ASSIGNMENT 4

Assignment Date	19 September 2022
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Maximum Marks	2 Marks

ARTIFICIAL INTELLIGENCE

- HISTORICALLY, Individual parts have been a cause of disunity & resulted in countries being exploited by foreign nations. The decline of Mughals resulted in Individual kings fighting each other which led to the British Empire. The spirit of the 'whole' led to Independence.
- On the LEGAL front post-Independence, the constituent Assembly put together 395 Articles & 8 Schedules. Then began the clash of the parts -Fundamental Rights v/s Directive Principles. Finally, the Keshavanand Bharti case focused on the 'whole' and gave birth to the 'Basic Structure Doctrine' which represents the essence of the constitution in India.
- On the CULTURAL Front, India resembles a melting pot of diverse cultures. From ancient times, it has incorporated the Greek, Persian, Turkish, Mughal & European cultures. It stands for the principle of 'UNITY IN DIVERSITY. This has been possible because Individuals identity themselves as 'INDIAN' first.
- On the POLITICAL front, the recent repeal of the farm laws again proves the power of the whole. If Individual farmers had tried to protest, the farm laws might never have been repealed. The 'Sanyukt Kisan Morcha' formed by the farmers acted as a 'united pressure group' & brought the change which they intended.

- On the ECONOMIC front, India ranks 6th in terms of nominal GDP. Post COVID-19, the Indian Economy had contracted by 8%. This could have been worse without the 20% growth registered by the agricultural sector during the COVID period. This is an important lesson for policymakers to focus on the development of the economy as a 'whole' with a balanced focus on its parts.
- Finally, in SCIENCE & TECHNOLOGY front, the Industrial Revolution 4.0 has begun on the 'foundation of the Internet'. The Internet is one the best example of the power of the 'whole'. It has given hope for universal access to technology thereby enabling an egalitarian society.

```
#
coding:
utf-8
          # In[1]:
          import gensim, logging
          # In[2]:
          logging.basicConfig(format='%(asctime)s : %(levelname)s : %(message)s',
          level=logging.INFO)
          # In[3]:
          from gensim.models.doc2vec import LabeledSentence
          from gensim.models import Doc2Vec
          # In[4]:
          with open("sentiment labelled sentences/yelp_labelled.txt") as f:
              for item_no, line in enumerate(f):
                  print(item_no, line)
          # In[19]:
          sentences = []
          sentiments = []
          with open("sentiment labelled sentences/yelp_labelled.txt") as f:
              for item_no, line in enumerate(f):
                  line_split = line.strip().split('\t')
                  sentences.append((line_split[0], "yelp_%d" % item_no))
```

```
        sentiments.append(int(line_split[1]))
# In[21]:
len(sentences), sentences
# In[37]:
import re
sentences = []
sentiments = []
for fname in ["yelp", "amazon_cells", "imdb"]:
    with open("sentiment labelled sentences/%s_labelled.txt" % fname) as f:
        for item_no, line in enumerate(f):
            line_split = line.strip().split('\t')
            sent = line_split[0].lower()
            sent = re.sub(r'\'', '', sent)
            sent = re.sub(r'\W', ' ', sent)
            sent = re.sub(r'\s+', ' ', sent).strip()
            sentences.append(LabeledSentence(sent.split(), ["%s_%d" % (fname,
item_no)]))
            sentiments.append(int(line_split[1]))
# In[38]:
sentences
# In[43]:
import random
class PermuteSentences(object):
    def __iter__(self):
        shuffled = list(sentences)
        random.shuffle(shuffled)
        for sent in shuffled:
            yield sent
permuter = PermuteSentences()

model = Doc2Vec(permuter, min_count=1)
# In[44]:
model.most_similar('tasty')
```

```
import numpy as np
from keras.models import Sequential, load_model
from keras.layers import Dropout, Flatten, Conv2D, MaxPooling2D, Dense,
Activation
from keras.utils import np_utils
from keras.preprocessing.image import ImageDataGenerator
```

```
from keras.callbacks import TensorBoard
import itertools
```

```
Using TensorFlow backend.
```

In [2]:

```
# all images will be converted to this size
ROWS = 256
COLS = 256
CHANNELS = 3
```

In [3]:

```
train_image_generator = ImageDataGenerator(horizontal_flip=True,
rescale=1./255, rotation_range=45)
test_image_generator = ImageDataGenerator(horizontal_flip=False,
rescale=1./255, rotation_range=0)

train_generator = train_image_generator.flow_from_directory('train',
target_size=(ROWS, COLS), class_mode='categorical')
test_generator = test_image_generator.flow_from_directory('test',
target_size=(ROWS, COLS), class_mode='categorical')
```

```
Found 5994 images belonging to 200 classes.
Found 5794 images belonging to 200 classes.
```

In [12]:

```
train_generator.reset()
test_generator.reset()

model = Sequential()
model.add(Conv2D(64, (3,3), input_shape=(ROWS, COLS, CHANNELS)))
model.add(Activation('relu'))
model.add(Conv2D(64, (3,3)))
model.add(Activation('relu'))
model.add(MaxPooling2D(pool_size=(4,4)))
model.add(Conv2D(64, (3,3)))
model.add(Activation('relu'))
model.add(Conv2D(64, (3,3)))
model.add(Activation('relu'))
model.add(MaxPooling2D(pool_size=(4,4)))
model.add(Flatten())
model.add(Dropout(0.5))
model.add(Dense(400))
model.add(Activation('relu'))
model.add(Dropout(0.5))
model.add(Dense(200))
model.add(Activation('softmax'))

model.compile(loss='categorical_crossentropy', optimizer='adamax',
metrics=['accuracy'])

model.summary()
```

```
_________________________________________________________________
Layer (type)                 Output Shape              Param #
=================================================================
conv2d_19 (Conv2D)           (None, 254, 254, 64)      1792
```

```
activation_28 (Activation)   (None, 254, 254, 64)      0
conv2d_20 (Conv2D)           (None, 252, 252, 64)      36928
activation_29 (Activation)   (None, 252, 252, 64)      0
max_pooling2d_10 (MaxPooling (None, 63, 63, 64)        0
conv2d_21 (Conv2D)           (None, 61, 61, 64)        36928
activation_30 (Activation)   (None, 61, 61, 64)        0
conv2d_22 (Conv2D)           (None, 59, 59, 64)        36928
activation_31 (Activation)   (None, 59, 59, 64)        0
max_pooling2d_11 (MaxPooling (None, 14, 14, 64)        0
flatten_5 (Flatten)          (None, 12544)             0
dropout_8 (Dropout)          (None, 12544)             0
dense_9 (Dense)              (None, 400)               5018000
activation_32 (Activation)   (None, 400)               0
dropout_9 (Dropout)          (None, 400)               0
dense_10 (Dense)             (None, 200)               80200
activation_33 (Activation)   (None, 200)               0
=================================================================
Total params: 5,210,776
Trainable params: 5,210,776
Non-trainable params: 0
_________________________________________________________________
```

In [15]:

```
tensorboard = TensorBoard(log_dir='./logs/custom')

model.fit_generator(train_generator, steps_per_epoch=512, epochs=10,
callbacks=[tensorboard], verbose=2)
```

```
Epoch 1/10
 - 434s - loss: 4.4682 - acc: 0.0687
Epoch 2/10
 - 440s - loss: 4.1851 - acc: 0.0919
Epoch 3/10
 - 443s - loss: 3.9278 - acc: 0.1270
Epoch 4/10
 - 428s - loss: 3.6948 - acc: 0.1615
Epoch 5/10
 - 437s - loss: 3.4944 - acc: 0.1935
```

```
Epoch 6/10
 - 439s - loss: 3.3103 - acc: 0.2196
Epoch 7/10
 - 438s - loss: 3.1253 - acc: 0.2492
Epoch 8/10
 - 443s - loss: 2.9927 - acc: 0.2757
Epoch 9/10
 - 431s - loss: 2.8474 - acc: 0.2998
Epoch 10/10
 - 430s - loss: 2.7354 - acc: 0.3271
```

Out[15]:

In [16]:

```
print(model.evaluate_generator(test_generator, steps=1000))
```

```
[3.3455331521880121, 0.22266875981161696]
```

In [ ]: