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Lab: 16: Design of LSTM and GRU RNN for classification of IMDB reviews

```
M
In [1]:
import pandas as pd
import numpy as np
from nltk.corpus import stopwords
from sklearn.model_selection import train_test_split
from tensorflow.keras.preprocessing.text import Tokenizer
from tensorflow.keras.preprocessing.sequence import pad_sequences
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import Embedding, LSTM, Dense, Bidirectional, GRU
from tensorflow.keras.callbacks import ModelCheckpoint
from tensorflow.keras.models import load_model
In [2]:
                                                                                       M
import warnings
warnings.filterwarnings("ignore")
In [3]:
                                                                                       M
data = pd.read_csv('imdb.csv', encoding='latin1')
data.head()
Out[3]:
```

| | text | label |
|---|--|-------|
| 0 | It's been about 14 years since Sharon Stone aw | 0 |
| 1 | someone needed to make a car payment this i | 0 |
| 2 | The Guidelines state that a comment must conta | 0 |
| 3 | This movie is a muddled mish-mash of clichés f | 0 |
| 4 | Before Stan Laurel became the smaller half of | 0 |

```
import nltk
nltk.download('stopwords')
english_stops = set(stopwords.words('english'))
from nltk.corpus import stopwords

[nltk_data] Downloading package stopwords to
[nltk_data] C:\Users\arulk\AppData\Roaming\nltk_data...
[nltk_data] Package stopwords is already up-to-date!

In [5]:

english_sto = set(stopwords.words('english'))
```

In [6]: ▶

```
def load_dataset():
    df = pd.read_csv('imdb.csv', encoding='latin1')
    x_data = df['text']
    y_data = df['label']
    x_data = x_data.replace({'['A-Za-z]' : ' '}, regex = True)
    x_data = x_data.apply(lambda review: [w for w in review.split() if w not in english_x_data = x_data.apply(lambda review: [w.lower() for w in review])
    y_data = y_data.replace('positive', 1)
    y_data = y_data.replace('negative', 0)
    return x_data, y_data
    x_data, y_data = load_dataset()
    print('Reviews')
    print(x_data, '\n')
    print('Sentiment')
    print(y_data)
```

```
Reviews
      [', 14, -, ', ., ,, ,, ', ., ", 2", -, ", ", ....
0
1
      [..., ..., ', ..., ..., ., ?, ., ., $5.99...
2
                                            [., ., ,, ., .]
      [-, é, ., ,, ,, ', ., ,, (, ,, ,, ).<, /><,...
3
      [-, ,, ,, ., ,, ', (, ), .<, /><, />, ', ,, ',...
      [., ', ..., ., ', -, -, -, ., (, ).<, /><, />,...
81
      [,, 6, .<, /><, />, ', 2, ,, 11, 2001., ,, (, ...
82
      [., -, -, ., ,, ,, ., ., -, -, ., ', ,, ...
83
84 [.,,,,,',.,,,",",.,,',.,,',,',...
85 [,,',",,,,,,(,',),",",,,,...
Name: text, Length: 86, dtype: object
Sentiment
0
      0
1
      0
2
      0
3
      0
4
      0
81
      0
82
      1
83
      1
84
      0
85
Name: label, Length: 86, dtype: int64
```

```
In [7]:
                                                                                                 H
x_train, x_test, y_train, y_test = train_test_split(x_data, y_data, test_size = 0.4)
print('Train Set')
print(x_train, '\n')
print(x_test, '\n')
print('Test Set')
print(y_train, '\n')
print(y_test)
Train Set
40
      [,, ,, ., ,, .<, /><, />, ,, ,, ', ?, ,, ,, ,, ...
      [,, ., ', -, .<, /><, />, ', ., ,, .<, /><, />...
80
61
                  ', ., ,, ,, ,, (, ), ., ,, .<, />...
26
      [", ", .., ., ', ., ', !!, ..<, /><, />, 10...
[., ,, 5-7, ., ', ,, .<, />, />, ., ,, ,, ...
27
63
      64
      [., ,, ,, ', ., ,, ", ", ., ., ', ., ,, ', /, ...
84
51
                                   [., ,, .,
17
      [,, ,, ,, ., .<, /><, />, -, ..., ,,
25
      [', ., ', ., ., ,, ,, ., ', ., ', ,, ., ., ., ., ...
79
                     [,, ., ., ., ., ., ,, ,, ', ', '
78
      [-8,, (, ), ., 2, (, ), .<, /><, />, -, ?, ,, ...
      [-, ,, ", ", 100., ,, ,, (, ), , , -, ', ,, ., ...
[,, ', -, ', ., ,, ,, ', ', ., ., ,, ', ', ...
[5, 6, ., .<, />, />, ., ', !!!, ., ", "., !,...
15
77
76
      [,, ,, ,, ., ', ., (, )., ., ., ,, -, .,
24
      「.. 6. .<. /><. />. '. 2. .. 11. 2001.. .. (. ...
82
In [8]:
def get max length():
    review_length = []
```

```
def get_max_length():
    review_length = []
    for review in x_train:
        review_length.append(len(review))
    return int(np.ceil(np.mean(review_length)))
```

