LAB-14:TEXT DATASET CREATION AND DESIGN OF SIMPLE RNN FOR SENTIMENT ANALYSIS

```
In [1]: import csv
         import tensorflow as tf
         import numpy as np
         from tensorflow.keras.preprocessing.text import Tokenizer
         from tensorflow.keras.preprocessing.sequence import pad_sequences
         from nltk.corpus import stopwords
         STOPWORDS = set(stopwords.words ('english'))
         import pandas as pd
In [2]: import nltk
         nltk.download('stopwords')
         [nltk_data] Downloading package stopwords to
         [nltk data]
                           C:\Users\2mscdsa28\AppData\Roaming\nltk data...
                        Package stopwords is already up-to-date!
         [nltk data]
Out[2]: True
In [3]: df=pd.read csv('movie.csv')
In [4]: | df.head()
Out[4]:
             LABEL
                                                       TEXT
          0
                 1 films adapted from comic books have had plenty...
          1
                 1 to say moore and campbell thoroughly researche...
                 1
          2
                        in other words, don't dismiss this film becau...
          3
                      if you can get past the whole comic book thing...
                 1
                         the ghetto in question is, of course, whitec...
In [5]: | df.columns
Out[5]: Index(['LABEL', 'TEXT'], dtype='object')
In [6]: | df.size
Out[6]: 40
```

```
In [7]: y = df['LABEL']
x=[]
for review in df['TEXT']:
    filtered_sentence=[w.lower() for w in review.split() if not w in STOPWORDS]
    x.append(filtered_sentence)
x = pd.Series(x)
```

```
5. Dataset prepartaion
 In [8]: | from sklearn.model_selection import train_test_split
         X_train,X_val,y_train,y_val=train_test_split(x,y,train_size=0.7)
 In [9]: print(X_train.shape)
         print(X_val.shape)
         print(y_train.shape)
         print(y_val.shape)
         (14,)
         (6,)
         (14,)
         (6,)
In [10]: train_token = Tokenizer(num_words = 50,oov_token='<oov>')
         train_token.fit_on_texts(X_train)
         word index = train token.word index
         train_sequences=train_token.texts_to_sequences(X_train)
         dict(list(word_index.items())[0:10])
Out[10]: {'<oov>': 1,
           '.': 2,
          ',': 3,
           'nightclub': 4,
           'comic': 5,
           'story': 6,
           'getting': 7,
           'trouble': 8,
           'abberline': 9,
           'widower': 10}
In [11]: vocab_ze=len(train_token.word_index)+1
         vocab_ze
Out[11]: 97
In [12]: train_sequences[4]
Out[12]: [30, 31, 32, 3, 2, 2, 2, 33, 2]
In [13]: train_padded = pad_sequences(train_sequences,maxlen=100,padding='post')
```

```
In [14]: train padded[8]
Out[14]: array([1, 1, 1, 5, 1, 1, 3, 1, 1, 1, 1, 1, 1, 1, 3, 1, 1, 1, 2, 0, 0, 0,
              0, 0, 0, 0, 0, 0, 0, 0, 0, 0, 0])
In [15]: train_padded.shape
Out[15]: (14, 100)
In [16]: val token = Tokenizer(num words = 50,oov token='<oov>')
        val_token.fit_on_texts(X_val)
        val index = val token.word index
        val sequences=train token.texts to sequences(X val)
In [17]: val sequences[4]
Out[17]: [1, 1, 1, 3, 1, 1, 1, 2]
In [18]: val padded=pad sequences(val sequences,maxlen=100,padding='post')
In [19]: val_padded[2]
                                           0,
                                                            0,
Out[19]: array([ 1,
                      3, 44,
                             2,
                                0,
                                    0,
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                 1,
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In [20]: from tensorflow.keras.preprocessing.sequence import pad sequences
        from keras.models import Sequential
        from keras.layers import Dense, Embedding, SimpleRNN
        6.Model creation
In [21]: |model = Sequential()
        # Embedding Layer
        model.add(Embedding(300,70,input length=100))
        model.add(SimpleRNN(70,activation='relu'))
```

