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PDL Lab15: Text dataset creation and design of Simple RNN for Sentiment Analysis

1. Import libraries

```
In [1]:
                                                                                        M
import csv
import tensorflow as tf
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
from tensorflow.keras.preprocessing.text import Tokenizer
from tensorflow.keras.preprocessing.sequence import pad_sequences
In [2]:
                                                                                        M
from keras.models import Sequential
from keras.layers import Dense, Embedding, SimpleRNN
                                                                                        M
In [3]:
import nltk
nltk.download('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!
Out[3]:
True
In [4]:
                                                                                        H
from nltk.corpus import stopwords
STOPWORDS = set(stopwords.words('english'))
```

2. Creation of data

```
In [6]:
                                                                                                       M
df = pd.read_csv("data.csv",encoding="ISO-8859-1")
In [7]:
                                                                                                        M
df.head()
Out[7]:
   Labels
                                              Quotes
 0
           Look deep into nature, and then you will under...
                          Colors are the smiles of nature
 1
        0
 2
        0
                     I believe in God, only I spell it Nature
 3
           Nature does not hurry, yet everything is accom...
               If you truly love nature, you will find beauty...
 4
3. Opening your CSV file
In [8]:
                                                                                                       M
import csv
In [9]:
                                                                                                        M
file = open('data.csv')
type(file)
Out[9]:
_io.TextIOWrapper
                                                                                                       H
In [10]:
csvreader = csv.reader(file)
In [11]:
                                                                                                        M
header = []
header = next(csvreader)
header
```

Out[11]:

['Labels', 'Quotes']

```
In [24]:
                                                                                       M
with open('data.csv', 'r') as file:
    csvreader = csv.reader(file)
    rows = []
    for row in csvreader:
        rows.append(row)
row
Out[24]:
['1',
 'Technology brings convenience but often disconnects us from the natural
world']
In [25]:
                                                                                       M
file.close()
In [26]:
df.info()
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 12 entries, 0 to 11
Data columns (total 2 columns):
    Column Non-Null Count Dtype
0
     Labels 12 non-null
                             int64
     Quotes 12 non-null
                             object
dtypes: int64(1), object(1)
memory usage: 320.0+ bytes
4. Pre-processing the text
                                                                                       M
In [27]:
y = df['Labels']
X = df['Quotes']
5. Dataset Preparation
In [28]:
                                                                                       H
from sklearn.model selection import train test split
X_train, X_val, y_train, y_val = train_test_split(X, y, test_size=0.3)
```

```
In [29]:
                                                                                          M
print(X_train.shape)
print(X_val.shape)
print(y_train.shape)
print(y_val.shape)
(8,)
(4,)
(8,)
(4,)
                                                                                          H
In [30]:
# 4th step to be continue
train_token = Tokenizer(num_words=100,oov_token='<oov>')
train_token.fit_on_texts(X_train)
word_index = train_token.word_index
train_sequence = train_token.texts_to_sequences(X_train)
dict(list(word_index.items())[0:10])
Out[30]:
{'<oov>': 1,
 'the': 2,
 'is': 3,
 'of': 4,
 'nature': 5,
 'you': 6,
 'but': 7,
 'into': 8,
 'a': 9,
 'will': 10}
                                                                                          M
In [31]:
vocab = len(train_token.word_index) + 1
vocab
Out[31]:
63
In [32]:
                                                                                          M
train_sequence[3]
Out[32]:
[28, 29, 8, 5, 30, 31, 6, 10, 32, 33, 34]
```

```
In [33]:
                                                                                             M
train padded = pad sequences(train sequence, maxlen=100, padding='post')
train_padded[5]
Out[33]:
array([41, 3, 11, 42, 12, 43, 2, 44, 7, 12, 45, 46, 47,
                                                                2, 48,
0,
                     0,
                          0,
                              0,
                                  0,
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                                  0,
                                       0,
                                               0,
                                                    0,
                                                        0,
                                                            0,
                                                                     0])
In [34]:
                                                                                             M
train_padded.shape
Out[34]:
(8, 100)
                                                                                             M
In [35]:
val_token = Tokenizer(num_words=500,oov_token='<oov>')
val_token.fit_on_texts(X_val)
val_index = val_token.word_index
val_sequence = val_token.texts_to_sequences(X_val)
In [37]:
                                                                                             M
if len(val sequence) > 4:
    # Access the element at index 4
    value = val_sequence[4]
else:
    # Handle the case where the list doesn't have enough elements
    print("The list doesn't have enough elements.")
The list doesn't have enough elements.
In [38]:
                                                                                             M
val padded = pad sequences(val sequence,maxlen=100,padding='post')
```

