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#!/usr/bin/env python
# -*- coding: utf-8 -*-
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   Simple Neural Network with 1 hidden layer with the number
    of hidden units as a hyperparameter to calculate the XOR function
import numpy as np
def sigmoid(z):
    return 1 / (1 + np.exp(-z))
def initialize_parameters(n_x, n_h, n_y):
    W1 = np.random.randn(n_h, n_x)
     b1 = np.zeros((n_h, 1))
    W2 = np.random.randn(n_y, n_h)
    b2 = np.zeros((n_y, 1))
    parameters = {
         "W1": W1,
"b1": b1,
         "W2": W2,
         "b2": b2
     return parameters
def forward prop(X, parameters):
     W1 = parameters["W1"]
    b1 = parameters["b1"]
     W2 = parameters["W2"]
    b2 = parameters["b2"]
    Z1 = np.dot(W1, X) + b1
    A1 = np.tanh(Z1)

Z2 = np.dot(W2, A1) + b2
    A2 = sigmoid(Z2)
    cache = {
    "A1": A1,
        "A2": A2
     return A2, cache
def calculate_cost(A2, Y):
def model(X, Y, n_x, n_h, n_y, num_of_iters, learning_rate):
    parameters = initialize_parameters(n_x, n_h, n_y)
     for i in range(0, num_of_iters + 1):
        a2, cache = forward_prop(X, parameters)
         cost = calculate_cost(a2, Y)
        grads = backward prop(X, Y, cache, parameters)
        parameters = update_parameters(parameters, grads, learning_rate)
        if (i % 100 == 0):
             print('Cost after iteration# {:d}: {:f}'.format(i, cost))
     return parameters
def predict(X, parameters):
     a2, cache = forward_prop(X, parameters)
     vhat = a2
     yhat = np.squeeze(yhat)
    if (yhat >= 0.5):
        y_predict = 1
    else:
        y predict = 0
    return y_predict
np.random.seed(2)
# The 4 training examples by columns
X = np.array([[0, 0, 1, 1], [0, 1, 0, 1]])
\mbox{\it \#} The outputs of the XOR for every example in X
Y = np.array([[1, 0, 0, 1]])
# No. of training examples
m = X.shape[1]
# Set the hypernarameters
n_x = 2 # No. of neurons in first layer n_x = 2 # No. of neurons in hidden layer
n v = 1 # No. of neurons in output layer
num_of_iters = 1000
learning rate = 0.3
\label{eq:trained_parameters} \verb| = model(X, Y, n_x, n_h, n_y, num_of_iters, learning_rate)| \\
\ensuremath{\text{\# Test 2X1}} vector to calculate the XOR of its elements.
# Try (0, 0), (0, 1), (1, 0), (1, 1) X_{test} = np.array([[1], [1]])
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cost = -np.sum(np.multiply(Y, np.log(A2)) + np.multiply(1 - Y, np.log(1 - A2))) \ / \ m
     cost = np.squeeze(cost)
def backward_prop(X, Y, cache, parameters):
    A1 = cache["A1"]
     A2 = cache["A2"]
    W2 = parameters["W2"]
    dW2 = np.dot(dZ2, A1.T) / m
db2 = np.sum(dZ2, axis=1, keepdims=True) / m
dZ1 = np.multiply(np.dot(W2.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,
          "dh2": dh2
     return grads
def update_parameters(parameters, grads, learning_rate):
     W1 = parameters["W1"]
b1 = parameters["b1"]
     W2 = parameters["W2"
    b2 = parameters["b2"]
     dW1 = grads["dW1"]
     db1 = grads["db1"]
     dW2 = grads["dW2"
     db2 = grads["db2"]
    W1 = W1 - learning rate * dW1
     b1 = b1 - learning_rate * db1
     W2 = W2 - learning_rate * dW2
     b2 = b2 - learning_rate * db2
     new parameters = {
         "W1": W1,
"W2": W2,
         "b1": b1,
          "b2": b2
     return new parameters
y_predict = predict(X_test, trained_parameters)
\label{lem:print('Neural Network prediction for example ({:d}, {:d}) is {:d}'.format(
     X_test[0][0], X_test[1][0], y_predict))
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