Q1 XOR Neural Network with Gradient Descent from Scratch

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In [1]:
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import numpy as np
import random as rand
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In [2]:
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class Layer:
        init (self, layer neurons, inputs=False, input bias=False):
        self.layer_neurons = layer_neurons
        self.input_bias = input_bias
        self.inputs = inputs + input_bias # If bias input would increase by 1
        self.delta weights = 0
        if inputs:
            self.weights = np.array([[rand.uniform(-1,1) for in range(self.layer neurons)] for in range
(self.inputs)]) # (inputs, layer_neurons) shape of each layer's weights
   # This method initializes delta weights
   def initialize delta weights(self):
        self.delta weights = np.array([[0 for in range(self.layer neurons)] for in range(self.inputs)])
class NeuralNetwork:
        init (self, inputs):
   def
        self.inputs = inputs
       self.layers = None
   # This method is used to add another layer to neural network, you can use it multiple times to add many
many layers
   def add_layer(self, layer_neurons, input_bias=False):
        if self.layers is None: # If first layer
           self.layers = []
           self.layers.append( Layer(layer neurons, inputs=self.inputs, input bias=input bias) )
        else:
           prev layer neurons = self.layers[-1].layer neurons # New layer's input would be from previous l
ayer
           self.layers.append( Layer(layer neurons, inputs=prev layer neurons, input bias=input bias) )
   # This method returns neuron valyes at a layer
   def neuron values at layer(self, layer wanted, input values):
        input_values_copy = input_values.copy()
        if layer_wanted < 0: # As I have not made a layer object of Input so if input layer's neurons are r</pre>
equired it would not exist in layers
           if self.layers[0].input bias:
               input values copy.append(1)
            return np.array(input_values_copy)
        for ind layer, layer in enumerate(self.layers):
           if layer.input bias:
               input values copy.append(1)
           neuron_values = np.matmul(input_values_copy, layer.weights)
             neuron_values = [self.sigmoid(value) for value in neuron_values] # As this was creating issue
and weights were behaving abnormally
           input_values_copy = neuron_values
           if ind layer == layer wanted:
                if layer.input bias:
                    neuron values = list(neuron values)
                    neuron values.append(1)
                return np.array(neuron values)
   # Calculate sigmoid
   def sigmoid (self, value):
        return 1 / (1 + np.exp(-value))
   # Forward propogate input value to current network with current weights
   def forward_propogate(self, input_values):
        input_values_copy = input_values.copy()
        for layer in self.layers:
           if layer.input bias:
               input_values_copy.append(1)
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neuron_values = np.matmul(input_values_copy, layer.weights)
            neuron_values = [self.sigmoid(value) for value in neuron_values]
            input values copy = neuron values
        output = self.sigmoid(neuron values[0])
#
          output = neuron values[0]
        return output
   # Gradient descent algorithm
   def gradient descent(self, train examples, labels, learning rate, epochs):
        # Step 1: Initialize each weight to small random value
            # Already doing while initalizing layers
        # Step 2: Until Termination condition is met
        for in range(epochs):
            # Step 3: Initialize delta weights to 0
            for layer in self.layers:
                layer.initialize_delta_weights()
            # Step 4: For each individual example in all training examples
            for train example in train examples:
                # Step 5: Forward propogate and compute output
                prediction = self.forward_propogate(train_example)
                # Step 6: Compute delta_weights
                for ind, layer in enumerate(self.layers):
                    error gradient = (labels[ind]-prediction) * (prediction*(1-prediction)) * self.neuron va
lues_at_layer(ind-1, train_example) # see equation in slides/video lecture for sigmoid
                    layer.delta weights = ( layer.delta weights.T + (learning rate * error gradient) ).T
            # Step 7: Update weights using delta weights
            for layer in self.layers:
                  print(layer.weights, '----Updating---')
                layer.weights = layer.weights + layer.delta weights
                 print(layer.weights, '\n')
In [3]:
model = NeuralNetwork(inputs=2)
model.add layer(layer neurons=2)
model.add_layer(layer_neurons=1)
x = [0,0], [0,1], [1,0], [1,1]
y = [0, 1, 1, 0]
model.gradient_descent(x, y, 0.01, 3000)
In [4]:
for ind, inp in enumerate(x):
   out = model.forward propogate(inp)
   print(inp, ' -> ', \( \frac{1}{2} \). format(out), ' ', 'actual:', y[ind], 'predicted:', 1 if out>0.5 else 0)
[0, 0] \rightarrow 0.500000000
                           actual: 0 predicted: 0
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[0, 1] -> 0.620745754

[1, 0] -> 0.621908601

[1, 1] -> 0.622459153

In []:

actual: 1 predicted: 1

actual: 1 predicted: 1

actual: 0 predicted: 1