6. Model Creation

```
In [40]:

model = Sequential()
# Embedding Layer
model.add(Embedding(300,70,input_length=100))
model.add(SimpleRNN(70,activation='relu'))
model.add(Dense('1',activation='sigmoid'))

In [41]:

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

In [42]:

model.summary()
```

Model: "sequential"

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

Total params: 30941 (120.86 KB)

Trainable params: 30941 (120.86 KB) Non-trainable params: 0 (0.00 Byte)

```
In [43]: ▶
```

```
history=model.fit(train_padded,y_train,epochs=10,verbose=2,batch_size=15)
```

```
Epoch 1/10
1/1 - 3s - loss: 0.6932 - accuracy: 0.5000 - 3s/epoch - 3s/step
Epoch 2/10
1/1 - 0s - loss: 0.6932 - accuracy: 0.5000 - 33ms/epoch - 33ms/step
Epoch 3/10
1/1 - 0s - loss: 0.6932 - accuracy: 0.5000 - 25ms/epoch - 25ms/step
Epoch 4/10
1/1 - 0s - loss: 0.6932 - accuracy: 0.5000 - 25ms/epoch - 25ms/step
Epoch 5/10
1/1 - 0s - loss: 0.6932 - accuracy: 0.5000 - 35ms/epoch - 35ms/step
Epoch 6/10
1/1 - 0s - loss: 0.6932 - accuracy: 0.5000 - 31ms/epoch - 31ms/step
Epoch 7/10
1/1 - 0s - loss: 0.6932 - accuracy: 0.5000 - 45ms/epoch - 45ms/step
Epoch 8/10
1/1 - 0s - loss: 0.6931 - accuracy: 0.5000 - 32ms/epoch - 32ms/step
Epoch 9/10
1/1 - 0s - loss: 0.6932 - accuracy: 0.5000 - 30ms/epoch - 30ms/step
Epoch 10/10
1/1 - 0s - loss: 0.6932 - accuracy: 0.5000 - 27ms/epoch - 27ms/step
```

In [44]: ▶

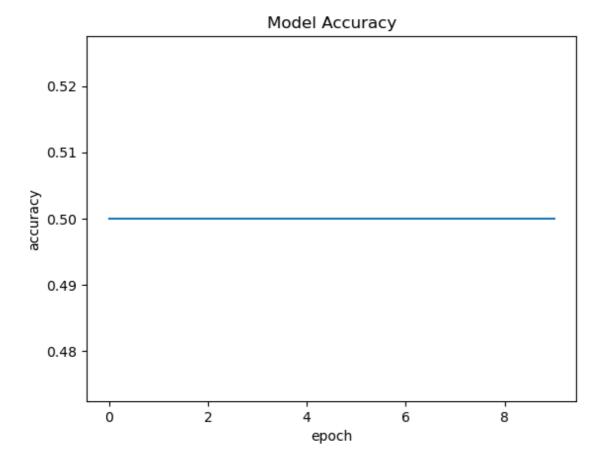
```
model.evaluate(val_padded,y_val)
```

Out[44]:

[0.693149745464325, 0.5]

