Lab 6

Aim: To implement a Recurrent Neural Network

```
docs = ['go india',
           'india india',
           'hip hip hurray',
           'india will win bro',
           'i love my india',
           'kohli is the greatest batsman',
           'sachin scored a lot of centuries',
           'dhoni won the second world cup for india',
           'will modi win the third term',
           'secular and diverse india'l
from keras.preprocessing.text import Tokenizer
#tokenizer = Tokenizer(oov token='<IoTSection>')
tokenizer = Tokenizer()
tokenizer.fit on texts(docs)
tokenizer.word index
{'india': 1,
 'the': 2,
 'hip': 3,
 'will': 4,
 'win': 5,
 'go': 6,
 'hurray': 7,
 'bro': 8,
 'i': 9,
 'love': 10,
 'my': 11,
 'kohli': 12,
 'is': 13,
 'greatest': 14,
 'batsman': 15,
 'sachin': 16,
 'scored': 17,
 'a': 18,
 'lot': 19,
 'of': 20,
 'centuries': 21,
 'dhoni': 22,
 'won': 23,
 'second': 24,
 'world': 25,
```

```
'cup': 26,
 'for': 27,
 'modi': 28,
 'third': 29,
 'term': 30,
 'secular': 31,
 'and': 32,
 'diverse': 33}
tokenizer.word_counts
OrderedDict([('go', 1),
                ('india', 7),
                ('hip', 2),
                ('hurray', 1),
('will', 2),
                ('win', 2),
('bro', 1),
                ('i', 1),
                ('love', 1),
                ('my', 1),
                ('kohli', 1),
                ('is', 1),
                ('the', 3),
                ('greatest', 1),
                ('batsman', 1),
                ('sachin', 1),
                ('scored', 1),
                ('a', 1),
                ('lot', 1),
('of', 1),
                ('centuries', 1),
                ('dhoni', 1),
                ('won', 1),
                ('second', 1),
                ('world', 1),
('cup', 1),
('for', 1),
('modi', 1),
                ('third', 1),
                ('term', 1),
                ('secular', 1),
                ('and', 1),
                ('diverse', 1)])
{\tt tokenizer.document\_count}
10
```

```
sequences = tokenizer.texts to sequences(docs)
sequences
[[6, 1],
 [1, 1],
 [3, 3, 7],
 [1, 4, 5, 8],
 [9, 10, 11, 1],
 [12, 13, 2, 14, 15],
 [16, 17, 18, 19, 20, 21],
 [22, 23, 2, 24, 25, 26, 27, 1],
 [4, 28, 5, 2, 29, 30],
 [31, 32, 33, 1]]
from keras.utils import pad sequences
sequences = pad_sequences(sequences,padding='post')
sequences
                  0,
                      0,
array([[ 6,
                          0,
                               0,
                                   0,
                                       0],
             1,
                      0,
                          0,
       [ 1,
              1,
                  0,
                               0,
                                   0,
                                       0],
                      0,
                          0,
       [ 3,
             3,
                  7,
                               0,
                                   0,
                                       0],
             4,
                      8,
                               0,
                  5,
       [ 1,
                          0,
                                   0,
                                       0],
                      1,
                          0,
                               0,
       [ 9, 10, 11,
                                   0,
                                       0],
       [12, 13,
                 2, 14, 15,
                               0,
                                   0,
                                       0],
       [16, 17, 18, 19, 20, 21,
                                   0,
                                       01,
                                  27,
       [22, 23, 2, 24, 25, 26,
                                       1],
                                   0,
                 5, 2, 29, 30,
       [ 4, 28,
                                       0],
                                   0,
       [31, 32, 33, 1, 0, 0,
                                       0]], dtype=int32)
len(sequences[0])
8
sequences
array([[ 6,
                  0,
                          0,
                               0,
                                   0,
                                       0],
             1,
                      0,
                      0,
             1,
       [ 1,
                  0,
                           0,
                               0,
                                   0,
                                       01,
       [ 3,
              3,
                  7,
                      0,
                           0,
                               0,
                                   0,
