```
In [1]:
         import pandas as pd
         import numpy as np
         import matplotlib.pyplot as plt
         import seaborn as sns
In [2]: data = pd.read_csv('dataANN.csv')
In [3]: data.shape
Out[3]: (70000, 14)
In [4]: data
Out[4]:
                     id age_days
                                   age_year gender height weight ap_hi ap_lo cholesterol gluc smok
                     0
                           18393 50.391781
                                                  2
                                                              62.0
              0
                                                       168
                                                                      110
                                                                             80
                                                                                          1
                                                                                               1
              1
                     1
                           20228 55.419178
                                                       156
                                                              85.0
                                                                     140
                                                                             90
                                                                                          3
                                                  1
                                                                                               1
              2
                     2
                           18857 51.663014
                                                       165
                                                              64.0
                                                                      130
                                                                             70
                                                                                          3
                                                                                               1
              3
                     3
                           17623 48.282192
                                                  2
                                                       169
                                                              82.0
                                                                      150
                                                                            100
                                                                                          1
                                                                                               1
                     4
              4
                           17474 47.873973
                                                              56.0
                                                                     100
                                                  1
                                                       156
                                                                             60
                                                                                          1
                                                                                               1
                                                        ...
                                                               ...
                                                                      ...
                                                                             ...
           69995
                 99993
                           19240 52.712329
                                                  2
                                                       168
                                                              76.0
                                                                     120
                                                                                          1
                                                                                               1
                                                                             80
           69996
                 99995
                           22601 61.920548
                                                       158
                                                             126.0
                                                                     140
                                                                             90
                                                                                          2
                                                                                               2
           69997
                 99996
                           19066 52.235616
                                                  2
                                                       183
                                                             105.0
                                                                     180
                                                                             90
                                                                                          3
                                                                                               1
                                                                                               2
           69998
                 99998
                                                              72.0
                                                                      135
                                                                                          1
                           22431 61.454795
                                                  1
                                                       163
                                                                             80
          69999
                 99999
                           20540 56.273973
                                                       170
                                                              72.0
                                                                      120
                                                                             80
                                                                                          2
                                                                                               1
         70000 rows × 14 columns
In [5]: | X = data.iloc[:, 1:13].values
         y = data.iloc[:, 13].values
```

```
In [6]: X
 Out[6]: array([[1.83930000e+04, 5.03917808e+01, 2.00000000e+00, ...,
                 0.00000000e+00, 0.00000000e+00, 1.00000000e+00],
                [2.02280000e+04, 5.54191781e+01, 1.00000000e+00, ...,
                 0.00000000e+00, 0.00000000e+00, 1.00000000e+00],
                [1.88570000e+04, 5.16630137e+01, 1.00000000e+00, ...,
                 0.00000000e+00, 0.00000000e+00, 0.00000000e+00],
                [1.90660000e+04, 5.22356164e+01, 2.00000000e+00, ...,
                 0.00000000e+00, 1.00000000e+00, 0.00000000e+00],
                [2.24310000e+04, 6.14547945e+01, 1.00000000e+00, ...,
                 0.00000000e+00, 0.00000000e+00, 0.00000000e+00],
                [2.05400000e+04, 5.62739726e+01, 1.00000000e+00, ...,
                 0.00000000e+00, 0.00000000e+00, 1.00000000e+00]])
 In [7]: y
 Out[7]: array([0, 1, 1, ..., 1, 1, 0], dtype=int64)
 In [8]: from sklearn.preprocessing import LabelEncoder
         labelencoder_X_1 = LabelEncoder()
         y = labelencoder X 1.fit transform(y)
 In [9]: y
 Out[9]: array([0, 1, 1, ..., 1, 1, 0], dtype=int64)
In [11]: from sklearn.model selection import train test split
         X_train, X_test, y_train, y_test = train_test_split(X, y, test_size = 0.2, random
In [12]: | from sklearn.preprocessing import StandardScaler
         sc = StandardScaler()
         X train = sc.fit transform(X train)
         X test = sc.transform(X test)
In [13]: X train
Out[13]: array([[ 1.5985301 , 1.5985301 , -0.73535519, ..., -0.31159702,
                 -0.23933665, 0.49416152],
                [0.01419727, 0.01419727, -0.73535519, ..., -0.31159702,
                 -0.23933665, 0.49416152],
                [0.56581583, 0.56581583, -0.73535519, \ldots, -0.31159702,
                 -0.23933665, -2.02362986],
                 . . . ,
                [1.17782468, 1.17782468, 1.35988704, ..., -0.31159702,
                 -0.23933665, 0.49416152],
                [-0.46933025, -0.46933025, -0.73535519, \ldots, -0.31159702,
                 -0.23933665, 0.49416152],
                [0.38302379, 0.38302379, 1.35988704, ..., -0.31159702,
                 -0.23933665, 0.49416152]])
```

```
In [14]: | X test
Out[14]: array([[-1.72779877, -1.72779877, -0.73535519, ..., -0.31159702,
                 -0.23933665, 0.49416152],
                [0.13619373, 0.13619373, -0.73535519, ..., -0.31159702,
                 -0.23933665, 0.49416152],
                [0.43936103, 0.43936103, -0.73535519, \ldots, -0.31159702,
                 -0.23933665, 0.49416152],
                . . . ,
                [1.21713916, 1.21713916, 1.35988704, ..., -0.31159702,
                 -0.23933665, -2.02362986],
                [-1.96652275, -1.96652275, 1.35988704, ..., 3.20927329,
                 -0.23933665, 0.49416152],
                [-0.22209489, -0.22209489, -0.73535519, ..., -0.31159702,
                 -0.23933665, 0.49416152]])
In [15]: X_train.shape
Out[15]: (56000, 12)
In [16]: X test.shape
Out[16]: (14000, 12)
In [17]:
         import keras
         from keras.models import Sequential
         from keras.layers import Dense, Dropout
         from keras import models
         from keras import layers
In [18]: | classifier = models.Sequential()
In [19]: | classifier.add(layers.Dense(12, activation='relu', use_bias=True, name = 'input'))
         classifier.add(layers.Dropout(rate=0.1))
         classifier.add(layers.Dense(12, activation='relu',use bias=True,name = 'hidden1')
In [20]:
         classifier.add(layers.Dropout(rate=0.1))
In [21]: classifier.add(layers.Dense(1, activation='sigmoid',use bias=True,name = 'output'
In [29]: classifier.compile(optimizer='adam', loss='binary crossentropy', metrics=['accura
```

```
In [30]: history = classifier.fit(X_train, y_train, batch_size=100, epochs=100,validation
        acy: 0.7341 - val loss: 0.5624 - val accuracy: 0.7223
        Epoch 79/100
        acy: 0.7334 - val_loss: 0.5613 - val_accuracy: 0.7223
        Epoch 80/100
        549/549 [============= ] - 1s 1ms/step - loss: 0.5460 - accur
        acy: 0.7337 - val loss: 0.5618 - val accuracy: 0.7196
        Epoch 81/100
        549/549 [============= ] - 1s 1ms/step - loss: 0.5459 - accur
        acy: 0.7328 - val loss: 0.5627 - val accuracy: 0.7223
        Epoch 82/100
        549/549 [============= ] - 1s 1ms/step - loss: 0.5456 - accur
        acy: 0.7342 - val_loss: 0.5620 - val_accuracy: 0.7170
        Epoch 83/100
        549/549 [============= ] - 1s 1ms/step - loss: 0.5453 - accur
        acy: 0.7340 - val_loss: 0.5607 - val_accuracy: 0.7214
        Epoch 84/100
        549/549 [============= ] - 1s 1ms/step - loss: 0.5457 - accur
        acy: 0.7339 - val_loss: 0.5619 - val_accuracy: 0.7250
        Fnoch 25/100
```

In [31]: classifier.summary()

Model: "sequential"

Layer (type)	Output Shape	Param #
input (Dense)	(None, 12)	156
dropout (Dropout)	(None, 12)	0
hidden1 (Dense)	(None, 12)	156
dropout_1 (Dropout)	(None, 12)	0
output (Dense)	(None, 1)	13

Total params: 325
Trainable params: 325
Non-trainable params: 0

```
In [32]: | classifier.weights
Out[32]: [<tf.Variable 'input/kernel:0' shape=(12, 12) dtype=float32, numpy=</pre>
          array([ 3.83899808e-01, 1.69875354e-01, 1.31449610e-01,
                  -5.87779820e-01, 2.34277979e-01, -1.80614993e-01,
                   -4.30697612e-02, -4.36798662e-01, 3.78667116e-01,
                   2.19583318e-01, 1.91744030e-01, -1.89905941e-01],
                 [ 5.99754393e-01, -8.85015503e-02, 3.74571681e-01,
                   2.10306183e-01, -2.19680190e-01, 1.69994906e-01,
                   1.52537206e-04, 4.21699017e-01, 2.50457585e-01,
                   -2.23337755e-01, -2.62025326e-01, -1.05706677e-01],
                 [-1.71571448e-01, 1.20864145e-03, 2.18342021e-01,
                  -2.10637208e-02, -7.51293777e-03, 1.07831676e-02,
                   3.61952223e-02, -2.58291513e-02, 2.91432198e-02,
                   7.97408447e-02, 1.52941672e-02, -9.51172411e-03],
                 [ 9.13591310e-02, 1.28161669e-01, -1.54117823e-01,
                  -5.98947778e-02, -1.02040470e-02, -1.29963206e-02,
                   4.14380021e-02, -1.83649827e-02, -2.07286794e-02,
                   6.34928867e-02, 1.15355207e-02, 2.87403492e-03],
                 [ 1.41857406e-02, -3.10321122e-01, 5.23836732e-01,
                   -2.32308537e-01,
                                    4.24790680e-02,
                                                     4.41529416e-02,
In [33]:
         plt.plot(history.history['accuracy'])
         plt.show()
          0.735
          0.734
          0.733
          0.732
          0.731
          0.730
                        20
                                                 80
                                                         100
In [34]: y pred = classifier.predict(X test)
         y_pred = (y_pred > 0.5)
In [35]: from sklearn.metrics import confusion_matrix
         cm = confusion_matrix(y_test, y_pred)
         cm
Out[35]: array([[5428, 1641],
                [2085, 4846]], dtype=int64)
         print("Our accuracy is {}%".format(((cm[0][0] + cm[1][1])/14000)*100))
In [36]:
         Our accuracy is 73.38571428571429%
```