In [22]: model.compile(optimizer='adam',loss='binary_crossentropy',metrics=['accuracy'])

model.add(Dense('1',activation='sigmoid'))

In [23]: model.summary()

Model: "sequential"

Layer (type)	Output Shape	Param #
embedding (Embedding)	(None, 100, 70)	21000
<pre>simple_rnn (SimpleRNN)</pre>	(None, 70)	9870
dense (Dense)	(None, 1)	71

Total params: 30,941 Trainable params: 30,941 Non-trainable params: 0

In [24]: history=model.fit(train_padded,y_train,epochs=10,verbose=2,batch_size=15)

```
Epoch 1/10
1/1 - 1s - loss: 0.6909 - accuracy: 0.5714 - 746ms/epoch - 746ms/step
Epoch 2/10
1/1 - 0s - loss: 0.6888 - accuracy: 0.5714 - 10ms/epoch - 10ms/step
Epoch 3/10
1/1 - 0s - loss: 0.6872 - accuracy: 0.5714 - 12ms/epoch - 12ms/step
Epoch 4/10
1/1 - 0s - loss: 0.6858 - accuracy: 0.5714 - 8ms/epoch - 8ms/step
Epoch 5/10
1/1 - 0s - loss: 0.6847 - accuracy: 0.5714 - 8ms/epoch - 8ms/step
Epoch 6/10
1/1 - 0s - loss: 0.6837 - accuracy: 0.5714 - 8ms/epoch - 8ms/step
Epoch 7/10
1/1 - 0s - loss: 0.6831 - accuracy: 0.5714 - 8ms/epoch - 8ms/step
Epoch 8/10
1/1 - 0s - loss: 0.6829 - accuracy: 0.5714 - 8ms/epoch - 8ms/step
Epoch 9/10
1/1 - 0s - loss: 0.6833 - accuracy: 0.5714 - 8ms/epoch - 8ms/step
Epoch 10/10
1/1 - 0s - loss: 0.6836 - accuracy: 0.5714 - 8ms/epoch - 8ms/step
```

In [25]: model.summary()

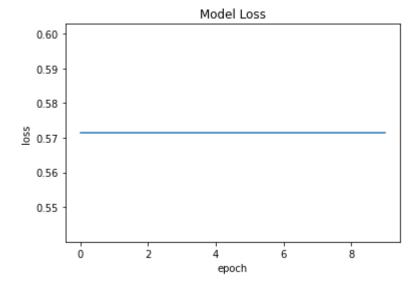
Model: "sequential"

100, 70)	21000
70)	9870
1)	71
7	0)

