neural-network-gradient-descent-from-scratch

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1 Q1 XOR Neural Network with Gradient Descent from Scratch

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[1]: import numpy as np
     import random as rand
[2]: class Layer:
        def __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:
        def __init__(self, inputs):
            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:
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prev_layer_neurons = self.layers[-1].layer_neurons # New layer'su
→ input would be from previous layer
           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 sou
→ if input layer's neurons are required 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 propagate 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)
           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]
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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 values 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')
[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)
[4]: for ind, inp in enumerate(x):
        out = model.forward_propogate(inp)
        print(inp, ' -> ', '{:.9f}'.format(out), ' ', 'actual:', y[ind], __
```

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[0, 0] -> 0.500000000 actual: 0 predicted: 0
[0, 1] -> 0.620745754 actual: 1 predicted: 1
[1, 0] -> 0.621908601 actual: 1 predicted: 1
[1, 1] -> 0.622459153 actual: 0 predicted: 1
```