Assignment -4

SMS SPAM Classification

ASSIGNMENT DATE	31 October 2022
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MAXIMUM MARKS	2 Marks

Question-1:

• Import required library

SOLUTION:

```
import pandas as pd
import numpy as np
from sklearn.model_selection import train_test_split
from sklearn.preprocessing import LabelEncoder
from tensorflow.keras.models import Sequential
from tensorflow.keras.layers import LSTM, Dense, Dropout, Embedding
from tensorflow.keras.optimizers import RMSprop
from tensorflow.keras.preprocessing.text import Tokenizer
from tensorflow.keras.preprocessing import sequence
```

Question-2:

• Read dataset and do pre-processing

SOLUTION:

Data Pre-Processing

```
df = pd.read_csv('/content/drive/MyDrive/ibm/spam.csv', delimiter=',', encoding='latin-1')
df.head()
```

```
v1
                                             v2 Unnamed: 2 Unnamed: 3 Unnamed: 4
                                                       NaN
   ham
           Go until jurong point, crazy.. Available only ...
1 ham
                           Ok lar... Joking wif u oni...
                                                       NaN
                                                                   NaN
                                                                               NaN
2 spam Free entry in 2 a wkly comp to win FA Cup fina...
                                                       NaN
                                                                   NaN
                                                                               NaN
         U dun say so early hor... U c already then say...
                                                       NaN
                                                                   NaN
                                                                               NaN
4 ham
          Nah I don't think he goes to usf, he lives aro...
                                                       NaN
                                                                   NaN
                                                                               NaN
df.drop(['Unnamed: 2', 'Unnamed: 3', 'Unnamed: 4'], axis=1, inplace=True)
df.info()
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 5572 entries, 0 to 5571
Data columns (total 2 columns):
# Column Non-Null Count Dtype
--- ----- ---------
0 v1 5572 non-null object
1 v2 5572 non-null object
dtypes: object(2)
memory usage: 87.2+ KB
X = df.v2
Y = df.v1
encoder = LabelEncoder()
Y = encoder.fit_transform(Y)
Y = Y.reshape(-1,1)
X_train, X_test, y_train, y_test = train_test_split(X, Y, test_size=0.2)
tokenizer = Tokenizer(num_words=2000, lower=True)
tokenizer.fit_on_texts(X_train)
sequences = tokenizer.texts_to_sequences(X_train)
X_train = sequence.pad_sequences(sequences, maxlen=200)
```

Question-3:

Create Model

SOLUTION:

Create Model

```
model = Sequential()
```

Question-4:

Create Model

SOLUTION:

Add layers

```
model.add(Embedding(2000, 50, input_length=200))
model.add(LSTM(64))
model.add(Dense(256, activation="relu"))
model.add(Dropout(0.5))
model.add(Dense(1,activation="sigmoid"))
model.summary()
Model: "sequential"
Layer (type)
                     Output Shape
-----
embedding (Embedding)
                     (None, 200, 50)
                                          100000
1stm (LSTM)
                     (None, 64)
                                         29440
                                         16640
dense (Dense)
                     (None, 256)
dropout (Dropout)
                     (None, 256)
dense 1 (Dense)
                     (None, 1)
                                          257
______
Total params: 146,337
Trainable params: 146,337
Non-trainable params: 0
```

Question-5:

• Compile the Model

SOLUTION:

Compile the Model

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

Question-6:

• Fit the Model

SOLUTION:

Fit the Model

```
model.fit(X_train, y_train, batch_size=128, epochs=10, validation_split=0.2)
Epoch 1/10
========] - 9s 317ms/step - loss: 0.1027 - accuracy: 0.9742 - val_loss: 0.0565 - val_accuracy: 0.9798
Epoch 3/10
         28/28 [====
Epoch 4/10
28/28 [====
Epoch 5/10
          =========] - 9s 332ms/step - loss: 0.0329 - accuracy: 0.9905 - val_loss: 0.0493 - val_accuracy: 0.9832
           ========] - 9s 325ms/step - loss: 0.0219 - accuracy: 0.9938 - val_loss: 0.0523 - val_accuracy: 0.9888
28/28 [====
28/28 [====
       Epoch 7/10
==========] - 9s 318ms/step - loss: 0.0091 - accuracy: 0.9969 - val_loss: 0.0693 - val_accuracy: 0.9798
28/28 [=====
28/28 [====
       Epoch 10/10
<keras.callbacks.History at 0x7f528e02bb90>
```

Question-7

• Save The Model

SOLUTION:

Save the Model

```
model.save("model.h5")
```

Question-8

• Test The Model

SOLUTION

Test the Model

```
test_sequences = tokenizer.texts_to_sequences(X_test)
X_test = sequence.pad_sequences(test_sequences, maxlen=200)
acc = model.evaluate(X_test, y_test)
def predict(message):
   txt = tokenizer.texts_to_sequences(message)
   txt = sequence.pad_sequences(txt, maxlen=200)
   preds = model.predict(txt)
  if preds > 0.5:
    print("Span")
   else:
     print("Not Span")
predict(["Sorry, I'll call after the meeting."])
1/1 [-----] - 1s 508ms/step
Not Span
1/1 [-----] - 0s 28ms/step
Span
predict(["you won rupess 10,0000"])
1/1 [-----] - 0s 32ms/step
Span
predict(["This is the very important problem"])
Not Span
```