
[]: # Model that uses an Embedding layer trained from scratch, 35.8s
     # Test acc: 0.614, improved.
     # embedding is very useful with a limited training size.
[26]: 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.keras",
                                  save best only=True)
    model.fit(int_train_ds, validation_data=int_val_ds, epochs=10,__
     ⇔callbacks=callbacks)
    model = keras.models.load_model("embeddings_bidir_gru.keras")
    print(f"Test acc: {model.evaluate(int_test_ds)[1]:.3f}")
    Model: "model_1"
```

5

```
Layer (type)
               Output Shape
                             Param #
______
input_2 (InputLayer)
               [(None, None)]
embedding 2 (Embedding)
               (None, None, 256)
                             2560000
bidirectional 1 (Bidirectio (None, 64)
                             73984
nal)
dropout_1 (Dropout)
               (None, 64)
                             0
dense_1 (Dense)
               (None, 1)
                             65
______
Total params: 2,634,049
Trainable params: 2,634,049
Non-trainable params: 0
Epoch 1/10
4/4 [============ - 5s 982ms/step - loss: 0.6927 - accuracy:
0.5000 - val_loss: 0.6926 - val_accuracy: 0.5221
Epoch 2/10
0.7700 - val_loss: 0.6957 - val_accuracy: 0.5372
Epoch 3/10
0.8000 - val_loss: 0.6995 - val_accuracy: 0.5451
Epoch 4/10
0.8300 - val_loss: 0.7030 - val_accuracy: 0.5499
Epoch 5/10
0.8600 - val_loss: 0.7497 - val_accuracy: 0.5459
Epoch 6/10
0.9100 - val_loss: 0.6955 - val_accuracy: 0.5895
Epoch 7/10
0.9400 - val_loss: 0.6625 - val_accuracy: 0.6187
Epoch 8/10
0.9900 - val_loss: 0.7034 - val_accuracy: 0.6130
1.0000 - val_loss: 0.8337 - val_accuracy: 0.5967
Epoch 10/10
1.0000 - val_loss: 0.8965 - val_accuracy: 0.6043
```

```
accuracy: 0.6142
    Test acc: 0.614
[]: # Using an Embedding layer with masking enabled, 47s
     # Test acc: 0.633 improve again.
     # mask is useful for this case.
[27]: 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.keras",
                                    save_best_only=True)
     model.fit(int_train_ds, validation_data=int_val_ds, epochs=10,_
     ⇔callbacks=callbacks)
     model = keras.models.load_model("embeddings_bidir_gru_with_masking.keras")
     print(f"Test acc: {model.evaluate(int_test_ds)[1]:.3f}")
    Model: "model_2"
     Layer (type) Output Shape Param #
    ______
     input_3 (InputLayer) [(None, None)]
```

embedding_3 (Embedding) (None, None, 256) 2560000

bidirectional_2 (Bidirectio (None, 64) 73984

nal)

dropout_2 (Dropout) (None, 64) 0

dense_2 (Dense) (None, 1) 65

Total params: 2,634,049 Trainable params: 2,634,049 Non-trainable params: 0

```
Epoch 1/10
   4/4 [============ ] - 9s 1s/step - loss: 0.6987 - accuracy:
   0.4400 - val_loss: 0.6926 - val_accuracy: 0.5149
   Epoch 2/10
   0.8900 - val_loss: 0.6914 - val_accuracy: 0.5444
   Epoch 3/10
   0.9900 - val_loss: 0.6900 - val_accuracy: 0.5292
   Epoch 4/10
   0.9800 - val_loss: 0.6861 - val_accuracy: 0.5351
   Epoch 5/10
   0.9300 - val_loss: 0.6773 - val_accuracy: 0.5723
   Epoch 6/10
   1.0000 - val_loss: 0.6471 - val_accuracy: 0.6290
   Epoch 7/10
   1.0000 - val_loss: 0.6415 - val_accuracy: 0.6401
   Epoch 8/10
   1.0000 - val_loss: 0.6507 - val_accuracy: 0.6182
   Epoch 9/10
   1.0000 - val_loss: 0.6579 - val_accuracy: 0.6358
   Epoch 10/10
   4/4 [============= ] - 3s 1s/step - loss: 0.0452 - accuracy:
   1.0000 - val_loss: 0.7007 - val_accuracy: 0.6351
   accuracy: 0.6328
   Test acc: 0.633
[]: # pretrained model
[29]: # Parsing the GloVe word-embeddings file
   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.
[30]: # Preparing the GloVe word-embeddings matrix, Os
      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:</pre>
              embedding_vector = embeddings_index.get(word)
          if embedding_vector is not None:
              embedding_matrix[i] = embedding_vector
[31]: embedding_layer = layers.Embedding(
          max_tokens,
          embedding_dim,
          embeddings_initializer=keras.initializers.Constant(embedding_matrix),
          trainable=False,
          mask_zero=True,
 []: # Model that uses a pretrained Embedding layer, 48.6s
      # Test acc: 0.571, bad performance which beyond my expectation, becasue Ithink
       →with li
      # limited data, pretrained model should really dominates among all trials.
      ##############need to increase the training size to find a "breakeven,
       ⇒point" for this
```

```
save_best_only=True)
]
model.fit(int_train_ds, validation_data=int_val_ds, epochs=10,_
 ⇔callbacks=callbacks)
model = keras.models.load_model("glove_embeddings_sequence_model.keras")
print(f"Test acc: {model.evaluate(int test ds)[1]:.3f}")
Model: "model_3"
Layer (type)
            Output Shape
                               Param #
______
input_4 (InputLayer)
                 [(None, None)]
embedding 4 (Embedding) (None, None, 100)
                                1000000
bidirectional_3 (Bidirectio (None, 64)
                                34048
nal)
dropout_3 (Dropout)
                 (None, 64)
                                0
dense_3 (Dense)
                 (None, 1)
                                65
______
Total params: 1,034,113
Trainable params: 34,113
Non-trainable params: 1,000,000
Epoch 1/10
0.4700 - val_loss: 0.6881 - val_accuracy: 0.5333
Epoch 2/10
0.6100 - val_loss: 0.6822 - val_accuracy: 0.5624
Epoch 3/10
0.4800 - val_loss: 0.6805 - val_accuracy: 0.5792
Epoch 4/10
0.6500 - val_loss: 0.6891 - val_accuracy: 0.5345
Epoch 5/10
0.5700 - val_loss: 0.6816 - val_accuracy: 0.5520
0.6300 - val_loss: 0.7068 - val_accuracy: 0.5142
Epoch 7/10
0.5900 - val_loss: 0.6753 - val_accuracy: 0.5744
```

hw4 part3

November 23, 2023

```
[]: # my part 2 code, which is with 100 10000 25000 size shows
     # even with embedding and pretrained model, the performance is not good.
     # this could be due to limited traing size
     # in this part 3 code, I will increase traing size to 10000(5000/5000)
     # then ppl in my part 4 code, I will increase training size to 3000(1500/1500)
     # please be noted, that max= 25000-10000(assign for valid)=15000
[8]: # set-up train valid test to 10000(5000/5000) 10000(5000/5000) 25000(12500/
      →12500)
     import os, pathlib, shutil, random
     from tensorflow import keras
     batch_size = 32
     base_dir = pathlib.Path("aclImdb")
     val dir = base dir / "val"
     train_dir = base_dir / "train"
     for category in ("neg", "pos"):
         os.makedirs(val_dir / category)
         files = os.listdir(train dir / category)
         random.Random(1337).shuffle(files)
         num val samples = int(0.4 * len(files))
         val_files = files[-num_val_samples:]
         for fname in val_files:
             shutil.move(train_dir / category / fname,
                         val_dir / category / fname)
     # I have no idea about how to keep the training set to a specific number
     # so I did it in a very straight non-fancy way
     # previous block(code) has transfer 20% training data to valid data (25000*0.2)
     # so here I only need to keep the neg and pos folder under train folder to 50_{\sqcup}
      ⇔obs randomly.
            # path of folder
     folder_path = r'C:\Users\zhong\Desktop\HW4_part3\aclImdb\train\neg'
```