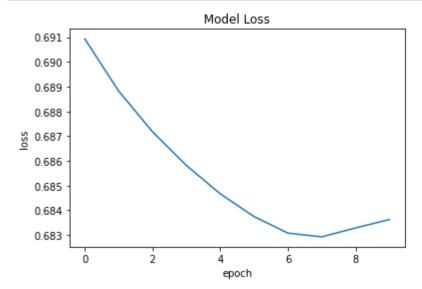
Total params: 30,941 Trainable params: 30,941 Non-trainable params: 0

```
In [26]: import matplotlib.pyplot as plt
```

```
In [27]: plt.plot(history.history['accuracy'])
    plt.title('Model Loss')
    plt.ylabel('loss')
    plt.xlabel('epoch')
    plt.show()
```



```
In [28]: plt.plot(history.history['loss'])
    plt.title('Model Loss')
    plt.ylabel('loss')
    plt.xlabel('epoch')
    plt.show()
```



```
In [29]: text = ['Education is what remains after one has forgotten what one has learned i
          #sent = [w.lower() for w in text.split() if not w in STOPWORDS]
          trail_token = Tokenizer()
          trail_token.fit_on_texts(text)
          #word index = trail token.word index
          trail seq = trail token.texts to sequences(text)
          #dict(list(word_index.items())[0:10])
          trail_pad = pad_sequences(trail_seq,maxlen=100,padding='post')
In [30]: trail_pad
Out[30]: array([[ 4,
                        5,
                                                               3,
                            1,
                                 6,
                                     7,
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                            0,
                                 0]])
```

```
In [31]: res = model.predict(trail pad)
         label = ['positive', 'negative']
         print(res,label[np.argmax(trail_pad)>50])
         [[0.59026337]] positive
         C:\Users\2mscdsa28\AppData\Local\Temp\ipykernel 12452\1535703611.py:3: Deprecat
         ionWarning: In future, it will be an error for 'np.bool_' scalars to be interpr
         eted as an index
           print(res,label[np.argmax(trail pad)>50])
In [32]: model1 = Sequential()
         # Embedding Layer
         model1.add(Embedding(5000,64,input length=100))
         model1.add(SimpleRNN(32,activation='tanh'))
         model1.add(Embedding(5000,32,input_length=100))
         model1.add(SimpleRNN(32,activation='tanh' ))
         model1.add(Dense('1',activation='sigmoid'))
In [33]: model1.summary()
         Model: "sequential 1"
          Layer (type)
                                     Output Shape
                                                              Param #
                                     (None, 100, 64)
          embedding_1 (Embedding)
                                                              320000
          simple rnn 1 (SimpleRNN)
                                     (None, 32)
                                                              3104
                                     (None, 32, 32)
          embedding_2 (Embedding)
                                                              160000
                                     (None, 32)
          simple rnn 2 (SimpleRNN)
                                                              2080
          dense 1 (Dense)
                                     (None, 1)
                                                              33
         Total params: 485,217
         Trainable params: 485,217
         Non-trainable params: 0
In [34]: model1.compile(optimizer='adam',loss='binary_crossentropy',metrics=['accuracy'])
```

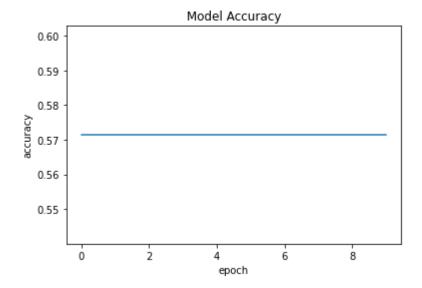
```
In [35]: history1=model1.fit(train padded,y train,epochs=10,verbose=2,batch size=15)
         Epoch 1/10
         WARNING:tensorflow:Gradients do not exist for variables ['embedding 1/embedding
         s:0', 'simple_rnn_1/simple_rnn_cell_1/kernel:0', 'simple_rnn_1/simple_rnn_cell_
         1/recurrent_kernel:0', 'simple_rnn_1/simple_rnn_cell_1/bias:0'] when minimizing
         the loss. If you're using `model.compile()`, did you forget to provide a `loss`
         argument?
         WARNING:tensorflow:Gradients do not exist for variables ['embedding 1/embedding
         s:0', 'simple_rnn_1/simple_rnn_cell_1/kernel:0', 'simple_rnn_1/simple_rnn_cell_
         1/recurrent_kernel:0', 'simple_rnn_1/simple_rnn_cell_1/bias:0'] when minimizing
         the loss. If you're using `model.compile()`, did you forget to provide a `loss`
         argument?
         1/1 - 1s - loss: 0.6840 - accuracy: 0.5714 - 754ms/epoch - 754ms/step
         Epoch 2/10
         1/1 - 0s - loss: 0.6853 - accuracy: 0.5714 - 6ms/epoch - 6ms/step
         Epoch 3/10
         1/1 - 0s - loss: 0.6839 - accuracy: 0.5714 - 6ms/epoch - 6ms/step
         Epoch 4/10
         1/1 - 0s - loss: 0.6829 - accuracy: 0.5714 - 4ms/epoch - 4ms/step
         Epoch 5/10
         1/1 - 0s - loss: 0.6832 - accuracy: 0.5714 - 5ms/epoch - 5ms/step
         Epoch 6/10
         1/1 - 0s - loss: 0.6835 - accuracy: 0.5714 - 6ms/epoch - 6ms/step
         Epoch 7/10
         1/1 - 0s - loss: 0.6833 - accuracy: 0.5714 - 5ms/epoch - 5ms/step
         Epoch 8/10
         1/1 - 0s - loss: 0.6830 - accuracy: 0.5714 - 5ms/epoch - 5ms/step
         Epoch 9/10
         1/1 - 0s - loss: 0.6830 - accuracy: 0.5714 - 5ms/epoch - 5ms/step
         Epoch 10/10
         1/1 - 0s - loss: 0.6832 - accuracy: 0.5714 - 4ms/epoch - 4ms/step
```

```
In [36]: |model1.evaluate(val_padded,y_val)
```

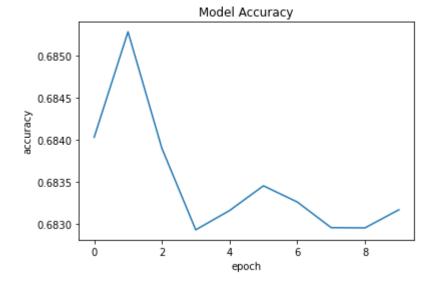
```
1/1 [================== ] - 0s 293ms/step - loss: 0.7401 - accuracy:
0.3333
```

Out[36]: [0.7401018142700195, 0.3333333432674408]

```
In [37]: plt.plot(history1.history['accuracy'])
    plt.title('Model Accuracy')
    plt.ylabel('accuracy')
    plt.xlabel('epoch')
    plt.show()
```



```
In [38]: plt.plot(history1.history['loss'])
    plt.title('Model Accuracy')
    plt.ylabel('accuracy')
    plt.xlabel('epoch')
    plt.show()
```



```
In [40]: model2 = Sequential()
# Embedding Layer
model2.add(Embedding(4000,128,input_length=100))
model2.add(SimpleRNN(64,activation='tanh'))
model2.add(Embedding(4000,128,input_length=100))
model2.add(SimpleRNN(64,activation='relu'))
model2.add(Embedding(4000,128,input_length=100))
model2.add(SimpleRNN(64,activation='tanh'))
model2.add(Dense('1',activation='sigmoid'))
```

In [41]: model2.summary()

Model: "sequential_2"

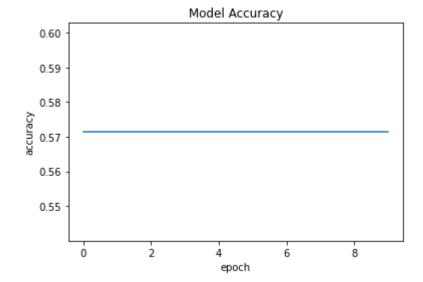
embedding_3 (Embedding) (None, 100, 128) 512000 simple_rnn_3 (SimpleRNN) (None, 64) 12352 embedding_4 (Embedding) (None, 64, 128) 512000 simple_rnn_4 (SimpleRNN) (None, 64) 12352
embedding_4 (Embedding) (None, 64, 128) 512000
simple_rnn_4 (SimpleRNN) (None, 64) 12352
embedding_5 (Embedding) (None, 64, 128) 512000
simple_rnn_5 (SimpleRNN) (None, 64) 12352
dense_2 (Dense) (None, 1) 65

