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```
import tensorflow as tf
In [1]:
             import numpy as np
             import pandas as pd
             import matplotlib.pyplot as plt
             import nltk
             import sklearn
             from sklearn.model selection import train test split
             from tensorflow.keras.preprocessing.text import Tokenizer
             from tensorflow.keras.preprocessing.sequence import pad sequences
             from tensorflow.keras.optimizers import RMSprop , Adam
             from keras.models import Sequential
             from keras.layers import *
             from nltk.corpus import stopwords
In [2]:
             data = pd.read_csv("news20.csv")
             data.head()
    Out[2]:
                    category
                                                               text
              0
                        tech
                              tv future in the hands of viewers with home th...
              1
                    business
                             worldcom boss left books alone former worldc...
              2
                                 tigers wary of farrell gamble leicester say ...
                       sport
              3
                       sport yeading face newcastle in fa cup premiership s...
                entertainment ocean s twelve raids box office ocean s twelve...
In [3]:
             data.shape
    Out[3]: (2225, 2)
In [4]:
          data.info()
             <class 'pandas.core.frame.DataFrame'>
             RangeIndex: 2225 entries, 0 to 2224
             Data columns (total 2 columns):
                             Non-Null Count Dtype
              #
                  Column
                             -----
              0
                  category 2225 non-null
                                               object
                             2225 non-null
                                               object
              1
                  text
             dtypes: object(2)
             memory usage: 34.9+ KB
```

```
english_stops = set(stopwords.words('english'))
        In [6]:
           X=[]
           for review in data['text']:
               filtered sentence = [w.lower() for w in review.split() if not w in en
               X.append(filtered sentence)
           X = pd.Series(X)
In [7]:
       y tokenizer = Tokenizer()
           y tokenizer.fit on texts(y)
           y_seq = np.array(y_tokenizer.texts_to_sequences (y))
           X_token = Tokenizer(num_words=5000,oov_token='<oov>')
           X_token.fit_on_texts(X)
           word index = X token.word index
           X_sequence = X_token.texts_to_sequences(X)
           dict(list(word index.items())[0:15])
   Out[7]: {'<oov>': 1,
             'said': 2,
             '-': 3,
             'mr': 4,
             'would': 5,
             'also': 6,
             'people': 7,
             'new': 8,
             'us': 9,
             'one': 10,
             'could': 11,
             'said.': 12,
             'year': 13,
             'last': 14,
             'first': 15}
In [8]:  X_padding= pad_sequences(X_sequence, maxlen=200, padding='post')
            print(y_seq.shape)
           print(X_padding.shape)
            (2225, 1)
            (2225, 200)
        | x_train, x_test, y_train, y_test = train_test_split(X_padding, y_seq, test)
In [9]:
```

Model: "sequential"

Layer (type)	Output Shape	Param #
embedding (Embedding)	(None, None, 64)	320000
lstm (LSTM)	(None, 64)	33024
dense (Dense)	(None, 64)	4160
dense_1 (Dense)	(None, 6)	390

Total params: 357574 (1.36 MB)
Trainable params: 357574 (1.36 MB)
Non-trainable params: 0 (0.00 Byte)

```
Epoch 1/20
45/45 - 14s - loss: 1.6435 - accuracy: 0.2514 - val_loss: 1.5901 - val_a
ccuracy: 0.2444 - 14s/epoch - 316ms/step
Epoch 2/20
45/45 - 7s - loss: 1.4629 - accuracy: 0.3581 - val_loss: 1.3056 - val_ac
curacy: 0.4270 - 7s/epoch - 159ms/step
Epoch 3/20
45/45 - 8s - loss: 1.2197 - accuracy: 0.4066 - val loss: 2.1563 - val ac
curacy: 0.3118 - 8s/epoch - 175ms/step
Epoch 4/20
45/45 - 10s - loss: 1.6528 - accuracy: 0.3329 - val loss: 1.3889 - val a
ccuracy: 0.3371 - 10s/epoch - 230ms/step
Epoch 5/20
45/45 - 7s - loss: 1.2053 - accuracy: 0.4424 - val_loss: 1.6691 - val_ac
curacy: 0.3202 - 7s/epoch - 151ms/step
Epoch 6/20
45/45 - 7s - loss: 1.1893 - accuracy: 0.4698 - val loss: 1.1787 - val ac
curacy: 0.4298 - 7s/epoch - 153ms/step
Epoch 7/20
45/45 - 7s - loss: 1.0962 - accuracy: 0.5414 - val loss: 1.2625 - val ac
curacy: 0.4242 - 7s/epoch - 154ms/step
45/45 - 7s - loss: 1.0884 - accuracy: 0.5485 - val loss: 1.8789 - val ac
curacy: 0.3006 - 7s/epoch - 152ms/step
Epoch 9/20
45/45 - 7s - loss: 1.0505 - accuracy: 0.5035 - val_loss: 1.1410 - val_ac
curacy: 0.4410 - 7s/epoch - 153ms/step
Epoch 10/20
45/45 - 7s - loss: 1.1096 - accuracy: 0.5049 - val loss: 1.3858 - val ac
curacy: 0.4045 - 7s/epoch - 151ms/step
Epoch 11/20
45/45 - 7s - loss: 0.9143 - accuracy: 0.5801 - val_loss: 1.1253 - val_ac
curacy: 0.4747 - 7s/epoch - 157ms/step
Epoch 12/20
45/45 - 7s - loss: 0.8332 - accuracy: 0.6138 - val_loss: 1.0587 - val_ac
curacy: 0.4888 - 7s/epoch - 150ms/step
Epoch 13/20
45/45 - 7s - loss: 0.7721 - accuracy: 0.6419 - val_loss: 1.0822 - val_ac
curacy: 0.4803 - 7s/epoch - 150ms/step
Epoch 14/20
45/45 - 7s - loss: 0.7011 - accuracy: 0.6931 - val loss: 1.1689 - val ac
curacy: 0.5056 - 7s/epoch - 153ms/step
Epoch 15/20
45/45 - 7s - loss: 0.7657 - accuracy: 0.6369 - val_loss: 1.0378 - val_ac
curacy: 0.5225 - 7s/epoch - 150ms/step
Epoch 16/20
45/45 - 7s - loss: 0.7298 - accuracy: 0.6798 - val_loss: 1.2224 - val_ac
curacy: 0.4860 - 7s/epoch - 150ms/step
Epoch 17/20
45/45 - 7s - loss: 0.8023 - accuracy: 0.6728 - val loss: 0.8225 - val ac
curacy: 0.7219 - 7s/epoch - 152ms/step
Epoch 18/20
45/45 - 7s - loss: 0.7452 - accuracy: 0.7240 - val loss: 1.0383 - val ac
curacy: 0.5758 - 7s/epoch - 150ms/step
Epoch 19/20
45/45 - 7s - loss: 0.6798 - accuracy: 0.7416 - val_loss: 0.8665 - val_ac
curacy: 0.6713 - 7s/epoch - 154ms/step
```

Epoch 20/20 45/45 - 7s - loss: 0.5761 - accuracy: 0.7704 - val_loss: 1.0178 - val_accuracy: 0.6461 - 7s/epoch - 152ms/step

Model: "sequential_1"

Layer (type)	Output Shape	Param #
embedding_1 (Embedding)	(None, None, 64)	320000
conv1d (Conv1D)	(None, None, 32)	10272
<pre>max_pooling1d (MaxPooling1 D)</pre>	(None, None, 32)	0
lstm_1 (LSTM)	(None, 64)	24832
dense_2 (Dense)	(None, 128)	8320
dense_3 (Dense)	(None, 6)	774

Total params: 364198 (1.39 MB)
Trainable params: 364198 (1.39 MB)
Non-trainable params: 0 (0.00 Byte)

```
Epoch 1/20
45/45 - 12s - loss: 1.6424 - accuracy: 0.2563 - val_loss: 1.5689 - val_a
ccuracy: 0.3343 - 12s/epoch - 262ms/step
Epoch 2/20
45/45 - 4s - loss: 1.3598 - accuracy: 0.3617 - val_loss: 1.2377 - val_ac
curacy: 0.4017 - 4s/epoch - 91ms/step
Epoch 3/20
45/45 - 5s - loss: 1.1110 - accuracy: 0.4171 - val loss: 1.2128 - val ac
curacy: 0.3399 - 5s/epoch - 102ms/step
Epoch 4/20
45/45 - 4s - loss: 1.0370 - accuracy: 0.4803 - val loss: 0.9303 - val ac
curacy: 0.5506 - 4s/epoch - 94ms/step
Epoch 5/20
45/45 - 4s - loss: 0.8156 - accuracy: 0.5730 - val_loss: 0.8693 - val_ac
curacy: 0.5871 - 4s/epoch - 94ms/step
Epoch 6/20
45/45 - 4s - loss: 0.7669 - accuracy: 0.5878 - val loss: 0.9167 - val ac
curacy: 0.5337 - 4s/epoch - 94ms/step
Epoch 7/20
45/45 - 4s - loss: 0.7271 - accuracy: 0.6124 - val_loss: 0.8524 - val_ac
curacy: 0.5871 - 4s/epoch - 91ms/step
45/45 - 4s - loss: 0.6842 - accuracy: 0.6390 - val loss: 0.8838 - val ac
curacy: 0.5899 - 4s/epoch - 90ms/step
Epoch 9/20
45/45 - 4s - loss: 0.6064 - accuracy: 0.7058 - val_loss: 0.7325 - val_ac
curacy: 0.7163 - 4s/epoch - 89ms/step
Epoch 10/20
45/45 - 4s - loss: 0.3846 - accuracy: 0.8694 - val loss: 1.3970 - val ac
curacy: 0.5646 - 4s/epoch - 89ms/step
Epoch 11/20
45/45 - 4s - loss: 0.5864 - accuracy: 0.7542 - val_loss: 0.7597 - val ac
curacy: 0.6938 - 4s/epoch - 88ms/step
Epoch 12/20
45/45 - 4s - loss: 0.5965 - accuracy: 0.7423 - val loss: 0.9662 - val ac
curacy: 0.5955 - 4s/epoch - 89ms/step
Epoch 13/20
45/45 - 4s - loss: 0.4547 - accuracy: 0.7746 - val_loss: 1.2960 - val_ac
curacy: 0.6601 - 4s/epoch - 88ms/step
Epoch 14/20
45/45 - 4s - loss: 0.3195 - accuracy: 0.9024 - val loss: 0.4736 - val ac
curacy: 0.8764 - 4s/epoch - 89ms/step
Epoch 15/20
45/45 - 4s - loss: 0.1583 - accuracy: 0.9572 - val_loss: 0.4431 - val_ac
curacy: 0.8904 - 4s/epoch - 89ms/step
Epoch 16/20
45/45 - 4s - loss: 0.0906 - accuracy: 0.9761 - val_loss: 0.5063 - val_ac
curacy: 0.8848 - 4s/epoch - 88ms/step
Epoch 17/20
45/45 - 5s - loss: 0.0713 - accuracy: 0.9874 - val loss: 0.5558 - val ac
curacy: 0.8708 - 5s/epoch - 101ms/step
Epoch 18/20
45/45 - 4s - loss: 0.0695 - accuracy: 0.9853 - val loss: 0.4693 - val ac
curacy: 0.8961 - 4s/epoch - 97ms/step
Epoch 19/20
45/45 - 4s - loss: 0.1698 - accuracy: 0.9670 - val_loss: 0.4143 - val_ac
curacy: 0.9017 - 4s/epoch - 91ms/step
```

```
Epoch 20/20
            45/45 - 4s - loss: 0.1892 - accuracy: 0.9487 - val_loss: 0.5028 - val_ac
             curacy: 0.8680 - 4s/epoch - 89ms/step
        In [21]:
            from gensim.scripts.glove2word2vec import glove2word2vec
         glove file = "glove.6B.100d.txt"
In [22]:
            glove word2vec file = "glove.6B.100d.txt.word2vec"
            glove2word2vec(glove_file, glove_word2vec_file)
            glove embeddings = KeyedVectors.load word2vec format(glove word2vec file,
            C:\Users\8mpra\AppData\Local\Temp\ipykernel 5460\889183730.py:3: Depreca
            tionWarning: Call to deprecated `glove2word2vec` (KeyedVectors.load_word
            2vec format(.., binary=False, no header=True) loads GLoVE text vector
            s.).
              glove2word2vec(glove_file, glove_word2vec_file)
In [26]:
            embedding_matrix = np.zeros((vocab_size, 300))
            for word, i in X_token.word_index.items():
                try:
                    embedding_vector = glove_embeddings[word]
                    if embedding_vector is not None:
                        embedding_matrix[i] = embedding_vector
                except:
                    pass
 In [ ]:  | model1 = Sequential()
            model1.add(Embedding(vocab_size, embedding_dim))
            model1.add(Conv1D(filters=32, kernel_size=5, strides=1, activation='relu'
            model1.add(MaxPooling1D((2)))
            model1.add(LSTM(embedding dim))
            model1.add(Dense(128, activation= 'relu'))
            model1.add(Dense(6, activation='softmax'))
            model1.summary()
```


Model: "sequential_2"

Layer (type)	Output Shape	Param #
embedding_2 (Embedding)	(None, 200, 300)	1500000
conv1d_1 (Conv1D)	(None, 196, 32)	48032
<pre>max_pooling1d_1 (MaxPoolin g1D)</pre>	(None, 98, 32)	0
lstm_2 (LSTM)	(None, 50)	16600
dense_4 (Dense)	(None, 128)	6528
dense_5 (Dense)	(None, 6)	774

Total params: 1571934 (6.00 MB)
Trainable params: 71934 (280.99 KB)
Non-trainable params: 1500000 (5.72 MB)


```
Epoch 1/5
45/45 - 12s - loss: 1.6754 - accuracy: 0.2338 - val_loss: 1.6813 - val_a ccuracy: 0.2022 - 12s/epoch - 263ms/step
Epoch 2/5
45/45 - 5s - loss: 1.6716 - accuracy: 0.2338 - val_loss: 1.6780 - val_ac curacy: 0.2022 - 5s/epoch - 116ms/step
Epoch 3/5
45/45 - 5s - loss: 1.6682 - accuracy: 0.2338 - val_loss: 1.6749 - val_ac curacy: 0.2022 - 5s/epoch - 116ms/step
Epoch 4/5
45/45 - 5s - loss: 1.6650 - accuracy: 0.2338 - val_loss: 1.6721 - val_ac curacy: 0.2022 - 5s/epoch - 115ms/step
Epoch 5/5
45/45 - 5s - loss: 1.6621 - accuracy: 0.2338 - val_loss: 1.6691 - val_ac curacy: 0.2022 - 5s/epoch - 122ms/step
```

In []: N