```
# go through all file in the folder
all_files = os.listdir(folder_path)
         # set up the number of file that I want to keep
num_files_to_keep = 5000
         # randomly keep the file
files_to_keep = random.sample(all_files, num_files_to_keep)
        # delete the file that I dont need.
for file_name in all_files:
   file_path = os.path.join(folder_path, file_name)
   if file_name not in files_to_keep:
        os.remove(file_path)
# Repeat the same thing for pos folder.
folder_path = r'C:\Users\zhong\Desktop\HW4_part3\aclImdb\train\pos'
all_files = os.listdir(folder_path)
num_files_to_keep = 5000
files_to_keep = random.sample(all_files, num_files_to_keep)
for file_name in all_files:
   file_path = os.path.join(folder_path, file_name)
   if file_name not in files_to_keep:
        os.remove(file_path)
# show the number of train valid test
train_ds = keras.utils.text_dataset_from_directory(
    "aclImdb/train", batch_size=batch_size
val_ds = keras.utils.text_dataset_from_directory(
   "aclImdb/val", batch_size=batch_size
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 10000 files belonging to 2 classes.
Found 10000 files belonging to 2 classes.
Found 25000 files belonging to 2 classes.
```

```
[9]: # Preparing integer sequence datasets
     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)
```

```
Model: "model"
```

```
tf.one_hot (TFOpLambda)
                         (None, None, 10000)
    bidirectional (Bidirectiona (None, 64)
                                            2568448
    1)
    dropout (Dropout)
                         (None, 64)
    dense (Dense)
                         (None, 1)
                                            65
   Total params: 2,568,513
   Trainable params: 2,568,513
   Non-trainable params: 0
    _____
[]: # Training a first basic sequence model 2m33s
    # Test acc: 0.832 , significantly improved compared to part 2 (100_{\sqcup}
     ⇔training size)
[11]: callbacks = [
       keras.callbacks.ModelCheckpoint("one_hot_bidir_lstm.keras",
                              save best only=True)
    model.fit(int_train_ds, validation_data=int_val_ds, epochs=10,
           callbacks=callbacks)
    model = keras.models.load_model("one_hot_bidir_lstm.keras")
    print(f"Test acc: {model.evaluate(int_test_ds)[1]:.3f}")
   Epoch 1/10
   accuracy: 0.6872 - val_loss: 0.4436 - val_accuracy: 0.8099
   Epoch 2/10
   313/313 [============= ] - 11s 34ms/step - loss: 0.4045 -
   accuracy: 0.8451 - val_loss: 0.4656 - val_accuracy: 0.8029
   Epoch 3/10
   accuracy: 0.8845 - val_loss: 0.3770 - val_accuracy: 0.8447
   Epoch 4/10
   accuracy: 0.9098 - val_loss: 0.5399 - val_accuracy: 0.8284
   Epoch 5/10
   accuracy: 0.9331 - val_loss: 0.3785 - val_accuracy: 0.8407
   Epoch 6/10
   accuracy: 0.9460 - val_loss: 0.4934 - val_accuracy: 0.8183
   Epoch 7/10
   313/313 [============== ] - 11s 34ms/step - loss: 0.1219 -
```

```
accuracy: 0.9597 - val_loss: 0.4484 - val_accuracy: 0.8242
    Epoch 8/10
    313/313 [============= ] - 11s 34ms/step - loss: 0.1035 -
    accuracy: 0.9682 - val_loss: 0.4944 - val_accuracy: 0.8067
    Epoch 9/10
    accuracy: 0.9757 - val loss: 1.4872 - val accuracy: 0.7528
    Epoch 10/10
    accuracy: 0.9771 - val_loss: 0.5549 - val_accuracy: 0.8114
    accuracy: 0.8324
    Test acc: 0.832
[12]: # Learning word embeddings with the Embedding layer
     # Instantiating an Embedding layer
    embedding_layer = layers.Embedding(input_dim=max_tokens, output_dim=256)
[]: | # Model that uses an Embedding layer trained from scratch, 1m25s
     # Test acc: 0.826, does not improved.
     # embedding is not that useful when you have a large size of training
[13]: 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.keras",
                                  save_best_only=True)
    model.fit(int_train_ds, validation_data=int_val_ds, epochs=10,_
     ⇒callbacks=callbacks)
    model = keras.models.load_model("embeddings_bidir_gru.keras")
    print(f"Test acc: {model.evaluate(int_test_ds)[1]:.3f}")
    Model: "model 1"
    Layer (type)
                          Output Shape
    ______
```

```
embedding_1 (Embedding)
                (None, None, 256)
                                  2560000
bidirectional_1 (Bidirectio (None, 64)
                                  73984
nal)
dropout_1 (Dropout)
                 (None, 64)
             (None, 1)
dense_1 (Dense)
                                  65
Total params: 2,634,049
Trainable params: 2,634,049
Non-trainable params: 0
-----
Epoch 1/10
accuracy: 0.7458 - val_loss: 0.3838 - val_accuracy: 0.8402
313/313 [============== ] - 7s 24ms/step - loss: 0.3290 -
accuracy: 0.8730 - val_loss: 0.4204 - val_accuracy: 0.8446
Epoch 3/10
accuracy: 0.9076 - val_loss: 0.5638 - val_accuracy: 0.8094
Epoch 4/10
accuracy: 0.9256 - val_loss: 0.4686 - val_accuracy: 0.8280
Epoch 5/10
accuracy: 0.9429 - val_loss: 0.4620 - val_accuracy: 0.8343
Epoch 6/10
accuracy: 0.9553 - val_loss: 0.4810 - val_accuracy: 0.8297
Epoch 7/10
accuracy: 0.9660 - val loss: 0.5283 - val accuracy: 0.8315
Epoch 8/10
accuracy: 0.9744 - val_loss: 0.9635 - val_accuracy: 0.7708
Epoch 9/10
313/313 [============ ] - 8s 24ms/step - loss: 0.0520 -
accuracy: 0.9838 - val_loss: 0.6769 - val_accuracy: 0.8233
Epoch 10/10
accuracy: 0.9870 - val_loss: 1.0744 - val_accuracy: 0.7632
782/782 [=========== ] - 7s 8ms/step - loss: 0.4010 -
accuracy: 0.8258
Test acc: 0.826
```

```
[]: # Using an Embedding layer with masking enabled, 1m49s
     # Test acc: 0.831 a little improve compared to the latest one.
     # mask is useful for this case.
[14]: 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.keras",
                                    save_best_only=True)
     model.fit(int_train_ds, validation_data=int_val_ds, epochs=10,_
      ⇔callbacks=callbacks)
     model = keras.models.load_model("embeddings_bidir_gru_with_masking.keras")
     print(f"Test acc: {model.evaluate(int_test_ds)[1]:.3f}")
    Model: "model_2"
     Layer (type) Output Shape
                                             Param #
                            [(None, None)]
     input_3 (InputLayer)
     embedding_2 (Embedding) (None, None, 256) 2560000
     bidirectional_2 (Bidirectio (None, 64)
                                                  73984
     nal)
     dropout_2 (Dropout)
                      (None, 64)
     dense_2 (Dense)
                             (None, 1)
                                                    65
    _____
    Total params: 2,634,049
    Trainable params: 2,634,049
    Non-trainable params: 0
                       -----
    Epoch 1/10
    accuracy: 0.7578 - val_loss: 0.4665 - val_accuracy: 0.7741
```

```
accuracy: 0.8821 - val_loss: 0.3893 - val_accuracy: 0.8472
   accuracy: 0.9168 - val_loss: 0.3992 - val_accuracy: 0.8480
   accuracy: 0.9444 - val_loss: 1.0728 - val_accuracy: 0.7050
   Epoch 5/10
   accuracy: 0.9599 - val_loss: 0.5367 - val_accuracy: 0.8292
   Epoch 6/10
   accuracy: 0.9735 - val_loss: 0.5818 - val_accuracy: 0.8225
   Epoch 7/10
   accuracy: 0.9821 - val_loss: 0.6493 - val_accuracy: 0.8176
   Epoch 8/10
   accuracy: 0.9893 - val_loss: 0.6908 - val_accuracy: 0.8194
   Epoch 9/10
   accuracy: 0.9949 - val_loss: 0.8471 - val_accuracy: 0.8105
   Epoch 10/10
   accuracy: 0.9955 - val_loss: 0.9377 - val_accuracy: 0.8186
   accuracy: 0.8312
   Test acc: 0.831
[]: # pretrained model
[15]: # Parsing the GloVe word-embeddings file
   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.