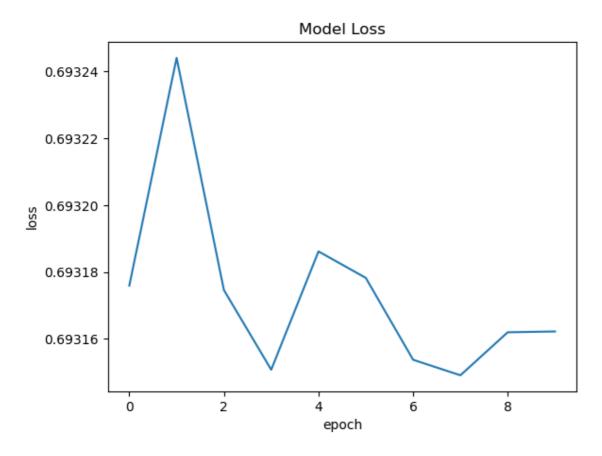
In [45]: ▶

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



```
In [46]: ▶
```

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



```
In [47]:
text = df["Quotes"]
```

```
In [48]: ▶
```

```
#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 [49]:
trail_pad
```

```
Out[49]:
```

```
array([[20, 21, 10, ..., 0, 0, 0], [26, 27, 1, ..., 0, 0, 0], [13, 29, 30, ..., 0, 0, 0], ..., [1, 66, 18, ..., 0, 0, 0], [2, 4, 71, ..., 0, 0, 0], [77, 78, 79, ..., 0, 0, 0]])
```

Step-7 and 8:

Step-8:

In [51]: ▶

```
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 [52]: ▶

```
model1.summary()
```

Model: "sequential_1"

Layer (type)	Output Shape	Param #		
embedding_1 (Embedding)	(None, 100, 64)	320000		
<pre>simple_rnn_1 (SimpleRNN)</pre>	(None, 32)	3104		
<pre>embedding_2 (Embedding)</pre>	(None, 32, 32)	160000		
simple_rnn_2 (SimpleRNN)	(None, 32)	2080		
dense_1 (Dense)	(None, 1)	33		
======================================				

Non-trainable params: 0 (0.00 Byte)

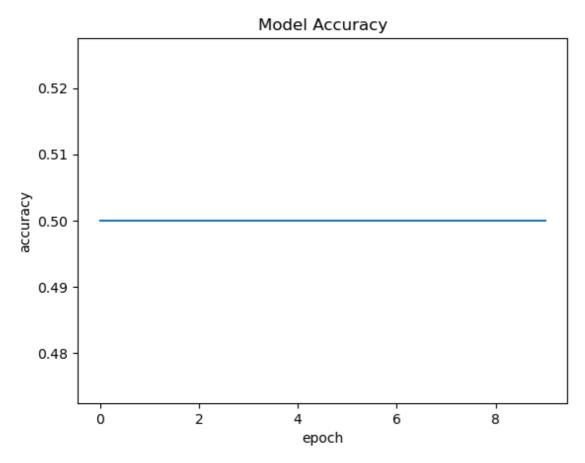
```
9/24/23, 9:14 PM
                                         225229103 PDL LAB 15 - Jupyter Notebook
  In [53]:
 model1.compile(optimizer='adam',loss='binary_crossentropy',metrics=['accuracy'])
  In [54]:
 history1=model1.fit(train padded,y train,epochs=10,verbose=2,batch size=15)
 Epoch 1/10
 WARNING:tensorflow:Gradients do not exist for variables ['embedding_1/emb
  eddings:0', 'simple_rnn_1/simple_rnn_cell/kernel:0', 'simple_rnn_1/simple
  _rnn_cell/recurrent_kernel:0', 'simple_rnn_1/simple_rnn_cell/bias:0'] whe
 n 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/emb
  eddings:0', 'simple_rnn_1/simple_rnn_cell/kernel:0', 'simple_rnn_1/simple
  _rnn_cell/recurrent_kernel:0', 'simple_rnn_1/simple_rnn_cell/bias:0'] whe
  n 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/emb
 eddings:0', 'simple_rnn_1/simple_rnn_cell/kernel:0', 'simple_rnn_1/simple
  _rnn_cell/recurrent_kernel:0', 'simple_rnn_1/simple_rnn_cell/bias:0'] whe
 n 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/emb
```

eddings:0', 'simple_rnn_1/simple_rnn_cell/kernel:0', 'simple_rnn_1/simple _rnn_cell/recurrent_kernel:0', 'simple_rnn_1/simple_rnn_cell/bias:0'] whe n minimizing the loss. If you're using `model.compile()`, did you forget to provide a `loss` argument?

```
1/1 - 3s - loss: 0.6933 - accuracy: 0.5000 - 3s/epoch - 3s/step
Epoch 2/10
1/1 - 0s - loss: 0.6933 - accuracy: 0.5000 - 26ms/epoch - 26ms/step
Epoch 3/10
1/1 - 0s - loss: 0.6932 - accuracy: 0.5000 - 17ms/epoch - 17ms/step
Epoch 4/10
1/1 - 0s - loss: 0.6931 - accuracy: 0.5000 - 27ms/epoch - 27ms/step
Epoch 5/10
1/1 - 0s - loss: 0.6932 - accuracy: 0.5000 - 17ms/epoch - 17ms/step
Epoch 6/10
1/1 - 0s - loss: 0.6932 - accuracy: 0.5000 - 15ms/epoch - 15ms/step
Epoch 7/10
1/1 - 0s - loss: 0.6932 - accuracy: 0.5000 - 16ms/epoch - 16ms/step
Epoch 8/10
1/1 - 0s - loss: 0.6931 - accuracy: 0.5000 - 22ms/epoch - 22ms/step
Epoch 9/10
1/1 - 0s - loss: 0.6932 - accuracy: 0.5000 - 22ms/epoch - 22ms/step
Epoch 10/10
1/1 - 0s - loss: 0.6932 - accuracy: 0.5000 - 28ms/epoch - 28ms/step
```

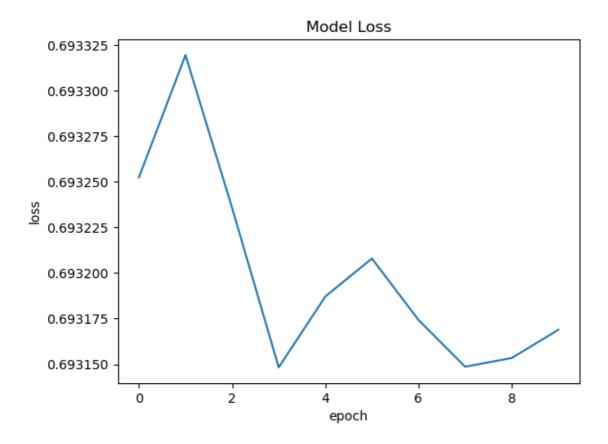
M

M



```
In [57]: ▶
```

```
plt.plot(history1.history['loss'])
plt.title('Model Loss')
plt.ylabel('loss')
plt.xlabel('epoch')
plt.show()
```



```
In [59]:

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(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 [60]:

model2.summary()

Model: "sequential_2"

Layer (type)	Output Shape	Param #
embedding_3 (Embedding)	(None, 100, 128)	512000
<pre>simple_rnn_3 (SimpleRNN)</pre>	(None, 64)	12352
<pre>embedding_4 (Embedding)</pre>	(None, 64, 128)	512000
<pre>simple_rnn_4 (SimpleRNN)</pre>	(None, 64)	12352
<pre>embedding_5 (Embedding)</pre>	(None, 64, 128)	512000
<pre>simple_rnn_5 (SimpleRNN)</pre>	(None, 64)	12352
dense_2 (Dense)	(None, 1)	65

Total params: 1573121 (6.00 MB)
Trainable params: 1573121 (6.00 MB)
Non-trainable params: 0 (0.00 Byte)

In [61]: ▶

model2.compile(optimizer='adam',loss='binary_crossentropy',metrics=['accuracy'])

In [62]: ▶

```
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/emb eddings:0', 'simple_rnn_3/simple_rnn_cell/kernel:0', 'simple_rnn_3/simple_rnn_cell/recurrent_kernel:0', 'simple_rnn_3/simple_rnn_cell/bias:0', 'em bedding_4/embeddings:0', 'simple_rnn_4/simple_rnn_cell/kernel:0', 'simple_rnn_4/simple_rnn_cell/recurrent_kernel:0', 'simple_rnn_4/simple_rnn_cell/bias:0'] when minimizing the loss. If you're using `model.compile()`, d id you forget to provide a `loss` argument?

WARNING:tensorflow:Gradients do not exist for variables ['embedding_3/emb eddings:0', 'simple_rnn_3/simple_rnn_cell/kernel:0', 'simple_rnn_3/simple _rnn_cell/recurrent_kernel:0', 'simple_rnn_3/simple_rnn_cell/bias:0', 'em bedding_4/embeddings:0', 'simple_rnn_4/simple_rnn_cell/kernel:0', 'simple_rnn_4/simple_rnn_cell/recurrent_kernel:0', 'simple_rnn_4/simple_rnn_cell/bias:0'] when minimizing the loss. If you're using `model.compile()`, d id you forget to provide a `loss` argument?

WARNING:tensorflow:Gradients do not exist for variables ['embedding_3/emb eddings:0', 'simple_rnn_3/simple_rnn_cell/kernel:0', 'simple_rnn_3/simple _rnn_cell/recurrent_kernel:0', 'simple_rnn_3/simple_rnn_cell/bias:0', 'em bedding_4/embeddings:0', 'simple_rnn_4/simple_rnn_cell/kernel:0', 'simple_rnn_4/simple_rnn_cell/recurrent_kernel:0', 'simple_rnn_4/simple_rnn_cell/bias:0'] when minimizing the loss. If you're using `model.compile()`, d id you forget to provide a `loss` argument?

WARNING:tensorflow:Gradients do not exist for variables ['embedding_3/emb eddings:0', 'simple_rnn_3/simple_rnn_cell/kernel:0', 'simple_rnn_3/simple_rnn_cell/recurrent_kernel:0', 'simple_rnn_3/simple_rnn_cell/bias:0', 'em bedding_4/embeddings:0', 'simple_rnn_4/simple_rnn_cell/kernel:0', 'simple_rnn_4/simple_rnn_cell/recurrent_kernel:0', 'simple_rnn_4/simple_rnn_cell/bias:0'] when minimizing the loss. If you're using `model.compile()`, d id you forget to provide a `loss` argument?

```
1/1 - 3s - loss: 0.6957 - accuracy: 0.5000 - 3s/epoch - 3s/step Epoch 2/10
```

1/1 - 0s - loss: 0.6996 - accuracy: 0.5000 - 42ms/epoch - 42ms/step Epoch 3/10

1/1 - 0s - loss: 0.6949 - accuracy: 0.5000 - 31ms/epoch - 31ms/step Epoch 4/10

1/1 - 0s - loss: 0.6932 - accuracy: 0.5000 - 50ms/epoch - 50ms/step Epoch 5/10

1/1 - 0s - loss: 0.6944 - accuracy: 0.5000 - 37ms/epoch - 37ms/step Epoch 6/10

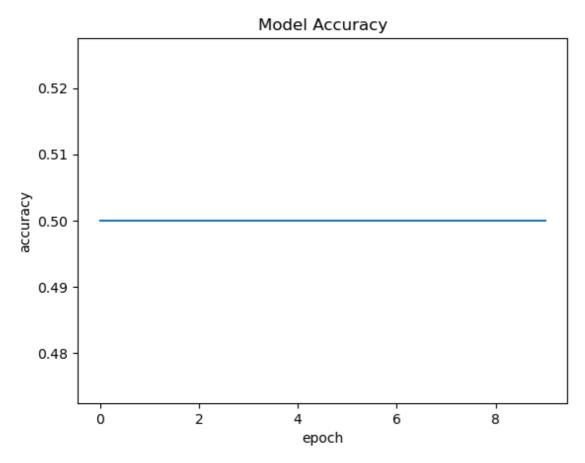
1/1 - 0s - loss: 0.6947 - accuracy: 0.5000 - 37ms/epoch - 37ms/step

Epoch 7/10 1/1 - 0s - loss: 0.6938 - accuracy: 0.5000 - 47ms/epoch - 47ms/step

Epoch 8/10 1/1 - 0s - loss: 0.6932 - accuracy: 0.5000 - 45ms/epoch - 45ms/step

Epoch 9/10 1/1 - 0s - loss: 0.6935 - accuracy: 0.5000 - 52ms/epoch - 52ms/step Epoch 10/10

1/1 - 0s - loss: 0.6940 - accuracy: 0.5000 - 49ms/epoch - 49ms/step



In [65]: ▶

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

