

imdb reviews classification with BERT

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Loading R packages

```
library(uum1)
```

Python packages

```
import tensorflow as tf
from tensorflow import keras
from tensorflow.keras.datasets import imdb
from tensorflow.keras import preprocessing
from tensorflow.keras.models import Sequential
from keras import layers
from keras import models
import tensorflow_hub as hub
import tensorflow_text as text
from tensorflow.keras import Model
from tensorflow.keras.layers import Flatten, Dense, Dropout, Embedding
from tensorflow.keras.layers import BatchNormalization
import matplotlib.pyplot as plt

# Number of words to consider as features
max_features = 10000
# First 30 words(among the most frequent 10,000) would be considered of a single review
maxlen = 30
(x_train, y_train), (x_test, y_test) = imdb.load_data(num_words=max_features)
x_train = preprocessing.sequence.pad_sequences(x_train, maxlen=maxlen)
x_test = preprocessing.sequence.pad_sequences(x_test, maxlen=maxlen)
```

1 Implementing an BERT Model

Needed Packages

```
# Bert prepossessing layer
bert_preprocess = hub.KerasLayer("https://tfhub.dev/tensorflow/bert_en_uncased_preprocess/3")
```

```

# Corresponding encoding layer
bert_encoder = hub.KerasLayer("https://tfhub.dev/tensorflow/bert_en_uncased_L-12_H-768_A-12/4")

def decode_imdb_reviews(text_data):
    result = [0 for x in range(len(text_data))]
    word_index = imdb.get_word_index()
    reverse_word_index = dict([(value, key) for (key, value) in word_index.items()])
    for review in range(0, len(text_data)):
        for index in enumerate(text_data[review]):
            decoded_review = ' '.join([reverse_word_index.get(index - 3, '?') for index in text_data[review]])
            result[review] = decoded_review
    return result

x_train1 = decode_imdb_reviews(x_train)

# Bert layers
text_input = tf.keras.layers.Input(shape=(), dtype=tf.string, name='text')
preprocessed_text = bert_preprocess(text_input)
outputs = bert_encoder(preprocessed_text)

# Neural network layers
l = tf.keras.layers.Dropout(0.1, name="dropout")(outputs['pooled_output'])
l = tf.keras.layers.Flatten()(l)
l = tf.keras.layers.Dense(64, activation='sigmoid')(l)
l = tf.keras.layers.Dense(1, activation='sigmoid', name="output")(l)

# Use inputs and outputs to construct a final model
model = tf.keras.Model(inputs=[text_input], outputs = [l])
model.summary()

```

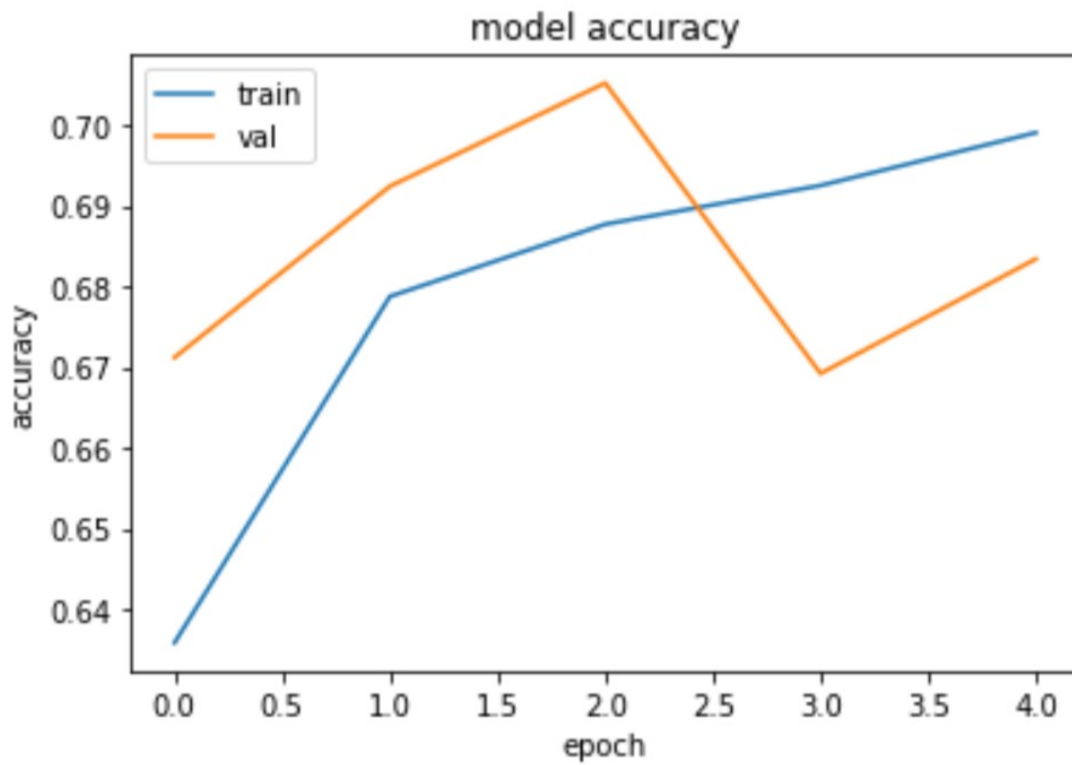
Model: "model_1"