Epoch 2/10

```
[16]: # Preparing the GloVe word-embeddings matrix, Os
      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:</pre>
              embedding_vector = embeddings_index.get(word)
          if embedding_vector is not None:
              embedding_matrix[i] = embedding_vector
[17]: embedding_layer = layers.Embedding(
          max_tokens,
          embedding_dim,
          embeddings initializer=keras.initializers.Constant(embedding matrix),
          trainable=False,
          mask_zero=True,
      )
 []: # Model that uses a pretrained Embedding layer, 1m51s
      # Test acc: 0.820, same-flat
      ##############need to increase the training size to find a "breakeven"
       ⇒point" for this
[18]: 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"])
      model.summary()
      callbacks = [
          keras.callbacks.ModelCheckpoint("glove_embeddings_sequence_model.keras",
                                          save best only=True)
      model.fit(int_train_ds, validation_data=int_val_ds, epochs=10,__
       ⇒callbacks=callbacks)
      model = keras.models.load_model("glove_embeddings_sequence_model.keras")
      print(f"Test acc: {model.evaluate(int_test_ds)[1]:.3f}")
```

Model: "model_3"

· · · · · · · · · · · · · · · · · · ·	Output Shape	
input_4 (InputLayer)		0
embedding_3 (Embedding)	(None, None, 100)	1000000
<pre>bidirectional_3 (Bidirectional)</pre>	(None, 64)	34048
dropout_3 (Dropout)	(None, 64)	0
dense_3 (Dense)	(None, 1)	65
Total params: 1,034,113		=======
Trainable params: 34,113 Non-trainable params: 1,000,		
Epoch 1/10 313/313 [===================================	======] - 15s 36ms	/step - loss: 0.6426 -
Epoch 2/10 313/313 [===================================		-
Epoch 3/10 313/313 [===================================	·	
accuracy: 0.7718 - val_loss: Epoch 4/10		•
313/313 [===================================		-
Epoch 5/10 313/313 [===================================		
accuracy: 0.8061 - val_loss: 0.4281 - val_accuracy: 0.7989 Epoch 6/10 313/313 [===================================		
accuracy: 0.8139 - val_loss: Epoch 7/10		-
313/313 [===================================		-
313/313 [===================================		-
Epoch 9/10 313/313 [===================================		-

accuracy: 0.8499 - val_loss: 0.3958 - val_accuracy: 0.8197

accuracy: 0.8196 Test acc: 0.820

hw4 part4

November 23, 2023

```
[]: # then ppl in my part 4 code, I will increase training size to 3000(1500/1500)
     # please be noted, that max= 25000-10000(assign for valid)=15000
[1]: # set-up train valid test to 10000(5000/5000) 10000(5000/5000) 25000(12500/
      →12500)
     import os, pathlib, shutil, random
     from tensorflow import keras
     batch_size = 32
     base_dir = pathlib.Path("aclImdb")
     val_dir = base_dir / "val"
     train_dir = base_dir / "train"
     for category in ("neg", "pos"):
         os.makedirs(val_dir / category)
         files = os.listdir(train_dir / category)
         random.Random(1337).shuffle(files)
         num_val_samples = int(0.4 * len(files))
         val_files = files[-num_val_samples:]
         for fname in val files:
             shutil.move(train_dir / category / fname,
                         val_dir / category / fname)
     # I have no idea about how to keep the training set to a specific number
     # so I did it in a very straight non-fancy way
     # previous block(code) has transfer 20% training data to valid data (25000*0.2)
     # so here I only need to keep the neg and pos folder under train folder to 50_{\sqcup}
      ⇔obs randomly.
            # path of folder
     folder_path = r'C:\Users\zhong\Desktop\HW4_part4\aclImdb\train\neg'
              # go through all file in the folder
     all_files = os.listdir(folder_path)
     #
              # set up the number of file that I want to keep
```

```
num_files_to_keep = 1500
         # randomly keep the file
files_to_keep = random.sample(all_files, num_files_to_keep)
        # delete the file that I dont need.
for file_name in all_files:
   file_path = os.path.join(folder_path, file_name)
    if file_name not in files_to_keep:
        os.remove(file_path)
# Repeat the same thing for pos folder.
folder_path = r'C:\Users\zhong\Desktop\HW4_part4\aclImdb\train\pos'
all_files = os.listdir(folder_path)
num_files_to_keep = 1500
files_to_keep = random.sample(all_files, num_files_to_keep)
for file_name in all_files:
   file_path = os.path.join(folder_path, file_name)
   if file_name not in files_to_keep:
        os.remove(file_path)
# show the number of train valid test
train_ds = keras.utils.text_dataset_from_directory(
   "aclImdb/train", batch_size=batch_size
val_ds = keras.utils.text_dataset_from_directory(
   "aclImdb/val", batch_size=batch_size
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 3000 files belonging to 2 classes. Found 10000 files belonging to 2 classes. Found 25000 files belonging to 2 classes.

```
[2]: # Preparing integer sequence datasets
    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)
[3]: # A sequence model built on one-hot encoded vector sequences
    import tensorflow as tf
    inputs = keras.Input(shape=(None,), dtype="int64")
    embedded = tf.one_hot(inputs, depth=max_tokens)
    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()
    # RNN-LSTM
    Model: "model"
                                Output Shape
                                                         Param #
    Layer (type)
    ______
     input_1 (InputLayer)
                                [(None, None)]
```

2568448

(None, None, 10000)

tf.one_hot (TFOpLambda)

1)

bidirectional (Bidirectiona (None, 64)

```
dropout (Dropout)
                  (None, 64)
   dense (Dense)
                   (None, 1)
                                   65
  ______
  Total params: 2,568,513
  Trainable params: 2,568,513
  Non-trainable params: 0
[4]: # Training a first basic sequence model 1m47s
   # Test acc: 0.787
[5]: callbacks = [
     keras.callbacks.ModelCheckpoint("one_hot_bidir_lstm.keras",
                        save_best_only=True)
  ]
  model.fit(int_train_ds, validation_data=int_val_ds, epochs=10,
        callbacks=callbacks)
  model = keras.models.load_model("one_hot_bidir_lstm.keras")
  print(f"Test acc: {model.evaluate(int test ds)[1]:.3f}")
  Epoch 1/10
  accuracy: 0.5370 - val_loss: 0.6772 - val_accuracy: 0.6471
  Epoch 2/10
  0.6860 - val_loss: 0.5309 - val_accuracy: 0.7625
  Epoch 3/10
  0.8253 - val_loss: 0.5375 - val_accuracy: 0.7344
  Epoch 4/10
  0.8757 - val_loss: 0.5020 - val_accuracy: 0.7965
  Epoch 5/10
  0.9083 - val_loss: 0.6533 - val_accuracy: 0.7203
  Epoch 6/10
  0.9180 - val_loss: 0.5436 - val_accuracy: 0.7824
  Epoch 7/10
  0.9473 - val_loss: 0.6594 - val_accuracy: 0.7781
  0.9430 - val_loss: 0.5565 - val_accuracy: 0.7690
  Epoch 9/10
```

