

```
In [3]: import matplotlib.pyplot as plt
import numpy as np
import tensorflow as tf
from tensorflow import keras
from tensorflow.keras import layers
from tensorflow.keras.layers import Embedding, SimpleRNN, LSTM, GRU, Dense
from tensorflow.keras.models import Sequential
from tensorflow.keras.datasets import imdb
```

WARNING:tensorflow:From C:\Users\user\anaconda3\Lib\site-packages\keras\src\losses.py:2976: The name tf.losses.sparse_softmax_cross_entropy is deprecated. Please use tf.compat.v1.losses.sparse_softmax_cross_entropy instead.

```
In [49]: def decode_review(review):
return ' '.join([reverse_word_index.get(i-3, '?') for i in review])
```

```
In [8]: def load_imdb_data(max_features, maxlen):
(x_train, y_train), (x_test, y_test) = imdb.load_data(num_words=max_features)
x_train = tf.keras.preprocessing.sequence.pad_sequences(x_train, maxlen=maxlen)
x_test = tf.keras.preprocessing.sequence.pad_sequences(x_test, maxlen=maxlen)
return x_train, y_train, x_test, y_test
```

```
In [23]: def rnn_model(max_features, maxlen):
model = Sequential([
    Embedding(max_features, 32, input_length=maxlen),
    SimpleRNN(32),
    Dense(1, activation='sigmoid')
])
model.compile(loss='binary_crossentropy', optimizer='adam', metrics=['accuracy'])
return model
```

```
In [36]: def lstm_model(max_features,maxlen):
    model=Sequential([
        Embedding(max_features,32,input_length=maxlen),
        LSTM(32),
        Dense(1,activation='sigmoid')
    ])
    model.compile(loss='binary_crossentropy',optimizer='adam',metrics=['accuracy'])
    return model
```

```
In [37]: def gru_model(max_features,maxlen):
    model=Sequential([
        Embedding(max_features,32,input_length=maxlen),
        GRU(32),
        Dense(1,activation='sigmoid')
    ])
    model.compile(loss='binary_crossentropy',optimizer='adam',metrics=['accuracy'])
    return model
```

```
In [50]: def training_and_evaluation(model,x_train,y_train,x_test,y_test,epochs=3,batch_size=128):
    history=model.fit(x_train,y_train,epochs=epochs,batch_size=batch_size,validation_split=0.2,verbose=2)
    loss,accuracy=model.evaluate(x_test,y_test,verbose=0)
    for i in range(2):
        index=np.random.randint(0,len(x_test))
        review=x_test[index]
        actual_label=y_test[index]
        probab=model.predict(np.expand_dims(review,axis=0))[0,0]
        print(f'review{decode_review(review)}')
        label=1 if probab>0.5 else 0
        print(f'review{i+1}')
        print(f'actual_label{actual_label}')
        print(f'predicted_label{label}')
    return history,loss,accuracy
```

```
In [46]: max_features=10000  
maxlen=100
```

```
In [55]: def plotting(history,model_name):  
    print(f'model: {model_name}')  
    print(f'accuracy: {accuracy}')  
    print(f'loss: {loss}')  
    plt.title(f'accuracy vs validation accuracy')  
    plt.plot(history.history['accuracy'],label='training accuracy')  
    plt.plot(history.history['val_accuracy'],label='validation accuracy')  
    plt.xlabel('epochs')  
    plt.ylabel('accuracy')  
    plt.show()
```

```
In [56]: word_index=imdb.get_word_index()
reverse_word_index=dict([(value,key) for (key,value) in word_index.items()])

x_train,y_train,x_test,y_test=load_imdb_data(max_features,maxlen)
rnn=rnn_model(max_features,maxlen)
history,loss,accuracy=training_and_evaluation(rnn,x_train,y_train,x_test,y_test)
plotting(history,'rnn')
```

Epoch 1/3

157/157 - 11s - loss: 0.6040 - accuracy: 0.6504 - val_loss: 0.4256 - val_accuracy: 0.8124 - 11s/epoch - 67ms/step

Epoch 2/3

157/157 - 7s - loss: 0.3340 - accuracy: 0.8611 - val_loss: 0.3720 - val_accuracy: 0.8408 - 7s/epoch - 46ms/step

Epoch 3/3

157/157 - 7s - loss: 0.2247 - accuracy: 0.9157 - val_loss: 0.4037 - val_accuracy: 0.8318 - 7s/epoch - 42ms/step

