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[1]: import numpy as np
 [2]: def sigmoid(z):
    return 1 / (1 + np.exp(-z))
[3]: def initialization(input_feat, hidden_neurons, output_feat):
    W1 = np.random.randn(hidden_neurons, input_feat)
    W2 = np.random.randn(output_feat, hidden_neurons)
    b1 = np.zeros((hidden_neurons, 1))
    b2 = np.zeros((output_feat, 1))
               [4]: 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
[5]: def backward_propagation(X, Y, cache):
    m = X.shape[1]
    (Z1, A1, W1, b1, Z2, A2, W2, b2) = cache
                        \label{eq:dz2} \begin{array}{lll} dZ2 &= A2 - Y \\ dW2 &= np.dot(dZ2, A1.T) \ / \ m \\ db2 &= np.sum(dZ2, axis = 1, keepdims = True) \end{array} 
                       dA1 = np.dot(W2.T, dZ2)
dZ1 = np.multiply(dA1, A1 * (1- A1))
dW1 = np.dot(dZ1, X.T) / m
db1 = np.sum(dZ1, axis = 1, keepdims = True) / m
                      gradients = {"dZ2": dZ2, "dw2": dW2, "db2": db2, "dZ1": dZ1, "dw1": dW1, "db1": db1}
                    return gradients
[6]: 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
[7]: 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 = initlalization(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)
[8]: 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. 0. 0. 1.]]
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