```
0.9657 - val_loss: 0.6821 - val_accuracy: 0.7889
   Epoch 10/10
   0.9727 - val loss: 0.6785 - val accuracy: 0.7928
   accuracy: 0.7870
   Test acc: 0.787
[6]: # Learning word embeddings with the Embedding layer
    # Instantiating an Embedding layer
   embedding_layer = layers.Embedding(input_dim=max_tokens, output_dim=256)
[7]: # Model that uses an Embedding layer trained from scratch, 51s
    # Test acc: 0.760, does not improved.
    # embedding is not that useful when you have a large size of training
[8]: 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.keras",
                                save_best_only=True)
   model.fit(int_train_ds, validation_data=int_val_ds, epochs=10,_
    ⇔callbacks=callbacks)
   model = keras.models.load model("embeddings bidir gru.keras")
   print(f"Test acc: {model.evaluate(int_test_ds)[1]:.3f}")
   Model: "model_1"
   Layer (type)
                         Output Shape
                                              Param #
   ______
    embedding_1 (Embedding) (None, None, 256)
                                              2560000
    bidirectional_1 (Bidirectio (None, 64)
                                               73984
    nal)
```

```
dropout_1 (Dropout)
              (None, 64)
                                0
  dense_1 (Dense)
                  (None, 1)
                                65
  ______
  Total params: 2,634,049
  Trainable params: 2,634,049
  Non-trainable params: 0
       -----
  Epoch 1/10
  0.6060 - val_loss: 0.6115 - val_accuracy: 0.6587
  Epoch 2/10
  0.8377 - val_loss: 0.4763 - val_accuracy: 0.7736
  Epoch 3/10
  0.9170 - val_loss: 0.4873 - val_accuracy: 0.7749
  Epoch 4/10
  0.9510 - val_loss: 0.5740 - val_accuracy: 0.7849
  Epoch 5/10
  0.9827 - val_loss: 0.6868 - val_accuracy: 0.7910
  Epoch 6/10
  94/94 [============ ] - 4s 42ms/step - loss: 0.0566 - accuracy:
  0.9827 - val_loss: 0.7031 - val_accuracy: 0.8034
  Epoch 7/10
  0.9917 - val_loss: 0.7309 - val_accuracy: 0.7972
  Epoch 8/10
  94/94 [============ ] - 4s 42ms/step - loss: 0.0187 - accuracy:
  0.9943 - val_loss: 0.9624 - val_accuracy: 0.7948
  Epoch 9/10
  0.9927 - val_loss: 0.8571 - val_accuracy: 0.7741
  Epoch 10/10
  0.9960 - val_loss: 0.8434 - val_accuracy: 0.7871
  accuracy: 0.7605
  Test acc: 0.760
[9]: # Using an Embedding layer with masking enabled, 1m49s
  # Test acc: 0.800 a little improve compared to the latest one.
```

mask is useful for this case.

```
[10]: 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.keras",
                                   save_best_only=True)
     ]
     model.fit(int_train_ds, validation_data=int_val_ds, epochs=10,_
      ⇔callbacks=callbacks)
     model = keras.models.load_model("embeddings_bidir_gru_with_masking.keras")
     print(f"Test acc: {model.evaluate(int_test_ds)[1]:.3f}")
    Model: "model_2"
     Layer (type) Output Shape
    ______
     embedding_2 (Embedding) (None, None, 256)
                                                  2560000
     bidirectional_2 (Bidirectio (None, 64)
                                                  73984
     nal)
     dropout_2 (Dropout)
                      (None, 64)
     dense_2 (Dense)
                             (None, 1)
                                                   65
    ______
    Total params: 2,634,049
```

Trainable params: 2,634,049 Non-trainable params: 0

```
Epoch 1/10
accuracy: 0.6037 - val_loss: 0.5286 - val_accuracy: 0.7475
Epoch 2/10
0.8327 - val_loss: 0.4258 - val_accuracy: 0.8047
Epoch 3/10
```

```
0.9117 - val_loss: 0.4841 - val_accuracy: 0.7983
   Epoch 4/10
   0.9587 - val_loss: 0.6003 - val_accuracy: 0.7968
   Epoch 5/10
   0.9823 - val_loss: 0.6838 - val_accuracy: 0.7925
   Epoch 6/10
   0.9877 - val_loss: 0.8371 - val_accuracy: 0.7939
   Epoch 7/10
   0.9947 - val_loss: 0.7747 - val_accuracy: 0.8006
   0.9970 - val_loss: 0.9324 - val_accuracy: 0.8028
   Epoch 9/10
   0.9990 - val_loss: 1.1798 - val_accuracy: 0.7969
   Epoch 10/10
   0.9963 - val_loss: 0.9934 - val_accuracy: 0.7939
   accuracy: 0.7999
   Test acc: 0.800
[11]: # pretrained model
[17]: # Parsing the GloVe word-embeddings file
   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.
[18]: # Preparing the GloVe word-embeddings matrix, Os
   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:</pre>
              embedding_vector = embeddings_index.get(word)
          if embedding_vector is not None:
              embedding_matrix[i] = embedding_vector
[19]: embedding_layer = layers.Embedding(
          max_tokens,
          embedding_dim,
          embeddings_initializer=keras.initializers.Constant(embedding_matrix),
          trainable=False,
          mask_zero=True,
      )
 []: # Model that uses a pretrained Embedding layer, 1m9.9s
      # Test acc: 0.779
      ##############need to increase the training size to find a "breakeven,
       ⇔point" for this
[20]: 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"])
      model.summary()
      callbacks = \Gamma
          keras.callbacks.ModelCheckpoint("glove_embeddings_sequence_model.keras",
                                          save best only=True)
      model.fit(int_train_ds, validation_data=int_val_ds, epochs=10,_
       ⇔callbacks=callbacks)
      model = keras.models.load_model("glove embeddings sequence model.keras")
      print(f"Test acc: {model.evaluate(int_test_ds)[1]:.3f}")
     Model: "model 4"
      Layer (type)
                                Output Shape
                                                            Param #
```

```
input_5 (InputLayer)
           [(None, None)]
embedding_4 (Embedding)
              (None, None, 100)
                            1000000
bidirectional 4 (Bidirectio (None, 64)
                            34048
nal)
dropout_4 (Dropout)
           (None, 64)
dense_4 (Dense)
               (None, 1)
                            65
______
Total params: 1,034,113
Trainable params: 34,113
Non-trainable params: 1,000,000
Epoch 1/10
accuracy: 0.5433 - val_loss: 0.6708 - val_accuracy: 0.5738
Epoch 2/10
0.6067 - val_loss: 0.5988 - val_accuracy: 0.6891
Epoch 3/10
0.6707 - val_loss: 0.6339 - val_accuracy: 0.6473
Epoch 4/10
0.7007 - val_loss: 0.5753 - val_accuracy: 0.6982
94/94 [=========== ] - 5s 57ms/step - loss: 0.5535 - accuracy:
0.7227 - val_loss: 0.5182 - val_accuracy: 0.7556
Epoch 6/10
0.7440 - val_loss: 0.7002 - val_accuracy: 0.6547
Epoch 7/10
0.7563 - val loss: 0.4837 - val accuracy: 0.7727
Epoch 8/10
0.7690 - val_loss: 0.4892 - val_accuracy: 0.7675
Epoch 9/10
0.7880 - val_loss: 0.4848 - val_accuracy: 0.7710
Epoch 10/10
0.7977 - val_loss: 0.4621 - val_accuracy: 0.7825
accuracy: 0.7795
```

Test acc: 0.779

hw4_part5

November 23, 2023

