# In [18]:

```
import numpy as np
```

# In [19]:

```
def sigmoid(z):
    return 1 / (1 + np.exp(-z))
```

### In [20]:

## In [21]:

```
def forward_propagation(X, Y, parameters):
    m = X.shape[1]
    W1 = parameters["W1"]
    W2 = parameters["W2"]
    b1 = parameters["b1"]
    b2 = parameters["b2"]

Z1 = np.dot(W1, X) + b1
    A1 = sigmoid(Z1)
    Z2 = np.dot(W2, A1) + b2
    A2 = sigmoid(Z2)

cache = (Z1, A1, W1, b1, Z2, A2, W2, b2)
    logprobs = np.multiply(np.log(A2), Y) + np.multiply(np.log(1 - A2), (1 - Y))
    cost = -np.sum(logprobs) / m
    return cost, cache, A2
```

#### In [22]:

## In [23]:

```
def updateParameters(parameters, gradients, learningRate):
    parameters["W1"] = parameters["W1"] - learningRate * gradients["dW1"]
    parameters["W2"] = parameters["W2"] - learningRate * gradients["dW2"]
    parameters["b1"] = parameters["b1"] - learningRate * gradients["db1"]
    parameters["b2"] = parameters["b2"] - learningRate * gradients["db2"]
    return parameters
```

# In [24]:

```
X = np.array([[0, 0, 1, 1], [0, 1, 0, 1]])
Y = np.array([[0, 1, 1, 0]])

hidden_neurons = 2
input_feat = X.shape[0]
output_feat = Y.shape[0]
parameters = initialization(input_feat, hidden_neurons, output_feat)
epoch = 100000
learningRate = 0.01
losses = np.zeros((epoch, 1))

for i in range(epoch):
    losses[i, 0], cache, A2 = forward_propagation(X, Y, parameters)
    gradients = backward_propagation(X, Y, cache)
    parameters = updateParameters(parameters, gradients, learningRate)
```

### In [25]:

```
X = np.array([[1, 1, 0, 0], [0, 1, 0, 1]]) # XOR input
cost, _, A2 = forward_propagation(X, Y, parameters)
prediction = (A2 > 0.5) * 1.0
print(prediction)
```

```
[[1. 1. 0. 0.]]
```