generateAmazonDataset

April 19, 2023

```
[ ]: def remove_prefix(text, prefix):
        if text.startswith(prefix):
             return text[len(prefix):]
        return text
[]: import os
     import numpy as np
     import tensorflow as tf
     from tensorflow import keras
     from tensorflow.keras import layers
     import matplotlib.pyplot as plt
    2023-04-19 18:48:05.498513: I tensorflow/core/platform/cpu_feature_guard.cc:193]
    This TensorFlow binary is optimized with oneAPI Deep Neural Network Library
    (oneDNN) to use the following CPU instructions in performance-critical
    operations: AVX2 AVX VNNI FMA
    To enable them in other operations, rebuild TensorFlow with the appropriate
    compiler flags.
    2023-04-19 18:48:05.552992: I tensorflow/core/util/port.cc:104] oneDNN custom
    operations are on. You may see slightly different numerical results due to
    floating-point round-off errors from different computation orders. To turn them
    off, set the environment variable `TF_ENABLE_ONEDNN_OPTS=0`.
    2023-04-19 18:48:05.826900: W
    tensorflow/compiler/xla/stream_executor/platform/default/dso_loader.cc:64] Could
    not load dynamic library 'libnvinfer.so.7'; dlerror: libnvinfer.so.7: cannot
    open shared object file: No such file or directory; LD_LIBRARY_PATH:
    :/home/victorxesus.barreiro/anaconda3/envs/deepgpu4/lib/
    2023-04-19 18:48:05.826932: W
    tensorflow/compiler/xla/stream executor/platform/default/dso loader.cc:64] Could
    not load dynamic library 'libnvinfer_plugin.so.7'; dlerror:
    libnvinfer_plugin.so.7: cannot open shared object file: No such file or
    directory; LD_LIBRARY_PATH:
    :/home/victorxesus.barreiro/anaconda3/envs/deepgpu4/lib/
    2023-04-19 18:48:05.826934: W
    tensorflow/compiler/tf2tensorrt/utils/py_utils.cc:38] TF-TRT Warning: Cannot
    dlopen some TensorRT libraries. If you would like to use Nvidia GPU with
    TensorRT, please make sure the missing libraries mentioned above are installed
    properly.
```

```
#Returns a list with the text and a list with the labels
     def readData(fname):
         with open(fname, 'r', encoding="utf-8") as f:
             fileData = f.read()
         lines = fileData.split("\n")
         textData = list()
         textLabel = list()
         lineLength = np.zeros(len(lines))
         for i, aLine in enumerate(lines):
             if not aLine:
                 break
             label = aLine.split(" ")[0]
             lineLength[i] = len(aLine.split(" "))
             if(label == "__label__1"):
                 textLabel.append(0)
                 textData.append(remove_prefix(aLine, "__label__1 "))
             elif(label == "__label__2"):
                 textLabel.append(1)
                 textData.append(remove_prefix(aLine, "__label__2 "))
             else:
                 print("\nError in readData: ", i, aLine)
                 exit()
         f.close()
         return textData, textLabel, int(np.average(lineLength)+2*np.std(lineLength))
[]: from tensorflow.keras import layers
     def transformData(x_train, y_train, x_test, y_test, maxFeatures, seqLength):
         #transforms text input to int input based on the vocabulary
         #max_tokens = maxFeatures is the size of the vocabulary
         #output_sequence_length = seqLength is the maximum length of the
      →transformed text. Adds 0 is text length is shorter
         precLayer = layers.experimental.preprocessing.TextVectorization(max_tokens⊔
      ⇒= maxFeatures,
         standardize = 'lower_and_strip_punctuation', split = 'whitespace', u

output_mode = 'int',
         output_sequence_length = seqLength)
         precLayer.adapt(x_train)
         #print(precLayer.get_vocabulary())
         x_train_int = precLayer(x_train)
```

[]: #reads a file. Each line has the format: label text

```
y_train = tf.convert_to_tensor(y_train)
#print(x_train_int)
#print(y_train)
x_test_int= precLayer(x_test)
y_test = tf.convert_to_tensor(y_test)
#print(x_test_int)
#print(y_test)

return x_train_int, y_train, x_test_int, y_test
```

```
[]: import sys
[]: x_train, y_train, seqLength = readData("amazon/train_small.txt")
     x_test, y_test, tmp = readData("amazon/test_small.txt")
     del tmp
     print(sys.getsizeof(x_train))
     print(sys.getsizeof(y_train))
     print(sys.getsizeof(x_test))
     print(sys.getsizeof(y_test))
     #maxFeatures is a hyperparameter
     maxFeatures = 5500
     x_train_int, y_train, x_test_int, y_test = transformData(x_train, y_train, __
      sx_test, y_test, maxFeatures, seqLength)
     print("O obxecto")
     print(sys.getsizeof(transformData))
     del x_train
     del x_test
     del transformData
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```

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2023-04-19 18:48:06.403634: I
tensorflow/compiler/xla/stream_executor/cuda/cuda_gpu_executor.cc:981]
successful NUMA node read from SysFS had negative value (-1), but there must be at least one NUMA node, so returning NUMA node zero
2023-04-19 18:48:06.405835: I
tensorflow/compiler/xla/stream_executor/cuda/cuda_gpu_executor.cc:981]
successful NUMA node read from SysFS had negative value (-1), but there must be

```
at least one NUMA node, so returning NUMA node zero
    2023-04-19 18:48:06.405901: I
    tensorflow/compiler/xla/stream executor/cuda/cuda gpu executor.cc:981]
    successful NUMA node read from SysFS had negative value (-1), but there must be
    at least one NUMA node, so returning NUMA node zero
    2023-04-19 18:48:06.406092: I tensorflow/core/platform/cpu feature guard.cc:193]
    This TensorFlow binary is optimized with oneAPI Deep Neural Network Library
    (oneDNN) to use the following CPU instructions in performance-critical
    operations: AVX2 AVX VNNI FMA
    To enable them in other operations, rebuild TensorFlow with the appropriate
    compiler flags.
    2023-04-19 18:48:06.406461: I
    tensorflow/compiler/xla/stream executor/cuda/cuda gpu executor.cc:981]
    successful NUMA node read from SysFS had negative value (-1), but there must be
    at least one NUMA node, so returning NUMA node zero
    2023-04-19 18:48:06.406515: I
    tensorflow/compiler/xla/stream_executor/cuda/cuda_gpu_executor.cc:981]
    successful NUMA node read from SysFS had negative value (-1), but there must be
    at least one NUMA node, so returning NUMA node zero
    2023-04-19 18:48:06.406551: I
    tensorflow/compiler/xla/stream executor/cuda/cuda gpu executor.cc:981]
    successful NUMA node read from SysFS had negative value (-1), but there must be
    at least one NUMA node, so returning NUMA node zero
    2023-04-19 18:48:06.699228: I
    tensorflow/compiler/xla/stream_executor/cuda/cuda_gpu_executor.cc:981]
    successful NUMA node read from SysFS had negative value (-1), but there must be
    at least one NUMA node, so returning NUMA node zero
    2023-04-19 18:48:06.699305: I
    tensorflow/compiler/xla/stream_executor/cuda/cuda_gpu_executor.cc:981]
    successful NUMA node read from SysFS had negative value (-1), but there must be
    at least one NUMA node, so returning NUMA node zero
    2023-04-19 18:48:06.699348: I
    tensorflow/compiler/xla/stream executor/cuda/cuda gpu executor.cc:981]
    successful NUMA node read from SysFS had negative value (-1), but there must be
    at least one NUMA node, so returning NUMA node zero
    2023-04-19 18:48:06.699392: I
    tensorflow/core/common runtime/gpu/gpu device.cc:1613] Created device
    /job:localhost/replica:0/task:0/device:GPU:0 with 20000 MB memory:
    O, name: NVIDIA GeForce RTX 4090, pci bus id: 0000:01:00.0, compute capability:
    8.9
    O obxecto
    136
[]: from keras.layers import Input, Conv2D, MaxPooling2D, Flatten, Dense, Dropout
     from keras import Input, Model
     from keras.utils import plot_model
```

```
from tensorflow.keras.callbacks import ModelCheckpoint
import matplotlib.pyplot as plt
import numpy as np
import time
# Directory where the checkpoints will be saved.
dir = "models/"
def visualize_fit(history):
    """Visualize the fit of a model.
    Args:
        history (list): list of metrics along the epochs.
    history_dict = history.history
    print(history_dict.keys())
    history_dict.keys()
    loss_values = history_dict['loss']
    val_loss_values = history_dict['val_loss']
    epochs = range(1, len(loss_values) + 1)
    plt.figure(figsize=(10,5))
    plt.subplot(1, 2, 1)
    plt.plot(epochs, loss_values, 'b-o', label='Training loss')
    plt.plot(epochs, val_loss_values, 'r-o', label='Validation loss')
    plt.title('Training and validation loss')
    plt.xlabel('Epochs')
    plt.ylabel('Loss')
    plt.legend()
    plt.subplot(1, 2, 2)
    acc = history_dict['accuracy']
    val_acc = history_dict['val_accuracy']
    plt.plot(epochs, acc, 'b-o', label='Training Accuracy')
    plt.plot(epochs, val_acc, 'r-o', label='Validation Accuracy')
    plt.title('Training and validation Accuracy')
    plt.xlabel('Epochs')
    plt.ylabel('MAE')
    plt.ylim([0, 1])
    plt.legend()
    plt.tight_layout()
```

```
plt.show()
```

```
[]:|def fitModel(model, x_train, y_train, ds_val, num_epochs=20,__
      -monitor='val_accuracy', model_name='best_model.h5', callbacks=[],u
      ⇔batch_size=32):
         """Function to train a model. It saves the best model in a file. It also,
      ⇔prints the evolution of the training process.
         Args:
             model (Model): The model to be trained.
             ds train ( type ): The training dataset.
             ds_val (_type_): The validation dataset.
             num_epochs (int, optional): Defaults to 20.
             monitor (str, optional): Metric to monitor and save the best model. \Box
      → Defaults to 'val_mean_absolute_error'.
             model\_name (str, optional): Name of the file where the best model will_{\sqcup}
      ⇒be saved. Defaults to 'best model.h5'.
             callbacks (list, optional): List of callbacks to be used during
      \hookrightarrow training. Defaults to [].
         Returns:
             final_metrics (list): List with the final metrics of the model
         checkpoint = ModelCheckpoint(dir + model_name, save_best_only=True,_
      save weights only=False, monitor=monitor, mode='auto', verbose=1)
         history = model.fit(x_train, y_train, verbose = 1, epochs=num_epochs,__
      ⇒callbacks=callbacks+[checkpoint], validation_data=ds_val,
      ⇒batch_size=batch_size)
         visualize_fit(history)
         return history
```

1 Contexto

Lo primeo que debemos tener en mente al abordar un problema como este, en el que tengamos procesado del lenguaje natural, es el proceso de tokenización. En este caso tenemos fijados bastantes de estos aspectos y quedan de nuestra mano el número de caracteristicas que este representa y el tamaño de secuencia. En este sentido, hemos comprobado que el númeor de caracteristicas supone un aspecto muy importante en este problema, hemos visto que a partir de 5 000 caracteristicas no mejoramos los resultados, sin embargo, debemos tener en cuenta que para grandes corpus de texto se nos recomenda emplear un número muy superior del orden de 50 000 a 100 000, esto es un importante indicativo de la variabilidad de los datos y la complejidad de nuestro problema. Por ello, estamos ante un porblema que muy probablmemente este acotado al tipo de usuario que hace las reviws y los paises en los que se obtuvo por lo que no debemos esperar una buena generalización en otros contexto. En cuanto a la longitud de secuencia debemos ser cautos y generosos en este problema, las reviews son textos cortos por lo que no debemos cortarlos más

cuando puede supone un comportamiento inesperado del sistema, en definitiva, debemos emplear un tamaño suficientemente grande como 256 en el que quepan todas nuestras muestras, dado que esto no tiene un grave impacto computacional pero si lo puede tener en el comportamiento del sistema.

2 Recurrent Units

2.1 SimpleRNN

```
[]: from tensorflow.keras.callbacks import ReduceLROnPlateau
     np.random.seed(423423)
     tf.random.set_seed(1232413)
     reduce_lr = ReduceLROnPlateau(monitor='val_loss', factor=0.5, patience=5,__
      min_lr=0.000000000000000000000000001, verbose=1)
     start_time = time.time()
     inputs = keras.Input(shape=(seqLength, ))
     x = layers.Embedding(maxFeatures, 16, mask_zero=True,_
      →input_length=seqLength)(inputs)
     x = layers.SimpleRNN(64, return_sequences=False)(x)
     x = keras.layers.BatchNormalization()(x)
     outputs = layers.Dense(1, activation="sigmoid")(x)
     model = keras.Model(inputs, outputs)
     model.summary()
     optimizer = keras.optimizers.Adam(learning_rate=0.001)
     model.compile(optimizer=optimizer, loss="binary_crossentropy", __
      →metrics=["accuracy"])
     result = fitModel(model, x_train_int, y_train, (x_test_int, y_test),_
      onum_epochs=20, model_name="simple.h5", batch_size=1024,__
      ⇔callbacks=[reduce_lr])
     end_time = time.time()
     print("Elapsed time: ", end_time - start_time)
```

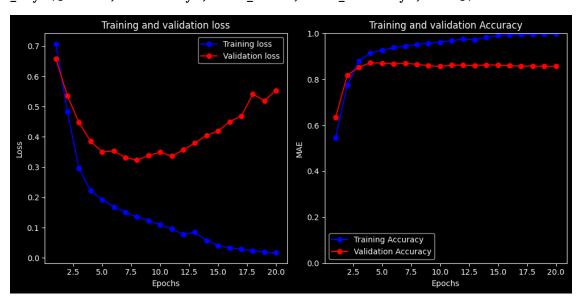
Model: "model"

Layer (type)	Output Shape	Param #
input_1 (InputLayer)	[(None, 166)]	0
embedding (Embedding)	(None, 166, 16)	80000

```
simple_rnn (SimpleRNN) (None, 64)
                                                5184
batch_normalization (BatchN (None, 64)
                                                256
ormalization)
                         (None, 1)
dense (Dense)
                                                65
Total params: 85,505
Trainable params: 85,377
Non-trainable params: 128
Epoch 1/20
2023-04-19 18:02:24.311930: I
tensorflow/compiler/xla/stream executor/cuda/cuda blas.cc:630] TensorFloat-32
will be used for the matrix multiplication. This will only be logged once.
2023-04-19 18:02:24.330526: I tensorflow/compiler/xla/service/service.cc:173]
XLA service 0x7f61d8021940 initialized for platform CUDA (this does not
guarantee that XLA will be used). Devices:
2023-04-19 18:02:24.330543: I tensorflow/compiler/xla/service/service.cc:181]
StreamExecutor device (0): NVIDIA GeForce RTX 4090, Compute Capability 8.9
2023-04-19 18:02:24.333169: I
tensorflow/compiler/mlir/tensorflow/utils/dump_mlir_util.cc:268] disabling MLIR
crash reproducer, set env var `MLIR CRASH REPRODUCER DIRECTORY` to enable.
2023-04-19 18:02:24.366949: I tensorflow/tsl/platform/default/subprocess.cc:304]
Start cannot spawn child process: No such file or directory
2023-04-19 18:02:24.386670: I
tensorflow/compiler/jit/xla_compilation_cache.cc:477] Compiled cluster using
XLA! This line is logged at most once for the lifetime of the process.
Epoch 1: val_accuracy improved from -inf to 0.63524, saving model to
models/simple.h5
accuracy: 0.5467 - val_loss: 0.6588 - val_accuracy: 0.6352 - lr: 0.0010
Epoch 2/20
25/25 [============== ] - ETA: Os - loss: 0.4827 - accuracy:
Epoch 2: val_accuracy improved from 0.63524 to 0.81820, saving model to
models/simple.h5
0.7758 - val_loss: 0.5356 - val_accuracy: 0.8182 - lr: 0.0010
Epoch 3/20
25/25 [============= ] - ETA: Os - loss: 0.2950 - accuracy:
Epoch 3: val_accuracy improved from 0.81820 to 0.85308, saving model to
models/simple.h5
```

```
0.8800 - val_loss: 0.4475 - val_accuracy: 0.8531 - lr: 0.0010
Epoch 4/20
0.9148
Epoch 4: val_accuracy improved from 0.85308 to 0.87216, saving model to
models/simple.h5
0.9148 - val_loss: 0.3852 - val_accuracy: 0.8722 - lr: 0.0010
Epoch 5/20
0.9261
Epoch 5: val_accuracy did not improve from 0.87216
0.9261 - val_loss: 0.3504 - val_accuracy: 0.8708 - lr: 0.0010
Epoch 6/20
0.9381
Epoch 6: val_accuracy did not improve from 0.87216
0.9381 - val_loss: 0.3537 - val_accuracy: 0.8677 - lr: 0.0010
Epoch 7/20
25/25 [=============== ] - ETA: Os - loss: 0.1508 - accuracy:
0.9449
Epoch 7: val_accuracy did not improve from 0.87216
0.9449 - val_loss: 0.3317 - val_accuracy: 0.8695 - lr: 0.0010
Epoch 8/20
25/25 [============= ] - ETA: Os - loss: 0.1353 - accuracy:
0.9512
Epoch 8: val_accuracy did not improve from 0.87216
0.9512 - val_loss: 0.3231 - val_accuracy: 0.8655 - lr: 0.0010
Epoch 9/20
0.9571
Epoch 9: val accuracy did not improve from 0.87216
0.9571 - val_loss: 0.3382 - val_accuracy: 0.8591 - lr: 0.0010
Epoch 10/20
25/25 [============== ] - ETA: Os - loss: 0.1094 - accuracy:
0.9624
Epoch 10: val_accuracy did not improve from 0.87216
0.9624 - val_loss: 0.3496 - val_accuracy: 0.8563 - lr: 0.0010
Epoch 11/20
25/25 [============== ] - ETA: Os - loss: 0.0956 - accuracy:
0.9677
```

```
Epoch 11: val_accuracy did not improve from 0.87216
0.9677 - val_loss: 0.3360 - val_accuracy: 0.8619 - lr: 0.0010
Epoch 12/20
0.9758
Epoch 12: val accuracy did not improve from 0.87216
0.9757 - val_loss: 0.3572 - val_accuracy: 0.8617 - lr: 0.0010
Epoch 13/20
Epoch 13: ReduceLROnPlateau reducing learning rate to 0.00050000000237487257.
Epoch 13: val_accuracy did not improve from 0.87216
0.9722 - val_loss: 0.3793 - val_accuracy: 0.8592 - lr: 0.0010
Epoch 14/20
25/25 [============== ] - ETA: Os - loss: 0.0577 - accuracy:
0.9839
Epoch 14: val_accuracy did not improve from 0.87216
0.9839 - val_loss: 0.4049 - val_accuracy: 0.8635 - lr: 5.0000e-04
Epoch 15/20
0.9907
Epoch 15: val_accuracy did not improve from 0.87216
0.9907 - val_loss: 0.4190 - val_accuracy: 0.8615 - lr: 5.0000e-04
Epoch 16/20
Epoch 16: val_accuracy did not improve from 0.87216
0.9935 - val loss: 0.4506 - val accuracy: 0.8592 - lr: 5.0000e-04
Epoch 17/20
0.9949
Epoch 17: val_accuracy did not improve from 0.87216
0.9949 - val_loss: 0.4684 - val_accuracy: 0.8578 - lr: 5.0000e-04
Epoch 18/20
0.9967
Epoch 18: ReduceLROnPlateau reducing learning rate to 0.0002500000118743628.
Epoch 18: val_accuracy did not improve from 0.87216
```



Elapsed time: 35.81513690948486

```
reduce_lr = ReduceLROnPlateau(monitor='val_loss', factor=0.5, patience=5,__
    min_lr=0.0000000000000000000000000001, verbose=1)
start_time = time.time()
inputs = keras.Input(shape=(seqLength, ))
x = layers.Embedding(maxFeatures, 16, mask_zero=True,_
  →input_length=seqLength)(inputs)
x = layers.Bidirectional(layers.SimpleRNN(64, return_sequences=False))(x)
x = keras.layers.BatchNormalization()(x)
outputs = layers.Dense(1, activation="sigmoid")(x)
model = keras.Model(inputs, outputs)
model.summary()
optimizer = keras.optimizers.Adam(learning_rate=0.001)
model.compile(optimizer=optimizer, loss="binary_crossentropy", __
   →metrics=["accuracy"])
result = fitModel(model, x_train_int, y_train, (x_test_int, y_test),__
   onum_epochs=20, model_name="simple_bi.h5", batch_size=1024, onum_epochs=20, onum_

callbacks=[reduce_lr])
end_time = time.time()
print("Elapsed time: ", end_time - start_time)
```

Model: "model_1"

Non-trainable params: 256

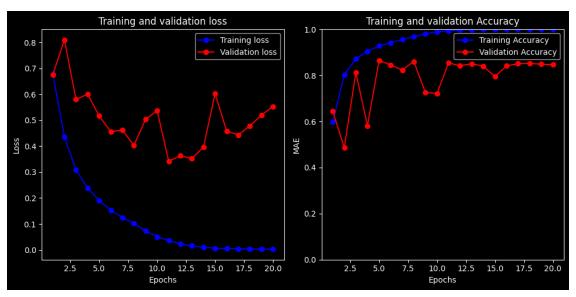
Layer (type)	Output Shape	Param #
input_2 (InputLayer)	[(None, 166)]	0
embedding_1 (Embedding)	(None, 166, 16)	80000
<pre>bidirectional (Bidirections 1)</pre>	n (None, 128)	10368
<pre>batch_normalization_1 (Batch hNormalization)</pre>	(None, 128)	512
dense_1 (Dense)	(None, 1)	129
Total params: 91,009 Trainable params: 90,753		

```
Epoch 1/20
25/25 [============= ] - ETA: Os - loss: 0.6776 - accuracy:
0.5968
Epoch 1: val_accuracy improved from -inf to 0.64556, saving model to
models/simple bi.h5
accuracy: 0.5968 - val_loss: 0.6760 - val_accuracy: 0.6456 - lr: 0.0010
Epoch 2/20
25/25 [============= ] - ETA: Os - loss: 0.4369 - accuracy:
0.8029
Epoch 2: val_accuracy did not improve from 0.64556
accuracy: 0.8029 - val_loss: 0.8095 - val_accuracy: 0.4863 - lr: 0.0010
Epoch 3/20
25/25 [============= ] - ETA: Os - loss: 0.3077 - accuracy:
0.8722
Epoch 3: val_accuracy improved from 0.64556 to 0.81312, saving model to
models/simple_bi.h5
accuracy: 0.8722 - val_loss: 0.5799 - val_accuracy: 0.8131 - lr: 0.0010
Epoch 4: val_accuracy did not improve from 0.81312
accuracy: 0.9047 - val_loss: 0.6006 - val_accuracy: 0.5810 - lr: 0.0010
Epoch 5/20
25/25 [============= ] - ETA: Os - loss: 0.1905 - accuracy:
Epoch 5: val_accuracy improved from 0.81312 to 0.86484, saving model to
models/simple_bi.h5
accuracy: 0.9281 - val_loss: 0.5164 - val_accuracy: 0.8648 - lr: 0.0010
Epoch 6/20
0.9426
Epoch 6: val accuracy did not improve from 0.86484
accuracy: 0.9426 - val_loss: 0.4556 - val_accuracy: 0.8460 - lr: 0.0010
Epoch 7/20
25/25 [============== ] - ETA: Os - loss: 0.1250 - accuracy:
0.9562
Epoch 7: val_accuracy did not improve from 0.86484
accuracy: 0.9562 - val_loss: 0.4615 - val_accuracy: 0.8233 - lr: 0.0010
25/25 [============== ] - ETA: Os - loss: 0.1020 - accuracy:
0.9680
```

```
Epoch 8: val_accuracy did not improve from 0.86484
accuracy: 0.9680 - val_loss: 0.4032 - val_accuracy: 0.8609 - lr: 0.0010
25/25 [============== ] - ETA: Os - loss: 0.0740 - accuracy:
0.9803
Epoch 9: val accuracy did not improve from 0.86484
accuracy: 0.9803 - val_loss: 0.5030 - val_accuracy: 0.7259 - lr: 0.0010
Epoch 10/20
25/25 [============= ] - ETA: Os - loss: 0.0507 - accuracy:
Epoch 10: val_accuracy did not improve from 0.86484
accuracy: 0.9897 - val_loss: 0.5367 - val_accuracy: 0.7218 - lr: 0.0010
Epoch 11/20
25/25 [============= ] - ETA: Os - loss: 0.0363 - accuracy:
Epoch 11: val_accuracy did not improve from 0.86484
accuracy: 0.9944 - val_loss: 0.3425 - val_accuracy: 0.8542 - lr: 0.0010
Epoch 12/20
25/25 [=============== ] - ETA: Os - loss: 0.0234 - accuracy:
0.9973
Epoch 12: val_accuracy did not improve from 0.86484
accuracy: 0.9973 - val_loss: 0.3624 - val_accuracy: 0.8424 - lr: 0.0010
Epoch 13/20
25/25 [============== ] - ETA: Os - loss: 0.0156 - accuracy:
0.9992
Epoch 13: val_accuracy did not improve from 0.86484
0.9992 - val_loss: 0.3521 - val_accuracy: 0.8509 - lr: 0.0010
Epoch 14/20
25/25 [============= ] - ETA: Os - loss: 0.0100 - accuracy:
0.9997
Epoch 14: val accuracy did not improve from 0.86484
0.9997 - val_loss: 0.3966 - val_accuracy: 0.8403 - lr: 0.0010
Epoch 15/20
25/25 [============== ] - ETA: Os - loss: 0.0067 - accuracy:
1.0000
Epoch 15: val_accuracy did not improve from 0.86484
accuracy: 1.0000 - val_loss: 0.6020 - val_accuracy: 0.7951 - lr: 0.0010
25/25 [============== ] - ETA: Os - loss: 0.0054 - accuracy:
1.0000
```

Epoch 16: ReduceLROnPlateau reducing learning rate to 0.0005000000237487257.

```
Epoch 16: val_accuracy did not improve from 0.86484
1.0000 - val loss: 0.4583 - val accuracy: 0.8415 - lr: 0.0010
Epoch 17/20
25/25 [============== ] - ETA: Os - loss: 0.0041 - accuracy:
1.0000
Epoch 17: val_accuracy did not improve from 0.86484
accuracy: 1.0000 - val_loss: 0.4431 - val_accuracy: 0.8512 - lr: 5.0000e-04
Epoch 18/20
1.0000
Epoch 18: val_accuracy did not improve from 0.86484
1.0000 - val_loss: 0.4778 - val_accuracy: 0.8526 - lr: 5.0000e-04
Epoch 19/20
1.0000
Epoch 19: val_accuracy did not improve from 0.86484
accuracy: 1.0000 - val_loss: 0.5193 - val_accuracy: 0.8490 - lr: 5.0000e-04
Epoch 20/20
1.0000
Epoch 20: val_accuracy did not improve from 0.86484
accuracy: 1.0000 - val_loss: 0.5531 - val_accuracy: 0.8463 - lr: 5.0000e-04
dict_keys(['loss', 'accuracy', 'val_loss', 'val_accuracy', 'lr'])
```



```
Elapsed time: 54.40829849243164
[]: model = keras.models.load_model(dir + "simple_bi.h5")
    model.evaluate(x_test_int, y_test)
    accuracy: 0.8649
[]: [0.5163887143135071, 0.8648800253868103]
    2.2 GRU
[]: from tensorflow.keras.callbacks import ReduceLROnPlateau
    np.random.seed(423423)
    tf.random.set_seed(1232413)
    reduce_lr = ReduceLROnPlateau(monitor='val_loss', factor=0.5, patience=5,_
     →min_lr=0.0000000000000000000001, verbose=1)
    start_time = time.time()
    inputs = keras.Input(shape=(seqLength, ))
    x = layers.Embedding(maxFeatures, 16, mask_zero=True,_
     →input_length=seqLength)(inputs)
    x = layers.GRU(64, return_sequences=False)(x)
    x = keras.layers.BatchNormalization()(x)
    outputs = layers.Dense(1, activation="sigmoid")(x)
    model = keras.Model(inputs, outputs)
    model.summary()
    optimizer = keras.optimizers.Adam(learning_rate=0.001)
    model.compile(optimizer=optimizer, loss="binary_crossentropy", __
     →metrics=["accuracy"])
    result = fitModel(model, x_train_int, y_train, (x_test_int, y_test),__
     onum_epochs=20, model_name="gru.h5", batch_size=1024, callbacks=[reduce_lr])
    end_time = time.time()
    print("Elapsed time: ", end_time - start_time)
    Model: "model 2"
    Layer (type)
                              Output Shape
                                                       Param #
```

0

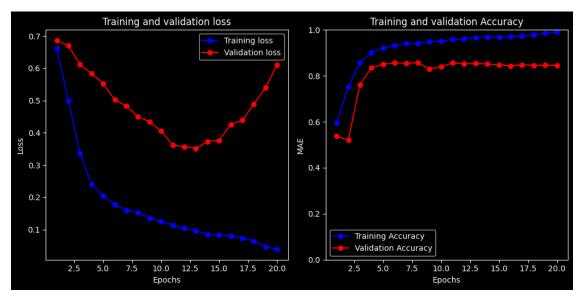
[(None, 166)]

input_3 (InputLayer)

```
embedding_2 (Embedding)
                    (None, 166, 16)
                                       80000
gru (GRU)
                     (None, 64)
                                        15744
batch_normalization_2 (Batc (None, 64)
                                        256
hNormalization)
dense 2 (Dense)
                     (None, 1)
                                        65
_____
Total params: 96,065
Trainable params: 95,937
Non-trainable params: 128
      ______
Epoch 1/20
2023-04-19 18:04:09.874587: I
tensorflow/compiler/xla/stream_executor/cuda/cuda_dnn.cc:428] Loaded cuDNN
version 8100
25/25 [============== ] - ETA: Os - loss: 0.6621 - accuracy:
0.5944
Epoch 1: val_accuracy improved from -inf to 0.53772, saving model to
models/gru.h5
0.5944 - val_loss: 0.6868 - val_accuracy: 0.5377 - lr: 0.0010
Epoch 2/20
25/25 [============= ] - ETA: Os - loss: 0.5002 - accuracy:
0.7534
Epoch 2: val_accuracy did not improve from 0.53772
0.7534 - val_loss: 0.6699 - val_accuracy: 0.5200 - lr: 0.0010
Epoch 3/20
0.8564
Epoch 3: val_accuracy improved from 0.53772 to 0.76196, saving model to
models/gru.h5
0.8566 - val_loss: 0.6119 - val_accuracy: 0.7620 - lr: 0.0010
Epoch 4/20
25/25 [============== ] - ETA: Os - loss: 0.2407 - accuracy:
0.9026
Epoch 4: val_accuracy improved from 0.76196 to 0.83520, saving model to
models/gru.h5
25/25 [============= ] - 1s 37ms/step - loss: 0.2407 - accuracy:
0.9026 - val_loss: 0.5834 - val_accuracy: 0.8352 - lr: 0.0010
Epoch 5/20
25/25 [============= ] - ETA: Os - loss: 0.2043 - accuracy:
```

```
0.9214
Epoch 5: val_accuracy improved from 0.83520 to 0.85024, saving model to
models/gru.h5
0.9214 - val_loss: 0.5530 - val_accuracy: 0.8502 - lr: 0.0010
Epoch 6/20
Epoch 6: val_accuracy improved from 0.85024 to 0.85644, saving model to
models/gru.h5
0.9328 - val_loss: 0.5025 - val_accuracy: 0.8564 - lr: 0.0010
Epoch 7/20
Epoch 7: val_accuracy did not improve from 0.85644
25/25 [============= ] - 1s 25ms/step - loss: 0.1597 - accuracy:
0.9408 - val_loss: 0.4832 - val_accuracy: 0.8546 - lr: 0.0010
Epoch 8/20
25/25 [============= ] - ETA: Os - loss: 0.1529 - accuracy:
0.9406
Epoch 8: val_accuracy improved from 0.85644 to 0.85732, saving model to
models/gru.h5
0.9406 - val_loss: 0.4496 - val_accuracy: 0.8573 - lr: 0.0010
Epoch 9/20
0.9493
Epoch 9: val_accuracy did not improve from 0.85732
0.9493 - val_loss: 0.4345 - val_accuracy: 0.8302 - lr: 0.0010
Epoch 10/20
25/25 [============== ] - ETA: Os - loss: 0.1262 - accuracy:
0.9519
Epoch 10: val accuracy did not improve from 0.85732
0.9519 - val_loss: 0.4057 - val_accuracy: 0.8393 - lr: 0.0010
Epoch 11/20
25/25 [============== ] - ETA: Os - loss: 0.1141 - accuracy:
0.9576
Epoch 11: val_accuracy did not improve from 0.85732
0.9576 - val_loss: 0.3613 - val_accuracy: 0.8564 - lr: 0.0010
Epoch 12/20
Epoch 12: val_accuracy did not improve from 0.85732
```

```
0.9613 - val_loss: 0.3572 - val_accuracy: 0.8527 - lr: 0.0010
Epoch 13/20
Epoch 13: val accuracy did not improve from 0.85732
0.9658 - val_loss: 0.3511 - val_accuracy: 0.8537 - lr: 0.0010
Epoch 14/20
0.9694
Epoch 14: val_accuracy did not improve from 0.85732
0.9693 - val_loss: 0.3733 - val_accuracy: 0.8515 - lr: 0.0010
Epoch 15/20
25/25 [============= ] - ETA: Os - loss: 0.0826 - accuracy:
0.9696
Epoch 15: val_accuracy did not improve from 0.85732
0.9696 - val_loss: 0.3764 - val_accuracy: 0.8476 - lr: 0.0010
Epoch 16/20
0.9727
Epoch 16: val_accuracy did not improve from 0.85732
0.9711 - val_loss: 0.4258 - val_accuracy: 0.8436 - lr: 0.0010
Epoch 17/20
0.9735
Epoch 17: val_accuracy did not improve from 0.85732
25/25 [============= ] - Os 19ms/step - loss: 0.0735 - accuracy:
0.9735 - val_loss: 0.4393 - val_accuracy: 0.8483 - lr: 0.0010
Epoch 18/20
0.9780
Epoch 18: ReduceLROnPlateau reducing learning rate to 0.00050000000237487257.
Epoch 18: val accuracy did not improve from 0.85732
0.9779 - val_loss: 0.4883 - val_accuracy: 0.8458 - lr: 0.0010
Epoch 19/20
25/25 [============== ] - ETA: Os - loss: 0.0473 - accuracy:
0.9867
Epoch 19: val_accuracy did not improve from 0.85732
0.9867 - val_loss: 0.5395 - val_accuracy: 0.8463 - lr: 5.0000e-04
Epoch 20/20
25/25 [============== ] - ETA: Os - loss: 0.0387 - accuracy:
0.9909
```



Elapsed time: 15.89973258972168

```
[]: model = keras.models.load_model(dir + "gru.h5")
model.evaluate(x_test_int, y_test)
```

[]: [0.44965192675590515, 0.8572800159454346]

Model: "model_3"

type_id: TFT_PRODUCT

type_id: TFT_TENSOR

args {

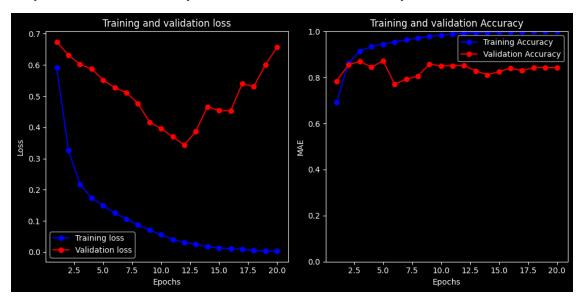
_		
Layer (type)	Output Shape	Param #
input_4 (InputLayer)	[(None, 166)]	0
<pre>embedding_3 (Embedding)</pre>	(None, 166, 16)	80000
<pre>bidirectional_1 (Bidirectio nal)</pre>	(None, 128)	31488
<pre>batch_normalization_3 (Batc hNormalization)</pre>	(None, 128)	512
dense_3 (Dense)	(None, 1)	129
Total params: 112,129 Trainable params: 111,873 Non-trainable params: 256 Epoch 1/20		
2023-04-19 18:04:29.218655: With tensorflow/core/common_runtime This indicates an invalid gradinal INVALID_ARGUMENT: expected control type_id: TFT_OPTIONAL args {	ne/type_inference.cc:339] caph that escaped type check	king. Error

```
args {
    type_id: TFT_INT32
 }
}
is neither a subtype nor a supertype of the combined inputs preceding it:
type id: TFT OPTIONAL
args {
 type_id: TFT_PRODUCT
 args {
  type_id: TFT_TENSOR
  args {
    type_id: TFT_FLOAT
  }
 }
}
     while inferring type of node 'cond_41/output/_22'
25/25 [============== ] - ETA: Os - loss: 0.5911 - accuracy:
0.6920
Epoch 1: val_accuracy improved from -inf to 0.78416, saving model to
models/gru bi.h5
0.6920 - val_loss: 0.6740 - val_accuracy: 0.7842 - lr: 0.0010
Epoch 2/20
25/25 [============== ] - ETA: Os - loss: 0.3267 - accuracy:
Epoch 2: val accuracy improved from 0.78416 to 0.85444, saving model to
models/gru_bi.h5
0.8618 - val_loss: 0.6317 - val_accuracy: 0.8544 - lr: 0.0010
Epoch 3/20
25/25 [============= ] - ETA: Os - loss: 0.2179 - accuracy:
0.9145
Epoch 3: val_accuracy improved from 0.85444 to 0.87052, saving model to
models/gru_bi.h5
0.9145 - val_loss: 0.6019 - val_accuracy: 0.8705 - lr: 0.0010
Epoch 4/20
25/25 [============== ] - ETA: Os - loss: 0.1736 - accuracy:
0.9338
Epoch 4: val_accuracy did not improve from 0.87052
0.9338 - val_loss: 0.5882 - val_accuracy: 0.8445 - lr: 0.0010
Epoch 5/20
25/25 [============= ] - ETA: Os - loss: 0.1491 - accuracy:
0.9444
```

```
Epoch 5: val_accuracy improved from 0.87052 to 0.87248, saving model to
models/gru_bi.h5
0.9444 - val_loss: 0.5516 - val_accuracy: 0.8725 - lr: 0.0010
Epoch 6/20
25/25 [============== ] - ETA: Os - loss: 0.1254 - accuracy:
0.9548
Epoch 6: val_accuracy did not improve from 0.87248
0.9548 - val_loss: 0.5273 - val_accuracy: 0.7707 - lr: 0.0010
Epoch 7/20
25/25 [============== ] - ETA: Os - loss: 0.1060 - accuracy:
0.9633
Epoch 7: val_accuracy did not improve from 0.87248
0.9633 - val_loss: 0.5115 - val_accuracy: 0.7925 - lr: 0.0010
Epoch 8/20
0.9717
Epoch 8: val accuracy did not improve from 0.87248
0.9717 - val_loss: 0.4767 - val_accuracy: 0.8055 - lr: 0.0010
Epoch 9/20
25/25 [============== ] - ETA: Os - loss: 0.0709 - accuracy:
0.9780
Epoch 9: val_accuracy did not improve from 0.87248
0.9780 - val_loss: 0.4169 - val_accuracy: 0.8579 - lr: 0.0010
Epoch 10/20
25/25 [============== ] - ETA: Os - loss: 0.0557 - accuracy:
0.9831
Epoch 10: val_accuracy did not improve from 0.87248
0.9831 - val_loss: 0.3959 - val_accuracy: 0.8508 - lr: 0.0010
Epoch 11/20
0.9890
Epoch 11: val_accuracy did not improve from 0.87248
0.9890 - val_loss: 0.3698 - val_accuracy: 0.8511 - lr: 0.0010
Epoch 12/20
0.9918
Epoch 12: val_accuracy did not improve from 0.87248
0.9918 - val_loss: 0.3437 - val_accuracy: 0.8516 - lr: 0.0010
Epoch 13/20
25/25 [============= ] - ETA: Os - loss: 0.0246 - accuracy:
```

```
0.9947
Epoch 13: val_accuracy did not improve from 0.87248
0.9947 - val_loss: 0.3874 - val_accuracy: 0.8288 - lr: 0.0010
Epoch 14/20
25/25 [============== ] - ETA: Os - loss: 0.0177 - accuracy:
0.9968
Epoch 14: val_accuracy did not improve from 0.87248
0.9968 - val_loss: 0.4663 - val_accuracy: 0.8116 - lr: 0.0010
Epoch 15/20
25/25 [============== ] - ETA: Os - loss: 0.0134 - accuracy:
0.9979
Epoch 15: val_accuracy did not improve from 0.87248
0.9979 - val_loss: 0.4537 - val_accuracy: 0.8256 - lr: 0.0010
Epoch 16/20
0.9984
Epoch 16: val accuracy did not improve from 0.87248
0.9984 - val_loss: 0.4532 - val_accuracy: 0.8402 - lr: 0.0010
Epoch 17/20
25/25 [============= ] - ETA: Os - loss: 0.0091 - accuracy:
0.9986
Epoch 17: ReduceLROnPlateau reducing learning rate to 0.00050000000237487257.
Epoch 17: val_accuracy did not improve from 0.87248
0.9986 - val_loss: 0.5399 - val_accuracy: 0.8298 - lr: 0.0010
Epoch 18/20
0.9996
Epoch 18: val_accuracy did not improve from 0.87248
0.9996 - val_loss: 0.5304 - val_accuracy: 0.8432 - lr: 5.0000e-04
Epoch 19/20
0.9999
Epoch 19: val_accuracy did not improve from 0.87248
0.9999 - val_loss: 0.6011 - val_accuracy: 0.8429 - lr: 5.0000e-04
Epoch 20/20
25/25 [============= ] - ETA: Os - loss: 0.0030 - accuracy:
0.9999
Epoch 20: val_accuracy did not improve from 0.87248
0.9999 - val_loss: 0.6564 - val_accuracy: 0.8426 - lr: 5.0000e-04
```

dict_keys(['loss', 'accuracy', 'val_loss', 'val_accuracy', 'lr'])



Elapsed time: 20.982495546340942

```
[]: model = keras.models.load_model(dir + "gru_bi.h5")
model.evaluate(x_test_int, y_test)
```

[]: [0.551629364490509, 0.8723999857902527]

2.3 LSTM

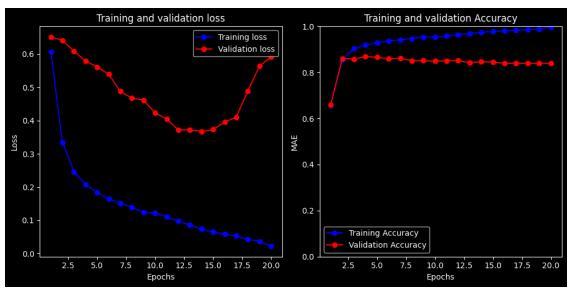
Model: "model_4"

Layer (type)	Output Shape	Param #	
input_5 (InputLayer)			
embedding_4 (Embedding)	(None, 166, 16)	80000	
lstm (LSTM)	(None, 64)	20736	
<pre>batch_normalization_4 (Batc hNormalization)</pre>	(None, 64)	256	
dense_4 (Dense)	(None, 1)	65	
Total params: 101,057 Trainable params: 100,929 Non-trainable params: 128			
Non-trainable params: 128			
Epoch 1/20 25/25 [====================================			
Epoch 1: val_accuracy improved from -inf to 0.66028, saving model to models/lstm.h5			
25/25 [====================================			
Epoch 2/20 25/25 [====================================	======] - ETA: Os - los	ss: 0.3353 - accuracy:	
Epoch 2: val_accuracy improved from 0.66028 to 0.86104, saving model to models/lstm.h5			
25/25 [============	======] - 1s 46ms/step	- loss: 0.3353 - accuracy:	

```
0.8561 - val_loss: 0.6414 - val_accuracy: 0.8610 - lr: 0.0010
Epoch 3/20
Epoch 3: val accuracy did not improve from 0.86104
0.9037 - val_loss: 0.6091 - val_accuracy: 0.8582 - lr: 0.0010
Epoch 4/20
25/25 [============= ] - ETA: Os - loss: 0.2082 - accuracy:
0.9184
Epoch 4: val accuracy improved from 0.86104 to 0.86868, saving model to
models/lstm.h5
0.9184 - val_loss: 0.5788 - val_accuracy: 0.8687 - lr: 0.0010
Epoch 5/20
25/25 [============== ] - ETA: Os - loss: 0.1842 - accuracy:
0.9302
Epoch 5: val_accuracy did not improve from 0.86868
0.9302 - val_loss: 0.5613 - val_accuracy: 0.8662 - lr: 0.0010
Epoch 6/20
Epoch 6: val_accuracy did not improve from 0.86868
0.9366 - val_loss: 0.5393 - val_accuracy: 0.8588 - lr: 0.0010
Epoch 7/20
0.9418
Epoch 7: val_accuracy did not improve from 0.86868
0.9418 - val_loss: 0.4878 - val_accuracy: 0.8621 - lr: 0.0010
Epoch 8/20
0.9476
Epoch 8: val_accuracy did not improve from 0.86868
0.9476 - val_loss: 0.4675 - val_accuracy: 0.8510 - lr: 0.0010
Epoch 9/20
0.9541
Epoch 9: val_accuracy did not improve from 0.86868
0.9541 - val_loss: 0.4616 - val_accuracy: 0.8512 - lr: 0.0010
Epoch 10/20
25/25 [============== ] - ETA: Os - loss: 0.1219 - accuracy:
0.9538
Epoch 10: val_accuracy did not improve from 0.86868
```

```
0.9538 - val_loss: 0.4229 - val_accuracy: 0.8488 - lr: 0.0010
Epoch 11/20
0.9583
Epoch 11: val_accuracy did not improve from 0.86868
0.9583 - val_loss: 0.4051 - val_accuracy: 0.8512 - lr: 0.0010
Epoch 12/20
0.9647
Epoch 12: val_accuracy did not improve from 0.86868
0.9642 - val_loss: 0.3714 - val_accuracy: 0.8514 - lr: 0.0010
Epoch 13/20
25/25 [============== ] - ETA: Os - loss: 0.0864 - accuracy:
0.9696
Epoch 13: val_accuracy did not improve from 0.86868
0.9696 - val_loss: 0.3728 - val_accuracy: 0.8423 - lr: 0.0010
Epoch 14/20
Epoch 14: val_accuracy did not improve from 0.86868
0.9745 - val_loss: 0.3668 - val_accuracy: 0.8465 - lr: 0.0010
Epoch 15/20
25/25 [============= ] - ETA: Os - loss: 0.0655 - accuracy:
0.9781
Epoch 15: val_accuracy did not improve from 0.86868
0.9781 - val_loss: 0.3732 - val_accuracy: 0.8450 - lr: 0.0010
Epoch 16/20
0.9816
Epoch 16: val_accuracy did not improve from 0.86868
0.9802 - val_loss: 0.3954 - val_accuracy: 0.8394 - lr: 0.0010
Epoch 17/20
0.9824
Epoch 17: val_accuracy did not improve from 0.86868
0.9824 - val_loss: 0.4104 - val_accuracy: 0.8405 - lr: 0.0010
Epoch 18/20
0.9867
Epoch 18: val_accuracy did not improve from 0.86868
```

```
0.9865 - val_loss: 0.4893 - val_accuracy: 0.8392 - lr: 0.0010
Epoch 19/20
25/25 [============== ] - ETA: Os - loss: 0.0371 - accuracy:
0.9887
Epoch 19: ReduceLROnPlateau reducing learning rate to 0.0005000000237487257.
Epoch 19: val_accuracy did not improve from 0.86868
0.9887 - val_loss: 0.5635 - val_accuracy: 0.8396 - lr: 0.0010
Epoch 20/20
25/25 [============= ] - ETA: Os - loss: 0.0223 - accuracy:
0.9957
Epoch 20: val_accuracy did not improve from 0.86868
0.9957 - val_loss: 0.5920 - val_accuracy: 0.8400 - lr: 5.0000e-04
dict_keys(['loss', 'accuracy', 'val_loss', 'val_accuracy', 'lr'])
```



Elapsed time: 15.766591787338257

```
[]: model = keras.models.load_model(dir + "lstm.h5")
model.evaluate(x_test_int, y_test)
```

[]: [0.5788054466247559, 0.8686800003051758]

```
[]: from tensorflow.keras.callbacks import ReduceLROnPlateau
     np.random.seed(423423)
     tf.random.set_seed(1232413)
     reduce_lr = ReduceLROnPlateau(monitor='val_loss', factor=0.5, patience=5,__
      min_lr=0.00000000000000000000000001, verbose=1)
     start_time = time.time()
     inputs = keras.Input(shape=(seqLength, ))
     x = layers.Embedding(maxFeatures, 16, mask_zero=True,_
      →input_length=seqLength)(inputs)
     x = layers.Bidirectional(layers.LSTM(64, return_sequences=False))(x)
     x = keras.layers.BatchNormalization()(x)
     outputs = layers.Dense(1, activation="sigmoid")(x)
     model = keras.Model(inputs, outputs)
     model.summary()
     optimizer = keras.optimizers.Adam(learning_rate=0.001)
     model.compile(optimizer=optimizer, loss="binary_crossentropy", __
      →metrics=["accuracy"])
     result = fitModel(model, x_train_int, y_train, (x_test_int, y_test),__
      onum_epochs=20, model_name="lstm_bi.h5", batch_size=1024, ___
      ⇔callbacks=[reduce_lr])
     end_time = time.time()
     print("Elapsed time: ", end_time - start_time)
```

Model: "model_5"

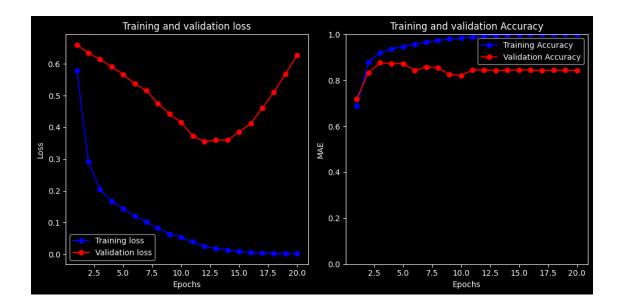
Layer (type)	Output Shape	Param #
input_6 (InputLayer)	[(None, 166)]	0
embedding_5 (Embedding)	(None, 166, 16)	80000
<pre>bidirectional_2 (Bidirectio nal)</pre>	(None, 128)	41472
<pre>batch_normalization_5 (Batc hNormalization)</pre>	(None, 128)	512
dense_5 (Dense)	(None, 1)	129

```
Non-trainable params: 256
______
Epoch 1/20
25/25 [============== ] - ETA: Os - loss: 0.5797 - accuracy:
Epoch 1: val_accuracy improved from -inf to 0.71900, saving model to
models/lstm bi.h5
accuracy: 0.6893 - val_loss: 0.6602 - val_accuracy: 0.7190 - lr: 0.0010
25/25 [============= ] - ETA: Os - loss: 0.2929 - accuracy:
Epoch 2: val_accuracy improved from 0.71900 to 0.83356, saving model to
models/lstm_bi.h5
25/25 [============= ] - 1s 54ms/step - loss: 0.2929 - accuracy:
0.8786 - val_loss: 0.6344 - val_accuracy: 0.8336 - lr: 0.0010
Epoch 3/20
25/25 [============= ] - ETA: Os - loss: 0.2040 - accuracy:
Epoch 3: val_accuracy improved from 0.83356 to 0.87624, saving model to
models/lstm bi.h5
0.9198 - val_loss: 0.6140 - val_accuracy: 0.8762 - lr: 0.0010
Epoch 4/20
25/25 [============== ] - ETA: Os - loss: 0.1666 - accuracy:
0.9366
Epoch 4: val_accuracy did not improve from 0.87624
0.9366 - val_loss: 0.5917 - val_accuracy: 0.8730 - lr: 0.0010
Epoch 5/20
25/25 [============== ] - ETA: Os - loss: 0.1442 - accuracy:
0.9465
Epoch 5: val accuracy did not improve from 0.87624
0.9465 - val_loss: 0.5669 - val_accuracy: 0.8729 - lr: 0.0010
Epoch 6/20
25/25 [============== ] - ETA: Os - loss: 0.1202 - accuracy:
0.9565
Epoch 6: val_accuracy did not improve from 0.87624
0.9565 - val_loss: 0.5371 - val_accuracy: 0.8431 - lr: 0.0010
Epoch 7/20
Epoch 7: val_accuracy did not improve from 0.87624
```

Total params: 122,113 Trainable params: 121,857