```
[1]: # set-up train valid test to
     import os, pathlib, shutil, random
     from tensorflow import keras
     batch_size = 32
     base dir = pathlib.Path("aclImdb")
     val_dir = base_dir / "val"
     train dir = base dir / "train"
     for category in ("neg", "pos"):
         os.makedirs(val_dir / category)
         files = os.listdir(train_dir / category)
         random.Random(1337).shuffle(files)
         num_val_samples = int(0.4 * len(files))
         val_files = files[-num_val_samples:]
         for fname in val_files:
             shutil.move(train_dir / category / fname,
                         val_dir / category / fname)
     # I have no idea about how to keep the training set to a specific number
     # so I did it in a very straight non-fancy way
     # previous block(code) has transfer 20% training data to valid data (25000*0.2)
     # so here I only need to keep the neg and pos folder under train folder to 50_{\sqcup}
      ⇔obs randomly.
            # path of folder
     folder_path = r'C:\Users\zhong\Desktop\HW4_part5\aclImdb\train\neg'
              # go through all file in the folder
     all_files = os.listdir(folder_path)
              # set up the number of file that I want to keep
     num_files_to_keep = 300
              # randomly keep the file
     files_to_keep = random.sample(all_files, num_files_to_keep)
```

```
# delete the file that I dont need.
for file_name in all_files:
    file_path = os.path.join(folder_path, file_name)
    if file_name not in files_to_keep:
        os.remove(file_path)
# Repeat the same thing for pos folder.
folder_path = r'C:\Users\zhong\Desktop\HW4_part5\aclImdb\train\pos'
all_files = os.listdir(folder_path)
num_files_to_keep = 300
files_to_keep = random.sample(all_files, num_files_to_keep)
#
for file_name in all_files:
    file_path = os.path.join(folder_path, file_name)
    if file_name not in files_to_keep:
        os.remove(file_path)
# show the number of train valid test
train_ds = keras.utils.text_dataset_from_directory(
    "aclImdb/train", batch_size=batch_size
val_ds = keras.utils.text_dataset_from_directory(
    "aclImdb/val", batch_size=batch_size
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 600 files belonging to 2 classes.
Found 10000 files belonging to 2 classes.
Found 25000 files belonging to 2 classes.
```

```
[3]: # Preparing integer sequence datasets

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)
```

Model: "model 1"

Layer (type)	Output Shape	Param #
input_2 (InputLayer)	[(None, None)]	0
tf.one_hot_1 (TFOpLambda)	(None, None, 10000)	0
<pre>bidirectional_1 (Bidirectio nal)</pre>	(None, 64)	2568448
dropout_1 (Dropout)	(None, 64)	0

```
dense_1 (Dense)
                    (None, 1)
                                     65
  ______
  Total params: 2,568,513
  Trainable params: 2,568,513
  Non-trainable params: 0
  _____
[]: # Training a first basic sequence model 1m22s
   # Test acc: 0.704
[5]: callbacks = [
     keras.callbacks.ModelCheckpoint("one_hot_bidir_lstm.keras",
                         save_best_only=True)
   model.fit(int_train_ds, validation_data=int_val_ds, epochs=10,
         callbacks=callbacks)
   model = keras.models.load_model("one_hot_bidir_lstm.keras")
   print(f"Test acc: {model.evaluate(int_test_ds)[1]:.3f}")
  Epoch 1/10
  accuracy: 0.5217 - val_loss: 0.6925 - val_accuracy: 0.5497
  Epoch 2/10
  19/19 [============ ] - 4s 239ms/step - loss: 0.6864 -
  accuracy: 0.6117 - val_loss: 0.6903 - val_accuracy: 0.5142
  Epoch 3/10
  accuracy: 0.7700 - val_loss: 0.6807 - val_accuracy: 0.5929
  Epoch 4/10
  accuracy: 0.7700 - val_loss: 0.6168 - val_accuracy: 0.6567
  Epoch 5/10
  accuracy: 0.8500 - val_loss: 0.7401 - val_accuracy: 0.6278
  Epoch 6/10
  accuracy: 0.9133 - val_loss: 0.5876 - val_accuracy: 0.7016
  Epoch 7/10
  accuracy: 0.9283 - val_loss: 1.3416 - val_accuracy: 0.5455
  accuracy: 0.9250 - val_loss: 0.6312 - val_accuracy: 0.7165
  accuracy: 0.9700 - val_loss: 0.7277 - val_accuracy: 0.7084
  Epoch 10/10
```

```
accuracy: 0.9867 - val_loss: 0.6871 - val_accuracy: 0.7206
   accuracy: 0.7036
   Test acc: 0.704
[6]: # Learning word embeddings with the Embedding layer
    # Instantiating an Embedding layer
    embedding_layer = layers.Embedding(input_dim=max_tokens, output_dim=256)
[]: # Model that uses an Embedding layer trained from scratch, 37.5s
    # Test acc: 0.718, improved a little
    # embedding is useful when you have a large size of training
[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.keras",
                                   save_best_only=True)
    model.fit(int_train_ds, validation_data=int_val_ds, epochs=10,_
     ⇔callbacks=callbacks)
    model = keras.models.load_model("embeddings_bidir_gru.keras")
    print(f"Test acc: {model.evaluate(int_test_ds)[1]:.3f}")
   Model: "model_2"
    Layer (type)
                            Output Shape
   ______
    input_3 (InputLayer)
                            [(None, None)]
    embedding_1 (Embedding)
                           (None, None, 256)
                                                  2560000
    bidirectional_2 (Bidirectio (None, 64)
                                                   73984
    nal)
    dropout_2 (Dropout)
                            (None, 64)
    dense_2 (Dense)
                             (None, 1)
                                                   65
```

```
Total params: 2,634,049
  Trainable params: 2,634,049
  Non-trainable params: 0
               _____
  Epoch 1/10
  accuracy: 0.5417 - val_loss: 0.6912 - val_accuracy: 0.5391
  Epoch 2/10
  accuracy: 0.7383 - val_loss: 0.6518 - val_accuracy: 0.6327
  Epoch 3/10
  accuracy: 0.8333 - val_loss: 0.6203 - val_accuracy: 0.6678
  Epoch 4/10
  19/19 [========== ] - 3s 149ms/step - loss: 0.2613 -
  accuracy: 0.9517 - val_loss: 0.5567 - val_accuracy: 0.7168
  Epoch 5/10
  accuracy: 0.9850 - val_loss: 0.7610 - val_accuracy: 0.6816
  Epoch 6/10
  accuracy: 0.9933 - val_loss: 0.9884 - val_accuracy: 0.6560
  Epoch 7/10
  accuracy: 0.9967 - val_loss: 0.7608 - val_accuracy: 0.7176
  Epoch 8/10
  accuracy: 0.9983 - val_loss: 0.7435 - val_accuracy: 0.6951
  Epoch 9/10
  accuracy: 0.9983 - val_loss: 0.7072 - val_accuracy: 0.6920
  Epoch 10/10
  accuracy: 1.0000 - val_loss: 0.8769 - val_accuracy: 0.7179
  accuracy: 0.7178
  Test acc: 0.718
[]: # Using an Embedding layer with masking enabled, 1m49s
   # Test acc: 0.716 same-flat
   # mask is not that useful for this case.
[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.keras",
                              save_best_only=True)
]
model.fit(int_train_ds, validation_data=int_val_ds, epochs=10,_
 ⇔callbacks=callbacks)
model = keras.models.load_model("embeddings_bidir_gru_with_masking.keras")
print(f"Test acc: {model.evaluate(int_test_ds)[1]:.3f}")
Model: "model_3"
Layer (type)
                Output Shape
                                     Param #
______
input_4 (InputLayer)
                      [(None, None)]
embedding_2 (Embedding) (None, None, 256) 2560000
bidirectional_3 (Bidirectio (None, 64)
                                             73984
nal)
dropout_3 (Dropout)
                 (None, 64)
dense_3 (Dense)
                       (None, 1)
                                             65
Total params: 2,634,049
Trainable params: 2,634,049
Non-trainable params: 0
Epoch 1/10
accuracy: 0.5300 - val_loss: 0.6886 - val_accuracy: 0.5595
Epoch 2/10
19/19 [============= ] - 3s 179ms/step - loss: 0.6157 -
accuracy: 0.7950 - val_loss: 0.6234 - val_accuracy: 0.6742
Epoch 3/10
accuracy: 0.8683 - val_loss: 0.6520 - val_accuracy: 0.6890
Epoch 4/10
```

