# Text and Sequence

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# 1 Text and Sequence Assignment

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We will be using IMDB data for this text and sequence problem. Firstly, we need to create a validation set with 80% of training dataset and setting apart 20% for training.

#### 1.1.1 Reading Data

Making a small training sample as well:

Reading our datasets:

```
[3]: from tensorflow import keras

batch_size = 32

train = keras.utils.

text_dataset_from_directory(train_dir_1,batch_size=batch_size)

validation=keras.utils.

text_dataset_from_directory(val_dir,batch_size=batch_size)

test=keras.utils.text_dataset_from_directory(base_dir/

test",batch_size=batch_size)
```

```
Found 1000 files belonging to 2 classes. Found 10000 files belonging to 2 classes. Found 25000 files belonging to 2 classes.
```

# 1.1.2 Trying sequencing model

Preparing dataset for this model:

```
[4]: from tensorflow.keras import layers
     max_length = 150 # Cutting off values after 150 words
     max_tokens = 10000 # Considering only top 10,000 words
     text_vectorization = layers.TextVectorization(
         max_tokens=max_tokens,
         output_mode="int",
         output_sequence_length=max_length,
     text_only_train_ds = train.map(lambda x, y: x)
     # Turning text to vectors
     text_vectorization.adapt(text_only_train_ds)
     int_train_ds = train.map(
     lambda x, y: (text_vectorization(x), y), num_parallel_calls=4)
     int_val_ds = validation.map(
     lambda x, y: (text_vectorization(x), y), num_parallel_calls=4)
     int test ds = test.map(
     lambda x, y: (text_vectorization(x), y), num_parallel_calls=4)
```

2024-04-27 13:19:52.616018: W tensorflow/core/framework/local\_rendezvous.cc:404] Local rendezvous is aborting with status: OUT\_OF\_RANGE: End of sequence

#### Model Construction - Embedding Layer

Model: "functional\_1"

Layer (type)	Output Shape	Param #	Connected to
<pre>input_layer (InputLayer)</pre>	(None, None)	0	-
embedding (Embedding)	(None, None, 256)	2,560,000	<pre>input_layer[0][0]</pre>
<pre>not_equal (NotEqual)</pre>	(None, None)	0	<pre>input_layer[0][0]</pre>
bidirectional (Bidirectional)	(None, 64)	73,984	<pre>embedding[0][0], not_equal[0][0]</pre>
dropout (Dropout)	(None, 64)	0	bidirectional[0]
dense (Dense)	(None, 1)	65	dropout[0][0]

Total params: 2,634,049 (10.05 MB)

Trainable params: 2,634,049 (10.05 MB)

Non-trainable params: 0 (0.00 B)

Fitting the model on our testing dataset

#### callbacks=callbacks)

```
Epoch 1/10
4/4
               6s 2s/step -
accuracy: 0.4818 - loss: 0.6944 - val_accuracy: 0.5032 - val_loss: 0.6934
Epoch 2/10
4/4
               5s 2s/step -
accuracy: 0.7362 - loss: 0.6836 - val_accuracy: 0.5099 - val_loss: 0.6930
Epoch 3/10
4/4
                5s 2s/step -
accuracy: 0.8605 - loss: 0.6745 - val_accuracy: 0.5021 - val_loss: 0.6929
Epoch 4/10
4/4
                5s 2s/step -
accuracy: 0.8587 - loss: 0.6659 - val_accuracy: 0.5110 - val_loss: 0.6927
Epoch 5/10
4/4
                8s 3s/step -
accuracy: 0.7935 - loss: 0.6541 - val_accuracy: 0.5189 - val_loss: 0.6922
Epoch 6/10
4/4
                8s 3s/step -
accuracy: 0.9400 - loss: 0.6336 - val_accuracy: 0.5178 - val_loss: 0.6921
Epoch 7/10
4/4
                8s 3s/step -
accuracy: 0.9265 - loss: 0.6022 - val_accuracy: 0.5318 - val_loss: 0.6906
Epoch 8/10
4/4
                8s 3s/step -
accuracy: 0.9448 - loss: 0.5635 - val_accuracy: 0.5279 - val_loss: 0.6902
Epoch 9/10
4/4
                8s 3s/step -
accuracy: 0.9569 - loss: 0.5023 - val_accuracy: 0.5012 - val_loss: 0.7443
Epoch 10/10
4/4
                8s 3s/step -
accuracy: 0.8550 - loss: 0.4527 - val_accuracy: 0.5146 - val_loss: 0.7370
```

[7]: <keras.src.callbacks.history.History at 0x1766aec10>

Testing this model

```
[8]: print("\n Model's accuracy:",round(model.evaluate(int_test_ds)[1]*100,2),"%")
```

```
782/782
                    18s 23ms/step -
accuracy: 0.5164 - loss: 0.7372
```

Model's accuracy: 51.42 %

Hence, our first model's accuracy with LSTM and embedding is just 51.42% which is quite low. We will now try a pre-trained word embedding.