In [9]: ▶

```
token = Tokenizer(lower=False)
token.fit_on_texts(x_train)
x_train = token.texts_to_sequences(x_train)
x_test = token.texts_to_sequences(x_test)
max_length = get_max_length()
x_train=pad_sequences(x_train, maxlen=max_length, padding='post', truncating='post')
x_test=pad_sequences(x_test, maxlen=max_length, padding='post', truncating='post')
total_words = len(token.word_index) + 1

print(' Encoded X Train\n', x_train, '\n')
print(' Encoded X Test\n', x_test, '\n')
print('Maximum review length: ', max_length)
```

```
Encoded X Train
     1
         2 ... 3
         3 ... 23
[ 1
      2
                  6
                       7]
[ 18
      1 11 ... 41 41
                      11]
1
        3 ...
                    3
                       5]
 1
                       1]
      1 32 ...
  1
                1
                    1
  2 116
         2 ...
                0
                    0
                       0]]
Encoded X Test
[[ 3 5 3 ... 1 1 2]
[ 3 1 8 ... 1 1 3]
[ 2 3 11 ... 1 12 2]
[ 1 2 3 ... 2 4 4]
             0 0 0]
[11 11 3 ...
[ 2 5 5 ...
             0 0
                  0]]
```

Maximum review length: 45

In [10]: ▶

```
EMBED_DIM = 32
LSTM_OUT = 64

# Assuming x_data is a list of lists, where each inner list contains words
sentences = [' '.join(sentence) for sentence in x_data]

# Calculate the maximum sequence length
max_sequence_length = max(len(sentence.split()) for sentence in sentences)

# Set your_input_length to the maximum sequence length
your_input_length = max_sequence_length

model = Sequential()
model.add(Embedding(total_words, EMBED_DIM, input_length=your_input_length))
model.add(Dense(32, activation='relu'))
model.add(Dense(32, activation='relu'))
model.add(Dense(1, activation='sigmoid'))
model.compile(optimizer='adam', loss='binary_crossentropy', metrics=['accuracy'])
print(model.summary())
```

Model: "sequential"

| Layer (type) | Output Shape | Param # |
|-----------------------|-----------------|---------|
| embedding (Embedding) | (None, 156, 32) | 3776 |
| lstm (LSTM) | (None, 64) | 24832 |
| dense (Dense) | (None, 32) | 2080 |
| dense_1 (Dense) | (None, 1) | 33 |
| | | |

Total params: 30721 (120.00 KB)
Trainable params: 30721 (120.00 KB)
Non-trainable params: 0 (0.00 Byte)

None

In [12]:

```
checkpoint = ModelCheckpoint('model_weights.h5', save_best_only=True, save_weights_only=
```