1/1 [=====] - 0s 330ms/step

reviewcould easily have been ? in his ultimately caring attitude ? keeps him sharp enough to keep him from being too sweet but the revelation is stephen robertson as michael not since ? ? ? in gilbert ? has anyone so perfectly captured a person with an ? that i began to believe he really was an actor with cerebral ? and his eyes my god he can rip you apart with them br br this movie is to ? ? f ? amazing go see it take a box of ? and enjoy every well earned tear and laugh

review1

actual_label1

predicted_label1

1/1 [=====] - 0s 63ms/step

reviewonly area of the film i enjoyed was the commentary on film making for the most part this film seemed random and somewhat ? i don't say that in a ? way however and just silly it was as if he was mixing fantasy with everyday life which may sound intriguing in some films but the fantasy merely seemed needlessly perverse br br my criticism of this film is not upon the actors rather the story itself i found it boring and ? i wanted my money back but considering it was a film festival that wasn't about to happen

review2

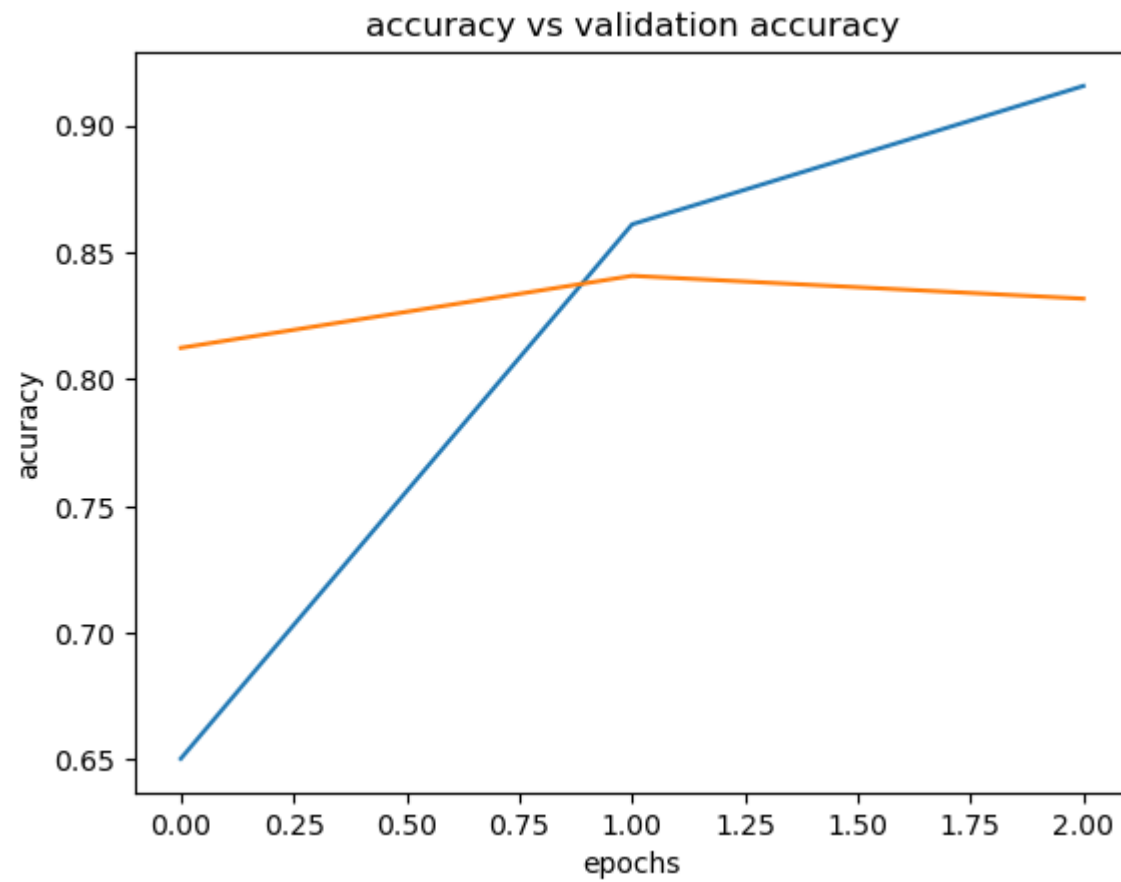
actual_label0

predicted_label0

model: rnn

accuracy: 0.8271999955177307

loss: 0.40923866629600525



In [57]:

```
lstm=lstm_model(max_features,maxlen)
history,loss,accuracy=training_and_evaluation(lstm,x_train,y_train,x_test,y_test)
plotting(history,'lstm')
```

Epoch 1/3

157/157 - 21s - loss: 0.5020 - accuracy: 0.7435 - val_loss: 0.3513 - val_accuracy: 0.8446 - 21s/epoch - 131ms/step

Epoch 2/3

157/157 - 15s - loss: 0.2771 - accuracy: 0.8892 - val_loss: 0.3942 - val_accuracy: 0.8350 - 15s/epoch - 95ms/step