### Model Construction - Pretrained word embedded

Parsing after downloading the glove pretrained word-embedding.

Preparing a matrix of GloVe:

```
[10]: 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>
```

Making an embedding layer with this embedded matrix:

Making a final model with pretrained work-embedding:

]

Model: "functional\_3"

Layer (type)	Output Shape	Param #	Connected to
<pre>input_layer_1 (InputLayer)</pre>	(None, None)	0	_
<pre>embedding_1 (Embedding)</pre>	(None, None, 100)	1,000,000	input_layer_1[0]
<pre>not_equal_1 (NotEqual)</pre>	(None, None)	0	input_layer_1[0]
<pre>bidirectional_1 (Bidirectional)</pre>	(None, 64)	34,048	embedding_1[0][0 not_equal_1[0][0]
<pre>dropout_1 (Dropout)</pre>	(None, 64)	0	bidirectional_1[
dense_1 (Dense)	(None, 1)	65	dropout_1[0][0]

Total params: 1,034,113 (3.94 MB)

Trainable params: 1,034,113 (3.94 MB)

Non-trainable params: 0 (0.00 B)

Training this model on our training dataset:

```
Epoch 1/10
4/4 6s 2s/step -
accuracy: 0.4599 - loss: 0.7884 - val_accuracy: 0.5023 - val_loss: 0.7113
Epoch 2/10
4/4 5s 2s/step -
accuracy: 0.5549 - loss: 0.6722 - val_accuracy: 0.5100 - val_loss: 0.7009
Epoch 3/10
4/4 5s 2s/step -
accuracy: 0.5099 - loss: 0.6993 - val_accuracy: 0.5084 - val_loss: 0.6960
Epoch 4/10
4/4 5s 2s/step -
```

```
accuracy: 0.5520 - loss: 0.6879 - val_accuracy: 0.5189 - val_loss: 0.6929
     Epoch 5/10
     4/4
                     5s 2s/step -
     accuracy: 0.6635 - loss: 0.6248 - val_accuracy: 0.5227 - val_loss: 0.6922
     Epoch 6/10
     4/4
                     5s 2s/step -
     accuracy: 0.5994 - loss: 0.6553 - val accuracy: 0.5264 - val loss: 0.6912
     Epoch 7/10
                     5s 2s/step -
     accuracy: 0.6260 - loss: 0.6428 - val_accuracy: 0.5284 - val_loss: 0.6954
     Epoch 8/10
     4/4
                     5s 2s/step -
     accuracy: 0.6689 - loss: 0.6680 - val_accuracy: 0.5079 - val_loss: 0.7074
     Epoch 9/10
     4/4
                     5s 2s/step -
     accuracy: 0.6194 - loss: 0.6576 - val_accuracy: 0.5259 - val_loss: 0.6979
     Epoch 10/10
                     5s 2s/step -
     4/4
     accuracy: 0.7012 - loss: 0.5911 - val_accuracy: 0.5364 - val_loss: 0.6909
[13]: <keras.src.callbacks.history.History at 0x284cd6fd0>
```

Testing this model on our testing data:

```
[14]: print("\n Model's Accuracy:", round(model.evaluate(int_test_ds)[1]*100,2))
```

```
782/782
                    12s 15ms/step -
accuracy: 0.5329 - loss: 0.6908
Model's Accuracy: 53.69
```

Pre-trained embedding is not really helpful in this case. Hence, training from scratch worked better for this dataset. Now, we will try to increase training sample size and then train our model again.

# Increasing training size by 7000 samples

```
[15]: for category in ("neg","pos"):
          files= os.listdir(train_dir/category)
          random.Random(1337).shuffle(files)
          num_train_samples = 3500
          train_file = files[:num_train_samples]
          for fname in train_file:
              shutil.move(train_dir/category/fname,
                         train dir 1/category, fname)
```

Making a training dataset again:

```
[16]: train = keras.utils.
       stext_dataset_from_directory(train_dir_1,batch_size=batch_size)
      int_train_ds = train.map(
```

```
Found 7100 files belonging to 2 classes.
     Training the last pretrained embedding model with new training dataset:
[17]: model.fit(int_train_ds, validation_data=int_val_ds, epochs=10,
      callbacks=callbacks)
     Epoch 1/10
                         19s 84ms/step -
     222/222
     accuracy: 0.5677 - loss: 0.6838 - val_accuracy: 0.6883 - val_loss: 0.5900
     Epoch 2/10
                         20s 91ms/step -
     222/222
     accuracy: 0.6891 - loss: 0.5843 - val_accuracy: 0.7678 - val_loss: 0.4902
     Epoch 3/10
     222/222
                         20s 91ms/step -
     accuracy: 0.7569 - loss: 0.5130 - val_accuracy: 0.7687 - val_loss: 0.4837
     Epoch 4/10
     222/222
                         20s 90ms/step -
     accuracy: 0.7942 - loss: 0.4484 - val_accuracy: 0.8093 - val_loss: 0.4346
     Epoch 5/10
     222/222
                         20s 88ms/step -
     accuracy: 0.8222 - loss: 0.4099 - val_accuracy: 0.8072 - val_loss: 0.4569
     Epoch 6/10
     222/222
                         20s 90ms/step -
     accuracy: 0.8476 - loss: 0.3613 - val_accuracy: 0.8168 - val_loss: 0.4379
     Epoch 7/10
     222/222
                         20s 92ms/step -
     accuracy: 0.8602 - loss: 0.3317 - val_accuracy: 0.8075 - val_loss: 0.4466
     Epoch 8/10
     222/222
                         20s 90ms/step -
     accuracy: 0.8772 - loss: 0.3005 - val_accuracy: 0.7936 - val_loss: 0.5010
     Epoch 9/10
     222/222
                         20s 88ms/step -
     accuracy: 0.8870 - loss: 0.2743 - val_accuracy: 0.8126 - val_loss: 0.5556
     Epoch 10/10
     222/222
                         20s 89ms/step -
     accuracy: 0.9015 - loss: 0.2440 - val_accuracy: 0.8118 - val_loss: 0.5086
[17]: <keras.src.callbacks.history.History at 0x2d68b8410>
     Testing the model now:
[18]: print("\n Model's Accuracy:", round(model.evaluate(int_test_ds)[1]*100,2))
     782/782
                         12s 15ms/step -
     accuracy: 0.8226 - loss: 0.4714
      Model's Accuracy: 82.05
```

lambda x, y: (text\_vectorization(x), y), num\_parallel\_calls=4)

Increasing samples did not really increase any acuracy.

#### Increasing training sample again by 7000

Reading new training set:

Found 14100 files belonging to 2 classes.

Training this model again:

```
[21]: model.fit(int_train_ds, validation_data=int_val_ds, epochs=10)
     Epoch 1/10
     441/441
                         35s 79ms/step -
     accuracy: 0.8645 - loss: 0.3277 - val_accuracy: 0.8279 - val_loss: 0.3957
     Epoch 2/10
     441/441
                         34s 77ms/step -
     accuracy: 0.8832 - loss: 0.2903 - val_accuracy: 0.8309 - val_loss: 0.4049
     Epoch 3/10
     441/441
                         34s 77ms/step -
     accuracy: 0.8924 - loss: 0.2706 - val_accuracy: 0.8308 - val_loss: 0.4207
     Epoch 4/10
     441/441
                         35s 80ms/step -
     accuracy: 0.9031 - loss: 0.2490 - val accuracy: 0.8290 - val loss: 0.4348
     Epoch 5/10
     441/441
                         38s 85ms/step -
     accuracy: 0.9088 - loss: 0.2264 - val_accuracy: 0.8125 - val_loss: 0.4916
     Epoch 6/10
     441/441
                         36s 82ms/step -
     accuracy: 0.9225 - loss: 0.2067 - val_accuracy: 0.8277 - val_loss: 0.4868
     Epoch 7/10
     441/441
                         39s 89ms/step -
     accuracy: 0.9311 - loss: 0.1781 - val_accuracy: 0.8179 - val_loss: 0.4934
     Epoch 8/10
     441/441
                         38s 87ms/step -
     accuracy: 0.9386 - loss: 0.1605 - val_accuracy: 0.8258 - val_loss: 0.5785
```

Epoch 9/10

441/441 38s 87ms/step -

accuracy: 0.9477 - loss: 0.1361 - val\_accuracy: 0.8122 - val\_loss: 0.5701

Epoch 10/10

441/441 40s 91ms/step -

accuracy: 0.9593 - loss: 0.1171 - val\_accuracy: 0.8180 - val\_loss: 0.6679

[21]: <keras.src.callbacks.history.History at 0x2d3403910>

Testing this model:

[22]: print("\n Model's Accuracy:",round(model.evaluate(int\_test\_ds)[1]\*100,2))

Model's Accuracy: 82.2

There is not any improvement with the second increase of training size. This explains that beyond a certain point, increasing training set size does not affect accuracy.