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## PDL Lab16. Design of LSTM and GRU RNN for classification of IMDB reviews

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]:

```
import warnings
warnings.filterwarnings("ignore")
```

In [3]:

```
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

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_stop])
    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

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

```

In [7]:

```
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)
```

[illegible]

In [8]:

```
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  1  2 ...  3  2  0]
 [ 1  2  3 ... 23  6  7]
 [18  1 11 ... 41 41 11]
...
 [ 1  1  3 ...  9  3  5]
 [ 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]
 [11 11  3 ...  0  0  0]
 [ 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(LSTM(LSTM_OUT))
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=True, monitor='val_loss', mode='min', verbose=1)
```

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  
1/1 [=====] - ETA: 0s - loss: 0.6929 - accuracy: 0.5098WARNING:tensorflow:Can save best model only with val\_loss available, skipping.  
1/1 [=====] - 5s 5s/step - loss: 0.6929 - accuracy: 0.5098  
Epoch 2/10  
1/1 [=====] - ETA: 0s - loss: 0.6921 - accuracy: 0.5686WARNING:tensorflow:Can save best model only with val\_loss available, skipping.  
1/1 [=====] - 0s 41ms/step - loss: 0.6921 - accuracy: 0.5686  
Epoch 3/10  
1/1 [=====] - ETA: 0s - loss: 0.6913 - accuracy: 0.5490WARNING:tensorflow:Can save best model only with val\_loss available, skipping.  
1/1 [=====] - 0s 32ms/step - loss: 0.6913 - accuracy: 0.5490  
Epoch 4/10  
1/1 [=====] - ETA: 0s - loss: 0.6907 - accuracy: 0.5490WARNING:tensorflow:Can save best model only with val\_loss available, skipping.  
1/1 [=====] - 0s 32ms/step - loss: 0.6907 - accuracy: 0.5490  
Epoch 5/10  
1/1 [=====] - ETA: 0s - loss: 0.6900 - accuracy: 0.5490WARNING:tensorflow:Can save best model only with val\_loss available, skipping.  
1/1 [=====] - 0s 35ms/step - loss: 0.6900 - accuracy: 0.5490  
Epoch 6/10  
1/1 [=====] - ETA: 0s - loss: 0.6894 - accuracy: 0.5490WARNING:tensorflow:Can save best model only with val\_loss available, skipping.  
1/1 [=====] - 0s 31ms/step - loss: 0.6894 - accuracy: 0.5490  
Epoch 7/10  
1/1 [=====] - ETA: 0s - loss: 0.6887 - accuracy: 0.5490WARNING:tensorflow:Can save best model only with val\_loss available, skipping.  
1/1 [=====] - 0s 35ms/step - loss: 0.6887 - accuracy: 0.5490  
Epoch 8/10  
1/1 [=====] - ETA: 0s - loss: 0.6881 - accuracy: 0.5490WARNING:tensorflow:Can save best model only with val\_loss available, skipping.  
1/1 [=====] - 0s 34ms/step - loss: 0.6881 - accuracy: 0.5490  
Epoch 9/10  
1/1 [=====] - ETA: 0s - loss: 0.6875 - accuracy: 0.5490WARNING:tensorflow:Can save best model only with val\_loss available, skipping.  
1/1 [=====] - 0s 33ms/step - loss: 0.6875 - accuracy: 0.5490  
Epoch 10/10  
1/1 [=====] - ETA: 0s - loss: 0.6869 - accuracy: 0.5490WARNING:tensorflow:Can save best model only with val\_loss available, skipping.  
1/1 [=====] - 0s 35ms/step - loss: 0.6869 - accuracy: 0.5490  
2/2 [=====] - 1s 13ms/step - loss: 0.6999 - accuracy: 0.5429  
Model: "sequential\_3"

Layer (type)	Output Shape	Param #
=====		
embedding_3 (Embedding)	(None, 45, 32)	3776
lstm_3 (LSTM)	(None, 32)	8320
dense_6 (Dense)	(None, 32)	1056
dense_7 (Dense)	(None, 1)	33
=====		
Total params: 13185 (51.50 KB)		
Trainable params: 13185 (51.50 KB)		
Non-trainable params: 0 (0.00 Byte)		

None

In [15]:

```
model2 = Sequential()
model2.add(Embedding(total_words, 32, input_length=max_length))
model2.add(Bidirectional(LSTM(32)))
model2.add(Dense(32, activation='relu'))
model2.add(Dense(1, activation='sigmoid'))
model2.compile(optimizer = 'adam', loss = 'binary_crossentropy', metrics = ['accuracy'])
model2.fit(x_train, y_train, batch_size = 128, epochs = 10)
model2.evaluate(x_test, y_test)
print(model2.summary())
```

Epoch 1/10  
1/1 [=====] - 7s 7s/step - loss: 0.6933 - accuracy: 0.4706  
Epoch 2/10  
1/1 [=====] - 0s 106ms/step - loss: 0.6920 - accuracy: 0.5490  
Epoch 3/10  
1/1 [=====] - 0s 74ms/step - loss: 0.6910 - accuracy: 0.5490  
Epoch 4/10  
1/1 [=====] - 0s 41ms/step - loss: 0.6901 - accuracy: 0.5490  
Epoch 5/10  
1/1 [=====] - 0s 46ms/step - loss: 0.6893 - accuracy: 0.5490  
Epoch 6/10  
1/1 [=====] - 0s 43ms/step - loss: 0.6885 - accuracy: 0.5490  
Epoch 7/10  
1/1 [=====] - 0s 47ms/step - loss: 0.6877 - accuracy: 0.5490  
Epoch 8/10  
1/1 [=====] - 0s 86ms/step - loss: 0.6869 - accuracy: 0.5490  
Epoch 9/10  
1/1 [=====] - 0s 65ms/step - loss: 0.6861 - accuracy: 0.5490  
Epoch 10/10  
1/1 [=====] - 0s 66ms/step - loss: 0.6853 - accuracy: 0.5490  
2/2 [=====] - 1s 17ms/step - loss: 0.6992 - accuracy: 0.5429  
Model: "sequential\_4"

Layer (type)	Output Shape	Param #
=====		
embedding_4 (Embedding)	(None, 45, 32)	3776
bidirectional (Bidirectional)	(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		

In [16]:

```
model2.evaluate(x_test, y_test)
```

2/2 [=====] - 0s 15ms/step - loss: 0.6992 - accuracy: 0.5429

Out[16]:  
[0.699213981628418, 0.5428571701049805]

In [17]:

```
from keras.preprocessing.text import Tokenizer

t = Tokenizer()
# De f i n i n g 4 d o c m e n t l i s t s
fit_text = ['Machine Learning Knowledge', 'Machine Learning',
'Deep Learning',
'Artificial Intelligence']
t.fit_on_texts(fit_text)
```