5\_1

```
In [2]: import numpy as np

def vectorize_sequences(sequences, dimensions=10000):
    results = np.zeros((len(sequences), dimensions))
    for i, sequence in enumerate(sequences):
        results[i, sequence] = 1.
    return results

x_train = vectorize_sequences(train_data)
x_test = vectorize_sequences(test_data)

y_train = np.asarray(train_labels).astype('float32')
y_test = np.asarray(test_labels).astype('float32')
```

```
In [4]: from keras import models, layers
        model = models.Sequential()
        model.add(layers.Dense(16, activation= 'relu', input_shape= (10000,)))
        model.add(layers.Dense(16, activation= 'relu'))
        model.add(layers.Dense(1, activation= 'sigmoid'))
        model.compile(optimizer= 'rmsprop',
                      loss= 'binary_crossentropy',
                     metrics= ['accuracy'])
        x_val = x_train[:10000]
        partial_x_train = x_train[10000:]
        y_val = y_train[:10000]
        partial_y_train = y_train[10000:]
        history = model.fit(partial_x_train,
                            partial_y_train,
                            epochs= 20,
                            batch size= 512,
                            validation_data= (x_val, y_val))
```

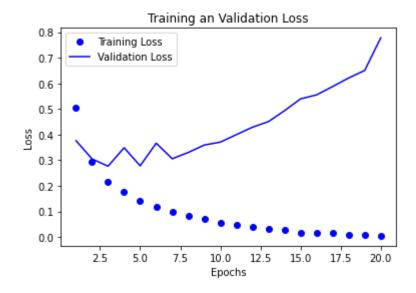
```
Epoch 1/20
30/30 [============== ] - 1s 44ms/step - loss: 0.5036 - accura
cy: 0.7875 - val_loss: 0.3765 - val_accuracy: 0.8631
cy: 0.9047 - val_loss: 0.3053 - val_accuracy: 0.8830
Epoch 3/20
cy: 0.9307 - val_loss: 0.2769 - val_accuracy: 0.8905
Epoch 4/20
30/30 [================ ] - 1s 30ms/step - loss: 0.1754 - accura
cy: 0.9417 - val_loss: 0.3489 - val_accuracy: 0.8591
Epoch 5/20
30/30 [================ ] - 1s 31ms/step - loss: 0.1434 - accura
cy: 0.9535 - val_loss: 0.2784 - val_accuracy: 0.8889
Epoch 6/20
30/30 [================== ] - 1s 28ms/step - loss: 0.1167 - accura
cy: 0.9645 - val_loss: 0.3667 - val_accuracy: 0.8603
Epoch 7/20
30/30 [================ ] - 1s 27ms/step - loss: 0.0980 - accura
cy: 0.9716 - val_loss: 0.3060 - val_accuracy: 0.8843
30/30 [========================= ] - 1s 27ms/step - loss: 0.0825 - accura
cy: 0.9771 - val_loss: 0.3306 - val_accuracy: 0.8809
Epoch 9/20
30/30 [================ ] - 1s 26ms/step - loss: 0.0697 - accura
cy: 0.9809 - val_loss: 0.3596 - val_accuracy: 0.8786
Epoch 10/20
30/30 [================ ] - 1s 30ms/step - loss: 0.0560 - accura
cy: 0.9871 - val_loss: 0.3709 - val_accuracy: 0.8775
Epoch 11/20
30/30 [================ ] - 1s 29ms/step - loss: 0.0483 - accura
cy: 0.9874 - val_loss: 0.4002 - val_accuracy: 0.8729
Epoch 12/20
30/30 [================ ] - 1s 29ms/step - loss: 0.0388 - accura
cy: 0.9919 - val loss: 0.4290 - val accuracy: 0.8695
Epoch 13/20
30/30 [================ ] - 1s 29ms/step - loss: 0.0306 - accura
cy: 0.9944 - val loss: 0.4511 - val accuracy: 0.8743
Epoch 14/20
30/30 [================ ] - 1s 42ms/step - loss: 0.0266 - accura
cy: 0.9941 - val_loss: 0.4935 - val_accuracy: 0.8728
Epoch 15/20
cy: 0.9975 - val_loss: 0.5397 - val_accuracy: 0.8706
Epoch 16/20
cy: 0.9976 - val_loss: 0.5554 - val_accuracy: 0.8666
Epoch 17/20
cy: 0.9976 - val loss: 0.5879 - val accuracy: 0.8666
Epoch 18/20
cy: 0.9997 - val_loss: 0.6213 - val_accuracy: 0.8681
Epoch 19/20
cy: 0.9986 - val_loss: 0.6507 - val_accuracy: 0.8656
```

```
In [5]: import matplotlib.pyplot as plt

history_dict = history.history
loss_values = history_dict['loss']
val_loss_values = history_dict['val_loss']
epochs = range(1, len(history_dict['accuracy']) + 1)

plt.plot(epochs, loss_values, 'bo', label= 'Training Loss')
plt.plot(epochs, val_loss_values, 'b', label= "Validation Loss")
plt.title('Training an Validation Loss')
plt.xlabel('Epochs')
plt.ylabel('Loss')
plt.legend()
```

Out[5]: <matplotlib.legend.Legend at 0x7f054e372f40>

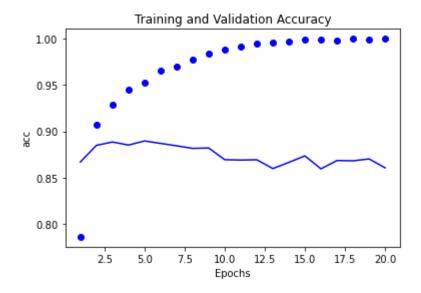


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```
In [17]: acc_values = history_dict['accuracy']
    val_acc_values = history_dict['val_accuracy']

    plt.plot(epochs, acc_values, 'bo', label= 'Training Acc')
    plt.plot(epochs, val_acc_values, 'b', label= 'Validation Acc')
    plt.title('Training and Validation Accuracy')
    plt.xlabel('Epochs')
    plt.ylabel('acc')
```

Out[17]: Text(0, 0.5, 'acc')



Out[18]: [0.7985910177230835, 0.843999981880188]

```
In [43]:
        model = models.Sequential()
         model.add(layers.Dense(32, activation= 'relu', input_shape= (10000,)))
         model.add(layers.Dense(16, activation= 'relu'))
         model.add(layers.Dense(1, activation= 'sigmoid'))
         model.compile(optimizer= 'rmsprop',
                     loss= 'binary crossentropy',
                     metrics= ['accuracy'])
         x_val = x_train[:10000]
         partial x train = x train[10000:]
         y_val = y_train[:10000]
         partial_y_train = y_train[10000:]
         history = model.fit(partial x train,
                           partial_y_train,
                           epochs= 4,
                           batch_size= 512,
                           validation_data= (x_val, y_val))
         Epoch 1/4
         30/30 [================ ] - 1s 41ms/step - loss: 0.4995 - accura
         cy: 0.7743 - val_loss: 0.3606 - val_accuracy: 0.8675
         Epoch 2/4
         30/30 [================ ] - 1s 29ms/step - loss: 0.2746 - accura
         cy: 0.9061 - val loss: 0.3846 - val accuracy: 0.8370
         30/30 [============= ] - 1s 29ms/step - loss: 0.2040 - accura
         cy: 0.9291 - val_loss: 0.2785 - val_accuracy: 0.8879
         Epoch 4/4
         30/30 [================ ] - 1s 28ms/step - loss: 0.1580 - accura
         cy: 0.9463 - val loss: 0.2820 - val accuracy: 0.8882
In [44]: model.evaluate(x test, y test)
         782/782 [================ ] - 2s 2ms/step - loss: 0.2990 - accur
         acy: 0.8807
Out[44]: [0.29896876215934753, 0.8806800246238708]
In [ ]:
In [ ]:
```