Total params: 1,573,121 Trainable params: 1,573,121 Non-trainable params: 0

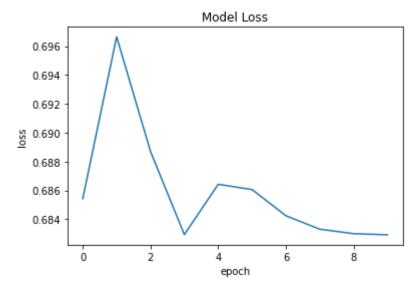
```
In [43]: history2=model2.fit(train_padded,y_train,epochs=10,verbose=2,batch_size=15)
         Epoch 1/10
         WARNING:tensorflow:Gradients do not exist for variables ['embedding_3/embedding
         s:0', 'simple_rnn_3/simple_rnn_cell_3/kernel:0', 'simple_rnn_3/simple_rnn_cell_
         3/recurrent_kernel:0', 'simple_rnn_3/simple_rnn_cell_3/bias:0', 'embedding_4/em
         beddings:0', 'simple_rnn_4/simple_rnn_cell_4/kernel:0', 'simple_rnn_4/simple_rn
         n_cell_4/recurrent_kernel:0', 'simple_rnn_4/simple_rnn_cell_4/bias:0'] when min
         imizing the loss. If you're using `model.compile()`, did you forget to provide
         a `loss` argument?
         WARNING:tensorflow:Gradients do not exist for variables ['embedding_3/embedding
         s:0', 'simple_rnn_3/simple_rnn_cell_3/kernel:0', 'simple_rnn_3/simple_rnn_cell_
         3/recurrent_kernel:0', 'simple_rnn_3/simple_rnn_cell_3/bias:0', 'embedding_4/em
         beddings:0', 'simple_rnn_4/simple_rnn_cell_4/kernel:0', 'simple_rnn_4/simple_rn
         n cell 4/recurrent kernel:0', 'simple rnn 4/simple rnn cell 4/bias:0'] when min
         imizing the loss. If you're using `model.compile()`, did you forget to provide
         a `loss` argument?
         1/1 - 1s - loss: 0.6854 - accuracy: 0.5714 - 834ms/epoch - 834ms/step
         Epoch 2/10
         1/1 - 0s - loss: 0.6966 - accuracy: 0.5714 - 10ms/epoch - 10ms/step
         Epoch 3/10
         1/1 - 0s - loss: 0.6887 - accuracy: 0.5714 - 10ms/epoch - 10ms/step
         Epoch 4/10
         1/1 - 0s - loss: 0.6829 - accuracy: 0.5714 - 9ms/epoch - 9ms/step
         Epoch 5/10
         1/1 - 0s - loss: 0.6864 - accuracy: 0.5714 - 10ms/epoch - 10ms/step
         Epoch 6/10
         1/1 - 0s - loss: 0.6861 - accuracy: 0.5714 - 10ms/epoch - 10ms/step
         Epoch 7/10
         1/1 - 0s - loss: 0.6842 - accuracy: 0.5714 - 10ms/epoch - 10ms/step
         Epoch 8/10
         1/1 - 0s - loss: 0.6833 - accuracy: 0.5714 - 9ms/epoch - 9ms/step
         Epoch 9/10
         1/1 - 0s - loss: 0.6830 - accuracy: 0.5714 - 10ms/epoch - 10ms/step
         Epoch 10/10
         1/1 - 0s - loss: 0.6829 - accuracy: 0.5714 - 10ms/epoch - 10ms/step
In [44]: |model2.evaluate(val_padded,y_val)
         1/1 [=============== ] - 0s 240ms/step - loss: 0.7520 - accuracy:
         0.3333
```

Out[44]: [0.7519607543945312, 0.3333333432674408]

```
In [45]: plt.plot(history2.history['accuracy'])
    plt.title('Model Accuracy')
    plt.ylabel('accuracy')
    plt.xlabel('epoch')
    plt.show()
```



```
In [46]: plt.plot(history2.history['loss'])
    plt.title('Model Loss')
    plt.ylabel('loss')
    plt.xlabel('epoch')
    plt.show()
```



```
In [47]: res = model2.predict(trail_pad)
label = ['positive', 'negative']
print(res,label[np.argmax(trail_pad)>50])
```

```
1/1 [=======] - 0s 315ms/step [[0.5720007]] positive
```

C:\Users\2mscdsa28\AppData\Local\Temp\ipykernel_12452\479605851.py:3: Deprecati onWarning: In future, it will be an error for 'np.bool_' scalars to be interpre ted as an index

print(res,label[np.argmax(trail_pad)>50])