```
In [30]:
```

```
import numpy as n
```

In [31]:

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

In [32]:

```
def layers(X, Y,h):
    n_x = X.shape[0] # size of input layer
    n_h = h
    n_y = Y.shape[0] # size of output layer
    return (n_x, n_h, n_y)
```

In [34]:

```
a = input('Enter the weights: ').split(',')
a = list(map(float, a))
i = np.array([a])
X = i.T
h = int(input('Neuron in hidden layer: '))
y = np.array([1])
(n_x, n_h, n_y) = layers(X, y,h)
print("The size of the input layer is: n_x = " + str(n_x))
print("The size of the hidden layer is: n_h = " + str(n_h))
print("The size of the output layer is: n_y = " + str(n_y))
```

```
Enter the weights: 0.4,0.6,0.9

Neuron in hidden layer: 5

The size of the input layer is: n_x = 3

The size of the hidden layer is: n_y = 1

The size of the output layer is: n_y = 1
```

In [35]:

```
In [36]:
```

```
parameters = network_parameters(n_x, n_h, n_y)
print("Weights1 = " + str(parameters["W1"]))
print("bias1 = " + str(parameters["b1"]))
print("Weights2 = " + str(parameters["W2"]))
print("bias2 = " + str(parameters["b2"]))
Weights1 = [[ 0.0625245 -0.01605134 -0.07688364]
 [-0.02300307 0.07450563 0.19761108]
 [-0.12441233 -0.06264169 -0.08037661]
 [-0.24190832 -0.0923792 -0.10238758]
 [ 0.1123978 -0.01319142 -0.16232854]]
bias1 = [[ 0.64667545]
 [-0.35627076]
 [-1.74314104]
 [-0.59664964]
 [-0.58859438]]
Weights2 = [[-0.08738823  0.00297138 -0.22482578 -0.02677619  0.10131834]]
bias2 = [[0.85279784]]
In [37]:
#Forward propogation
```

In [38]:

In [39]:

In [40]:

```
#Compute cost of network
```

In [41]:

```
def compute_loss(A2, Y, parameters):
    if len(Y.shape)==2:
        m = Y.shape[1]
    else:
        m = 1
    logprobs = np.multiply(np.log(A2),Y) + np.multiply(np.log(1-A2),(1-Y))
    cost = - (np.sum(logprobs)/m)
    cost = float(np.squeeze(cost))
    return cost
```

In [42]:

```
print("cost = " + str(compute_loss(A2, y, parameters)))
```

cost = 0.372864724386385

In [43]:

```
#backpropogate the cost
```

In [44]:

```
def backward_propagation(parameters, cache, X, Y):
    m = X.shape[1]
    Weights1 = parameters["W1"]
    Weights2 = parameters["W2"]
    A1 = cache['A1']
    A2 = cache['A2']
    dZ2 = A2-Y
    dW2 = (np.dot(dZ2,A1.T)/m)
    db2 = (np.sum(dZ2, axis=1,keepdims=True)/m)
    dZ1 = np.dot(Weights2.T, dZ2) * (1-np.power(A1,2))
    dW1 = (np.dot(dZ1, X.T)/m)
    db1 = (np.sum(dZ1, axis=1,keepdims=True)/m)
    grads = {"dW1": dW1,}
             "db1": db1,
             "dW2": dW2,
             "db2": db2}
    return grads
```

```
In [45]:
```

```
grads = backward_propagation(parameters, cache, X, y)
print ("dW1 = "+ str(grads["dW1"]))
print ("db1 = "+ str(grads["db1"]))
print ("dW2 = "+ str(grads["dW2"]))
print ("db2 = "+ str(grads["db2"]))
dW1 = [[ 0.00636711  0.00955067  0.014326 ]]
 [-0.00029017 -0.00043525 -0.00065288]
 [ 0.02751856  0.04127784  0.06191676]
 [ 0.00303087  0.0045463
                           0.00681945]
 [-0.01122065 -0.01683097 -0.02524646]]
db1 = [[ 0.01591778]
 [-0.00072542]
 [ 0.0687964 ]
 [ 0.00757717]
 [-0.02805162]]
dW2 = [[-0.2004461 -0.14451913 -0.04039559 -0.09378533 -0.10343629]]
db2 = [[-0.3112416]]
In [46]:
```

```
#update parameters
```

