Lab sheet 11 - Deep Learning

1. Neural Network Language Model

Model: "sequential_1"			
Layer (type)	Output	Shape	Param #
dense_1 (Dense)	(None,	100)	500200
dense_2 (Dense)	(None,	100)	10100
dense_3 (Dense)	(None,	100)	10100
dense_4 (Dense)	(None,		505101
Total params: 1,025,501			
Trainable params: 1,025,501			
Non-trainable params: 0			

Model fitting

```
i . __newline_
Train on 81369 samples, validate on 23828 samples
Epoch 2/2
         ------] • 97s lms/step • loss: 5.0930 • acc: 0.1934 • val_loss: 4.9334 • val_acc: 0.1646
81369/81369 [----
1 have , __newline__
Train on 81369 samples, validate on 23828 samples
Epoch 3/3
1 have , __newline_
Train on 81369 samples, validate on 23828 samples
Epoch 4/4
            81369/81369 [-----
i am a _unk_ , _newline_
Train on 81369 samples, validate on 23828 samples
Epoch 5/5
I have , __newline_
Train on 81369 samples, validate on 23828 samples
Epoch 6/6
I have been absent friends , __newline_
Train on 81369 samples, validate on 23828 samples
1 am a __unk__ , __newline__
```

Bigram model

```
81368/81368 [====
          i have not , __newline_
Train on 81368 samples, validate on 23827 samples
Epoch 3/3
i have 'ss,__newline_
Train on 81368 samples, validate on 23827 samples
Epoch 4/4
i have 't . __newline__
Train on 81368 samples, validate on 23827 samples
i have 't . __newline__
Train on 81368 samples, validate on 23827 samples
         81368/81368 [====
Train on 81368 samples, validate on 23827 samples
              :=======] - 73s 899us/step - loss: 4.2191 - acc: 0.2485 - val_loss: 5.0076 - val_acc: 0.1656
81368/81368 [====
i have 't . __newline__
Train on 81368 samples, validate on 23827 samples
81368/81368 [====
           i have seen her die . __newline__
Train on 81368 samples, validate on 23827 samples
Epoch 50/50
i have not ended if this thou perceiv 'st , __newline__
```

Trigram model

```
Train on $1367 samples, validate on 23826 samples
Epoch 1/1
81367/81367 [---
           i have to , __newline__
Train on $1367 samples, validate on 23826 samples
Epoch 2/2
            81367/81367 [----
i have to not , __newline_
Train on $1367 samples, validate on 23826 samples
Epoch 3/3
81367/81367 [----
            I have to have , __newline_
Train on 81367 samples, validate on 23826 samples
         I have to not the _unk _ of _unk _ newline
Train on $1367 samples, validate on 23826 samples
Epoch 5/5
i have to __unk__ . __newline_
Train on 81367 samples, validate on 23826 samples
Epoch 6/6
I have to not the _unk_ of _unk_ , _newline_
Train on 81367 samples, validate on 23826 samples
i have to sworn me . __newline_
Train on 81367 samples, validate on 23826 samples
Epoch 58/58
```

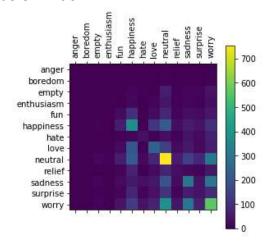
2. Sentiment Classification using Neural Network

Model architecture

Model: "sequential"			
Layer (type)	Output	Shape	Param #
dense (Dense)	(None,	25)	375075
dropout (Dropout)	(None,	25)	0
dense_1 (Dense)	(None,	100)	2600
dropout_1 (Dropout)	(None,	100)	0
dense_2 (Dense)	(None,	50)	5050
dense_3 (Dense)	(None,	13)	663
softmax (Softmax)	(None,	13)	0
Total params: 383,388		**************	
Trainable params: 383,388			
Non-trainable params: θ			

Test accuracy

Confusion matrix



3. Per timestep prediction using Neural Network - POS tagging

Vocabulary building

```
Vocabulary size: 54562
Most frequent tokens
       the: 189915
       to: 68599
       af: 64398
       and: 59512
       that: 48974
       1: 44784
Least frequent tokens
       incapacitates: 1
       backruptcies: I
       IPOs: 1
       est 1
       Separate: 1
       raking: 1
       Heroin: 1
```

Model Definition

```
Model: "sequential"
Layer (type)
                            Output Shape
                                                      Param #
embedding (Embedding)
                            (None, 100, 25)
                                                      250050
simple_rnn (SimpleRNN)
                            (None, 100, 30)
                                                      1680
time_distributed (TimeDistri (None, 100, 44)
                                                      1364
time_distributed_1 (TimeDist (None, 100, 44)
                                                      0
Total params: 253,094
Trainable params: 253,094
Non-trainable params: 0
```

Test set accuracy

4. Pre trained embedding

Pre trained vectors

Training and testing accurcy

5. Recurrent Neural Network

Data Loading

Data preparing

```
[[ 0 0 0 0 ... 15 17 12]
[ 0 0 0 ... 505 17 12]
[ 0 0 0 ... 11 17 12]
...
[ 0 0 0 ... 254 17 12]
[ 0 0 0 ... 2735 17 12]
[ 0 0 0 ... 4329 17 12]]
```

Simple RNN

```
Model: "sequential_1"
Layer (type)
                            Output Shape
                                                      Param #
embedding_1 (Embedding)
                            (None, 200, 10)
                                                      100000
simple_rnn_1 (SimpleRNN)
                            (None, 25)
                                                      900
dense_1 (Dense)
                            (None, 46)
                                                      1196
softmax_1 (Softmax)
                            (None, 46)
                                                      0
Total params: 102,096
Trainable params: 102,096
Non-trainable params: 0
```