```
0.9650 - val_loss: 0.5157 - val_accuracy: 0.8582 - lr: 0.0010
Epoch 8/20
Epoch 8: val accuracy did not improve from 0.87624
0.9727 - val_loss: 0.4755 - val_accuracy: 0.8546 - lr: 0.0010
Epoch 9/20
25/25 [============= ] - ETA: Os - loss: 0.0637 - accuracy:
0.9806
Epoch 9: val_accuracy did not improve from 0.87624
0.9806 - val_loss: 0.4422 - val_accuracy: 0.8264 - lr: 0.0010
Epoch 10/20
0.9842
Epoch 10: val_accuracy did not improve from 0.87624
0.9842 - val_loss: 0.4157 - val_accuracy: 0.8209 - lr: 0.0010
Epoch 11/20
0.9894
Epoch 11: val_accuracy did not improve from 0.87624
0.9894 - val_loss: 0.3733 - val_accuracy: 0.8456 - lr: 0.0010
Epoch 12/20
0.9949
Epoch 12: val_accuracy did not improve from 0.87624
0.9946 - val_loss: 0.3549 - val_accuracy: 0.8455 - lr: 0.0010
Epoch 13/20
25/25 [============== ] - ETA: Os - loss: 0.0178 - accuracy:
0.9967
Epoch 13: val accuracy did not improve from 0.87624
0.9967 - val_loss: 0.3595 - val_accuracy: 0.8427 - lr: 0.0010
Epoch 14/20
25/25 [============== ] - ETA: Os - loss: 0.0133 - accuracy:
0.9979
Epoch 14: val_accuracy did not improve from 0.87624
0.9979 - val_loss: 0.3600 - val_accuracy: 0.8446 - lr: 0.0010
Epoch 15/20
Epoch 15: val_accuracy did not improve from 0.87624
```

```
0.9989 - val_loss: 0.3855 - val_accuracy: 0.8446 - lr: 0.0010
Epoch 16/20
Epoch 16: val accuracy did not improve from 0.87624
0.9996 - val_loss: 0.4119 - val_accuracy: 0.8453 - lr: 0.0010
Epoch 17/20
25/25 [============= ] - ETA: Os - loss: 0.0037 - accuracy:
0.9998
Epoch 17: ReduceLROnPlateau reducing learning rate to 0.00050000000237487257.
Epoch 17: val_accuracy did not improve from 0.87624
0.9998 - val_loss: 0.4609 - val_accuracy: 0.8423 - lr: 0.0010
Epoch 18/20
Epoch 18: val_accuracy did not improve from 0.87624
0.9999 - val_loss: 0.5103 - val_accuracy: 0.8448 - lr: 5.0000e-04
Epoch 19/20
25/25 [=============== ] - ETA: Os - loss: 0.0017 - accuracy:
1.0000
Epoch 19: val_accuracy did not improve from 0.87624
1.0000 - val_loss: 0.5683 - val_accuracy: 0.8440 - lr: 5.0000e-04
Epoch 20/20
25/25 [============ ] - ETA: Os - loss: 0.0015 - accuracy:
1.0000
Epoch 20: val_accuracy did not improve from 0.87624
1.0000 - val_loss: 0.6275 - val_accuracy: 0.8433 - lr: 5.0000e-04
dict_keys(['loss', 'accuracy', 'val_loss', 'val_accuracy', 'lr'])
```



Elapsed time: 22.130324840545654

3 Intentado mejorar el rendimiento

```
x = layers.Bidirectional(layers.GRU(16, return_sequences=False))(x)
x = keras.layers.BatchNormalization()(x)
x = keras.layers.Dense(128, activation="relu")(x)
x = keras.layers.BatchNormalization()(x)
x = keras.layers.Dense(32, activation="relu")(x)
x = keras.layers.BatchNormalization()(x)
x = keras.layers.Dense(8, activation="relu")(x)
x = keras.layers.BatchNormalization()(x)
outputs = layers.Dense(1, activation="sigmoid")(x)
model = keras.Model(inputs, outputs)
model.summary()
optimizer = keras.optimizers.Adam(learning_rate=0.001)
model.compile(optimizer=optimizer, loss="binary_crossentropy", __
 ⇔metrics=["accuracy"])
result = fitModel(model, x_train_int, y_train, (x_test_int, y_test),__
 onum_epochs=50, model_name="gru_bi_a0.h5", batch_size=1024, □
⇔callbacks=[reduce_lr])
end_time = time.time()
print("Elapsed time: ", end_time - start_time)
```

Model: "model_1"

Layer (type)	Output Shape	Param #
input_2 (InputLayer)	[(None, 166)]	0
<pre>embedding_1 (Embedding)</pre>	(None, 166, 128)	704000
<pre>batch_normalization_7 (Batc hNormalization)</pre>	(None, 166, 128)	512
<pre>bidirectional_3 (Bidirectio nal)</pre>	(None, 166, 1024)	1972224
<pre>batch_normalization_8 (BatchNormalization)</pre>	(None, 166, 1024)	4096
<pre>bidirectional_4 (Bidirectio nal)</pre>	(None, 166, 256)	886272
<pre>batch_normalization_9 (Batc hNormalization)</pre>	(None, 166, 256)	1024

```
bidirectional_5 (Bidirectio (None, 32)
                                                    26304
 nal)
 batch_normalization_10 (Bat (None, 32)
                                                    128
 chNormalization)
dense 4 (Dense)
                         (None, 128)
                                                    4224
 batch_normalization_11 (Bat (None, 128)
                                                    512
 chNormalization)
                           (None, 32)
 dense_5 (Dense)
                                                    4128
 batch_normalization_12 (Bat (None, 32)
                                                    128
 chNormalization)
 dense_6 (Dense)
                           (None, 8)
                                                    264
 batch_normalization_13 (Bat (None, 8)
                                                    32
 chNormalization)
 dense 7 (Dense)
                           (None, 1)
Total params: 3,603,857
Trainable params: 3,600,641
Non-trainable params: 3,216
             _____
Epoch 1/50
2023-04-19 18:48:44.414353: W
tensorflow/core/common runtime/type_inference.cc:339] Type inference failed.
This indicates an invalid graph that escaped type checking. Error message:
INVALID ARGUMENT: expected compatible input types, but input 1:
type_id: TFT_OPTIONAL
args {
 type_id: TFT_PRODUCT
 args {
   type_id: TFT_TENSOR
   args {
     type_id: TFT_FLOAT
 }
}
 is neither a subtype nor a supertype of the combined inputs preceding it:
type_id: TFT_OPTIONAL
args {
 type_id: TFT_PRODUCT
```

```
args {
   type_id: TFT_TENSOR
   args {
     type_id: TFT_INT8
   }
 }
}
      while inferring type of node 'cond_43/output/_22'
2023-04-19 18:48:44.866746: I
tensorflow/compiler/xla/stream executor/cuda/cuda dnn.cc:428] Loaded cuDNN
version 8100
2023-04-19 18:48:45.775788: I
tensorflow/compiler/xla/stream executor/cuda/cuda blas.cc:630] TensorFloat-32
will be used for the matrix multiplication. This will only be logged once.
2023-04-19 18:48:45.777527: I tensorflow/compiler/xla/service/service.cc:173]
XLA service 0x7f7950039440 initialized for platform CUDA (this does not
guarantee that XLA will be used). Devices:
2023-04-19 18:48:45.777537: I tensorflow/compiler/xla/service/service.cc:181]
StreamExecutor device (0): NVIDIA GeForce RTX 4090, Compute Capability 8.9
2023-04-19 18:48:45.779699: I
tensorflow/compiler/mlir/tensorflow/utils/dump mlir util.cc:268] disabling MLIR
crash reproducer, set env var `MLIR_CRASH_REPRODUCER_DIRECTORY` to enable.
2023-04-19 18:48:45.813118: I tensorflow/tsl/platform/default/subprocess.cc:304]
Start cannot spawn child process: No such file or directory
2023-04-19 18:48:45.835131: I
tensorflow/compiler/jit/xla_compilation_cache.cc:477] Compiled cluster using
    This line is logged at most once for the lifetime of the process.
Epoch 1: val_accuracy improved from -inf to 0.48628, saving model to
models/gru_bi_a0.h5
accuracy: 0.6955 - val_loss: 0.6889 - val_accuracy: 0.4863 - lr: 0.0010
Epoch 2/50
0.8714
Epoch 2: val_accuracy did not improve from 0.48628
accuracy: 0.8714 - val_loss: 0.7445 - val_accuracy: 0.4863 - lr: 0.0010
Epoch 3/50
25/25 [============== ] - ETA: Os - loss: 0.1856 - accuracy:
0.9324
Epoch 3: val_accuracy did not improve from 0.48628
accuracy: 0.9324 - val_loss: 0.8110 - val_accuracy: 0.4863 - lr: 0.0010
Epoch 4/50
```

```
25/25 [============== ] - ETA: Os - loss: 0.1056 - accuracy:
0.9694
Epoch 4: val accuracy improved from 0.48628 to 0.48668, saving model to
models/gru_bi_a0.h5
accuracy: 0.9694 - val_loss: 0.7903 - val_accuracy: 0.4867 - lr: 0.0010
25/25 [=============== ] - ETA: Os - loss: 0.0633 - accuracy:
0.9854
Epoch 5: val_accuracy improved from 0.48668 to 0.49128, saving model to
models/gru_bi_a0.h5
accuracy: 0.9854 - val_loss: 0.7562 - val_accuracy: 0.4913 - lr: 0.0010
Epoch 6/50
25/25 [============= ] - ETA: Os - loss: 0.0355 - accuracy:
0.9941
Epoch 6: ReduceLROnPlateau reducing learning rate to 0.0005000000237487257.
Epoch 6: val_accuracy improved from 0.49128 to 0.52432, saving model to
models/gru bi a0.h5
accuracy: 0.9941 - val_loss: 0.7245 - val_accuracy: 0.5243 - lr: 0.0010
Epoch 7/50
25/25 [============== ] - ETA: Os - loss: 0.0205 - accuracy:
0.9978
Epoch 7: val accuracy improved from 0.52432 to 0.54636, saving model to
models/gru_bi_a0.h5
accuracy: 0.9978 - val_loss: 0.7324 - val_accuracy: 0.5464 - lr: 5.0000e-04
Epoch 8/50
25/25 [============== ] - ETA: Os - loss: 0.0142 - accuracy:
Epoch 8: val accuracy improved from 0.54636 to 0.55388, saving model to
models/gru_bi_a0.h5
accuracy: 0.9991 - val_loss: 0.7996 - val_accuracy: 0.5539 - lr: 5.0000e-04
Epoch 9/50
25/25 [=============== ] - ETA: Os - loss: 0.0112 - accuracy:
0.9993
Epoch 9: val_accuracy improved from 0.55388 to 0.58840, saving model to
models/gru_bi_a0.h5
accuracy: 0.9993 - val_loss: 0.7922 - val_accuracy: 0.5884 - lr: 5.0000e-04
Epoch 10/50
25/25 [============= ] - ETA: Os - loss: 0.0095 - accuracy:
Epoch 10: val_accuracy improved from 0.58840 to 0.61616, saving model to
models/gru_bi_a0.h5
```

```
accuracy: 0.9995 - val_loss: 0.8130 - val_accuracy: 0.6162 - lr: 5.0000e-04
Epoch 11/50
0.9995
Epoch 11: ReduceLROnPlateau reducing learning rate to 0.0002500000118743628.
Epoch 11: val_accuracy improved from 0.61616 to 0.65148, saving model to
models/gru bi a0.h5
25/25 [=========== ] - 5s 211ms/step - loss: 0.0082 -
accuracy: 0.9995 - val_loss: 0.7976 - val_accuracy: 0.6515 - lr: 5.0000e-04
Epoch 12/50
25/25 [============= ] - ETA: Os - loss: 0.0073 - accuracy:
0.9996
Epoch 12: val_accuracy improved from 0.65148 to 0.69696, saving model to
models/gru_bi_a0.h5
25/25 [=========== ] - 5s 214ms/step - loss: 0.0073 -
accuracy: 0.9996 - val_loss: 0.7296 - val_accuracy: 0.6970 - lr: 2.5000e-04
Epoch 13/50
25/25 [============= ] - ETA: Os - loss: 0.0069 - accuracy:
0.9996
Epoch 13: val_accuracy improved from 0.69696 to 0.73912, saving model to
models/gru_bi_a0.h5
25/25 [========== ] - 5s 213ms/step - loss: 0.0069 -
accuracy: 0.9996 - val_loss: 0.6680 - val_accuracy: 0.7391 - lr: 2.5000e-04
Epoch 14/50
25/25 [============= ] - ETA: Os - loss: 0.0065 - accuracy:
0.9997
Epoch 14: val_accuracy improved from 0.73912 to 0.76892, saving model to
models/gru_bi_a0.h5
accuracy: 0.9997 - val_loss: 0.6262 - val_accuracy: 0.7689 - lr: 2.5000e-04
Epoch 15/50
0.9997
Epoch 15: val_accuracy improved from 0.76892 to 0.79552, saving model to
models/gru bi a0.h5
25/25 [=========== ] - 5s 209ms/step - loss: 0.0061 -
accuracy: 0.9997 - val_loss: 0.5993 - val_accuracy: 0.7955 - lr: 2.5000e-04
Epoch 16/50
25/25 [============== ] - ETA: Os - loss: 0.0058 - accuracy:
0.9997
Epoch 16: val_accuracy improved from 0.79552 to 0.81120, saving model to
models/gru_bi_a0.h5
accuracy: 0.9997 - val_loss: 0.5977 - val_accuracy: 0.8112 - lr: 2.5000e-04
Epoch 17/50
25/25 [============= ] - ETA: Os - loss: 0.0055 - accuracy:
```

```
0.9998
Epoch 17: val_accuracy improved from 0.81120 to 0.82324, saving model to
models/gru_bi_a0.h5
25/25 [============ ] - 5s 207ms/step - loss: 0.0055 -
accuracy: 0.9998 - val loss: 0.5985 - val accuracy: 0.8232 - lr: 2.5000e-04
Epoch 18/50
25/25 [============= ] - ETA: Os - loss: 0.0050 - accuracy:
0.9998
Epoch 18: val_accuracy improved from 0.82324 to 0.83300, saving model to
models/gru_bi_a0.h5
accuracy: 0.9998 - val_loss: 0.6039 - val_accuracy: 0.8330 - lr: 2.5000e-04
Epoch 19/50
25/25 [============ ] - ETA: Os - loss: 0.0048 - accuracy:
Epoch 19: val_accuracy improved from 0.83300 to 0.84036, saving model to
models/gru_bi_a0.h5
accuracy: 0.9998 - val_loss: 0.6105 - val_accuracy: 0.8404 - lr: 2.5000e-04
Epoch 20/50
25/25 [============== ] - ETA: Os - loss: 0.0046 - accuracy:
0.9998
Epoch 20: val_accuracy improved from 0.84036 to 0.84568, saving model to
models/gru_bi_a0.h5
accuracy: 0.9998 - val_loss: 0.6062 - val_accuracy: 0.8457 - lr: 2.5000e-04
Epoch 21/50
25/25 [============== ] - ETA: Os - loss: 0.0043 - accuracy:
0.9998
Epoch 21: val_accuracy improved from 0.84568 to 0.85204, saving model to
models/gru_bi_a0.h5
accuracy: 0.9998 - val_loss: 0.5888 - val_accuracy: 0.8520 - lr: 2.5000e-04
Epoch 22/50
25/25 [============= ] - ETA: Os - loss: 0.0042 - accuracy:
0.9998
Epoch 22: val_accuracy improved from 0.85204 to 0.85844, saving model to
models/gru_bi_a0.h5
accuracy: 0.9998 - val_loss: 0.5634 - val_accuracy: 0.8584 - lr: 2.5000e-04
Epoch 23/50
25/25 [============== ] - ETA: Os - loss: 0.0039 - accuracy:
Epoch 23: val_accuracy improved from 0.85844 to 0.86480, saving model to
models/gru_bi_a0.h5
accuracy: 0.9998 - val_loss: 0.5414 - val_accuracy: 0.8648 - lr: 2.5000e-04
Epoch 24/50
```

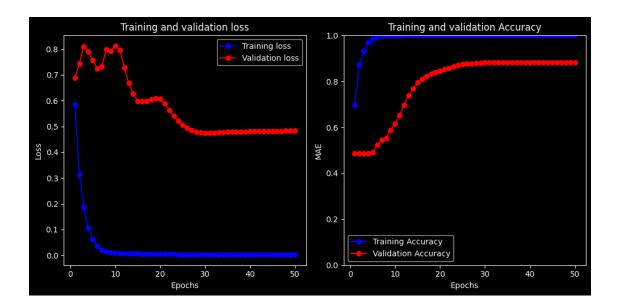
```
25/25 [============== ] - ETA: Os - loss: 0.0037 - accuracy:
0.9998
Epoch 24: val_accuracy improved from 0.86480 to 0.87020, saving model to
models/gru_bi_a0.h5
accuracy: 0.9998 - val_loss: 0.5217 - val_accuracy: 0.8702 - lr: 2.5000e-04
Epoch 25/50
25/25 [=============== ] - ETA: Os - loss: 0.0036 - accuracy:
0.9999
Epoch 25: val_accuracy improved from 0.87020 to 0.87420, saving model to
models/gru_bi_a0.h5
accuracy: 0.9999 - val_loss: 0.5064 - val_accuracy: 0.8742 - lr: 2.5000e-04
Epoch 26/50
0.9999
Epoch 26: val_accuracy improved from 0.87420 to 0.87640, saving model to
models/gru_bi_a0.h5
accuracy: 0.9999 - val_loss: 0.4945 - val_accuracy: 0.8764 - lr: 2.5000e-04
Epoch 27/50
25/25 [============== ] - ETA: Os - loss: 0.0033 - accuracy:
Epoch 27: val_accuracy improved from 0.87640 to 0.87796, saving model to
models/gru_bi_a0.h5
accuracy: 0.9999 - val_loss: 0.4851 - val_accuracy: 0.8780 - lr: 2.5000e-04
Epoch 28/50
25/25 [============== ] - ETA: Os - loss: 0.0032 - accuracy:
0.9999
Epoch 28: val_accuracy improved from 0.87796 to 0.87980, saving model to
models/gru_bi_a0.h5
25/25 [========== ] - 5s 210ms/step - loss: 0.0032 -
accuracy: 0.9999 - val_loss: 0.4798 - val_accuracy: 0.8798 - lr: 2.5000e-04
Epoch 29/50
25/25 [============== ] - ETA: Os - loss: 0.0031 - accuracy:
Epoch 29: val_accuracy improved from 0.87980 to 0.88056, saving model to
models/gru_bi_a0.h5
25/25 [=========== ] - 5s 211ms/step - loss: 0.0031 -
accuracy: 0.9999 - val_loss: 0.4769 - val_accuracy: 0.8806 - lr: 2.5000e-04
Epoch 30/50
25/25 [============= ] - ETA: Os - loss: 0.0029 - accuracy:
0.9999
Epoch 30: val_accuracy improved from 0.88056 to 0.88120, saving model to
models/gru_bi_a0.h5
accuracy: 0.9999 - val_loss: 0.4745 - val_accuracy: 0.8812 - lr: 2.5000e-04
```

```
Epoch 31/50
25/25 [============= ] - ETA: Os - loss: 0.0028 - accuracy:
0.9999
Epoch 31: val_accuracy improved from 0.88120 to 0.88156, saving model to
models/gru bi a0.h5
25/25 [============== ] - 5s 211ms/step - loss: 0.0028 -
accuracy: 0.9999 - val loss: 0.4746 - val accuracy: 0.8816 - lr: 2.5000e-04
Epoch 32/50
25/25 [============= ] - ETA: Os - loss: 0.0028 - accuracy:
0.9999
Epoch 32: val_accuracy improved from 0.88156 to 0.88224, saving model to
models/gru_bi_a0.h5
25/25 [=========== ] - 5s 209ms/step - loss: 0.0028 -
accuracy: 0.9999 - val_loss: 0.4747 - val_accuracy: 0.8822 - lr: 2.5000e-04
Epoch 33/50
25/25 [============== ] - ETA: Os - loss: 0.0026 - accuracy:
0.9999
Epoch 33: val_accuracy improved from 0.88224 to 0.88236, saving model to
models/gru_bi_a0.h5
accuracy: 0.9999 - val_loss: 0.4761 - val_accuracy: 0.8824 - lr: 2.5000e-04
Epoch 34/50
25/25 [=============== ] - ETA: Os - loss: 0.0025 - accuracy:
0.9999
Epoch 34: val_accuracy did not improve from 0.88236
accuracy: 0.9999 - val_loss: 0.4761 - val_accuracy: 0.8820 - lr: 2.5000e-04
Epoch 35/50
25/25 [============ ] - ETA: Os - loss: 0.0024 - accuracy:
0.9999
Epoch 35: ReduceLROnPlateau reducing learning rate to 0.0001250000059371814.
Epoch 35: val_accuracy improved from 0.88236 to 0.88276, saving model to
models/gru_bi_a0.h5
25/25 [=========== ] - 5s 204ms/step - loss: 0.0024 -
accuracy: 0.9999 - val_loss: 0.4791 - val_accuracy: 0.8828 - lr: 2.5000e-04
Epoch 36/50
25/25 [=============== ] - ETA: Os - loss: 0.0023 - accuracy:
Epoch 36: val_accuracy improved from 0.88276 to 0.88280, saving model to
models/gru_bi_a0.h5
accuracy: 0.9999 - val_loss: 0.4786 - val_accuracy: 0.8828 - lr: 1.2500e-04
Epoch 37/50
Epoch 37: val_accuracy did not improve from 0.88280
```

```
accuracy: 0.9999 - val_loss: 0.4785 - val_accuracy: 0.8821 - lr: 1.2500e-04
Epoch 38/50
Epoch 38: val accuracy did not improve from 0.88280
accuracy: 0.9999 - val_loss: 0.4789 - val_accuracy: 0.8819 - lr: 1.2500e-04
Epoch 39/50
25/25 [============= ] - ETA: Os - loss: 0.0022 - accuracy:
0.9999
Epoch 39: val_accuracy did not improve from 0.88280
accuracy: 0.9999 - val_loss: 0.4795 - val_accuracy: 0.8816 - lr: 1.2500e-04
Epoch 40/50
25/25 [============= ] - ETA: Os - loss: 0.0021 - accuracy:
0.9999
Epoch 40: ReduceLROnPlateau reducing learning rate to 6.25000029685907e-05.
Epoch 40: val_accuracy did not improve from 0.88280
accuracy: 0.9999 - val_loss: 0.4804 - val_accuracy: 0.8818 - lr: 1.2500e-04
Epoch 41/50
25/25 [=============== ] - ETA: Os - loss: 0.0021 - accuracy:
0.9999
Epoch 41: val_accuracy did not improve from 0.88280
accuracy: 0.9999 - val_loss: 0.4807 - val_accuracy: 0.8819 - lr: 6.2500e-05
Epoch 42/50
25/25 [============== ] - ETA: Os - loss: 0.0021 - accuracy:
0.9999
Epoch 42: val_accuracy did not improve from 0.88280
accuracy: 0.9999 - val_loss: 0.4811 - val_accuracy: 0.8822 - lr: 6.2500e-05
Epoch 43/50
0.9999
Epoch 43: val accuracy did not improve from 0.88280
accuracy: 0.9999 - val_loss: 0.4815 - val_accuracy: 0.8818 - lr: 6.2500e-05
Epoch 44/50
25/25 [============== ] - ETA: Os - loss: 0.0021 - accuracy:
0.9999
Epoch 44: val_accuracy did not improve from 0.88280
accuracy: 0.9999 - val_loss: 0.4820 - val_accuracy: 0.8822 - lr: 6.2500e-05
25/25 [============== ] - ETA: Os - loss: 0.0020 - accuracy:
0.9999
```

```
Epoch 45: ReduceLROnPlateau reducing learning rate to 3.125000148429535e-05.
```

```
Epoch 45: val_accuracy did not improve from 0.88280
25/25 [============== ] - 5s 204ms/step - loss: 0.0020 -
accuracy: 0.9999 - val_loss: 0.4823 - val_accuracy: 0.8819 - lr: 6.2500e-05
Epoch 46/50
25/25 [============= ] - ETA: Os - loss: 0.0020 - accuracy:
0.9999
Epoch 46: val_accuracy did not improve from 0.88280
accuracy: 0.9999 - val_loss: 0.4824 - val_accuracy: 0.8819 - lr: 3.1250e-05
Epoch 47/50
25/25 [============= ] - ETA: Os - loss: 0.0020 - accuracy:
0.9999
Epoch 47: val_accuracy did not improve from 0.88280
accuracy: 0.9999 - val_loss: 0.4825 - val_accuracy: 0.8819 - lr: 3.1250e-05
Epoch 48/50
25/25 [============== ] - ETA: Os - loss: 0.0020 - accuracy:
0.9999
Epoch 48: val_accuracy did not improve from 0.88280
accuracy: 0.9999 - val_loss: 0.4827 - val_accuracy: 0.8820 - lr: 3.1250e-05
Epoch 49/50
0.9999
Epoch 49: val_accuracy did not improve from 0.88280
25/25 [=========== ] - 5s 206ms/step - loss: 0.0020 -
accuracy: 0.9999 - val_loss: 0.4829 - val_accuracy: 0.8820 - lr: 3.1250e-05
Epoch 50/50
25/25 [============== ] - ETA: Os - loss: 0.0020 - accuracy:
Epoch 50: ReduceLROnPlateau reducing learning rate to 1.5625000742147677e-05.
Epoch 50: val accuracy did not improve from 0.88280
accuracy: 0.9999 - val_loss: 0.4832 - val_accuracy: 0.8820 - lr: 3.1250e-05
dict_keys(['loss', 'accuracy', 'val_loss', 'val_accuracy', 'lr'])
```



Elapsed time: 275.1228439807892

```
[]: model = keras.models.load_model(dir + "gru_bi_a0.h5")
results = model.evaluate(x_test_int, y_test)
print(results)
```

```
[]: from tensorflow.keras.callbacks import ReduceLROnPlateau
     np.random.seed(423423)
     tf.random.set seed(1232413)
     reduce_lr = ReduceLROnPlateau(monitor='val_loss', factor=0.5, patience=5, __
      min_lr=0.000000000000000000000000001, verbose=1)
     start_time = time.time()
     inputs = keras.Input(shape=(seqLength, ))
     x = layers.Embedding(maxFeatures, 128, mask_zero=True, input_length=256)(inputs)
     x = keras.layers.BatchNormalization()(x)
     x = layers.Bidirectional(layers.GRU(512, return_sequences=True))(x)
     x = keras.layers.BatchNormalization()(x)
     x = layers.Bidirectional(layers.GRU(256, return_sequences=False))(x)
     x = keras.layers.BatchNormalization()(x)
     x = keras.layers.Dense(128, activation="relu")(x)
     x = keras.layers.BatchNormalization()(x)
     x = keras.layers.Dense(32, activation="relu")(x)
```

Model: "model_10"

Layer (type)		
input_11 (InputLayer)		0
embedding_10 (Embedding)	(None, 166, 128)	704000
<pre>batch_normalization_25 (Bat chNormalization)</pre>	(None, 166, 128)	512
<pre>bidirectional_8 (Bidirectio nal)</pre>	(None, 166, 1024)	1972224
<pre>batch_normalization_26 (Bat chNormalization)</pre>	(None, 166, 1024)	4096
<pre>bidirectional_9 (Bidirectio nal)</pre>	(None, 512)	1969152
<pre>batch_normalization_27 (Bat chNormalization)</pre>	(None, 512)	2048
dense_20 (Dense)	(None, 128)	65664
<pre>batch_normalization_28 (Bat chNormalization)</pre>	(None, 128)	512

```
dense_21 (Dense)
                                       4128
                    (None, 32)
batch_normalization_29 (Bat (None, 32)
                                       128
chNormalization)
dense 22 (Dense)
                    (None, 8)
                                       264
batch_normalization_30 (Bat (None, 8)
                                       32
chNormalization)
                    (None, 1)
dense_23 (Dense)
_____
Total params: 4,722,769
Trainable params: 4,719,105
Non-trainable params: 3,664
      .....
Epoch 1/40
25/25 [============== ] - ETA: Os - loss: 0.4815 - accuracy:
0.7684
Epoch 1: val_accuracy improved from -inf to 0.51372, saving model to
models/gru bi a.h5
25/25 [============== ] - 13s 264ms/step - loss: 0.4815 -
accuracy: 0.7684 - val_loss: 0.7493 - val_accuracy: 0.5137 - lr: 0.0010
Epoch 2/40
0.9144
Epoch 2: val_accuracy did not improve from 0.51372
accuracy: 0.9144 - val_loss: 0.7416 - val_accuracy: 0.5137 - lr: 0.0010
Epoch 3/40
25/25 [============== ] - ETA: Os - loss: 0.1251 - accuracy:
0.9630
Epoch 3: val_accuracy did not improve from 0.51372
accuracy: 0.9630 - val_loss: 0.7213 - val_accuracy: 0.5137 - lr: 0.0010
Epoch 4/40
25/25 [============= ] - ETA: Os - loss: 0.0618 - accuracy:
0.9878
Epoch 4: val_accuracy did not improve from 0.51372
accuracy: 0.9878 - val_loss: 0.7389 - val_accuracy: 0.5137 - lr: 0.0010
25/25 [============== ] - ETA: Os - loss: 0.0322 - accuracy:
0.9968
Epoch 5: val_accuracy improved from 0.51372 to 0.51376, saving model to
models/gru_bi_a.h5
```

```
accuracy: 0.9968 - val_loss: 0.6925 - val_accuracy: 0.5138 - lr: 0.0010
Epoch 6/40
Epoch 6: val_accuracy improved from 0.51376 to 0.51380, saving model to
models/gru bi a.h5
accuracy: 0.9992 - val_loss: 0.7140 - val_accuracy: 0.5138 - lr: 0.0010
Epoch 7/40
25/25 [============= ] - ETA: Os - loss: 0.0119 - accuracy:
0.9998
Epoch 7: val_accuracy improved from 0.51380 to 0.52916, saving model to
models/gru_bi_a.h5
accuracy: 0.9998 - val_loss: 0.6385 - val_accuracy: 0.5292 - lr: 0.0010
Epoch 8/40
25/25 [============= ] - ETA: Os - loss: 0.0091 - accuracy:
0.9999
Epoch 8: val_accuracy improved from 0.52916 to 0.57464, saving model to
models/gru bi a.h5
25/25 [============= ] - 5s 186ms/step - loss: 0.0091 -
accuracy: 0.9999 - val_loss: 0.6117 - val_accuracy: 0.5746 - lr: 0.0010
Epoch 9/40
25/25 [============== ] - ETA: Os - loss: 0.0074 - accuracy:
1.0000
Epoch 9: val accuracy improved from 0.57464 to 0.62336, saving model to
models/gru_bi_a.h5
accuracy: 1.0000 - val_loss: 0.5964 - val_accuracy: 0.6234 - lr: 0.0010
Epoch 10/40
25/25 [============== ] - ETA: Os - loss: 0.0063 - accuracy:
Epoch 10: val accuracy improved from 0.62336 to 0.69020, saving model to
models/gru_bi_a.h5
accuracy: 1.0000 - val_loss: 0.5481 - val_accuracy: 0.6902 - lr: 0.0010
Epoch 11/40
25/25 [============== ] - ETA: Os - loss: 0.0054 - accuracy:
1.0000
Epoch 11: val_accuracy improved from 0.69020 to 0.75192, saving model to
models/gru_bi_a.h5
accuracy: 1.0000 - val_loss: 0.4864 - val_accuracy: 0.7519 - lr: 0.0010
Epoch 12/40
25/25 [============= ] - ETA: Os - loss: 0.0047 - accuracy:
Epoch 12: val_accuracy improved from 0.75192 to 0.78004, saving model to
models/gru_bi_a.h5
```

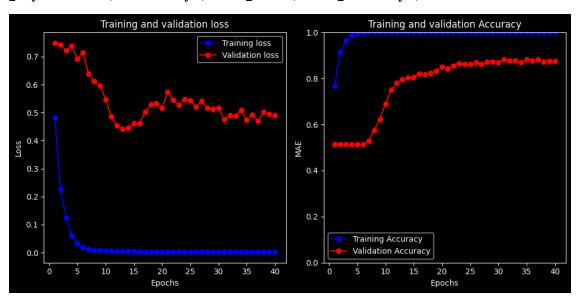
```
accuracy: 1.0000 - val_loss: 0.4546 - val_accuracy: 0.7800 - lr: 0.0010
Epoch 13/40
25/25 [============= ] - ETA: Os - loss: 0.0042 - accuracy:
1.0000
Epoch 13: val_accuracy improved from 0.78004 to 0.79536, saving model to
models/gru bi a.h5
accuracy: 1.0000 - val_loss: 0.4414 - val_accuracy: 0.7954 - lr: 0.0010
Epoch 14/40
1.0000
Epoch 14: val_accuracy improved from 0.79536 to 0.80176, saving model to
models/gru bi a.h5
25/25 [============ ] - 5s 185ms/step - loss: 0.0037 -
accuracy: 1.0000 - val_loss: 0.4457 - val_accuracy: 0.8018 - lr: 0.0010
Epoch 15/40
25/25 [============== ] - ETA: Os - loss: 0.0033 - accuracy:
1.0000
Epoch 15: val_accuracy improved from 0.80176 to 0.80612, saving model to
models/gru bi a.h5
accuracy: 1.0000 - val_loss: 0.4610 - val_accuracy: 0.8061 - lr: 0.0010
Epoch 16/40
Epoch 16: val_accuracy improved from 0.80612 to 0.81972, saving model to
models/gru_bi_a.h5
accuracy: 1.0000 - val_loss: 0.4609 - val_accuracy: 0.8197 - lr: 0.0010
Epoch 17/40
25/25 [============= ] - ETA: Os - loss: 0.0027 - accuracy:
1.0000
Epoch 17: val_accuracy did not improve from 0.81972
25/25 [============= ] - 4s 180ms/step - loss: 0.0027 -
accuracy: 1.0000 - val_loss: 0.5030 - val_accuracy: 0.8188 - lr: 0.0010
Epoch 18/40
25/25 [=============== ] - ETA: Os - loss: 0.0024 - accuracy:
1.0000
Epoch 18: ReduceLROnPlateau reducing learning rate to 0.00050000000237487257.
Epoch 18: val_accuracy improved from 0.81972 to 0.82316, saving model to
models/gru_bi_a.h5
accuracy: 1.0000 - val_loss: 0.5285 - val_accuracy: 0.8232 - lr: 0.0010
Epoch 19/40
25/25 [============== ] - ETA: Os - loss: 0.0023 - accuracy:
1.0000
```

```
Epoch 19: val_accuracy improved from 0.82316 to 0.83380, saving model to
models/gru_bi_a.h5
accuracy: 1.0000 - val_loss: 0.5339 - val_accuracy: 0.8338 - lr: 5.0000e-04
Epoch 20/40
25/25 [============= ] - ETA: Os - loss: 0.0022 - accuracy:
1.0000
Epoch 20: val_accuracy improved from 0.83380 to 0.85104, saving model to
models/gru bi a.h5
25/25 [=========== ] - 5s 185ms/step - loss: 0.0022 -
accuracy: 1.0000 - val_loss: 0.5162 - val_accuracy: 0.8510 - lr: 5.0000e-04
Epoch 21/40
25/25 [============= ] - ETA: Os - loss: 0.0021 - accuracy:
1.0000
Epoch 21: val_accuracy did not improve from 0.85104
accuracy: 1.0000 - val_loss: 0.5733 - val_accuracy: 0.8428 - lr: 5.0000e-04
Epoch 22/40
25/25 [============== ] - ETA: Os - loss: 0.0021 - accuracy:
1.0000
Epoch 22: val_accuracy improved from 0.85104 to 0.85516, saving model to
models/gru bi a.h5
25/25 [============= ] - 4s 182ms/step - loss: 0.0021 -
accuracy: 1.0000 - val_loss: 0.5446 - val_accuracy: 0.8552 - lr: 5.0000e-04
Epoch 23/40
1.0000
Epoch 23: ReduceLROnPlateau reducing learning rate to 0.0002500000118743628.
Epoch 23: val_accuracy improved from 0.85516 to 0.86480, saving model to
models/gru_bi_a.h5
accuracy: 1.0000 - val_loss: 0.5269 - val_accuracy: 0.8648 - lr: 5.0000e-04
Epoch 24/40
1.0000
Epoch 24: val accuracy did not improve from 0.86480
accuracy: 1.0000 - val_loss: 0.5471 - val_accuracy: 0.8612 - lr: 2.5000e-04
Epoch 25/40
25/25 [============== ] - ETA: Os - loss: 0.0018 - accuracy:
1.0000
Epoch 25: val_accuracy did not improve from 0.86480
accuracy: 1.0000 - val_loss: 0.5435 - val_accuracy: 0.8635 - lr: 2.5000e-04
Epoch 26/40
25/25 [============== ] - ETA: Os - loss: 0.0018 - accuracy:
1.0000
```

