```
In [1]:
           1 import pandas as pd
           2 from sklearn.model_selection import train_test_split
           3 from keras.models import Sequential
             from keras.layers import Activation,Dense
         C:\Users\Anusha V\anaconda3\lib\site-packages\scipy\__init__.py:155: UserW
         arning: A NumPy version >=1.18.5 and <1.25.0 is required for this version
         of SciPy (detected version 1.26.2
           warnings.warn(f"A NumPy version >={np_minversion} and <{np_maxversion}"</pre>
In [6]:
           1 import pandas as pd
           2 from sklearn.model_selection import train_test_split
           3 from keras.models import Sequential
           4 from keras.layers import Activation, Dense
In [7]:
              import pandas as pd
           2 data=pd.read_csv(r"C:\Users\Anusha V\Downloads\heart1.csv")
              data.head()
Out[7]:
             age sex
                         trestbps chol fbs
                                           restecg thalach exang oldpeak slope ca thal targe
          0
              52
                       0
                             125
                                  212
                                                     168
                                                                    1.0
                                                                           2
                                                                               2
                                                                                   3
                   1
                                        0
                                                1
                                                              0
              53
                       0
                                  203
                                                     155
                                                                               0
          1
                   1
                             140
                                        1
                                                0
                                                              1
                                                                    3.1
                                                                           0
                                                                                   3
          2
              70
                       0
                             145
                                  174
                                        0
                                                     125
                   1
                                                1
                                                              1
                                                                    2.6
                                                                           0
                                                                               0
                                                                                   3
          3
              61
                       0
                             148
                                  203
                                        0
                                                1
                                                     161
                                                              0
                                                                    0.0
                                                                                   3
                   1
                                                                           2
                                                                               1
              62
                   0
                             138
                                  294
                                                     106
                                                              0
                                                                    1.9
                                                                               3
                                                                                   2
              x = data.drop(columns=['age'])
In [8]:
              y = data['sex']
In [9]:
              x_train, x_test, y_train, y_test = train_test_split(x, y, test_size=0.2
In [11]:
              model = Sequential()
           1
              model.add(Dense(32, activation='relu', input_shape=(x_train.shape[1],))
              model.add(Dense(16, activation='relu'))
              model.add(Dense(1, activation='sigmoid'))
           4
           5
```

```
Epoch 1/20
21/21 [============= ] - 3s 41ms/step - loss: 14.4664 - ac
curacy: 0.3095 - val_loss: 1.3308 - val_accuracy: 0.5915
Epoch 2/20
uracy: 0.7088 - val_loss: 1.3201 - val_accuracy: 0.6585
Epoch 3/20
21/21 [============= ] - 0s 14ms/step - loss: 0.8365 - acc
uracy: 0.6159 - val_loss: 0.8179 - val_accuracy: 0.6280
Epoch 4/20
uracy: 0.6585 - val_loss: 0.7886 - val_accuracy: 0.5610
Epoch 5/20
racy: 0.6784 - val_loss: 0.7662 - val_accuracy: 0.5427
Epoch 6/20
racy: 0.6723 - val_loss: 0.7319 - val_accuracy: 0.6098
Epoch 7/20
uracy: 0.6768 - val_loss: 0.7145 - val_accuracy: 0.6463
Epoch 8/20
uracy: 0.6768 - val_loss: 0.6915 - val_accuracy: 0.6098
Epoch 9/20
uracy: 0.7088 - val_loss: 0.6830 - val_accuracy: 0.6159
Epoch 10/20
uracy: 0.7134 - val_loss: 0.6702 - val_accuracy: 0.6280
Epoch 11/20
21/21 [============= ] - 0s 13ms/step - loss: 0.5848 - acc
uracy: 0.7073 - val_loss: 0.6567 - val_accuracy: 0.6402
Epoch 12/20
uracy: 0.7119 - val_loss: 0.6549 - val_accuracy: 0.6585
Epoch 13/20
21/21 [============ ] - 0s 11ms/step - loss: 0.5685 - acc
uracy: 0.7165 - val_loss: 0.6417 - val_accuracy: 0.6280
Epoch 14/20
21/21 [=============== ] - 0s 15ms/step - loss: 0.5608 - acc
uracy: 0.7195 - val_loss: 0.6283 - val_accuracy: 0.6402
uracy: 0.7088 - val_loss: 0.6176 - val_accuracy: 0.6524
Epoch 16/20
uracy: 0.7256 - val_loss: 0.6260 - val_accuracy: 0.6707
Epoch 17/20
uracy: 0.7317 - val_loss: 0.6032 - val_accuracy: 0.6768
21/21 [=========== ] - 0s 11ms/step - loss: 0.5299 - acc
uracy: 0.7271 - val_loss: 0.5844 - val_accuracy: 0.6585
Epoch 19/20
uracy: 0.7393 - val_loss: 0.5889 - val_accuracy: 0.6707
Epoch 20/20
uracy: 0.7332 - val_loss: 0.5672 - val_accuracy: 0.6768
```

Out[12]: <keras.src.callbacks.History at 0x25b49c5eca0>

```
1 test_loss, test_acc = model.evaluate(x_test, y_test)
In [13]:
             print('Test accuracy:', test_acc)
        cy: 0.7220
        Test accuracy: 0.7219512462615967
In [ ]:
In [ ]:
In [14]:
          1 from nltk import ngrams
          2 from nltk.tokenize import word_tokenize
          3 | text = "The greatest glory in Living lies not in never falling but in r
            words = word_tokenize(text)
          5 def generate_ngrams(tokens, n):
                n_grams = ngrams(tokens, n)
          6
                return [' '.join(gram) for gram in n_grams]
          8 bi_grams = generate_ngrams(words, 2)
          9 print("Bi-grams:", bi_grams)
         10 | tri_grams = generate_ngrams(words, 3)
         11 print("Tri-grams:", tri_grams)
         12
        Bi-grams: ['The greatest', 'greatest glory', 'glory in', 'in Living', 'Liv
        ing lies', 'lies not', 'not in', 'in never', 'never falling', 'falling bu
        t', 'but in', 'in raising', 'raising every', 'every Lies']
        Tri-grams: ['The greatest glory', 'greatest glory in', 'glory in Living',
        'in Living lies', 'Living lies not', 'lies not in', 'not in never', 'in ne
        ver falling', 'never falling but', 'falling but in', 'but in raising', 'in
        raising every', 'raising every Lies']
In [ ]:
```