```
accuracy: 0.9817 - val_loss: 0.6230 - val_accuracy: 0.7237
  Epoch 5/10
  accuracy: 0.9917 - val_loss: 0.7655 - val_accuracy: 0.7199
  Epoch 6/10
  accuracy: 0.9983 - val_loss: 0.9429 - val_accuracy: 0.7016
  Epoch 7/10
  19/19 [============ ] - 3s 186ms/step - loss: 0.0128 -
  accuracy: 0.9983 - val_loss: 1.0571 - val_accuracy: 0.6894
  Epoch 8/10
  accuracy: 0.9817 - val_loss: 0.9275 - val_accuracy: 0.7149
  19/19 [========== ] - 3s 179ms/step - loss: 0.0084 -
  accuracy: 0.9983 - val_loss: 0.8543 - val_accuracy: 0.7378
  Epoch 10/10
  accuracy: 1.0000 - val_loss: 1.0246 - val_accuracy: 0.7283
  accuracy: 0.7158
  Test acc: 0.716
[]: # pretrained model
[9]: # Parsing the GloVe word-embeddings file
   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.

```
[10]: # Preparing the GloVe word-embeddings matrix, Os
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:</pre>
             embedding_vector = embeddings_index.get(word)
         if embedding_vector is not None:
             embedding_matrix[i] = embedding_vector
[11]: embedding_layer = layers.Embedding(
         max_tokens,
         embedding_dim,
         embeddings initializer=keras.initializers.Constant(embedding matrix),
         trainable=False,
         mask_zero=True,
     )
 []: # Model that uses a pretrained Embedding layer, 55.6s
     # Test acc: 0.645
     ###############need to increase the training size to find a "breakeven"
      →point" for this
[12]: 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"])
     model.summary()
     callbacks = [
         keras.callbacks.ModelCheckpoint("glove_embeddings_sequence_model.keras",
                                        save_best_only=True)
     model.fit(int_train_ds, validation_data=int_val_ds, epochs=10,_
      ⇔callbacks=callbacks)
     model = keras.models.load_model("glove embeddings_sequence_model.keras")
     print(f"Test acc: {model.evaluate(int_test_ds)[1]:.3f}")
     Model: "model_4"
     Layer (type)
                      Output Shape
                                                        Param #
     ______
      input_5 (InputLayer)
                             [(None, None)]
      embedding_3 (Embedding) (None, None, 100) 1000000
```

```
bidirectional_4 (Bidirectio (None, 64)
                             34048
nal)
dropout 4 (Dropout) (None, 64)
                             0
dense 4 (Dense)
               (None, 1)
                             65
______
Total params: 1,034,113
Trainable params: 34,113
Non-trainable params: 1,000,000
______
Epoch 1/10
accuracy: 0.4917 - val_loss: 0.6839 - val_accuracy: 0.5608
Epoch 2/10
accuracy: 0.6017 - val_loss: 0.6772 - val_accuracy: 0.5869
Epoch 3/10
accuracy: 0.5817 - val_loss: 0.7025 - val_accuracy: 0.5209
Epoch 4/10
accuracy: 0.5883 - val_loss: 0.6925 - val_accuracy: 0.5426
Epoch 5/10
accuracy: 0.6400 - val_loss: 0.6643 - val_accuracy: 0.6011
accuracy: 0.6833 - val_loss: 0.6436 - val_accuracy: 0.6321
Epoch 7/10
accuracy: 0.6683 - val_loss: 0.6630 - val_accuracy: 0.6081
Epoch 8/10
accuracy: 0.6717 - val loss: 0.6219 - val accuracy: 0.6604
Epoch 9/10
accuracy: 0.6883 - val_loss: 0.6363 - val_accuracy: 0.6394
Epoch 10/10
accuracy: 0.7500 - val_loss: 0.6720 - val_accuracy: 0.6312
782/782 [============= ] - 10s 10ms/step - loss: 0.6312 -
accuracy: 0.6454
Test acc: 0.645
```

```
[13]: # Parsing the GloVe word-embeddings file
      import numpy as np
      path_to_glove_file = "glove.6B.300d.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.
[15]: # Preparing the GloVe word-embeddings matrix, Os
      embedding_dim = 300
      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:</pre>
              embedding_vector = embeddings_index.get(word)
         if embedding_vector is not None:
              embedding_matrix[i] = embedding_vector
[16]: embedding_layer = layers.Embedding(
         max_tokens,
         embedding_dim,
         embeddings_initializer=keras.initializers.Constant(embedding_matrix),
         trainable=False,
         mask_zero=True,
      )
 []: # Model that uses a pretrained Embedding layer, 1m4.4s
      # Test acc: 0.715
      ##############need to increase the training size to find a "breakeven"
       ⇔point" for this
[17]: 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"])
model.summary()
callbacks = [
   keras.callbacks.ModelCheckpoint("glove_embeddings_sequence_model.keras",
                           save best only=True)
model.fit(int_train_ds, validation_data=int_val_ds, epochs=10,__
 ⇔callbacks=callbacks)
model = keras.models.load_model("glove embeddings sequence model.keras")
print(f"Test acc: {model.evaluate(int_test_ds)[1]:.3f}")
Model: "model 5"
Layer (type)
                    Output Shape
______
embedding_4 (Embedding) (None, None, 300)
                                        3000000
bidirectional_5 (Bidirectio (None, 64)
                                         85248
nal)
dense 5 (Dense)
                    (None, 1)
                                         65
Total params: 3,085,313
Trainable params: 85,313
Non-trainable params: 3,000,000
        .....
Epoch 1/10
19/19 [=========== ] - 12s 394ms/step - loss: 0.7039 -
accuracy: 0.4867 - val_loss: 0.6856 - val_accuracy: 0.5371
Epoch 2/10
accuracy: 0.6050 - val_loss: 0.6639 - val_accuracy: 0.6082
Epoch 3/10
19/19 [============= ] - 5s 288ms/step - loss: 0.6229 -
accuracy: 0.6600 - val_loss: 0.6427 - val_accuracy: 0.6327
Epoch 4/10
```

```
accuracy: 0.6983 - val_loss: 0.6230 - val_accuracy: 0.6444
    Epoch 5/10
    accuracy: 0.7267 - val_loss: 0.6454 - val_accuracy: 0.6390
    Epoch 6/10
    accuracy: 0.7633 - val loss: 0.5930 - val accuracy: 0.6903
    Epoch 7/10
    19/19 [============ ] - 5s 284ms/step - loss: 0.4500 -
    accuracy: 0.8033 - val_loss: 0.5605 - val_accuracy: 0.7131
    Epoch 8/10
    accuracy: 0.8200 - val_loss: 0.5819 - val_accuracy: 0.7102
    Epoch 9/10
    accuracy: 0.8367 - val_loss: 0.5980 - val_accuracy: 0.6909
    Epoch 10/10
    accuracy: 0.8133 - val_loss: 0.5638 - val_accuracy: 0.7092
    782/782 [============== ] - 9s 10ms/step - loss: 0.5601 -
    accuracy: 0.7146
    Test acc: 0.715
[18]: # try glove6b200d
    # Parsing the GloVe word-embeddings file
    import numpy as np
    path_to_glove_file = "glove.6B.200d.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.")
    # Preparing the GloVe word-embeddings matrix, Os
    embedding_dim = 200
    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:</pre>
```

```
embedding_vector = embeddings_index.get(word)
if embedding_vector is not None:
    embedding_matrix[i] = embedding_vector

embedding_layer = layers.Embedding(
    max_tokens,
    embedding_dim,
    embeddings_initializer=keras.initializers.Constant(embedding_matrix),
    trainable=False,
    mask_zero=True,
)
```