Epoch 3/3

157/157 - 15s - loss: 0.2114 - accuracy: 0.9214 - val_loss: 0.3571 - val_accuracy: 0.8422 - 15s/epoch - 97ms/step

1/1 [=====] - 1s 1s/step

reviewand i did not understand the film my good ? friend wow what a long name explained every thing to me what a great movie after watching this movie i felt i should have watched many more movies from ? ? film industry the war scenes were amazing camera work excellent and plot beautiful the actress what a beauty give her an award for best looking someone ? ? ? come on i smell a oscar winner i didnt understand the songs but they were excellent ? is a great director and i hope his next film was a success

review1

actual_label1

predicted_label1

1/1 [=====] - 0s 68ms/step

reviewfeminine areas secondly ? her breasts after new york and new jersey ? some confusion as to whether miss is actually ? ? br br scene four ? is running she falls down this gives her ? cancer we always knew it could come back her father says right ? cancer from ? perhaps i missed something as i said it's been a few years surely i missed something didn't i for the love of god please tell me the girl did not contract ? cancer from falling down br br that scream you just heard was my soul dying

review2

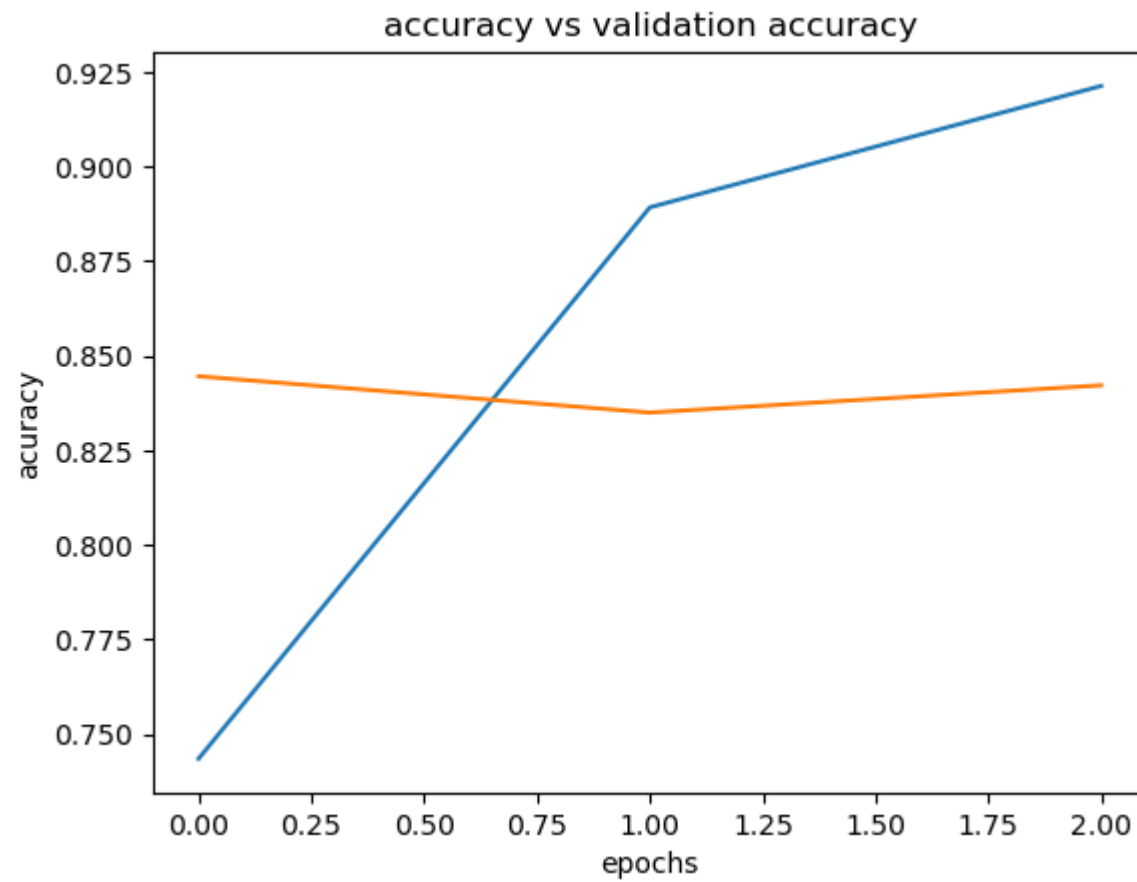
actual_label0

predicted_label0

model: lstm

accuracy: 0.8444399833679199

loss: 0.36172762513160706



In [58]:

```
gru=gru_model(max_features,maxlen)
history,loss,accuracy=training_and_evaluation(gru,x_train,y_train,x_test,y_test)
plotting(history,'gru')
```

Epoch 1/3

157/157 - 21s - loss: 0.5277 - accuracy: 0.7103 - val_loss: 0.3646 - val_accuracy: 0.8376 - 21s/epoch - 132ms/step

Epoch 2/3

157/157 - 14s - loss: 0.2875 - accuracy: 0.8835 - val_loss: 0.3711 - val_accuracy: 0.8442 - 14s/epoch - 90ms/step

Epoch 3/3

157/157 - 14s - loss: 0.2163 - accuracy: 0.9191 - val_loss: 0.3737 - val_accuracy: 0.8412 - 14s/epoch - 89ms/step

1/1 [=====] - 1s 880ms/step

reviewboy who is not in the military and has not trained to be a jet pilot takes off for a foreign country to rescue his dad if this is not ridiculous enough he talks a colonel in the air force into helping him get his hands on a jet wow to make the picture even more absurd the colonel risks his career and life by giving the ? ? some hands on aid t hey not only don't make like this anymore but they never did this sappy corny film should be tossed into the air and blown away by a

review1

actual_label0

predicted_label0

1/1 [=====] - 0s 62ms/step

reviewchallenging rap ? into the camera filmed in mtv style fast cuts crazy camera angles tight close ups and animation ? the dialogue might have been ? in the original languages of ? and than in the subtitles and i have no doubt th at some of the contemporary ? references will slip over western heads as well as the cultural and political context unless of course you are familiar with ? this kind of teen film may be a first for ? but it has been done before and done better in other western countries la ? 1995 for example

review2

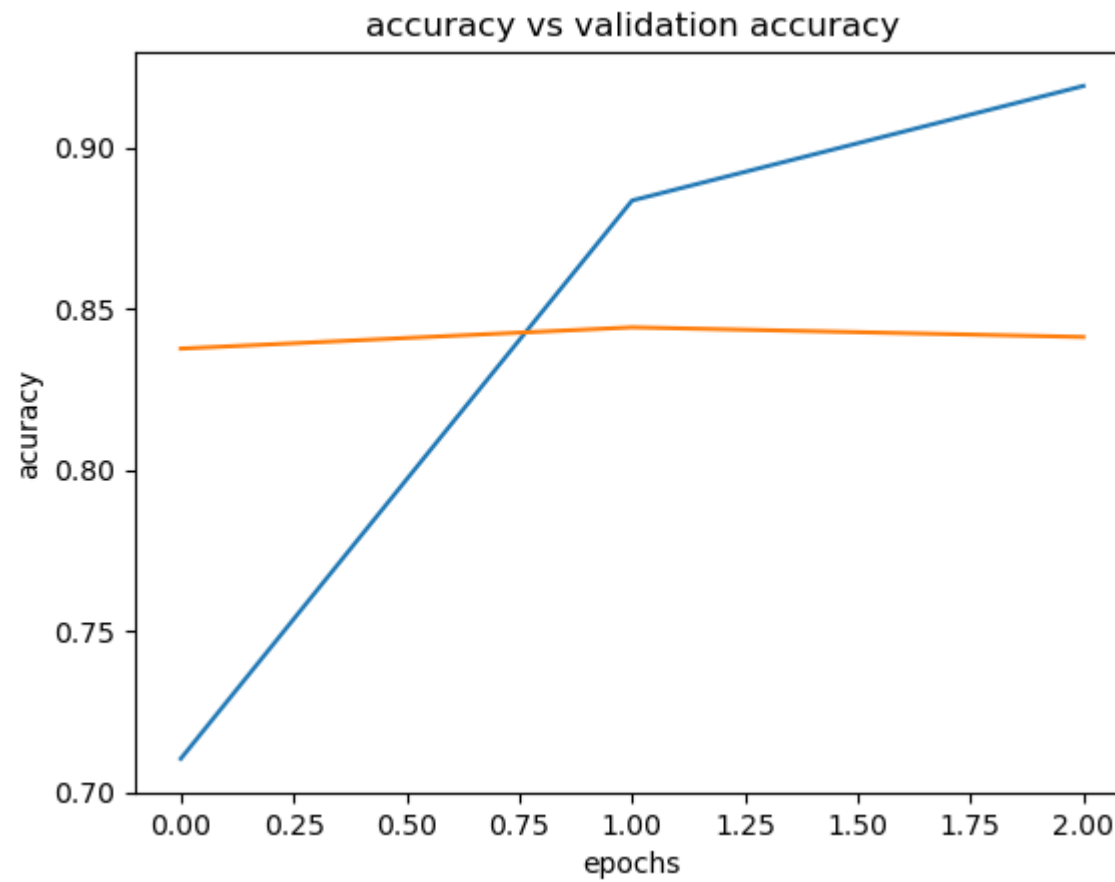
actual_label0

predicted_label1

model: gru

accuracy: 0.8432000279426575

loss: 0.3779047131538391



In []:

In []: