```
In [1]:
         # Vishakha Dhonde
         # COBB055
         # Importing Necesarry Packages
         import numpy as np
         from numpy.ma.core import argmax
         import pandas as pd
         from matplotlib import cm
         import matplotlib.pyplot as plt
         %matplotlib inline
         import seaborn as sns
         #import os
         import time
         from sklearn.metrics import confusion_matrix, accuracy_score, auc
         from keras.preprocessing import sequence
         from keras.models import Sequential
         from keras.layers import Dense, Dropout, Activation
         from keras.layers import Embedding
         from keras.layers import Conv1D, GlobalMaxPooling1D
         from keras.callbacks import EarlyStopping
         from keras import models
         from keras import layers
         from keras.datasets import imdb
```

```
In [2]:
         # Loading the dataset
         (X_train, y_train), (X_test, y_test) = imdb.load_data()
         X = np.concatenate((X_train, X_test), axis=0)
         y = np.concatenate((y_train, y_test), axis=0)
         # Exploring the Data
         print("Training data: ")
         print(X.shape)
         print(y.shape)
         print("Classes: ")
         print(np.unique(y))
         print("Number of words: ")
         print(len(np.unique(np.hstack(X))))
         print("Review length: ")
         result = [len(x) for x in X]
         print("Mean %.2f words (%f)" % (np.mean(result), np.std(result))) # Ploting the review
         plt.boxplot(result)
         plt.show()
```

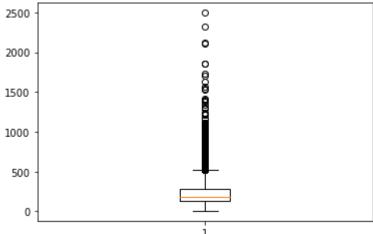
<\_array\_function\_\_ internals>:5: VisibleDeprecationWarning: Creating an ndarray from ra
gged nested sequences (which is a list-or-tuple of lists-or-tuples-or ndarrays with diff
erent lengths or shapes) is deprecated. If you meant to do this, you must specify 'dtype
=object' when creating the ndarray.

C:\Users\i-net computer\anaconda3\envs\tensorflow\lib\site-packages\tensorflow\python\ke ras\datasets\imdb.py:159: VisibleDeprecationWarning: Creating an ndarray from ragged nes ted sequences (which is a list-or-tuple of lists-or-tuples-or ndarrays with different le ngths or shapes) is deprecated. If you meant to do this, you must specify 'dtype=object' when creating the ndarray.

x\_train, y\_train = np.array(xs[:idx]), np.array(labels[:idx])

C:\Users\i-net computer\anaconda3\envs\tensorflow\lib\site-packages\tensorflow\python\ke ras\datasets\imdb.py:160: VisibleDeprecationWarning: Creating an ndarray from ragged nes ted sequences (which is a list-or-tuple of lists-or-tuples-or ndarrays with different le ngths or shapes) is deprecated. If you meant to do this, you must specify 'dtype=object'

```
when creating the ndarray.
  x_test, y_test = np.array(xs[idx:]), np.array(labels[idx:])
Training data:
(50000,)
(50000,)
Classes:
[0 1]
Number of words:
88585
Review length:
Mean 234.76 words (172.911495)
```



```
def vectorize_sequences(sequences, dimension=5000): # Function for vectorising data
    results = np.zeros((len(sequences), dimension)) # Creating an all-zero matrix of sh
    for i, sequence in enumerate(sequences):
        results[i, sequence] = 1. # Set specific indices of results[i] to 1s
    return results
```

```
# Creating Training and Testing Sets and Preprocessing them
(train_data, train_labels), (test_data, test_labels) = imdb.load_data(num_words=5000)
# Our vectorized training data
x_train = vectorize_sequences(train_data)
# Our vectorized test data
x_test = vectorize_sequences(test_data)
# Our vectorized Labels one-hot encoder
y_train = np.asarray(train_labels).astype('float32')
y_test = np.asarray(test_labels).astype('float32')
```

<\_array\_function\_\_ internals>:5: VisibleDeprecationWarning: Creating an ndarray from ra
gged nested sequences (which is a list-or-tuple of lists-or-tuples-or ndarrays with diff
erent lengths or shapes) is deprecated. If you meant to do this, you must specify 'dtype
=object' when creating the ndarray.

C:\Users\i-net computer\anaconda3\envs\tensorflow\lib\site-packages\tensorflow\python\ke ras\datasets\imdb.py:159: VisibleDeprecationWarning: Creating an ndarray from ragged nes ted sequences (which is a list-or-tuple of lists-or-tuples-or ndarrays with different le ngths or shapes) is deprecated. If you meant to do this, you must specify 'dtype=object' when creating the ndarray.

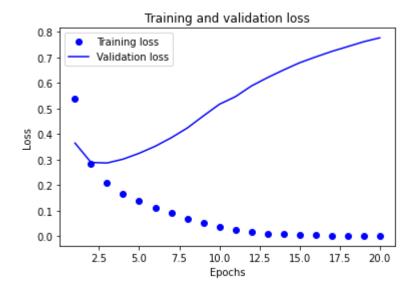
x\_train, y\_train = np.array(xs[:idx]), np.array(labels[:idx])

C:\Users\i-net computer\anaconda3\envs\tensorflow\lib\site-packages\tensorflow\python\ke ras\datasets\imdb.py:160: VisibleDeprecationWarning: Creating an ndarray from ragged nes ted sequences (which is a list-or-tuple of lists-or-tuples-or ndarrays with different le ngths or shapes) is deprecated. If you meant to do this, you must specify 'dtype=object'

```
when creating the ndarray.
      x test, y test = np.array(xs[idx:]), np.array(labels[idx:])
In [5]:
     # Creating the DNN Model
     model = models.Sequential()
     model.add(layers.Dense(32, activation='relu', input shape=(5000,)))
     model.add(layers.Dense(32, activation='relu',))
     model.add(layers.Dense(1, activation='sigmoid'))
In [6]:
     #Set validation set aside
     x \text{ val} = x \text{ train}[:10000]
     partial_x_train = x_train[10000:]
     y val = y train[:10000]
     partial y train = y train[10000:]
In [7]:
     # Compiling Model
     model.compile(optimizer='adam', loss='binary_crossentropy', metrics=['acc'])
     start_time_m1 = time.time()
     history = model.fit(partial x train,
                 partial y train,
                 epochs=20,
                 batch size=512,
                 validation_data=(x_val, y_val))
     total_time_m1 = time.time() - start_time_m1
     print("The Dense Convolutional Neural Network 1 layer took %.4f seconds to train." % (t
     Epoch 1/20
     loss: 0.3651 - val acc: 0.8600
     Epoch 2/20
     loss: 0.2893 - val acc: 0.8866
     Epoch 3/20
     loss: 0.2872 - val acc: 0.8849
     Epoch 4/20
     _loss: 0.3022 - val_acc: 0.8806
     Epoch 5/20
     loss: 0.3252 - val acc: 0.8768
     Epoch 6/20
     _loss: 0.3527 - val_acc: 0.8738
     Epoch 7/20
     loss: 0.3859 - val acc: 0.8693
     loss: 0.4242 - val acc: 0.8672
     Epoch 9/20
     _loss: 0.4713 - val_acc: 0.8616
     Epoch 10/20
     _loss: 0.5170 - val_acc: 0.8586
```

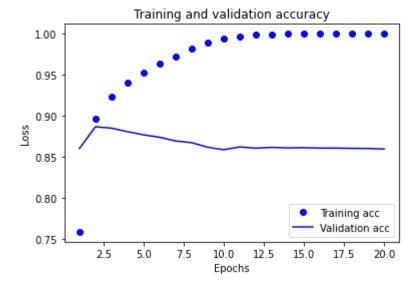
```
loss: 0.5468 - val acc: 0.8620
    Epoch 12/20
    loss: 0.5887 - val acc: 0.8605
    Epoch 13/20
    _loss: 0.6214 - val_acc: 0.8615
    Epoch 14/20
    _loss: 0.6513 - val_acc: 0.8608
    Epoch 15/20
    loss: 0.6795 - val acc: 0.8611
    Epoch 16/20
    _loss: 0.7022 - val_acc: 0.8606
    Epoch 17/20
    _loss: 0.7238 - val_acc: 0.8606
    Epoch 18/20
    loss: 0.7428 - val acc: 0.8603
    Epoch 19/20
    _loss: 0.7619 - val_acc: 0.8601
    Epoch 20/20
    loss: 0.7772 - val acc: 0.8595
    The Dense Convolutional Neural Network 1 layer took 100.5009 seconds to train.
In [8]:
     history dict = history.history
     history dict.keys()
    dict keys(['loss', 'acc', 'val loss', 'val acc'])
Out[8]:
In [9]:
     acc = history.history['acc']
     val acc = history.history['val acc']
     loss = history.history['loss']
     val loss = history.history['val loss']
     epochs = range(1, len(acc) + 1)
     # Plotting model loss
     plt.plot(epochs, loss, 'bo', label='Training loss') # "bo" is for "blue dot"
     plt.plot(epochs, val loss, 'b', label='Validation loss') # b is for "solid blue line"
     plt.title('Training and validation loss')
     plt.xlabel('Epochs')
     plt.ylabel('Loss')
     plt.legend()
     plt.show()
```

Epoch 11/20



```
In [10]:
    plt.clf() # clear figure
    acc_values = history_dict['acc']
    val_acc_values = history_dict['val_acc']

# Plotting model accuracy
    plt.plot(epochs, acc, 'bo', label='Training acc')
    plt.plot(epochs, val_acc, 'b', label='Validation acc')
    plt.title('Training and validation accuracy')
    plt.xlabel('Epochs')
    plt.ylabel('Loss')
    plt.legend()
    plt.show()
```



```
In [11]: # Model Summary
    print(model.summary())

# Predictions
    pred = model.predict(x_test)
    classes_x=np.argmax(pred,axis=1)
    accuracy_score(y_test,classes_x)

#Confusion Matrix
```

```
conf_mat = confusion_matrix(y_test, classes_x)
print(conf_mat)

conf_mat_normalized = conf_mat.astype('float') / conf_mat.sum(axis=1)[:, np.newaxis]
sns.heatmap(conf_mat_normalized)
plt.ylabel('True label')
plt.xlabel('Predicted label')
```

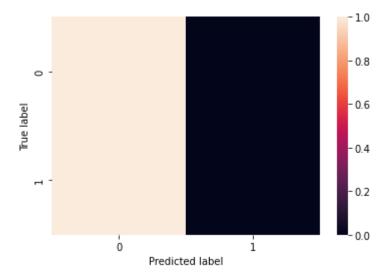
Model: "sequential"

Layer (type)	Output Shape	Param #
dense (Dense)	(None, 32)	160032
dense_1 (Dense)	(None, 32)	1056
dense_2 (Dense)	(None, 1)	33
Total names, 161, 121		========

Total params: 161,121 Trainable params: 161,121 Non-trainable params: 0

[[12500 0] [12500 0]] Out[11]: Text(0.5, 15.0, 'Predicted label')

None



```
In [12]: #Dense with Two Layer
    model2 = models.Sequential()
    model2.add(layers.Dense(32, activation='relu', input_shape=(5000,)))
    model2.add(layers.Dense(32, activation='relu'))
    model2.add(layers.Dense(32, activation='relu'))
    model2.add(layers.Dense(1, activation='sigmoid'))
```

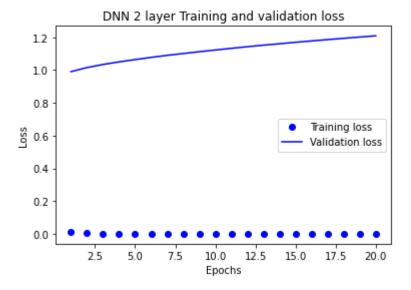
```
validation_data=(x_val, y_val))
total_time_m2 = time.time() - start_time_m2
print("The Dense Convolutional Neural Network 2 layers took %.4f seconds to train." % (
Epoch 1/20
l loss: 0.9904 - val acc: 0.8529
Epoch 2/20
loss: 1.0154 - val acc: 0.8521
Epoch 3/20
30/30 [============= ] - 1s 31ms/step - loss: 9.8876e-04 - acc: 1.0000 -
val loss: 1.0341 - val acc: 0.8521
Epoch 4/20
30/30 [============= ] - 1s 39ms/step - loss: 6.3576e-04 - acc: 1.0000 -
val_loss: 1.0499 - val_acc: 0.8519
30/30 [============= ] - 1s 26ms/step - loss: 4.9221e-04 - acc: 1.0000 -
val_loss: 1.0641 - val_acc: 0.8515
Epoch 6/20
val loss: 1.0774 - val acc: 0.8515
Epoch 7/20
val_loss: 1.0900 - val_acc: 0.8525
Epoch 8/20
val loss: 1.1010 - val acc: 0.8525
Epoch 9/20
val loss: 1.1121 - val acc: 0.8530
Epoch 10/20
30/30 [============= ] - 1s 23ms/step - loss: 2.3421e-04 - acc: 1.0000 -
val_loss: 1.1225 - val_acc: 0.8530
Epoch 11/20
30/30 [============== ] - 1s 24ms/step - loss: 2.0866e-04 - acc: 1.0000 -
val_loss: 1.1327 - val_acc: 0.8530
Epoch 12/20
30/30 [================= ] - 1s 25ms/step - loss: 1.8680e-04 - acc: 1.0000 -
val_loss: 1.1427 - val_acc: 0.8528
Epoch 13/20
30/30 [============= ] - 1s 26ms/step - loss: 1.6813e-04 - acc: 1.0000 -
val_loss: 1.1521 - val_acc: 0.8526
Epoch 14/20
val_loss: 1.1606 - val_acc: 0.8526
Epoch 15/20
30/30 [============== ] - 1s 23ms/step - loss: 1.3893e-04 - acc: 1.0000 -
val loss: 1.1695 - val acc: 0.8531
Epoch 16/20
30/30 [============= ] - 1s 22ms/step - loss: 1.2706e-04 - acc: 1.0000 -
val_loss: 1.1778 - val_acc: 0.8533
Epoch 17/20
30/30 [============= ] - 1s 28ms/step - loss: 1.1684e-04 - acc: 1.0000 -
val loss: 1.1861 - val acc: 0.8531
Epoch 18/20
val loss: 1.1937 - val acc: 0.8529
```

30/30 [============= ] - 1s 28ms/step - loss: 9.9063e-05 - acc: 1.0000 -

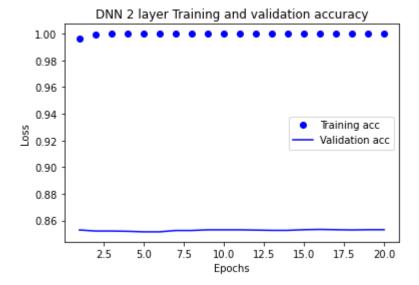
Epoch 19/20

```
In [18]:
    acc = history.history['acc']
    val_acc = history.history['val_acc']
    loss = history.history['loss']
    val_loss = history.history['val_loss']
    epochs = range(1, len(acc) + 1)

# Plotting Loss
plt.plot(epochs, loss, 'bo', label='Training loss') # "bo" is for "blue dot"
plt.plot(epochs, val_loss, 'b', label='Validation loss') # b is for "solid blue line"
plt.title('DNN 2 layer Training and validation loss')
plt.xlabel('Epochs')
plt.ylabel('Loss')
plt.legend()
plt.show()
```



```
In [19]: plt.clf() # clear figure
    acc_values = history_dict['acc']
    val_acc_values = history_dict['val_acc']
# Plotting Accuracy
plt.plot(epochs, acc, 'bo', label='Training acc')
plt.plot(epochs, val_acc, 'b', label='Validation acc')
plt.title('DNN 2 layer Training and validation accuracy')
plt.xlabel('Epochs')
plt.ylabel('Loss')
plt.legend()
plt.show()
```



In [20]:

```
print(model2.summary())
# Predictions
pred = model2.predict(x_test)
classes_x=np.argmax(pred,axis=-1)
accuracy_score(y_test,classes_x)
```

Model: "sequential\_1"

Layer (type)	Output Shape	Param #
dense_3 (Dense)	(None, 32)	160032
dense_4 (Dense)	(None, 32)	1056
dense_5 (Dense)	(None, 32)	1056
dense_6 (Dense)	(None, 1)	33

Total params: 162,177 Trainable params: 162,177 Non-trainable params: 0

None

Out[20]: 0.5

In [ ]:	