

```
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```



```
import numpy as np

import json
import tensorflow as tf

from tensorflow.keras.preprocessing.text import Tokenizer
from tensorflow.keras.preprocessing.sequence import pad_sequences

!wget --no-check-certificate \
    https://storage.googleapis.com/laurencemoroney-blog.appspot.com/sarcasm.json \
    -O /tmp/sarcasm.json

vocab_size = 10000
embedding_dim = 16
max_length = 120
trunc_type='post'
padding_type='post'
oov_tok = "<OOV>"
training_size = 20000

with open("/tmp/sarcasm.json", 'r') as f:
    datastore = json.load(f)

sentences = []
labels = []
urls = []
for item in datastore:
    sentences.append(item['headline'])
    labels.append(item['is_sarcastic'])

training_sentences = sentences[0:training_size]
testing_sentences = sentences[training_size:]
training_labels = labels[0:training_size]
```

```
training_labels = labels[:training_size]
testing_labels = labels[training_size:]

tokenizer = Tokenizer(num_words=vocab_size, oov_token=oov_tok)
tokenizer.fit_on_texts(training_sentences)

word_index = tokenizer.word_index

training_sequences = tokenizer.texts_to_sequences(training_sentences)
training_padded = pad_sequences(training_sequences, maxlen=max_length, padding=padding_type,

testing_sequences = tokenizer.texts_to_sequences(testing_sentences)
testing_padded = pad_sequences(testing_sequences, maxlen=max_length, padding=padding_type, tr

model = tf.keras.Sequential([
    tf.keras.layers.Embedding(vocab_size, embedding_dim, input_length=max_length),
    tf.keras.layers.Conv1D(128, 5, activation='relu'),
    tf.keras.layers.GlobalMaxPooling1D(),
    tf.keras.layers.Dense(24, activation='relu'),
    tf.keras.layers.Dense(1, activation='sigmoid')
])
model.compile(loss='binary_crossentropy', optimizer='adam', metrics=['accuracy'])
model.summary()

num_epochs = 50

training_padded = np.array(training_padded)
training_labels = np.array(training_labels)
testing_padded = np.array(testing_padded)
testing_labels = np.array(testing_labels)

history = model.fit(training_padded, training_labels, epochs=num_epochs, validation_data=(tes
```



```
--2020-09-20 10:04:13-- https://storage.googleapis.com/laurencemoroney-blog.appspot.com
Resolving storage.googleapis.com (storage.googleapis.com)... 108.177.97.128, 108.177.128.10
Connecting to storage.googleapis.com (storage.googleapis.com)|108.177.97.128|:443... connected
HTTP request sent, awaiting response... 200 OK
Length: 5643545 (5.4M) [application/json]
Saving to: '/tmp/sarcasm.json'
```

```
/tmp/sarcasm.json 100%[=====>] 5.38M --.-KB/s in 0.07s
```

```
2020-09-20 10:04:14 (75.8 MB/s) - '/tmp/sarcasm.json' saved [5643545/5643545]
```

```
Model: "sequential"
```

Layer (type)	Output Shape	Param #
embedding (Embedding)	(None, 120, 16)	16000
conv1d (Conv1D)	(None, 116, 128)	10368
global_max_pooling1d (Global	(None, 128)	0
dense (Dense)	(None, 24)	3096
dense_1 (Dense)	(None, 1)	25

```
Total params: 29,489
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Trainable params: 29,489
```

```
Non-trainable params: 0
```

```
Epoch 1/50
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```
625/625 [=====] - 4s 6ms/step - loss: 0.4688 - accuracy: 0.7671
```

```
Epoch 2/50
```

```
625/625 [=====] - 4s 6ms/step - loss: 0.3568 - accuracy: 0.8401
```

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Epoch 3/50
```

```
625/625 [=====] - 4s 6ms/step - loss: 0.3155 - accuracy: 0.8619
```

```
Epoch 4/50
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```
625/625 [=====] - 4s 6ms/step - loss: 0.2782 - accuracy: 0.8808
```

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Epoch 5/50
```

```
625/625 [=====] - 4s 6ms/step - loss: 0.2428 - accuracy: 0.8971
```

```
Epoch 6/50
```

```
625/625 [=====] - 4s 6ms/step - loss: 0.2083 - accuracy: 0.9151
```

```
Epoch 7/50
```

```
625/625 [=====] - 4s 6ms/step - loss: 0.1764 - accuracy: 0.9319
```

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Epoch 8/50
```

```
625/625 [=====] - 4s 6ms/step - loss: 0.1470 - accuracy: 0.9431
```

```
Epoch 9/50
```

```
625/625 [=====] - 4s 6ms/step - loss: 0.1222 - accuracy: 0.9556
```

```
Epoch 10/50
```

```
625/625 [=====] - 4s 6ms/step - loss: 0.1028 - accuracy: 0.9629
```

```
Epoch 11/50
```

```
625/625 [=====] - 4s 6ms/step - loss: 0.0846 - accuracy: 0.9701
```

```
Epoch 12/50
```

```
625/625 [=====] - 4s 6ms/step - loss: 0.0730 - accuracy: 0.9741
```

```
Epoch 13/50
```

```
625/625 [=====] - 4s 6ms/step - loss: 0.0642 - accuracy: 0.9774
```

```
Epoch 14/50
```

```
625/625 [=====] - 4s 6ms/step - loss: 0.0555 - accuracy: 0.9816
```

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Epoch 15/50
625/625 [=====] - 4s 6ms/step - loss: 0.0485 - accuracy: 0.9825
Epoch 16/50
625/625 [=====] - 4s 6ms/step - loss: 0.0437 - accuracy: 0.9842
Epoch 17/50
625/625 [=====] - 4s 6ms/step - loss: 0.0436 - accuracy: 0.9836
Epoch 18/50
625/625 [=====] - 4s 6ms/step - loss: 0.0398 - accuracy: 0.9846
Epoch 19/50
625/625 [=====] - 4s 6ms/step - loss: 0.0414 - accuracy: 0.9837
Epoch 20/50
625/625 [=====] - 4s 6ms/step - loss: 0.0372 - accuracy: 0.9851
Epoch 21/50
625/625 [=====] - 4s 6ms/step - loss: 0.0343 - accuracy: 0.9857
Epoch 22/50
625/625 [=====] - 4s 6ms/step - loss: 0.0308 - accuracy: 0.9872
Epoch 23/50
625/625 [=====] - 4s 6ms/step - loss: 0.0317 - accuracy: 0.9867
Epoch 24/50
625/625 [=====] - 4s 6ms/step - loss: 0.0313 - accuracy: 0.9868
Epoch 25/50
625/625 [=====] - 4s 6ms/step - loss: 0.0313 - accuracy: 0.9861
Epoch 26/50
625/625 [=====] - 4s 6ms/step - loss: 0.0293 - accuracy: 0.9875
Epoch 27/50
625/625 [=====] - 4s 6ms/step - loss: 0.0307 - accuracy: 0.9865
Epoch 28/50
625/625 [=====] - 4s 6ms/step - loss: 0.0231 - accuracy: 0.9904
Epoch 29/50
625/625 [=====] - 4s 6ms/step - loss: 0.0254 - accuracy: 0.9896
Epoch 30/50
625/625 [=====] - 4s 6ms/step - loss: 0.0298 - accuracy: 0.9876
Epoch 31/50
625/625 [=====] - 4s 6ms/step - loss: 0.0231 - accuracy: 0.9897
Epoch 32/50
625/625 [=====] - 4s 6ms/step - loss: 0.0265 - accuracy: 0.9883
Epoch 33/50
625/625 [=====] - 4s 6ms/step - loss: 0.0274 - accuracy: 0.9886
Epoch 34/50
625/625 [=====] - 4s 6ms/step - loss: 0.0266 - accuracy: 0.9882
Epoch 35/50
625/625 [=====] - 4s 6ms/step - loss: 0.0272 - accuracy: 0.9886
Epoch 36/50
625/625 [=====] - 4s 6ms/step - loss: 0.0209 - accuracy: 0.9908
Epoch 37/50
625/625 [=====] - 4s 6ms/step - loss: 0.0176 - accuracy: 0.9918
Epoch 38/50
625/625 [=====] - 4s 6ms/step - loss: 0.0163 - accuracy: 0.9923
Epoch 39/50
625/625 [=====] - 4s 6ms/step - loss: 0.0271 - accuracy: 0.9888
Epoch 40/50
625/625 [=====] - 4s 6ms/step - loss: 0.0334 - accuracy: 0.9872
Epoch 41/50

```

```
import matplotlib.pyplot as plt
```

```
def plot_graphs(history, string):
```