```
Epoch 26: val accuracy improved from 0.86480 to 0.87060, saving model to
models/gru_bi_a.h5
accuracy: 1.0000 - val_loss: 0.5213 - val_accuracy: 0.8706 - lr: 2.5000e-04
Epoch 27/40
25/25 [============== ] - ETA: Os - loss: 0.0018 - accuracy:
1.0000
Epoch 27: val_accuracy did not improve from 0.87060
accuracy: 1.0000 - val_loss: 0.5411 - val_accuracy: 0.8624 - lr: 2.5000e-04
Epoch 28/40
25/25 [============== ] - ETA: Os - loss: 0.0017 - accuracy:
1.0000
Epoch 28: ReduceLROnPlateau reducing learning rate to 0.0001250000059371814.
Epoch 28: val_accuracy improved from 0.87060 to 0.87112, saving model to
models/gru_bi_a.h5
accuracy: 1.0000 - val_loss: 0.5176 - val_accuracy: 0.8711 - lr: 2.5000e-04
Epoch 29/40
25/25 [============= ] - ETA: Os - loss: 0.0017 - accuracy:
1.0000
Epoch 29: val_accuracy did not improve from 0.87112
25/25 [============= ] - 4s 180ms/step - loss: 0.0017 -
accuracy: 1.0000 - val_loss: 0.5133 - val_accuracy: 0.8711 - lr: 1.2500e-04
Epoch 30/40
1.0000
Epoch 30: val_accuracy did not improve from 0.87112
25/25 [============ ] - 4s 180ms/step - loss: 0.0017 -
accuracy: 1.0000 - val_loss: 0.5174 - val_accuracy: 0.8686 - lr: 1.2500e-04
Epoch 31/40
25/25 [============== ] - ETA: Os - loss: 0.0016 - accuracy:
1.0000
Epoch 31: val_accuracy improved from 0.87112 to 0.88192, saving model to
models/gru bi a.h5
accuracy: 1.0000 - val_loss: 0.4752 - val_accuracy: 0.8819 - lr: 1.2500e-04
Epoch 32/40
25/25 [============== ] - ETA: Os - loss: 0.0017 - accuracy:
1.0000
Epoch 32: val_accuracy did not improve from 0.88192
25/25 [============ ] - 4s 180ms/step - loss: 0.0017 -
accuracy: 1.0000 - val_loss: 0.4896 - val_accuracy: 0.8772 - lr: 1.2500e-04
Epoch 33/40
25/25 [============== ] - ETA: Os - loss: 0.0016 - accuracy:
1.0000
Epoch 33: ReduceLROnPlateau reducing learning rate to 6.25000029685907e-05.
```

```
Epoch 33: val_accuracy did not improve from 0.88192
accuracy: 1.0000 - val_loss: 0.4892 - val_accuracy: 0.8770 - lr: 1.2500e-04
Epoch 34/40
25/25 [============== ] - ETA: Os - loss: 0.0016 - accuracy:
Epoch 34: val_accuracy did not improve from 0.88192
accuracy: 1.0000 - val_loss: 0.5079 - val_accuracy: 0.8707 - lr: 6.2500e-05
Epoch 35/40
25/25 [============== ] - ETA: Os - loss: 0.0016 - accuracy:
1.0000
Epoch 35: val_accuracy did not improve from 0.88192
25/25 [============ ] - 4s 181ms/step - loss: 0.0016 -
accuracy: 1.0000 - val_loss: 0.4737 - val_accuracy: 0.8817 - lr: 6.2500e-05
Epoch 36/40
1.0000
Epoch 36: val accuracy did not improve from 0.88192
25/25 [============= ] - 4s 181ms/step - loss: 0.0016 -
accuracy: 1.0000 - val_loss: 0.4926 - val_accuracy: 0.8760 - lr: 6.2500e-05
Epoch 37/40
25/25 [============= ] - ETA: Os - loss: 0.0016 - accuracy:
1.0000
Epoch 37: val accuracy improved from 0.88192 to 0.88200, saving model to
models/gru_bi_a.h5
accuracy: 1.0000 - val_loss: 0.4709 - val_accuracy: 0.8820 - lr: 6.2500e-05
Epoch 38/40
25/25 [============== ] - ETA: Os - loss: 0.0016 - accuracy:
Epoch 38: ReduceLROnPlateau reducing learning rate to 3.125000148429535e-05.
Epoch 38: val accuracy did not improve from 0.88200
accuracy: 1.0000 - val_loss: 0.5029 - val_accuracy: 0.8721 - lr: 6.2500e-05
Epoch 39/40
25/25 [============== ] - ETA: Os - loss: 0.0015 - accuracy:
1.0000
Epoch 39: val_accuracy did not improve from 0.88200
accuracy: 1.0000 - val_loss: 0.4937 - val_accuracy: 0.8746 - lr: 3.1250e-05
Epoch 40/40
1.0000
Epoch 40: val_accuracy did not improve from 0.88200
```

accuracy: 1.0000 - val_loss: 0.4907 - val_accuracy: 0.8752 - lr: 3.1250e-05
dict_keys(['loss', 'accuracy', 'val_loss', 'val_accuracy', 'lr'])



Elapsed time: 191.2750096321106

```
[]: model = keras.models.load_model(dir + "gru_bi_a.h5")
results = model.evaluate(x_test_int, y_test)
print(results)
```

Pasando a los modelos en si mismos debemos distinguir tres etapas de nuestro pipeline:

El Embedding que trata de obtener la mejor representación para cada token en el espcaio vectorial dejinido a través de los parámetros propuestos, concretamente sera un espacio de la forma longitud_de_secuencia x embbeding_size. De este capa obtendremos un vector para cada token introduci en la entrada del modelo, debemos utilizar un tamaño de embbeding que sea duficientemente grande como para representar la varibilidad de nuestro corpus. Si el tamño del embbeding es demasiado pequeño aunque intentemos aumentar la complejidad del resto de nuestro modelo no seremos capaces de mjorar su comportamiento ya que estamos perdiendo demasiada infromación en este paso.

Con la salida del embedding debemos aplicar la capa recurrente, aqui existen tres cuestiones de especial impacto en nuestro problema. La primera cuestión es el tipo de unidades recurrentes, debemos tener en cuenta que a diferencia de otros paradigmas de deep Learning como las convoluciones, las unidades recurrentes no son equivalentes a un conjunto de caps densas apiladas, además las puertas que implementan muestra comportamientos distintos que serán mas o menos adecuados a nuestro problema. En este caso, al tratarse de texto es de esperar que el mejor comportamiento los muestren las LSTM o GRU, claramente mas compleajs que las SimpleRNN. El

segunodo aspecto a tener en cuenta es el número de unidades que se introducen, que representan los pasos que se están empleando al procesar el texto, un número muy bajo será insuficiente para captar los distintos bloques con significado dentro del fragmento de texto procesad y un número muy elevado no nos aprotará una mejora en el rendimiento y puede suponer un obstáculo en el proceso de entrenamiento. Este está muy condicinodo por el tamño de secuencia que estamos empleando y en definitiva el tamaño de fragmentos de textos que estamos empleando, dado que necesitaremos más etapas si nuestro fragmento es de mayor longitud. En nuestro caso observamos un comportamiento razonablemente bueno desde valores bajos, como 6, dado que nuestros fragmentos son pequeños y observamos lijeras mejoras hasta tamaños de 16 o 32. Esto realizando un diseño contenido y razonable sobre nuestro problema, ya que no debemos olvidar que otros aspectos como el tamaño de embedding o el número de caracteristictics juegan roles importantes y pueden contribuir a cambiar el comportamiento de esta etapa. Además al tratarse de un problema relativamente pequeño podriamos observar un comportamiento relativamente bueno al añadir más complejidad en otros elementos de nuestro modelo, sin que estas decisiones sean justificadas con una adecuada compresnión del modelo.

Del mismo modo, destacamos la aleatoriedad caracteristica del proceso de entrenamiento. Hemos visto como en alguna ocasion el modelo inciaial o tras una única época muestra un comportamiento extraordinariamente bueno sin una justificación a priori y pequeños cambios en la semilla hacen que no sea repetible un comportamiento global similar. A nuestro juicio, debemos intentar enteder porque surgen estos casos pero no deberiamos tomarlos como conclusiones válidas al ser puramente anecdóticos. En este sentido, conviene destacar que sobre nuestro problema no hemos percibido una mejora relevante al introducir varias capas recurrentes apiladas.

Por último, la tercera etapa de este tipo de modelos son la capa des salida y capas densas. Hemos visto como

4 Modelos pequeños

Model: "model_23"

Layer (type) ====================================	Output Shape			
<pre>input_24 (InputLayer)</pre>		0		
embedding_23 (Embedding)	(None, 166, 4)	22000		
<pre>batch_normalization_55 (Bat chNormalization)</pre>	(None, 166, 4)	16		
gru_19 (GRU)	(None, 8)	336		
batch_normalization_56 (Bat chNormalization)	(None, 8)	32		
dense_36 (Dense)	(None, 1)	9		
Epoch 1/80				
25/25 [====================================	======] - ETA: Os - los	s: 0.7437 - a		
0.5478 Epoch 1: val_accuracy improve models/gru_bi_r.h5 25/25 [====================================	======] - 4s 100ms/step	- loss: 0.7		
Epoch 2/80 25/25 [====================================	•			
Epoch 2: val_accuracy improved from 0.51372 to 0.51424, saving model to models/gru_bi_r.h5 25/25 [====================================				
25/25 [====================================	======] - 2s 61ms/step	- ross: 0.61		

```
0.6630 - val_loss: 0.6898 - val_accuracy: 0.5142 - lr: 0.0010
Epoch 3/80
Epoch 3: val accuracy improved from 0.51424 to 0.65596, saving model to
models/gru bi r.h5
0.7220 - val_loss: 0.6882 - val_accuracy: 0.6560 - lr: 0.0010
Epoch 4/80
25/25 [============= ] - ETA: Os - loss: 0.5009 - accuracy:
0.7644
Epoch 4: val_accuracy did not improve from 0.65596
0.7644 - val_loss: 0.6884 - val_accuracy: 0.4889 - lr: 0.0010
Epoch 5/80
25/25 [============== ] - ETA: Os - loss: 0.4544 - accuracy:
0.7996
Epoch 5: val_accuracy did not improve from 0.65596
0.7996 - val_loss: 0.7003 - val_accuracy: 0.4863 - lr: 0.0010
Epoch 6/80
Epoch 6: val_accuracy did not improve from 0.65596
0.8234 - val_loss: 0.7313 - val_accuracy: 0.4863 - lr: 0.0010
Epoch 7/80
25/25 [============= ] - ETA: Os - loss: 0.3636 - accuracy:
0.8542
Epoch 7: val_accuracy did not improve from 0.65596
0.8542 - val_loss: 0.7636 - val_accuracy: 0.4863 - lr: 0.0010
Epoch 8/80
0.8894
Epoch 8: ReduceLROnPlateau reducing learning rate to 0.0005000000237487257.
Epoch 8: val_accuracy did not improve from 0.65596
0.8894 - val_loss: 0.7737 - val_accuracy: 0.4863 - lr: 0.0010
Epoch 9/80
25/25 [============== ] - ETA: Os - loss: 0.2543 - accuracy:
0.9134
Epoch 9: val_accuracy did not improve from 0.65596
0.9134 - val_loss: 0.7136 - val_accuracy: 0.4892 - lr: 5.0000e-04
Epoch 10/80
25/25 [============= ] - ETA: Os - loss: 0.2302 - accuracy:
```

```
0.9235
Epoch 10: val_accuracy improved from 0.65596 to 0.77124, saving model to
models/gru_bi_r.h5
0.9235 - val_loss: 0.5495 - val_accuracy: 0.7712 - lr: 5.0000e-04
Epoch 11/80
0.9304
Epoch 11: val_accuracy did not improve from 0.77124
0.9304 - val_loss: 0.5547 - val_accuracy: 0.7294 - lr: 5.0000e-04
Epoch 12/80
25/25 [============= ] - ETA: Os - loss: 0.1924 - accuracy:
0.9366
Epoch 12: val_accuracy did not improve from 0.77124
0.9366 - val_loss: 0.7742 - val_accuracy: 0.6217 - lr: 5.0000e-04
Epoch 13/80
0.9418
Epoch 13: val_accuracy did not improve from 0.77124
0.9418 - val_loss: 0.8485 - val_accuracy: 0.6242 - lr: 5.0000e-04
Epoch 14/80
Epoch 14: val_accuracy did not improve from 0.77124
0.9478 - val_loss: 1.0109 - val_accuracy: 0.6024 - lr: 5.0000e-04
Epoch 15/80
Epoch 15: ReduceLROnPlateau reducing learning rate to 0.0002500000118743628.
Epoch 15: val accuracy did not improve from 0.77124
0.9518 - val_loss: 1.1097 - val_accuracy: 0.5985 - lr: 5.0000e-04
Epoch 16/80
25/25 [============== ] - ETA: Os - loss: 0.1398 - accuracy:
0.9583
Epoch 16: val_accuracy did not improve from 0.77124
0.9583 - val_loss: 1.2156 - val_accuracy: 0.5898 - lr: 2.5000e-04
Epoch 17/80
Epoch 17: val_accuracy did not improve from 0.77124
```

```
0.9607 - val_loss: 1.2943 - val_accuracy: 0.5867 - lr: 2.5000e-04
Epoch 18/80
Epoch 18: val accuracy did not improve from 0.77124
0.9633 - val_loss: 1.2655 - val_accuracy: 0.6000 - lr: 2.5000e-04
Epoch 19/80
25/25 [============= ] - ETA: Os - loss: 0.1267 - accuracy:
0.9621
Epoch 19: val_accuracy did not improve from 0.77124
0.9621 - val_loss: 1.2011 - val_accuracy: 0.6211 - lr: 2.5000e-04
Epoch 20/80
0.9656
Epoch 20: ReduceLROnPlateau reducing learning rate to 0.0001250000059371814.
Epoch 20: val_accuracy did not improve from 0.77124
0.9656 - val_loss: 1.1686 - val_accuracy: 0.6360 - lr: 2.5000e-04
Epoch 21/80
25/25 [=============== ] - ETA: Os - loss: 0.1151 - accuracy:
0.9680
Epoch 21: val_accuracy did not improve from 0.77124
0.9680 - val_loss: 1.1637 - val_accuracy: 0.6428 - lr: 1.2500e-04
Epoch 22/80
25/25 [============== ] - ETA: Os - loss: 0.1118 - accuracy:
0.9694
Epoch 22: val_accuracy did not improve from 0.77124
0.9694 - val_loss: 1.0419 - val_accuracy: 0.6744 - lr: 1.2500e-04
Epoch 23/80
25/25 [============= ] - ETA: Os - loss: 0.1092 - accuracy:
0.9706
Epoch 23: val accuracy did not improve from 0.77124
0.9706 - val_loss: 0.9522 - val_accuracy: 0.6999 - lr: 1.2500e-04
Epoch 24/80
25/25 [============== ] - ETA: Os - loss: 0.1068 - accuracy:
0.9712
Epoch 24: val_accuracy did not improve from 0.77124
0.9712 - val_loss: 0.8408 - val_accuracy: 0.7299 - lr: 1.2500e-04
Epoch 25/80
25/25 [============== ] - ETA: Os - loss: 0.1046 - accuracy:
0.9724
```

```
Epoch 25: ReduceLROnPlateau reducing learning rate to 6.25000029685907e-05.
Epoch 25: val_accuracy did not improve from 0.77124
0.9724 - val_loss: 0.7966 - val_accuracy: 0.7433 - lr: 1.2500e-04
Epoch 26/80
0.9730
Epoch 26: val_accuracy improved from 0.77124 to 0.77592, saving model to
models/gru_bi_r.h5
0.9730 - val_loss: 0.6747 - val_accuracy: 0.7759 - 1r: 6.2500e-05
Epoch 27/80
0.9730
Epoch 27: val_accuracy improved from 0.77592 to 0.79716, saving model to
models/gru_bi_r.h5
0.9730 - val_loss: 0.6005 - val_accuracy: 0.7972 - lr: 6.2500e-05
Epoch 28/80
0.9737
Epoch 28: val_accuracy improved from 0.79716 to 0.81768, saving model to
models/gru_bi_r.h5
0.9737 - val_loss: 0.5266 - val_accuracy: 0.8177 - lr: 6.2500e-05
Epoch 29/80
25/25 [============= ] - ETA: Os - loss: 0.0988 - accuracy:
0.9746
Epoch 29: val_accuracy improved from 0.81768 to 0.83040, saving model to
models/gru_bi_r.h5
0.9746 - val_loss: 0.4806 - val_accuracy: 0.8304 - 1r: 6.2500e-05
Epoch 30/80
0.9755
Epoch 30: val_accuracy improved from 0.83040 to 0.83868, saving model to
models/gru_bi_r.h5
0.9755 - val_loss: 0.4541 - val_accuracy: 0.8387 - lr: 6.2500e-05
Epoch 31/80
25/25 [============== ] - ETA: Os - loss: 0.0962 - accuracy:
Epoch 31: val_accuracy improved from 0.83868 to 0.84976, saving model to
models/gru_bi_r.h5
0.9750 - val_loss: 0.4233 - val_accuracy: 0.8498 - 1r: 6.2500e-05
```

Epoch 32/80

```
25/25 [============== ] - ETA: Os - loss: 0.0963 - accuracy:
0.9749
Epoch 32: val_accuracy improved from 0.84976 to 0.85336, saving model to
models/gru bi r.h5
0.9749 - val_loss: 0.4108 - val_accuracy: 0.8534 - lr: 6.2500e-05
Epoch 33/80
0.9772
Epoch 33: val_accuracy improved from 0.85336 to 0.85424, saving model to
models/gru_bi_r.h5
0.9766 - val_loss: 0.4060 - val_accuracy: 0.8542 - 1r: 6.2500e-05
Epoch 34/80
0.9763
Epoch 34: val_accuracy improved from 0.85424 to 0.85588, saving model to
models/gru_bi_r.h5
0.9763 - val_loss: 0.3987 - val_accuracy: 0.8559 - lr: 6.2500e-05
Epoch 35/80
Epoch 35: val_accuracy improved from 0.85588 to 0.85712, saving model to
models/gru_bi_r.h5
0.9766 - val_loss: 0.3934 - val_accuracy: 0.8571 - lr: 6.2500e-05
Epoch 36/80
25/25 [============== ] - ETA: Os - loss: 0.0916 - accuracy:
0.9772
Epoch 36: val_accuracy improved from 0.85712 to 0.85852, saving model to
models/gru_bi_r.h5
0.9772 - val_loss: 0.3914 - val_accuracy: 0.8585 - lr: 6.2500e-05
Epoch 37/80
0.9778
Epoch 37: val_accuracy did not improve from 0.85852
0.9778 - val_loss: 0.3940 - val_accuracy: 0.8580 - lr: 6.2500e-05
Epoch 38/80
25/25 [============== ] - ETA: Os - loss: 0.0905 - accuracy:
0.9771
Epoch 38: val_accuracy did not improve from 0.85852
0.9771 - val_loss: 0.3934 - val_accuracy: 0.8581 - lr: 6.2500e-05
Epoch 39/80
25/25 [============= ] - ETA: Os - loss: 0.0887 - accuracy:
```

```
0.9779
Epoch 39: val_accuracy improved from 0.85852 to 0.85948, saving model to
models/gru_bi_r.h5
0.9779 - val_loss: 0.3918 - val_accuracy: 0.8595 - lr: 6.2500e-05
Epoch 40/80
0.9784
Epoch 40: val_accuracy did not improve from 0.85948
0.9782 - val_loss: 0.3946 - val_accuracy: 0.8590 - lr: 6.2500e-05
Epoch 41/80
25/25 [============= ] - ETA: Os - loss: 0.0869 - accuracy:
0.9787
Epoch 41: ReduceLROnPlateau reducing learning rate to 3.125000148429535e-05.
Epoch 41: val_accuracy did not improve from 0.85948
0.9787 - val_loss: 0.3945 - val_accuracy: 0.8588 - lr: 6.2500e-05
Epoch 42/80
25/25 [============== ] - ETA: Os - loss: 0.0862 - accuracy:
0.9791
Epoch 42: val_accuracy did not improve from 0.85948
0.9791 - val_loss: 0.3959 - val_accuracy: 0.8586 - lr: 3.1250e-05
Epoch 43/80
0.9788
Epoch 43: val_accuracy did not improve from 0.85948
0.9788 - val_loss: 0.3970 - val_accuracy: 0.8583 - lr: 3.1250e-05
Epoch 44/80
25/25 [============== ] - ETA: Os - loss: 0.0856 - accuracy:
0.9786
Epoch 44: val accuracy did not improve from 0.85948
25/25 [============== ] - Os 18ms/step - loss: 0.0856 - accuracy:
0.9786 - val_loss: 0.3982 - val_accuracy: 0.8581 - lr: 3.1250e-05
Epoch 45/80
25/25 [============== ] - ETA: Os - loss: 0.0846 - accuracy:
0.9789
Epoch 45: val_accuracy did not improve from 0.85948
0.9789 - val_loss: 0.3981 - val_accuracy: 0.8579 - lr: 3.1250e-05
Epoch 46/80
25/25 [============= ] - ETA: Os - loss: 0.0842 - accuracy:
Epoch 46: ReduceLROnPlateau reducing learning rate to 1.5625000742147677e-05.
```

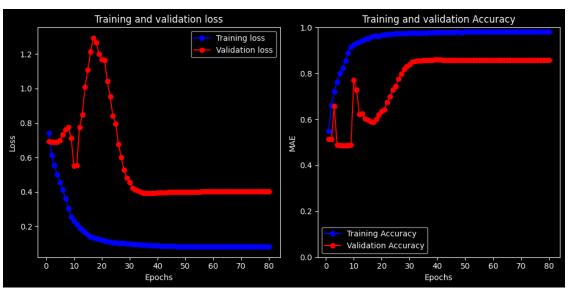
```
Epoch 46: val_accuracy did not improve from 0.85948
0.9797 - val_loss: 0.3992 - val_accuracy: 0.8580 - lr: 3.1250e-05
Epoch 47/80
25/25 [============== ] - ETA: Os - loss: 0.0836 - accuracy:
0.9800
Epoch 47: val accuracy did not improve from 0.85948
0.9800 - val_loss: 0.3990 - val_accuracy: 0.8579 - lr: 1.5625e-05
Epoch 48/80
Epoch 48: val_accuracy did not improve from 0.85948
25/25 [============ ] - Os 18ms/step - loss: 0.0838 - accuracy:
0.9794 - val_loss: 0.3996 - val_accuracy: 0.8577 - lr: 1.5625e-05
Epoch 49/80
Epoch 49: val_accuracy did not improve from 0.85948
0.9795 - val_loss: 0.3986 - val_accuracy: 0.8580 - lr: 1.5625e-05
Epoch 50/80
25/25 [=============== ] - ETA: Os - loss: 0.0829 - accuracy:
0.9801
Epoch 50: val_accuracy did not improve from 0.85948
0.9801 - val_loss: 0.3999 - val_accuracy: 0.8575 - lr: 1.5625e-05
Epoch 51/80
0.9802
Epoch 51: ReduceLROnPlateau reducing learning rate to 7.812500371073838e-06.
Epoch 51: val_accuracy did not improve from 0.85948
0.9801 - val loss: 0.4004 - val accuracy: 0.8578 - lr: 1.5625e-05
Epoch 52/80
0.9804
Epoch 52: val_accuracy did not improve from 0.85948
0.9800 - val_loss: 0.4002 - val_accuracy: 0.8577 - lr: 7.8125e-06
Epoch 53/80
0.9804
Epoch 53: val_accuracy did not improve from 0.85948
0.9804 - val_loss: 0.4004 - val_accuracy: 0.8577 - lr: 7.8125e-06
Epoch 54/80
```

```
0.9808
Epoch 54: val_accuracy did not improve from 0.85948
0.9802 - val_loss: 0.4006 - val_accuracy: 0.8577 - lr: 7.8125e-06
Epoch 55/80
0.9800
Epoch 55: val_accuracy did not improve from 0.85948
0.9800 - val_loss: 0.4006 - val_accuracy: 0.8579 - lr: 7.8125e-06
Epoch 56/80
0.9804
Epoch 56: ReduceLROnPlateau reducing learning rate to 3.906250185536919e-06.
Epoch 56: val_accuracy did not improve from 0.85948
0.9802 - val_loss: 0.4010 - val_accuracy: 0.8577 - lr: 7.8125e-06
Epoch 57/80
0.9806
Epoch 57: val_accuracy did not improve from 0.85948
0.9804 - val_loss: 0.4011 - val_accuracy: 0.8577 - lr: 3.9063e-06
Epoch 58/80
25/25 [============= ] - ETA: Os - loss: 0.0819 - accuracy:
0.9802
Epoch 58: val_accuracy did not improve from 0.85948
0.9802 - val_loss: 0.4011 - val_accuracy: 0.8578 - lr: 3.9063e-06
Epoch 59/80
25/25 [============== ] - ETA: Os - loss: 0.0819 - accuracy:
0.9806
Epoch 59: val accuracy did not improve from 0.85948
0.9806 - val_loss: 0.4014 - val_accuracy: 0.8576 - lr: 3.9063e-06
Epoch 60/80
0.9808
Epoch 60: val_accuracy did not improve from 0.85948
0.9808 - val_loss: 0.4013 - val_accuracy: 0.8577 - lr: 3.9063e-06
Epoch 61/80
Epoch 61: ReduceLROnPlateau reducing learning rate to 1.9531250927684596e-06.
```

```
Epoch 61: val_accuracy did not improve from 0.85948
0.9808 - val_loss: 0.4014 - val_accuracy: 0.8577 - lr: 3.9063e-06
Epoch 62/80
0.9802
Epoch 62: val accuracy did not improve from 0.85948
25/25 [============= ] - Os 12ms/step - loss: 0.0814 - accuracy:
0.9802 - val_loss: 0.4014 - val_accuracy: 0.8578 - lr: 1.9531e-06
Epoch 63/80
0.9804
Epoch 63: val_accuracy did not improve from 0.85948
0.9804 - val_loss: 0.4015 - val_accuracy: 0.8578 - lr: 1.9531e-06
Epoch 64/80
0.9808
Epoch 64: val_accuracy did not improve from 0.85948
0.9804 - val_loss: 0.4015 - val_accuracy: 0.8578 - lr: 1.9531e-06
Epoch 65/80
25/25 [=============== ] - ETA: Os - loss: 0.0821 - accuracy:
0.9800
Epoch 65: val_accuracy did not improve from 0.85948
0.9800 - val_loss: 0.4015 - val_accuracy: 0.8578 - lr: 1.9531e-06
Epoch 66/80
0.9808
Epoch 66: ReduceLROnPlateau reducing learning rate to 9.765625463842298e-07.
Epoch 66: val_accuracy did not improve from 0.85948
0.9807 - val loss: 0.4015 - val accuracy: 0.8578 - lr: 1.9531e-06
Epoch 67/80
0.9802
Epoch 67: val_accuracy did not improve from 0.85948
0.9802 - val_loss: 0.4015 - val_accuracy: 0.8578 - lr: 9.7656e-07
Epoch 68/80
25/25 [============= ] - ETA: Os - loss: 0.0816 - accuracy:
0.9802
Epoch 68: val_accuracy did not improve from 0.85948
0.9802 - val_loss: 0.4015 - val_accuracy: 0.8578 - lr: 9.7656e-07
Epoch 69/80
```

```
25/25 [============== ] - ETA: Os - loss: 0.0816 - accuracy:
0.9804
Epoch 69: val_accuracy did not improve from 0.85948
0.9804 - val_loss: 0.4015 - val_accuracy: 0.8578 - lr: 9.7656e-07
Epoch 70/80
25/25 [============== ] - ETA: Os - loss: 0.0817 - accuracy:
0.9805
Epoch 70: val_accuracy did not improve from 0.85948
0.9805 - val_loss: 0.4017 - val_accuracy: 0.8578 - lr: 9.7656e-07
Epoch 71/80
0.9804
Epoch 71: ReduceLROnPlateau reducing learning rate to 4.882812731921149e-07.
Epoch 71: val_accuracy did not improve from 0.85948
0.9804 - val_loss: 0.4016 - val_accuracy: 0.8578 - lr: 9.7656e-07
Epoch 72/80
0.9805
Epoch 72: val_accuracy did not improve from 0.85948
25/25 [============= ] - Os 12ms/step - loss: 0.0814 - accuracy:
0.9807 - val_loss: 0.4015 - val_accuracy: 0.8576 - lr: 4.8828e-07
Epoch 73/80
25/25 [============= ] - ETA: Os - loss: 0.0826 - accuracy:
0.9803
Epoch 73: val_accuracy did not improve from 0.85948
25/25 [============= ] - Os 15ms/step - loss: 0.0826 - accuracy:
0.9803 - val_loss: 0.4015 - val_accuracy: 0.8576 - lr: 4.8828e-07
Epoch 74/80
25/25 [============== ] - ETA: Os - loss: 0.0816 - accuracy:
0.9805
Epoch 74: val accuracy did not improve from 0.85948
0.9805 - val_loss: 0.4016 - val_accuracy: 0.8576 - lr: 4.8828e-07
Epoch 75/80
25/25 [============== ] - ETA: Os - loss: 0.0814 - accuracy:
0.9804
Epoch 75: val_accuracy did not improve from 0.85948
0.9804 - val_loss: 0.4016 - val_accuracy: 0.8578 - lr: 4.8828e-07
Epoch 76/80
Epoch 76: ReduceLROnPlateau reducing learning rate to 2.4414063659605745e-07.
```

```
Epoch 76: val_accuracy did not improve from 0.85948
0.9804 - val_loss: 0.4017 - val_accuracy: 0.8578 - lr: 4.8828e-07
Epoch 77/80
0.9804
Epoch 77: val_accuracy did not improve from 0.85948
0.9807 - val_loss: 0.4017 - val_accuracy: 0.8578 - lr: 2.4414e-07
Epoch 78/80
0.9800
Epoch 78: val_accuracy did not improve from 0.85948
0.9804 - val_loss: 0.4017 - val_accuracy: 0.8578 - lr: 2.4414e-07
Epoch 79/80
25/25 [============= ] - ETA: Os - loss: 0.0812 - accuracy:
0.9806
Epoch 79: val_accuracy did not improve from 0.85948
0.9806 - val_loss: 0.4016 - val_accuracy: 0.8578 - lr: 2.4414e-07
Epoch 80/80
25/25 [=============== ] - ETA: Os - loss: 0.0814 - accuracy:
0.9804
Epoch 80: val_accuracy did not improve from 0.85948
0.9804 - val_loss: 0.4016 - val_accuracy: 0.8578 - lr: 2.4414e-07
dict_keys(['loss', 'accuracy', 'val_loss', 'val_accuracy', 'lr'])
```



Elapsed time: 43.26756191253662

5 El impacto de los hiperparámetros

Batch size muy muy grande

```
[]: from tensorflow.keras.callbacks import ReduceLROnPlateau
     np.random.seed(423423)
     tf.random.set_seed(1232413)
     reduce_lr = ReduceLROnPlateau(monitor='val_loss', factor=0.5, patience=5,_
      min_lr=0.000000000000000000000000001, verbose=1)
     start time = time.time()
     inputs = keras.Input(shape=(seqLength, ))
     x = layers.Embedding(maxFeatures, 64, mask_zero=True,_
     →input_length=seqLength)(inputs)
     x = keras.layers.BatchNormalization()(x)
     x = layers.Bidirectional(layers.GRU(256, return_sequences=False))(x)
     x = keras.layers.BatchNormalization()(x)
     x = keras.layers.Dense(256, activation="relu")(x)
     x = keras.layers.BatchNormalization()(x)
     x = keras.layers.Dense(256, activation="relu")(x)
     x = keras.layers.BatchNormalization()(x)
     outputs = layers.Dense(1, activation="sigmoid")(x)
     model = keras.Model(inputs, outputs)
     model.summary()
     optimizer = keras.optimizers.Adam(learning_rate=0.01)
     model.compile(optimizer=optimizer, loss="binary_crossentropy", __
      →metrics=["accuracy"])
     result = fitModel(model, x_train_int, y_train, (x_test_int, y_test),_
      onum_epochs=30, model_name="gru_bi_a.h5", batch_size=512, ___
      ⇔callbacks=[reduce_lr])
     end_time = time.time()
```

print("Elapsed time: ", end_time - start_time)

Model: "model_24"

Layer (type)	1 1	Param # ======			
input_25 (InputLayer)		0			
embedding_24 (Embedding)	(None, 166, 64)	352000			
<pre>batch_normalization_57 (Bat chNormalization)</pre>	(None, 166, 64)	256			
<pre>bidirectional_15 (Bidirectional)</pre>	(None, 512)	494592			
<pre>batch_normalization_58 (Bat chNormalization)</pre>	(None, 512)	2048			
dense_37 (Dense)	(None, 256)	131328			
<pre>batch_normalization_59 (Bat chNormalization)</pre>	(None, 256)	1024			
dense_38 (Dense)	(None, 256)	65792			
<pre>batch_normalization_60 (Bat chNormalization)</pre>	(None, 256)	1024			
dense_39 (Dense)	(None, 1)	257			
		========			
Total params: 1,048,321					
Trainable params: 1,046,145 Non-trainable params: 2,176					
Epoch 1/30 49/49 [====================================					
0.8219					
Epoch 1: val_accuracy improved from -inf to 0.57620, saving model to					
models/gru_bi_a.h5					
49/49 [====================================					
Epoch 2/30					
49/49 [====================================					
Epoch 2: val_accuracy improved from 0.57620 to 0.65072, saving model to					

