```
In [1]:
            import numpy as np
In [2]:
          1 def sigmoid (x):
                 return 1/(1 + np.exp(-x))
In [3]:
          1 def sigmoid_derivative(x):
                 return x * (1 - x)
In [5]:
          1 #Input datasets
          3 inputs = np.array([[0,0],[0,1],[1,0],[1,1]])
          4 expected_output = np.array([[0],[1],[1],[0]])
In [6]:
          1 epochs = 10000
          2 | lr = 0.1
          3 inputLayerNeurons, hiddenLayerNeurons, outputLayerNeurons = 2,2,1
In [7]:
          1 #Random weights and bias initialization
          3 hidden_weights = np.random.uniform(size=(inputLayerNeurons, hiddenLayerN
          4 hidden_bias =np.random.uniform(size=(1,hiddenLayerNeurons))
          5 output weights = np.random.uniform(size=(hiddenLayerNeurons,outputLayer
          6 | output_bias = np.random.uniform(size=(1,outputLayerNeurons))
In [8]:
          1 print("Initial hidden weights: ",end='')
          2 print(*hidden_weights)
          3 print("Initial hidden biases: ",end='')
          4 print(*hidden_bias)
          5 print("Initial output weights: ",end='')
          6 print(*output_weights)
          7 print("Initial output biases: ",end='')
          8 print(*output_bias)
          9
        Initial hidden weights: [0.84730363 0.76859947] [0.91037732 0.31547999]
        Initial hidden biases: [0.21788549 0.33373447]
        Initial output weights: [0.41528087] [0.32743761]
        Initial output biases: [0.0497275]
```

```
In [10]:
           1 #Training algorithm
           2
           3
             for _ in range(epochs):
           4
           5
             #Forward Propagation
           6
                 hidden_layer_activation = np.dot(inputs,hidden_weights)
           7
                 hidden_layer_activation += hidden_bias
                 hidden_layer_output = sigmoid(hidden_layer_activation)
           8
           9
                 output layer activation = np.dot(hidden layer output,output weights
          10
                 output_layer_activation += output_bias
          11
                 predicted output = sigmoid(output layer activation)
          12
          13
          14 #Backpropagation
          15
                 error = expected_output - predicted_output
                 d_predicted_output = error * sigmoid_derivative(predicted_output)
          16
                 error hidden layer = d predicted output.dot(output weights.T)
          17
          18
                 d_hidden_layer = error_hidden_layer * sigmoid_derivative(hidden_lay
          19
          20
In [11]:
           1 #Updating Weights and Biases
           3 output weights += hidden layer output.T.dot(d predicted output) * lr
           4 output bias += np.sum(d predicted output,axis=0,keepdims=True) * lr
           5 hidden weights += inputs.T.dot(d hidden layer) * lr
           6 hidden bias += np.sum(d hidden layer,axis=0,keepdims=True) * lr
In [12]:
           1 print("Final hidden weights: ",end='')
           2 print(*hidden_weights)
           3 print("Final hidden bias: ",end='')
           4 print(*hidden bias)
           5 print("Final output weights: ",end='')
           6 print(*output_weights)
           7
             print("Final output bias: ",end='')
           8 print(*output_bias)
         Final hidden weights: [0.84728475 0.7683328 ] [0.9103537 0.31532894]
         Final hidden bias: [0.21700646 0.33291834]
         Final output weights: [0.40645002] [0.31851796]
         Final output bias: [0.0366907]
           1 print("\nOutput from neural network after 10,000 epochs: ",end='')
In [13]:
           2 print(*predicted_output)
         Output from neural network after 10,000 epochs: [0.61554638] [0.64073409]
         [0.64665556] [0.66329578]
In [ ]:
```