Result

Bidirectional RNN

Model: "sequential_2"		
Layer (type)	Output Shape	Param #
embedding_2 (Embedding)	(None, 200, 10)	100000
bidirectional (Bidirectional	(None, 25)	1800
dense_2 (Dense)	(None, 46)	1196
softmax_2 (Softmax)	(None, 46)	0
Total params: 102,996 Trainable params: 102,996 Non-trainable params: 0		

Result

Simple RNN with averaging

Model: "sequential_3"	, 	
Layer (type)	Output Shape	Param #
embedding_3 (Embedding)	(None, 200, 10)	100000
simple_rnn_3 (SimpleRNN)	(None, 200, 25)	900
global_average_pooling1d (Gl	(None, 25)	0
dense_3 (Dense)	(None, 46)	1196
softmax_3 (Softmax)	(None, 46)	0
Total params: 102,096		
Trainable params: 102,096		
Non-trainable params: 0		

Result

Multilayer Bidirectional RNN

Model: "sequential_4"		
Layer (type)	Output Shape	Param #
embedding_4 (Embedding)	(None, 200, 10)	100000
bidirectional_1 (Bidirection	(None, 200, 25)	7200
bidirectional_2 (Bidirection	(None, 25)	10200
dense_4 (Dense)	(None, 46)	1196
softmax_4 (Softmax)	(None, 46)	0
Total params: 118,596 Trainable params: 118,596 Non-trainable params: 0		

Result

6. Reuters News Classification

Data loading

Dense NN model

Vanilla RNN

```
Train on 5572 samples, validate on 620 samples
Epoch 1/10
Fnoch 2/10
5572/5572 [==========] - 18s 3ms/step - loss: 2.2394 - acc: 0.4072 - val_loss: 2.1670 - val_acc: 0.4145
Epoch 4/10
Epoch 5/10
5572/5572 [=======] - 14s 3ms/step - loss: 2.5381 - acc: 0.3487 - val_loss: 3.2278 - val_acc: 0.3871
Epoch 6/10
Epoch 7/10
5572/5572 [=======] - 15s 3ms/step - loss: 2.7480 - acc: 0.3320 - val_loss: 3.2958 - val_acc: 0.1903
Epoch 8/10
5572/5572 [=========] - 15s 3ms/step - loss: 2.7328 - acc: 0.3186 - val loss: 2.2404 - val acc: 0.4145
Epoch 9/10
Epoch 10/10
2654/2654 [=========] - 2s 665us/step
Test accuracy: 41.07%
```

Bidirectional RNN

```
Train on 5572 samples, validate on 620 samples
Epoch 1/10
5572/5572 [=========] - 15s 3ms/step - loss: 2.4509 - acc: 0.3835 - val_loss: 2.2217 - val_acc: 0.4145
Epoch 2/10
5572/5572 [==========] - 15s 3ms/step - loss: 2.2142 - acc: 0.4139 - val_loss: 2.1106 - val_acc: 0.4226
Epoch 3/10
5572/5572 [============ 1 - 19s 3ms/step - loss: 2.0135 - acc: 0.4883 - val loss: 1.8701 - val acc: 0.5403
5572/5572 [=======] - 19s 3ms/step - loss: 1.8171 - acc: 0.5558 - val_loss: 1.8466 - val_acc: 0.5435
Epoch 5/10
5572/5572 [====
           ===========] - 19s 3ms/step - loss: 1.6924 - acc: 0.5811 - val_loss: 1.8101 - val_acc: 0.5597
5572/5572 [===========] - 18s 3ms/step - loss: 1.6138 - acc: 0.5951 - val loss: 1.8306 - val acc: 0.5210
Epoch 7/10
Epoch 8/10
Epoch 9/10
Epoch 10/10
2654/2654 [==========] - 2s 798us/step
Test accuracy: 57.87%
```

Multilayer Bidirectional RNN

7. Text-Generation-with-LSTM

```
294283/286281 [----
                                                                                                                                                          - Generating with sent for that dream as reality, that is, we accept mar hypotheses "
on that drose as reality, that is, in
accept our hypotheses and the string the samplisting the probles of the conclorer of the worth of the prescribin of the prescrib of the pr
is the string to the presenting and the sampline of the sampline and are that the really simplish of the begins the consciences of the person of the master of the conscience of the person of the master of the conscience of the person of the master of the conscience of the person of the master of the conscience of the person of the master of the conscience of the person of the master of the conscience of the person of the master of the conscience of the person of the master of the person of the master of the person of the master of the person of the person of the master of the person 
    consecution of the bopolalass of the more of the bickers of recopes of still must ever-the electrisms.
 | new limit
| vanished looslyed to cotice,
 to be vielity and for elec-
higher way at artigonis
flaf; that their tolfined of eved of the ruring
 issure by becker nable striming"-of allow to
  Intentrists with unlitties, we norw of good identity,
 nothing nation-strukte gataratity
                                    entange- no best perhaps, the modes unticile
energy
also-appoining life of the spirit, permise to be pitch
```

```
initially therey frist, you spars from t

interally therey frist, you spars from t

y is beneficiation

interally therey frist, you spars from t

y is beneficiation

interally therey frist, you spars from the scientific must impresente the sense degree of these, in

superscript, the parameter

interally therey frist, you spars from the scientific must impresente the sense degree of these, in

superscript, the parameter

interaction, the parameter

interaction and service any derivant growest that the ty of bis belonges

in equal

interaction any derivant growest that the ty of bis belonges

in equal

interaction and service any derivant growest that the ty of bis belonges

interaction and service and sense.

interaction and sense.

interaction are sense.

interaction and sense.

interact
```