Found 400000 word vectors.

```
[]: # Model that uses a pretrained Embedding layer, 57,7s
# Test acc: 0.683

################need to increase the training size to find a "breakevenu opoint" for this
```

```
[19]: 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"])
      model.summary()
      callbacks = [
          keras.callbacks.ModelCheckpoint("glove embeddings sequence model.keras",
                                          save_best_only=True)
     model.fit(int_train_ds, validation_data=int_val_ds, epochs=10,__
       ⇔callbacks=callbacks)
      model = keras.models.load_model("glove_embeddings_sequence_model.keras")
      print(f"Test acc: {model.evaluate(int_test_ds)[1]:.3f}")
```

Model: "model_6"

```
Layer (type) Output Shape Param #

input_7 (InputLayer) [(None, None)] 0

embedding_5 (Embedding) (None, None, 200) 2000000
```

```
bidirectional_6 (Bidirectio (None, 64)
                                   59648
nal)
dropout_6 (Dropout)
                  (None, 64)
                                   0
dense 6 (Dense)
                  (None, 1)
                                   65
______
Total params: 2,059,713
Trainable params: 59,713
Non-trainable params: 2,000,000
       .____
Epoch 1/10
19/19 [=========== ] - 11s 332ms/step - loss: 0.6963 -
accuracy: 0.5350 - val_loss: 0.6814 - val_accuracy: 0.5705
Epoch 2/10
19/19 [============= ] - 5s 249ms/step - loss: 0.6677 -
accuracy: 0.5900 - val_loss: 0.6712 - val_accuracy: 0.5966
Epoch 3/10
accuracy: 0.6150 - val_loss: 0.6586 - val_accuracy: 0.6220
Epoch 4/10
19/19 [============ ] - 3s 187ms/step - loss: 0.6302 -
accuracy: 0.6700 - val_loss: 0.6737 - val_accuracy: 0.5696
Epoch 5/10
accuracy: 0.6750 - val_loss: 0.7103 - val_accuracy: 0.5591
Epoch 6/10
accuracy: 0.7100 - val_loss: 0.6275 - val_accuracy: 0.6518
Epoch 7/10
accuracy: 0.7317 - val_loss: 0.5839 - val_accuracy: 0.6940
Epoch 8/10
accuracy: 0.7533 - val_loss: 0.5865 - val_accuracy: 0.6865
Epoch 9/10
accuracy: 0.7833 - val_loss: 0.6129 - val_accuracy: 0.6547
Epoch 10/10
19/19 [========== ] - 3s 186ms/step - loss: 0.4544 -
accuracy: 0.7917 - val_loss: 0.7027 - val_accuracy: 0.6352
accuracy: 0.6834
Test acc: 0.683
```

hw4_part6

November 23, 2023

```
[]: # final trial will be used high D glove moedel with all availiable data to
     # build to expected best model
     # release some of the word length to 400
[3]: import os, pathlib, shutil, random
     from tensorflow import keras
     batch_size = 32
     base_dir = pathlib.Path("aclImdb")
     val_dir = base_dir / "val"
     train_dir = base_dir / "train"
     for category in ("neg", "pos"):
         os.makedirs(val_dir / category)
         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
     val_ds = keras.utils.text_dataset_from_directory(
         "aclImdb/val", batch_size=batch_size
     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.
[4]: from tensorflow.keras import layers
```

```
max_length = 400 # I think the max of length of my gpu that be handled is 400_{\square}
 →not 600
max_tokens = 20000
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)
```

[]: # A sequence model built on one-hot encoded vector sequences, 0.3s

Model: "model"

Layer (type)	Output Shape	Param #
input_1 (InputLayer)	[(None, None)]	0
tf.one_hot (TFOpLambda)	(None, None, 20000)	0
bidirectional (Bidirectiona 1)	(None, 64)	5128448
dropout (Dropout)	(None, 64)	0

```
dense (Dense)
                     (None, 1)
                                      65
  _____
  Total params: 5,128,513
  Trainable params: 5,128,513
  Non-trainable params: 0
  _____
[]: # Training a first basic sequence model 7m56.8s
   # Test acc: 0.878
[6]: callbacks = [
     keras.callbacks.ModelCheckpoint("one_hot_bidir_lstm.keras",
                          save best only=True)
   model.fit(int_train_ds, validation_data=int_val_ds, epochs=10,
         callbacks=callbacks)
   model = keras.models.load_model("one_hot_bidir_lstm.keras")
   print(f"Test acc: {model.evaluate(int_test_ds)[1]:.3f}")
  Epoch 1/10
  accuracy: 0.7498 - val_loss: 0.3549 - val_accuracy: 0.8650
  Epoch 2/10
  625/625 [============= ] - 44s 70ms/step - loss: 0.3636 -
  accuracy: 0.8664 - val_loss: 0.3023 - val_accuracy: 0.8748
  Epoch 3/10
  accuracy: 0.9020 - val_loss: 0.3188 - val_accuracy: 0.8710
  Epoch 4/10
  accuracy: 0.9191 - val_loss: 0.3021 - val_accuracy: 0.8878
  Epoch 5/10
  accuracy: 0.9314 - val_loss: 0.5378 - val_accuracy: 0.8586
  Epoch 6/10
  accuracy: 0.9398 - val_loss: 0.3078 - val_accuracy: 0.8832
  Epoch 7/10
  accuracy: 0.9480 - val_loss: 0.3476 - val_accuracy: 0.8798
  accuracy: 0.9559 - val_loss: 0.5844 - val_accuracy: 0.8706
  625/625 [============ ] - 44s 71ms/step - loss: 0.1137 -
  accuracy: 0.9626 - val_loss: 0.3568 - val_accuracy: 0.8602
```

Epoch 10/10

```
625/625 [============= ] - 44s 70ms/step - loss: 0.0918 -
   accuracy: 0.9713 - val_loss: 0.4962 - val_accuracy: 0.8804
   accuracy: 0.8779
   Test acc: 0.878
[7]: # Learning word embeddings with the Embedding layer
    # Instantiating an Embedding layer
    embedding_layer = layers.Embedding(input_dim=max_tokens, output_dim=256)
[]: # Model that uses an Embedding layer trained from scratch, 4m33s
    # Test acc: 0.849, embedding is not that useful when you have greate data_{f L}
     ⇔source for training.
[8]: 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.keras",
                                     save_best_only=True)
    model.fit(int_train_ds, validation_data=int_val_ds, epochs=10,_
     ⇔callbacks=callbacks)
    model = keras.models.load_model("embeddings_bidir_gru.keras")
    print(f"Test acc: {model.evaluate(int_test_ds)[1]:.3f}")
   Model: "model_1"
    Layer (type)
                              Output Shape
    ______
    input_2 (InputLayer)
                             [(None, None)]
    embedding_1 (Embedding)
                            (None, None, 256)
                                                     5120000
    bidirectional_1 (Bidirectio (None, 64)
                                                      73984
    nal)
    dropout_1 (Dropout)
                              (None, 64)
    dense_1 (Dense)
                              (None, 1)
                                                      65
```

```
Total params: 5,194,049
  Trainable params: 5,194,049
  Non-trainable params: 0
                _____
  Epoch 1/10
  accuracy: 0.7835 - val_loss: 0.3698 - val_accuracy: 0.8556
  Epoch 2/10
  625/625 [============ ] - 22s 35ms/step - loss: 0.3101 -
  accuracy: 0.8846 - val_loss: 0.4703 - val_accuracy: 0.7758
  Epoch 3/10
  accuracy: 0.9133 - val_loss: 0.3550 - val_accuracy: 0.8674
  Epoch 4/10
  625/625 [=========== ] - 22s 35ms/step - loss: 0.2037 -
  accuracy: 0.9309 - val_loss: 0.4144 - val_accuracy: 0.8630
  Epoch 5/10
  accuracy: 0.9423 - val_loss: 0.4548 - val_accuracy: 0.8706
  Epoch 6/10
  accuracy: 0.9526 - val_loss: 0.3856 - val_accuracy: 0.8706
  Epoch 7/10
  accuracy: 0.9653 - val_loss: 0.4173 - val_accuracy: 0.8740
  Epoch 8/10
  accuracy: 0.9722 - val_loss: 0.4445 - val_accuracy: 0.8708
  Epoch 9/10
  accuracy: 0.9765 - val_loss: 0.4931 - val_accuracy: 0.8604
  Epoch 10/10
  accuracy: 0.9815 - val_loss: 0.5164 - val_accuracy: 0.8708
  accuracy: 0.8490
  Test acc: 0.849
[]: # Using an Embedding layer with masking enabled, 4m29
             # Test acc: 0.874, a little better with the previous model, but
   \hookrightarrow same as the naive model.
[9]: 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.keras",
                              save_best_only=True)
]
model.fit(int_train_ds, validation_data=int_val_ds, epochs=10,_
 ⇔callbacks=callbacks)
model = keras.models.load_model("embeddings_bidir_gru_with_masking.keras")
print(f"Test acc: {model.evaluate(int_test_ds)[1]:.3f}")
Model: "model_2"
                Output Shape
Layer (type)
______
input_3 (InputLayer)
                      [(None, None)]
embedding_2 (Embedding) (None, None, 256) 5120000
bidirectional_2 (Bidirectio (None, 64)
                                             73984
nal)
dropout_2 (Dropout)
                 (None, 64)
dense_2 (Dense)
                       (None, 1)
                                             65
Total params: 5,194,049
Trainable params: 5,194,049
Non-trainable params: 0
Epoch 1/10
accuracy: 0.8156 - val_loss: 0.2924 - val_accuracy: 0.8816
Epoch 2/10
625/625 [============= ] - 25s 39ms/step - loss: 0.2356 -
accuracy: 0.9109 - val_loss: 0.3206 - val_accuracy: 0.8812
Epoch 3/10
accuracy: 0.9391 - val_loss: 0.5246 - val_accuracy: 0.8366
```

Epoch 4/10

```
625/625 [============= ] - 24s 39ms/step - loss: 0.1268 -
     accuracy: 0.9556 - val_loss: 0.3862 - val_accuracy: 0.8518
     Epoch 5/10
     625/625 [============ ] - 24s 39ms/step - loss: 0.0945 -
     accuracy: 0.9681 - val_loss: 0.4018 - val_accuracy: 0.8700
     Epoch 6/10
     625/625 [============ ] - 24s 39ms/step - loss: 0.0676 -
     accuracy: 0.9764 - val_loss: 0.5169 - val_accuracy: 0.8550
     Epoch 7/10
     625/625 [============ ] - 24s 39ms/step - loss: 0.0499 -
     accuracy: 0.9832 - val_loss: 0.5912 - val_accuracy: 0.8592
     Epoch 8/10
     625/625 [============ ] - 25s 40ms/step - loss: 0.0363 -
     accuracy: 0.9878 - val_loss: 0.5559 - val_accuracy: 0.8650
     625/625 [============] - 25s 41ms/step - loss: 0.0233 -
     accuracy: 0.9922 - val_loss: 0.6376 - val_accuracy: 0.8568
     Epoch 10/10
     625/625 [============ ] - 24s 39ms/step - loss: 0.0178 -
     accuracy: 0.9947 - val_loss: 0.6269 - val_accuracy: 0.8658
     782/782 [============= ] - 15s 17ms/step - loss: 0.3007 -
     accuracy: 0.8740
     Test acc: 0.874
[]: # Parsing the GloVe word-embeddings file, 27s
[10]: import numpy as np
     path_to_glove_file = "glove.6B.300d.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.
[]: # Preparing the GloVe word-embeddings matrix, Os
[11]: embedding_dim = 300
     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:</pre>
              embedding_vector = embeddings_index.get(word)
          if embedding_vector is not None:
              embedding_matrix[i] = embedding_vector
[12]: embedding_layer = layers.Embedding(
          max_tokens,
          embedding dim,
          embeddings_initializer=keras.initializers.Constant(embedding_matrix),
          trainable=False,
          mask_zero=True,
 []: # Model that uses a pretrained Embedding layer, 4m52s
      # Test acc: 0.879 , did not beat my nonpretrained model, ????!!!!!!!!! textbook_{\sqcup}
       \hookrightarrowgave one explanatin for this, for this serise of trials , my sample size is
       slarge enough to let even the most naive model to learn from embedding from
       \hookrightarrowscratch.
[13]: 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"])
      model.summary()
      callbacks = [
          keras.callbacks.ModelCheckpoint("glove embeddings sequence model.keras",
                                           save_best_only=True)
      model.fit(int_train_ds, validation_data=int_val_ds, epochs=10,__
      ⇔callbacks=callbacks)
      model = keras.models.load_model("glove_embeddings_sequence_model.keras")
      print(f"Test acc: {model.evaluate(int_test_ds)[1]:.3f}")
     Model: "model_3"
      Layer (type)
                                 Output Shape
      input_4 (InputLayer)
                                 [(None, None)]
      embedding_3 (Embedding) (None, None, 300) 6000000
```

```
bidirectional_3 (Bidirectio (None, 64)
                                  85248
nal)
dropout_3 (Dropout)
                  (None, 64)
                                   0
dense 3 (Dense)
                  (None, 1)
                                   65
______
Total params: 6,085,313
Trainable params: 85,313
Non-trainable params: 6,000,000
      -----
Epoch 1/10
625/625 [============ ] - 36s 51ms/step - loss: 0.5093 -
accuracy: 0.7470 - val_loss: 0.3996 - val_accuracy: 0.8236
Epoch 2/10
accuracy: 0.8298 - val_loss: 0.4476 - val_accuracy: 0.8210
Epoch 3/10
625/625 [============= ] - 29s 46ms/step - loss: 0.3469 -
accuracy: 0.8523 - val_loss: 0.3434 - val_accuracy: 0.8532
Epoch 4/10
accuracy: 0.8706 - val_loss: 0.3523 - val_accuracy: 0.8498
Epoch 5/10
accuracy: 0.8798 - val_loss: 0.3461 - val_accuracy: 0.8578
Epoch 6/10
accuracy: 0.8960 - val_loss: 0.3612 - val_accuracy: 0.8566
Epoch 7/10
accuracy: 0.9048 - val_loss: 0.3609 - val_accuracy: 0.8664
Epoch 8/10
accuracy: 0.9145 - val_loss: 0.3478 - val_accuracy: 0.8668
Epoch 9/10
625/625 [============== ] - 25s 40ms/step - loss: 0.1940 -
accuracy: 0.9238 - val_loss: 0.3710 - val_accuracy: 0.8668
Epoch 10/10
625/625 [============ ] - 29s 47ms/step - loss: 0.1740 -
accuracy: 0.9337 - val_loss: 0.3317 - val_accuracy: 0.8742
accuracy: 0.8794
Test acc: 0.879
```