

### Program 5

```
import numpy as np

X = np.array([[2, 9], [1, 5], [3, 6]], dtype=float)  # X = (hours sleeping, hours studying)

y = np.array([[92], [86], [89]], dtype=float)      # y = score on test

# scale units

X = X/np.amax(X, axis=0)    # maximum of X array

y = y/100                  # max test score is 100

class Neural_Network(object):

    def __init__(self):

        # Parameters

        self.inputSize = 2

        self.outputSize = 1

        self.hiddenSize = 3

        # Weights

        self.W1 = np.random.randn(self.inputSize, self.hiddenSize)  # (3x2) weight matrix from input
        to hidden layer

        self.W2 = np.random.randn(self.hiddenSize, self.outputSize)  # (3x1) weight matrix from
        hidden to output layer

    def forward(self, X):

        #forward propagation through our network

        self.z = np.dot(X, self.W1)    # dot product of X (input) and first set of 3x2 weights

        self.z2 = self.sigmoid(self.z)    # activation function

        self.z3 = np.dot(self.z2, self.W2)    # dot product of hidden layer (z2) and second set of 3x1
        weights

        o = self.sigmoid(self.z3)    # final activation function

        return o
```

```
def sigmoid(self, s):  
    return 1/(1+np.exp(-s))    # activation function
```

```
def sigmoidPrime(self, s):  
    return s * (1 - s)    # derivative of sigmoid
```

```
def backward(self, X, y, o):  
    # backward propagate through the network  
  
    self.o_error = y - o    # error in output  
  
    self.o_delta = self.o_error*self.sigmoidPrime(o) # applying derivative of sigmoid to  
    self.z2_error = self.o_delta.dot(self.W2.T)    # z2 error: how much our hidden layer weights  
    contributed to output error  
  
    self.z2_delta = self.z2_error*self.sigmoidPrime(self.z2) # applying derivative of sigmoid to z2  
    error  
  
    self.W1 += X.T.dot(self.z2_delta)    # adjusting first set (input --> hidden) weights  
  
    self.W2 += self.z2.T.dot(self.o_delta) # adjusting second set (hidden --> output) weights
```

```
def train (self, X, y):  
    o = self.forward(X)  
    self.backward(X, y, o)
```

```
NN = Neural_Network()  
  
print ("\nInput: \n" + str(X))  
  
print ("\nActual Output: \n" + str(y))  
  
print ("\nPredicted Output: \n" + str(NN.forward(X)))  
  
print ("\nLoss: \n" + str(np.mean(np.square(y - NN.forward(X)))))    # mean sum squared loss  
  
NN.train(X, y)
```

Input:

[[0.66666667 1. ]

[0.33333333 0.55555556]

[1. 0.66666667]]

Actual Output:

[[0.92]

[0.86]

[0.89]]

Predicted Output:

[[0.37569264]

[0.37059885]

[0.36376607]]

Loss:

0.2709020442986832