                                       0],
             4,
                  5,
                      8,
       [ 1,
                          0,
                               0,
                                   0,
                                       0],
                               0,
                                       0],
       [ 9, 10, 11,
                      1,
                          0,
                                   0,
                 2, 14, 15,
                                   0,
       [12, 13,
                               0,
                                       0],
       [16, 17, 18, 19, 20, 21,
                                   0,
                                       0],
       [22, 23, 2, 24, 25, 26,
                                  27,
                                       1],
       [ 4, 28,
                 5, 2, 29, 30,
                                   0,
                                       0],
                                   0,
       [31, 32, 33, 1, 0, 0,
                                       0]], dtype=int32)
from keras import Sequential
from keras.layers import Dense, SimpleRNN, Embedding
model = Sequential()
model.add(Embedding(34,output dim=2,input length=8))
```

```
model.add(Dense(units = 2, activation='softmax'))
model.summary()
Model: "sequential 4"
Laver (type)
                             Output Shape
                                                       Param #
                           _____
 embedding 4 (Embedding)
                          (None, 8, 2)
                                                       68
dense 4 (Dense)
                             (None, 8, 2)
                                                       6
Total params: 74 (296.00 Byte)
Trainable params: 74 (296.00 Byte)
Non-trainable params: 0 (0.00 Byte)
model.compile('adam', 'accuracy')
pred = model.predict(sequences)
print(pred)
pred.shape
1/1 [======= ] - 0s 244ms/step
[[[0.5128169 0.4871832 ]
  [0.50593483 0.49406517]
  [0.47728473 0.5227153 ]
  [0.47728473 0.5227153 ]
  [0.47728473 0.5227153 ]
  [0.47728473 0.5227153 ]
  [0.47728473 0.5227153 ]
  [0.47728473 0.5227153 11
 [[0.50593483 0.49406517]
  [0.50593483 0.49406517]
  [0.47728473 0.5227153 ]
  [0.47728473 0.5227153 ]
  [0.47728473 0.5227153 ]
  [0.47728473 0.5227153 ]
  [0.47728473 0.5227153 ]
  [0.47728473 0.5227153 ]]
 [[0.5012797 0.49872032]
  [0.5012797 0.49872032]
  [0.48481685 0.51518315]
  [0.47728473 0.5227153 ]
  [0.47728473 0.5227153 ]
  [0.47728473 0.5227153 ]
  [0.47728473 0.5227153 ]
  [0.47728473 0.5227153 ]]
```

```
[[0.50593483 0.49406517]
[0.5035462
             0.496453851
 [0.4906695
             0.5093305 1
 [0.49849063 0.50150937]
 [0.47728473 0.5227153 ]
 [0.47728473 0.5227153 ]
 [0.47728473 0.5227153 ]
 [0.47728473 0.5227153 ]]
[[0.49348876 0.5065113 ]
 [0.51095986 0.48904002]
 [0.5139852
             0.486014781
 [0.50593483 0.49406517]
 [0.47728473 0.5227153 ]
 [0.47728473 0.5227153 ]
 [0.47728473 0.5227153 ]
 [0.47728473 0.5227153 ]]
[[0.4952051
             0.5047948 1
 [0.5148093
             0.485190661
 [0.4865776]
             0.5134225 ]
 [0.4873007
             0.512699371
 [0.50977427 0.49022582]
 [0.47728473 0.5227153 ]
 [0.47728473 0.5227153 ]
 [0.47728473 0.5227153 ]]
[[0.4775283
             0.522471671
 [0.5062023
             0.493797661
 [0.5092295
             0.490770521
 [0.50273633 0.49726364]
 [0.5161321
             0.483867881
 [0.48259777 0.5174022 ]
 [0.47728473 0.5227153 ]
[0.47728473 0.5227153 ]]
[[0.50934523 0.49065474]
 [0.5105865
             0.489413561
 [0.4865776
             0.5134225 ]
 [0.50501454 0.4949855 ]
 [0.5084248
             0.4915752 1
 [0.48253918 0.5174609 ]
 [0.5083535
             0.4916465 1
 [0.50593483 0.49406517]]
[[0.5035462
             0.496453851
 [0.51851344 0.4814865 ]
 [0.4906695
             0.5093305 1
 [0.4865776
             0.5134225 ]
```

```
[0.48617196 0.513828 ]
  [0.5031508 0.49684924]
  [0.47728473 0.5227153 ]
  [0.47728473 0.5227153 ]]
 [[0.49673644 0.5032636 ]
  [0.5091956 0.49080437]
  [0.4768464 0.5231536 ]
  [0.50593483 0.49406517]
  [0.47728473 0.5227153 ]
  [0.47728473 0.5227153 ]
  [0.47728473 0.5227153 ]
  [0.47728473 0.5227153 ]]]
(10, 8, 2)
from keras.datasets import imdb
from keras import Sequential
from keras.layers import Dense, SimpleRNN, Embedding, Flatten
from keras.preprocessing.text import Tokenizer
from keras.utils import pad sequences
(X_train,y_train),(X_test,y_test) = imdb.load_data()
Downloading data from https://storage.googleapis.com/tensorflow/tf-
keras-datasets/imdb.npz
X train[0]
[1,
14,
22,
16,
43,
530,
973,
1622,
1385,
65,
458,
4468,
66,
3941,
4,
173,
36,
256.
5,
25,
 100,
```

```
43,
838,
112,
50,
670,
22665,
9,
35,
480,
284,
5,
150,
4,
172,
112,
167,
21631,
336,
385,
39,
4,
172,
4536,
1111,
17,
546,
38,
13,
447,
4,
192,
50,
16,
6,
147,
2025,
19,
14,
22,
4,
1920,
4613,
469,
4,
22,
71,
87,
12,
16,
```

```
43,
530,
38,
76,
15,
13,
1247,
4, 22,
17,
515,
17,
12,
16,
626,
18,
19193,
5,
62,
386,
12,
8,
316,
8,
106,
5,
4,
2223,
5244,
16,
480,
66,
3785,
33,
4,
130,
12,
16,
38,
619,
5,
25,
124,
51,
36,
135,
48,
25,
1415,
```

```
33,
6,
22,
12,
215,
28,
77,
52,
5,
14,
407,
16,
82,
10311,
8,
107,
117,
5952,
15,
256,
4,
31050,
7,
3766,
5,
723,
36,
71,
43,
530,
476,
26,
400,
317,
46,
7,
12118,
1029,
13,
104,
88,
4,
381,
15,
297,
98,
32,
```

```
2071,
56,
26,
141,
6,
194,
7486,
18,
4,
226,
22,
21,
134,
476,
26,
480,
5,
144,
30,
5535,
18,
51,
36,
28,
224,
92,
25,
104,
4,
226,
65,
16,
38,
1334,
88,
12,
16,
283,
5,
16,
4472,
113,
103,
32,
15,
16,
5345,
19,
```

```
178,
 32]
import numpy as np
max(X_train)
[1,
88325,
 733,
 7,
14,
 706,
 40354,
 1936,
 4,
4423,
 12322,
 23,
4,
419,
 97,
 252,
 25,
 332,
 12,
 11,
 420,
 25,
 92,
 6,
 604,
 7,
 3368,
 1828,
 125,
 83,
 4,
 933,
 1411,
7,
4,
 6024,
 10805,
 8,
 26519,
 187,
 19,
 6,
762,
 7,
```

```
6226,
11,
35,
1626,
6816,
4,
194,
2020,
223,
152,
193,
7880,
8,
14,
5,
4757,
8,
1261,
125,
4,
768,
31,
34,
31,
2256,
17683,
194,
2020,
309,
47,
6,
8849,
4191,
40,
393,
21,
9,
329,
629,
444,
74,
1335,
39,
1335,
5,
4,
48333,
91,
7,
```

```
4,
20,
47,
4,
1639,
4173,
156,
4470,
8,
30,
3368,
45514,
187,
11,
4,
1411,
5,
59784,
245,
14,
9,
6,
57,
352,
20,
11,
63,
55,
117,
571,
33,
222,
23,
268,
75,
81,
79,
8,
106,
4,
13537,
7,
4,
5186,
193,
6,
3014,
137,
31,
```

```
7,
 41,
 915,
 9732,
 3631,
 21,
 164,
 266,
 7,
 14]
X_train = pad_sequences(X_train,padding='post',maxlen=50)
X test = pad sequences(X test,padding='post',maxlen=50)
import numpy as np
np.max(X train)
88585
X train.shape
(25000, 50)
model = Sequential()
model.add(Embedding(90000,output dim=2,input length=50))
#model.summary()
model.add(SimpleRNN(32, return_sequences=False))
model.add(Dense(1,activation='sigmoid'))
model.summary()
Model: "sequential_5"
Layer (type)
                              Output Shape
                                                         Param #
 embedding 5 (Embedding)
                              (None, 50, 2)
                                                         180000
 simple rnn (SimpleRNN)
                              (None, 32)
                                                         1120
```

dense_5 (Dense) (None, 1) 33

Total params: 181153 (707.63 KB)
Trainable params: 181153 (707.63 KB)
Non-trainable params: 0 (0.00 Byte)

```
model.compile(loss='binary crossentropy',optimizer='adam',metrics=['ac
c'1)
model.fit(X train,y train,epochs=5,validation data=(X test,y test))
Epoch 1/5
782/782 [============= ] - 21s 25ms/step - loss:
0.5955 - acc: 0.6541 - val loss: 0.4638 - val acc: 0.7830
Epoch 2/5
782/782 [============= ] - 18s 23ms/step - loss:
0.3627 - acc: 0.8446 - val loss: 0.4292 - val acc: 0.8122
Epoch 3/5
782/782 [============= ] - 18s 24ms/step - loss:
0.2454 - acc: 0.9059 - val loss: 0.4469 - val acc: 0.8038
Epoch 4/5
0.1739 - acc: 0.9387 - val_loss: 0.6181 - val_acc: 0.7861
Epoch 5/5
0.1305 - acc: 0.9552 - val_loss: 0.6063 - val_acc: 0.7889
<keras.src.callbacks.History at 0x7ca2cc136140>
```