```
In [1]:
import numpy as np
In [2]:
input_value = np.array([[0,0],[0,1],[1,1],[1,0]])
print(input_value.shape)
input_value
(4, 2)
Out[2]:
array([[0, 0],
       [0, 1],
       [1, 1],
       [1, 0]])
In [3]:
output = np.array([0,1,1,0])
output = output.reshape(4,1)
output.shape
Out[3]:
(4, 1)
In [4]:
weights = np.array([[0.1],[0.2]])
weights
Out[4]:
array([[0.1],
       [0.2]])
In [5]:
bias = 0.3
In [6]:
# activation function
In [7]:
def sigmoid_func(x):
    return 1/(1+np.exp(-x))
In [8]:
def der(x):
    return sigmoid_func(x)*(1-sigmoid_func(x))
```

In [9]:

```
#udating the weights
```

In [10]:

```
for epochs in range(10000):
    input_arr = input_value
    weighted_sum = np.dot(input_arr,weights)+bias
    first_output = sigmoid_func(weighted_sum)
    error = first_output - output # error in the prediction
    total_error = np.square(np.subtract(first_output,output).mean())# ((sum of all error)/n
    #print(total_error)
    first_der = error
    second_der = der(first_output)
    derivative = first_der*second_der
    t_input = input_value.T # transpose
    final_derivative = np.dot(t_input,derivative)
# update weights
weights = weights - 0.05*final_derivative
#update bias
for i in derivative:
    bias = bias - 0.05 * i
print(weights)
print(bias)
```

```
[[0.09714683]
[0.20828807]]
[0.29481765]
```

In [11]:

```
#predictions
pred = np.array([0,1])
result = np.dot(pred,weights)+bias
res = sigmoid_func(result)
print(res)
```

[0.62318891]