In [47]:

```
def update_parameters(parameters, grads, learning_rate = 0.2):
   Weights1 = parameters['W1']
   bias1 = parameters['b1']
   Weights2 = parameters['W2']
   bias2 = parameters['b2']
   dW1 = grads['dW1']
   db1 = grads['db1']
   dW2 = grads['dW2']
   db2 = grads['db2']
   Weights1 = Weights1 - (learning_rate*dW1)
   bias1 = bias1 - (learning_rate*db1)
   Weights2 = Weights2 - (learning_rate*dW2)
   bias2 = bias2 - (learning_rate*db2)
   parameters = {"W1": Weights1,
                  "b1": bias1,
                  "W2": Weights2,
                  "b2": bias2}
   return parameters
```

In [48]:

```
parameters = update_parameters(parameters, grads)
print("Weights1 = " + str(parameters["W1"]))
print("bias1 = " + str(parameters["b1"]))
print("Weights2 = " + str(parameters["W2"]))
print("bias2 = " + str(parameters["b2"]))
Weights1 = [[ 0.06125107 -0.01796147 -0.07974883]
 [-0.02294504 0.07459268 0.19774165]
 [-0.12991604 -0.07089726 -0.09275996]
 [-0.24251449 -0.09328846 -0.10375147]
 [ 0.11464193 -0.00982523 -0.15727925]]
bias1 = [[ 0.6434919 ]]
 [-0.35612567]
 [-1.75690032]
 [-0.59816508]
 [-0.58298406]]
Weights2 = [[-0.04729901  0.03187521 -0.21674666 -0.00801912  0.1220056 ]]
bias2 = [[0.91504616]]
In [49]:
```

```
#Integrate everything
```

In [56]:

```
def nn_model(X, Y, n_h, num_iterations = 10000, print_cost=False):
    np.random.seed(3)
    n_x = layers(X, Y,n_h)[0]
    n_y = layers(X, Y,n_h)[2]

parameters = network_parameters(X.shape[0], n_h, Y.shape[0])

# Loop (gradient descent)

for i in range(0, num_iterations):
    A2, cache = propagate(X, parameters)
    cost = compute_loss(A2, Y, parameters)
    grads = backward_propagation(parameters, cache, X, Y)
    parameters = update_parameters(parameters, grads)
    if print_cost and i%1000 == 0:
        print ("Cost after iteration %i: %f" %(i, cost))
    return parameters
```

```
In [57]:
```

```
parameters = nn_model(X, y, h, num_iterations=10000, print_cost=True)
print("Weights1 = " + str(parameters["W1"]))
print("bias1 = " + str(parameters["b1"]))
print("Weights2 = " + str(parameters["W2"]))
print("bias2 = " + str(parameters["b2"]))
Cost after iteration 0: 1.141313
Cost after iteration 1000: 0.001117
Cost after iteration 2000: 0.000523
Cost after iteration 3000: 0.000338
Cost after iteration 4000: 0.000248
Cost after iteration 5000: 0.000196
Cost after iteration 6000: 0.000162
Cost after iteration 7000: 0.000137
Cost after iteration 8000: 0.000119
Cost after iteration 9000: 0.000106
Weights1 = [[0.54045013 0.58603192 0.82322114]
 [0.22171776 0.58436173 0.88267493]
 [0.47974685 0.66933143 1.09366543]
 [0.26922321 0.34403104 0.80158852]
 [0.41074391 0.65487547 0.73088061]]
bias1 = [[ 0.4992908 ]
 [ 0.47480764]
 [-0.32642482]
 [ 1.77472997]
 [-0.29453735]]
Weights2 = [[1.63965598 1.59481657 1.34490308 2.53282633 1.09274976]]
bias2 = [[2.2911678]]
In [58]:
def predict(parameters, X):
   A2, cache = propagate(X, parameters)
   predictions = np.where(A2 > 0.5, 1, 0)
   return predictions
In [59]:
predictions = predict(parameters, X)
print("predictions = " + str(predictions))
predictions = [[1]]
In [ ]:
```