## Classification using Multilayer Perceptron to solve XOR problem

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In [1]: import numpy as np
        def sigmoid(x):
            return 1.0/(1.0 + np.exp(-x))
        def sigmoid prime(x):
            return sigmoid(x)*(1.0-sigmoid(x))
        def tanh(x):
            return np.tanh(x)
        def tanh_prime(x):
            return 1.0 - x**2
In [2]: class MLP:
            def __init__(self, layers, activation='tanh'):
                if activation == 'sigmoid':
                    self.activation = sigmoid
                    self.activation_prime = sigmoid_prime
                elif activation == 'tanh':
                     self.activation = tanh
                    self.activation_prime = tanh_prime
                # Set weights
                self.weights = []
                for i in range(1, len(layers) - 1):
                     r = 2*np.random.random((layers[i-1] + 1, layers[i] + 1)) -1
                     self.weights.append(r)
                r = 2*np.random.random((layers[i] + 1, layers[i+1])) - 1
                self.weights.append(r)
            def fit(self, X, y, learning_rate=0.2, epochs=100000):
                ones = np.atleast 2d(np.ones(X.shape[0]))
                X = np.concatenate((ones.T, X), axis=1)
                for k in range(epochs):
                     if k % 10000 == 0: print('epochs:', k)
                     i = np.random.randint(X.shape[0])
                    a = [X[i]]
                     for 1 in range(len(self.weights)):
                         dot_value = np.dot(a[1], self.weights[1])
                         activation = self.activation(dot_value)
                         a.append(activation)
                     error = y[i] - a[-1]
                     deltas = [error * self.activation_prime(a[-1])]
                     for 1 in range(len(a) - 2, 0, -1):
                         deltas.append(deltas[-1].dot(self.weights[1].T)*self.activation_prime(a[1]))
                     deltas.reverse()
                     for i in range(len(self.weights)):
                         layer = np.atleast_2d(a[i])
                         delta = np.atleast_2d(deltas[i])
                         self.weights[i] += learning_rate * layer.T.dot(delta)
            def predict(self, x):
                a = np.hstack((np.ones(1).T,np.array(x)))
                for 1 in range(0, len(self.weights)):
                     a = self.activation(np.dot(a, self.weights[1]))
                return np.asscalar(a)
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In [3]: nn = MLP([2,2,1])
        ## defining the data
        X = np.array([[0, 0], [0, 1], [1, 0], [1, 1]])
        Y = np.array([0, 1, 1, 0])
        ## Fitting the MLP
        nn.fit(X, Y)
        epochs: 0
        epochs: 10000
        epochs: 20000
        epochs: 30000
        epochs: 40000
        epochs: 50000
        epochs: 60000
        epochs: 70000
        epochs: 80000
        epochs: 90000
In [4]: for e in X:
            print("X= %d Y=%d Prediction=%f" %(e[0],e[1],nn.predict(e)))
        X= 0 Y=0 Prediction=0.000044
        X= 0 Y=1 Prediction=0.996016
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X= 1 Y=0 Prediction=0.997521 X= 1 Y=1 Prediction=0.000021