Layer (type)	Output Shape	Param #	Connected to
text (InputLayer)	[(None,)]	0	[]
keras_layer (KerasLayer)	{'input_word_ids': (None, 128), 'input_type_ids': (None, 128), 'input_mask': (None, 128)}	0	['text[0][0]']
keras_layer_1 (KerasLayer)	{'default': (None, 768), 'encoder_outputs': [(None, 128, 768), (None, 128, 768), (None, 128, 768), (None, 128, 768), (None, 128, 768), (None, 128, 768), (None, 128, 768), (None, 128, 768), (None, 128, 768), (None, 128, 768), (None, 128, 768), (None, 128, 768)], 'sequence_output': (None, 128, 768), 'pooled_output': (None, 768)}	109482241	['keras_layer[1][0]', 'keras_layer[1][1]', 'keras_layer[1][2]']
dropout (Dropout)	(None, 768)	0	['keras_layer_1[1][13]']
flatten_1 (Flatten)	(None, 768)	0	['dropout[0][0]']
dense_1 (Dense)	(None, 64)	49216	['flatten_1[0][0]']
output (Dense)	(None, 1)	65	['dense_1[0][0]']
=====			
Total params: 109,531,522			
Trainable params: 49,281			
Non-trainable params: 109,482,241			

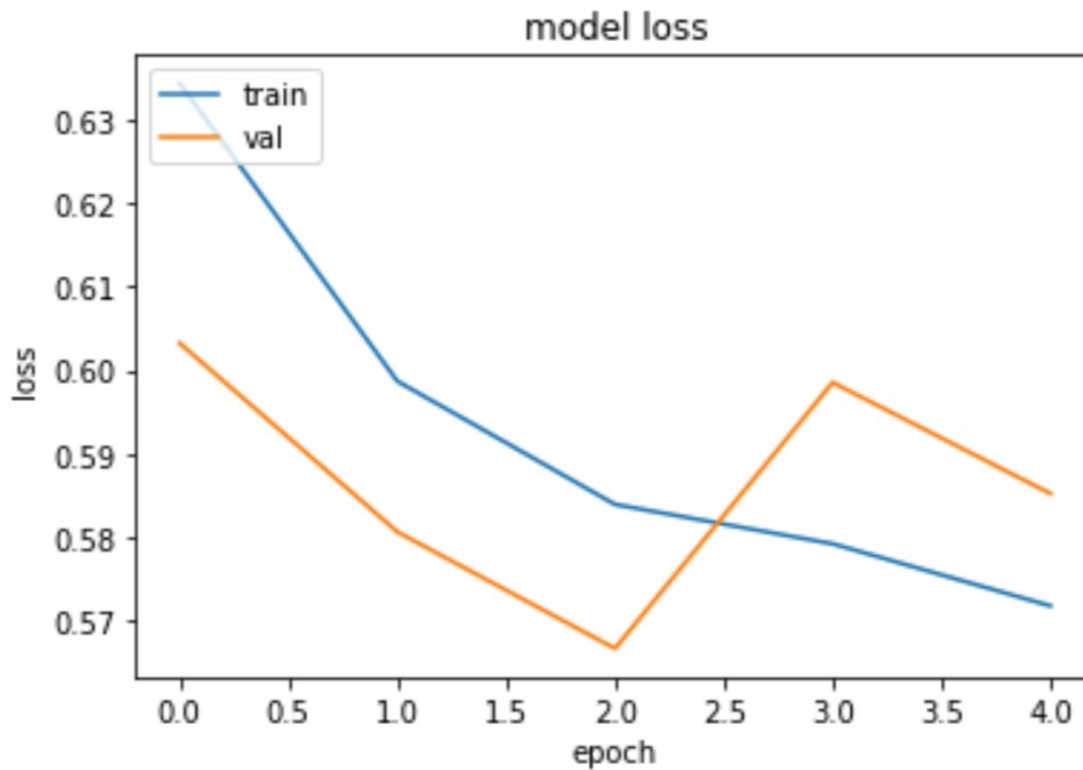
```
model.compile(optimizer='rmsprop', loss='binary_crossentropy', metrics=['acc'])
```

```
history = model.fit(x_train, y_train,  
                    epochs=10, batch_size=128,  
                    validation_split=0.2)
```

```
# summarize history for accuracy  
plt.plot(history.history['acc'])  
plt.plot(history.history['val_acc'])  
plt.title('model accuracy')  
plt.ylabel('accuracy')  
plt.xlabel('epoch')  
plt.legend(['train', 'val'], loc='upper left')  
plt.show()
```



```
# summarize history for loss
plt.plot(history.history['loss'])
plt.plot(history.history['val_loss'])
plt.title('model loss')
plt.ylabel('loss')
plt.xlabel('epoch')
plt.legend(['train', 'val'], loc='upper left')
plt.show()
```



```
x_test1 = decode_imdb_reviews(x_test)
y_test1 = tf.convert_to_tensor(y_test)
x_test2 = tf.convert_to_tensor(x_test1)

test_loss, test_acc = model.evaluate(x_test, y_test)
```

```
test_loss, test_acc = model.evaluate(x_test2, y_test1)
```

```
test_acc
```

```
0.6913599967956543
```

```
test_loss
```

```
0.5780715942382812
```

In this model we tried to use BERT encoding then we used neural networks with 64 units. The samples were preprocessed. We got them back into original form (maxlen= 30). Then we preprocess according to the specific BERT encoder. The BERT layer has L=12 hidden layers, a hidden size of H=768, and A=12 attention heads. We got the accuracy 69%.