```
models/gru_bi_a.h5
0.9312 - val_loss: 0.5930 - val_accuracy: 0.6507 - lr: 0.0100
Epoch 3/30
49/49 [=============== ] - ETA: Os - loss: 0.0985 - accuracy:
0.9643
Epoch 3: val_accuracy improved from 0.65072 to 0.79664, saving model to
models/gru_bi_a.h5
0.9643 - val_loss: 0.4437 - val_accuracy: 0.7966 - lr: 0.0100
Epoch 4/30
49/49 [============= ] - ETA: Os - loss: 0.0678 - accuracy:
0.9751
Epoch 4: val_accuracy improved from 0.79664 to 0.83320, saving model to
models/gru_bi_a.h5
0.9751 - val_loss: 0.4149 - val_accuracy: 0.8332 - lr: 0.0100
Epoch 5/30
49/49 [=================== ] - ETA: Os - loss: 0.0573 - accuracy:
0.9796
Epoch 5: val_accuracy did not improve from 0.83320
0.9796 - val_loss: 1.2609 - val_accuracy: 0.6974 - lr: 0.0100
Epoch 6/30
Epoch 6: val_accuracy improved from 0.83320 to 0.87512, saving model to
models/gru_bi_a.h5
0.9837 - val_loss: 0.4240 - val_accuracy: 0.8751 - lr: 0.0100
Epoch 7/30
49/49 [=============== ] - ETA: Os - loss: 0.0453 - accuracy:
0.9827
Epoch 7: val_accuracy did not improve from 0.87512
0.9827 - val_loss: 0.4793 - val_accuracy: 0.8743 - lr: 0.0100
Epoch 8/30
49/49 [=================== ] - ETA: Os - loss: 0.0390 - accuracy:
Epoch 8: val_accuracy improved from 0.87512 to 0.87832, saving model to
models/gru_bi_a.h5
0.9847 - val_loss: 0.5515 - val_accuracy: 0.8783 - lr: 0.0100
Epoch 9/30
Epoch 9: ReduceLROnPlateau reducing learning rate to 0.004999999888241291.
```

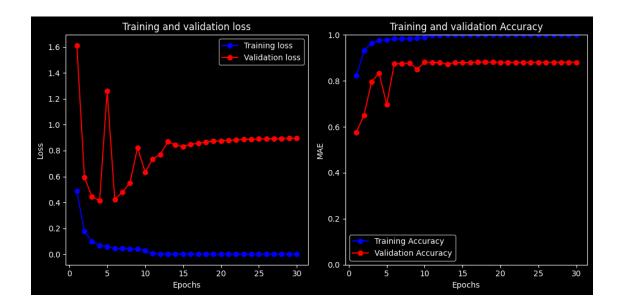
```
Epoch 9: val_accuracy did not improve from 0.87832
0.9856 - val_loss: 0.8223 - val_accuracy: 0.8510 - lr: 0.0100
Epoch 10/30
0.9896
Epoch 10: val_accuracy improved from 0.87832 to 0.88092, saving model to
models/gru_bi_a.h5
0.9896 - val_loss: 0.6342 - val_accuracy: 0.8809 - lr: 0.0050
Epoch 11/30
0.9982
Epoch 11: val_accuracy did not improve from 0.88092
0.9982 - val_loss: 0.7341 - val_accuracy: 0.8799 - lr: 0.0050
Epoch 12/30
0.9993
Epoch 12: val accuracy did not improve from 0.88092
0.9993 - val_loss: 0.7708 - val_accuracy: 0.8793 - lr: 0.0050
Epoch 13/30
49/49 [=============== ] - ETA: Os - loss: 0.0012 - accuracy:
0.9997
Epoch 13: val_accuracy did not improve from 0.88092
0.9997 - val_loss: 0.8693 - val_accuracy: 0.8720 - lr: 0.0050
Epoch 14/30
0.9996
Epoch 14: ReduceLROnPlateau reducing learning rate to 0.0024999999441206455.
Epoch 14: val_accuracy did not improve from 0.88092
0.9996 - val_loss: 0.8420 - val_accuracy: 0.8799 - lr: 0.0050
Epoch 15/30
0.9997
Epoch 15: val_accuracy did not improve from 0.88092
accuracy: 0.9997 - val_loss: 0.8318 - val_accuracy: 0.8787 - lr: 0.0025
0.9999
Epoch 16: val_accuracy did not improve from 0.88092
accuracy: 0.9999 - val_loss: 0.8478 - val_accuracy: 0.8798 - lr: 0.0025
```

```
Epoch 17/30
Epoch 17: val_accuracy improved from 0.88092 to 0.88148, saving model to
models/gru bi a.h5
accuracy: 1.0000 - val_loss: 0.8563 - val_accuracy: 0.8815 - lr: 0.0025
Epoch 18/30
1.0000
Epoch 18: val_accuracy did not improve from 0.88148
accuracy: 1.0000 - val_loss: 0.8650 - val_accuracy: 0.8810 - lr: 0.0025
Epoch 19/30
1.0000
Epoch 19: ReduceLROnPlateau reducing learning rate to 0.0012499999720603228.
Epoch 19: val_accuracy did not improve from 0.88148
accuracy: 1.0000 - val_loss: 0.8723 - val_accuracy: 0.8810 - lr: 0.0025
Epoch 20/30
1.0000
Epoch 20: val_accuracy did not improve from 0.88148
accuracy: 1.0000 - val_loss: 0.8756 - val_accuracy: 0.8808 - lr: 0.0012
Epoch 21/30
1.0000
Epoch 21: val_accuracy did not improve from 0.88148
accuracy: 1.0000 - val_loss: 0.8789 - val_accuracy: 0.8808 - lr: 0.0012
Epoch 22/30
1.0000
Epoch 22: val_accuracy did not improve from 0.88148
49/49 [============= ] - 1s 30ms/step - loss: 5.5817e-05 -
accuracy: 1.0000 - val_loss: 0.8821 - val_accuracy: 0.8807 - lr: 0.0012
Epoch 23/30
1.0000
Epoch 23: val_accuracy did not improve from 0.88148
accuracy: 1.0000 - val_loss: 0.8853 - val_accuracy: 0.8808 - lr: 0.0012
Epoch 24/30
1.0000
```

```
Epoch 24: ReduceLROnPlateau reducing learning rate to 0.0006249999860301614.
Epoch 24: val_accuracy did not improve from 0.88148
accuracy: 1.0000 - val loss: 0.8883 - val accuracy: 0.8806 - lr: 0.0012
Epoch 25/30
1.0000
Epoch 25: val_accuracy did not improve from 0.88148
accuracy: 1.0000 - val_loss: 0.8896 - val_accuracy: 0.8807 - lr: 6.2500e-04
Epoch 26/30
1.0000
Epoch 26: val_accuracy did not improve from 0.88148
accuracy: 1.0000 - val_loss: 0.8909 - val_accuracy: 0.8807 - lr: 6.2500e-04
Epoch 27/30
1.0000
Epoch 27: val_accuracy did not improve from 0.88148
accuracy: 1.0000 - val_loss: 0.8925 - val_accuracy: 0.8807 - lr: 6.2500e-04
Epoch 28/30
1.0000
Epoch 28: val_accuracy did not improve from 0.88148
49/49 [=========== ] - 1s 28ms/step - loss: 4.0223e-05 -
accuracy: 1.0000 - val_loss: 0.8939 - val_accuracy: 0.8807 - lr: 6.2500e-04
Epoch 29/30
Epoch 29: ReduceLROnPlateau reducing learning rate to 0.0003124999930150807.
Epoch 29: val accuracy did not improve from 0.88148
accuracy: 1.0000 - val loss: 0.8952 - val accuracy: 0.8809 - lr: 6.2500e-04
Epoch 30/30
```

dict_keys(['loss', 'accuracy', 'val_loss', 'val_accuracy', 'lr'])

Epoch 30: val_accuracy did not improve from 0.88148



Elapsed time: 57.49917149543762

```
[]: model = keras.models.load_model(dir + "lstm_bi_a.h5")
model.evaluate(x_test_int, y_test)
```

782/782 [============] - 4s 5ms/step - loss: 0.7453 - accuracy: 0.7994

[]: [0.7452947497367859, 0.7993999719619751]

```
[]: from tensorflow.keras.callbacks import ReduceLROnPlateau
    np.random.seed(423423)
    tf.random.set_seed(1232413)
    reduce_lr = ReduceLROnPlateau(monitor='val_loss', factor=0.5, patience=5,__
     start_time = time.time()
    inputs = keras.Input(shape=(seqLength, ))
    x = layers.Embedding(maxFeatures, 64, mask_zero=True,_
     →input_length=seqLength)(inputs)
    x = keras.layers.BatchNormalization()(x)
    x = layers.Bidirectional(layers.GRU(256, return_sequences=False))(x)
    x = keras.layers.BatchNormalization()(x)
    x = keras.layers.Dense(256, activation="relu")(x)
    x = keras.layers.BatchNormalization()(x)
    x = keras.layers.Dense(256, activation="relu")(x)
    x = keras.layers.BatchNormalization()(x)
```

Model: "model_25"

Layer (type)		Param #
input_26 (InputLayer)		0
embedding_25 (Embedding)	(None, 166, 64)	352000
<pre>batch_normalization_61 (Bat chNormalization)</pre>	(None, 166, 64)	256
<pre>bidirectional_16 (Bidirectional)</pre>	(None, 512)	494592
<pre>batch_normalization_62 (Bat chNormalization)</pre>	(None, 512)	2048
dense_40 (Dense)	(None, 256)	131328
<pre>batch_normalization_63 (Bat chNormalization)</pre>	(None, 256)	1024
dense_41 (Dense)	(None, 256)	65792
<pre>batch_normalization_64 (Bat chNormalization)</pre>	(None, 256)	1024
dense_42 (Dense)	(None, 1)	257

Total params: 1,048,321

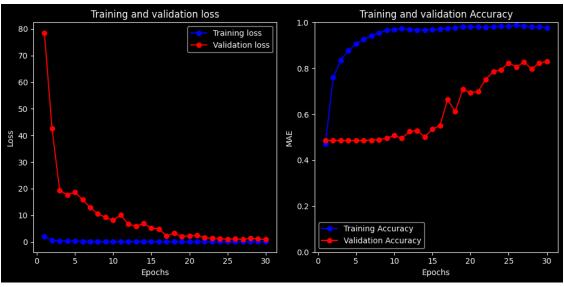
```
Non-trainable params: 2,176
          _____
Epoch 1/30
Epoch 1: val_accuracy improved from -inf to 0.48628, saving model to
models/gru_bi_a.h5
0.4715 - val_loss: 78.5632 - val_accuracy: 0.4863 - lr: 0.0400
Epoch 2/30
Epoch 2: val_accuracy did not improve from 0.48628
0.7608 - val_loss: 42.5607 - val_accuracy: 0.4863 - lr: 0.0400
Epoch 3/30
Epoch 3: val_accuracy did not improve from 0.48628
0.8348 - val_loss: 19.3976 - val_accuracy: 0.4863 - lr: 0.0400
Epoch 4/30
Epoch 4: val_accuracy did not improve from 0.48628
0.8761 - val_loss: 17.6954 - val_accuracy: 0.4863 - lr: 0.0400
Epoch 5/30
Epoch 5: val_accuracy improved from 0.48628 to 0.48644, saving model to
models/gru_bi_a.h5
0.9068 - val_loss: 18.6081 - val_accuracy: 0.4864 - lr: 0.0400
Epoch 6/30
Epoch 6: val accuracy improved from 0.48644 to 0.48648, saving model to
models/gru_bi_a.h5
0.9266 - val_loss: 15.8909 - val_accuracy: 0.4865 - lr: 0.0400
Epoch 7/30
Epoch 7: val_accuracy improved from 0.48648 to 0.48760, saving model to
models/gru_bi_a.h5
0.9419 - val_loss: 12.9067 - val_accuracy: 0.4876 - lr: 0.0400
Epoch 8: val_accuracy improved from 0.48760 to 0.49012, saving model to
models/gru bi a.h5
0.9548 - val_loss: 10.5438 - val_accuracy: 0.4901 - lr: 0.0400
```

Trainable params: 1,046,145

```
Epoch 9/30
7/7 [============ - ETA: Os - loss: 0.0964 - accuracy: 0.9653
Epoch 9: val accuracy improved from 0.49012 to 0.49584, saving model to
models/gru bi a.h5
0.9653 - val_loss: 9.1974 - val_accuracy: 0.4958 - lr: 0.0400
Epoch 10: val_accuracy improved from 0.49584 to 0.50780, saving model to
models/gru_bi_a.h5
0.9683 - val_loss: 8.1805 - val_accuracy: 0.5078 - lr: 0.0400
Epoch 11/30
Epoch 11: val_accuracy did not improve from 0.50780
0.9730 - val_loss: 10.0494 - val_accuracy: 0.4956 - lr: 0.0400
Epoch 12/30
7/7 [============ - ETA: Os - loss: 0.0802 - accuracy: 0.9694
Epoch 12: val accuracy improved from 0.50780 to 0.52568, saving model to
models/gru bi a.h5
0.9694 - val_loss: 6.6698 - val_accuracy: 0.5257 - lr: 0.0400
Epoch 13/30
Epoch 13: val accuracy improved from 0.52568 to 0.52952, saving model to
models/gru_bi_a.h5
0.9668 - val_loss: 5.8164 - val_accuracy: 0.5295 - lr: 0.0400
Epoch 14/30
7/7 [============ - ETA: Os - loss: 0.0884 - accuracy: 0.9662
Epoch 14: val_accuracy did not improve from 0.52952
0.9662 - val_loss: 6.9252 - val_accuracy: 0.5019 - lr: 0.0400
Epoch 15/30
Epoch 15: val_accuracy improved from 0.52952 to 0.53488, saving model to
models/gru bi a.h5
0.9689 - val_loss: 5.2468 - val_accuracy: 0.5349 - 1r: 0.0400
Epoch 16/30
Epoch 16: val_accuracy improved from 0.53488 to 0.55112, saving model to
models/gru_bi_a.h5
0.9712 - val_loss: 4.8231 - val_accuracy: 0.5511 - lr: 0.0400
Epoch 17/30
```

```
Epoch 17: val accuracy improved from 0.55112 to 0.66372, saving model to
models/gru_bi_a.h5
0.9742 - val_loss: 2.2540 - val_accuracy: 0.6637 - lr: 0.0400
Epoch 18/30
Epoch 18: val accuracy did not improve from 0.66372
0.9756 - val_loss: 3.2242 - val_accuracy: 0.6122 - lr: 0.0400
Epoch 19/30
Epoch 19: val_accuracy improved from 0.66372 to 0.70828, saving model to
models/gru_bi_a.h5
0.9817 - val_loss: 1.9893 - val_accuracy: 0.7083 - lr: 0.0400
Epoch 20/30
7/7 [============= - ETA: Os - loss: 0.0517 - accuracy: 0.9814
Epoch 20: val_accuracy did not improve from 0.70828
0.9814 - val_loss: 2.2629 - val_accuracy: 0.6935 - lr: 0.0400
Epoch 21/30
Epoch 21: val_accuracy did not improve from 0.70828
0.9810 - val_loss: 2.3855 - val_accuracy: 0.6995 - lr: 0.0400
Epoch 22/30
Epoch 22: val_accuracy improved from 0.70828 to 0.75224, saving model to
models/gru bi a.h5
0.9790 - val_loss: 1.5846 - val_accuracy: 0.7522 - lr: 0.0400
Epoch 23: val_accuracy improved from 0.75224 to 0.78652, saving model to
models/gru bi a.h5
0.9802 - val_loss: 1.3314 - val_accuracy: 0.7865 - lr: 0.0400
Epoch 24/30
Epoch 24: val_accuracy improved from 0.78652 to 0.79368, saving model to
models/gru_bi_a.h5
0.9834 - val_loss: 1.2335 - val_accuracy: 0.7937 - lr: 0.0400
Epoch 25/30
Epoch 25: val accuracy improved from 0.79368 to 0.82328, saving model to
models/gru_bi_a.h5
```

```
0.9829 - val_loss: 0.9464 - val_accuracy: 0.8233 - 1r: 0.0400
Epoch 26/30
Epoch 26: val_accuracy did not improve from 0.82328
0.9874 - val_loss: 1.1635 - val_accuracy: 0.8064 - lr: 0.0400
Epoch 27/30
Epoch 27: val_accuracy improved from 0.82328 to 0.82704, saving model to
models/gru bi a.h5
0.9841 - val_loss: 0.9156 - val_accuracy: 0.8270 - lr: 0.0400
Epoch 28/30
7/7 [============ ] - ETA: Os - loss: 0.0502 - accuracy: 0.9810
Epoch 28: val_accuracy did not improve from 0.82704
7/7 [========== - 1s 118ms/step - loss: 0.0502 - accuracy:
0.9810 - val_loss: 1.3372 - val_accuracy: 0.7977 - lr: 0.0400
Epoch 29/30
7/7 [============ - ETA: Os - loss: 0.0494 - accuracy: 0.9811
Epoch 29: val accuracy did not improve from 0.82704
7/7 [============ ] - 1s 123ms/step - loss: 0.0494 - accuracy:
0.9811 - val_loss: 1.0785 - val_accuracy: 0.8225 - lr: 0.0400
Epoch 30/30
7/7 [======
            ============== ] - ETA: Os - loss: 0.0582 - accuracy: 0.9773
Epoch 30: val_accuracy improved from 0.82704 to 0.83040, saving model to
models/gru_bi_a.h5
0.9773 - val_loss: 0.8671 - val_accuracy: 0.8304 - lr: 0.0400
dict_keys(['loss', 'accuracy', 'val_loss', 'val_accuracy', 'lr'])
```



Elapsed time: 30.094652891159058

6 Conclusiones

En general, lo primero que debemos tener en cuenta a la hora de proponer una solución a este problema es que estamos ante datos textuales por lo que el proceso de tokenización tiene un importante impacto en los resultados. En nuestro caso lo primero es el número de caracteristicas, que defini el dominio de las entradas del embedding y que tiene un importante impacto en el rendimiento pero bajo en la complejidad computacional, es mejor idea por exceso que por defecto. El segundo aspecto es la longitud de secuencai que deberemos ajustar al tamaño de los fragmentos de texto que queremos clasificar. en nuestro caso no es necesario que sea elevado y tiene un impacto importante en la complejidad computacional. Estos aspectos no sonn solo sensibles al problema sino tambien al modelo que vayamos a aplicar.

Respecto al diseño de la red es fundamental el empleo de una capa de Embedding y otra recurrente. En cuanto a la capa de embedding debemos tener especial cuidado con el tamaño que empleamos tine un importante imapcto en la complejidad computacional y en en los resultados al encargarse de representar la distacia semantica de de los tokens. Se debe cuidar la normalización, ua que esta capa tiene a sobreajustarse. Respecto a la etapa recurrente los mejores resultados se alcanzan con las unidades GRU de forma bidir3eccional, para este problema obtenemos resultados buenos con un número pequeño de 16 unidades, debemos ser cautos si deseamos redducirlo porque podriamos perder demasiada información contextual.

Respecto a las capas densas antes de la salida, hemos visto que no son necesaria para alcanzar el objetivo de este trabajo alcanzando un 85.6 % de accuracy de forma repetible. Para mejorar los resultados hemos visto que es necesario aumentar mucho la complejidad general de la red, aumentando tamañode embedding, unidades recurrentes, etc. Con esta estrategia hemos alcanzado un 88.6 % de accuracy. Se debe destacar que hemos pareciado una cierta tendencia al overfitting si aumentamos las caps densas y no las etapas anteriores.