In [14]: ▶

```
modell = Sequential()
modell.add(Embedding(total_words, 32, input_length=max_length))
modell.add(LSTM(32))
modell.add(Dense(32, activation='relu'))
modell.add(Dense(1, activation='sigmoid'))
modell.compile(optimizer = 'adam', loss = 'binary_crossentropy', metrics = ['accuracy'])
modell.fit(x_train, y_train, batch_size = 128, epochs = 10, callbacks=[checkpoint])
modell.evaluate(x_test, y_test)
print(modell.summary())
```

```
Epoch 1/10
0.5098WARNING:tensorflow:Can save best model only with val_loss availabl
e, skipping.
cy: 0.5098
Epoch 2/10
1/1 [============== ] - ETA: 0s - loss: 0.6921 - accuracy:
0.5686WARNING:tensorflow:Can save best model only with val_loss availabl
e, skipping.
1/1 [============ ] - 0s 41ms/step - loss: 0.6921 - accu
racy: 0.5686
Epoch 3/10
1/1 [=========== ] - ETA: 0s - loss: 0.6913 - accuracy:
0.5490WARNING:tensorflow:Can save best model only with val_loss availabl
e, skipping.
racy: 0.5490
Epoch 4/10
0.5490WARNING:tensorflow:Can save best model only with val_loss availabl
1/1 [============ ] - 0s 32ms/step - loss: 0.6907 - accu
racy: 0.5490
Epoch 5/10
1/1 [============== ] - ETA: 0s - loss: 0.6900 - accuracy:
0.5490WARNING:tensorflow:Can save best model only with val_loss availabl
e, skipping.
1/1 [=============== ] - 0s 35ms/step - loss: 0.6900 - accu
racy: 0.5490
Epoch 6/10
0.5490WARNING:tensorflow:Can save best model only with val_loss availabl
e, skipping.
1/1 [=============== ] - 0s 31ms/step - loss: 0.6894 - accu
racy: 0.5490
Epoch 7/10
0.5490WARNING:tensorflow:Can save best model only with val_loss availabl
e, skipping.
racy: 0.5490
Epoch 8/10
0.5490WARNING:tensorflow:Can save best model only with val_loss availabl
e, skipping.
racy: 0.5490
Epoch 9/10
0.5490WARNING:tensorflow:Can save best model only with val_loss availabl
e, skipping.
racy: 0.5490
Epoch 10/10
1/1 [============== ] - ETA: 0s - loss: 0.6869 - accuracy:
0.5490WARNING:tensorflow:Can save best model only with val_loss availabl
e, skipping.
racy: 0.5490
2/2 [============== ] - 1s 13ms/step - loss: 0.6999 - accu
```

```
racy: 0.5429
Model5] sequential_3"
```

H

Total params: 13185 (51.50 KB)
Trainable params: 13185 (51.50 KB)
Non-trainable params: 0 (0.00 Byte)

None

```
Epoch 1/10
cy: 0.4706
Epoch 2/10
uracy: 0.5490
Epoch 3/10
1/1 [========== ] - 0s 74ms/step - loss: 0.6910 - accu
racy: 0.5490
Epoch 4/10
1/1 [============ ] - 0s 41ms/step - loss: 0.6901 - accu
racy: 0.5490
Epoch 5/10
1/1 [============= ] - 0s 46ms/step - loss: 0.6893 - accu
racy: 0.5490
Epoch 6/10
racy: 0.5490
Epoch 7/10
1/1 [========== ] - 0s 47ms/step - loss: 0.6877 - accu
racy: 0.5490
Epoch 8/10
1/1 [============= ] - 0s 86ms/step - loss: 0.6869 - accu
racy: 0.5490
Epoch 9/10
1/1 [========== ] - 0s 65ms/step - loss: 0.6861 - accu
racy: 0.5490
Epoch 10/10
1/1 [============= ] - 0s 66ms/step - loss: 0.6853 - accu
racy: 0.5490
2/2 [========== ] - 1s 17ms/step - loss: 0.6992 - accu
racy: 0.5429
Model: "sequential_4"
```

| Layer (type) | Output Shape | Param # |
|---|----------------|---------|
| embedding_4 (Embedding) | (None, 45, 32) | 3776 |
| <pre>bidirectional (Bidirection al)</pre> | (None, 64) | 16640 |
| dense_8 (Dense) | (None, 32) | 2080 |
| dense_9 (Dense) | (None, 1) | 33 |

Total params: 22529 (88.00 KB)
Trainable params: 22529 (88.00 KB)
Non-trainable params: 0 (0.00 Byte)

None

```
M
In [16]:
model2.evaluate(x_test, y_test)
racy: 0.5429
Out[16]:
[0.699213981628418, 0.5428571701049805]
In [17]:
                                                                        M
from keras.preprocessing.text import Tokenizer
t = Tokenizer()
# De fin i ng 4 doc men t L i sts
fit_text = ['Machine Learning Knowledge', 'Machine Learning',
'Deep Learning',
'Artificial Intelligence']
t.fit_on_texts(fit_text)
```