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import pdb
import numpy as np
import random
import matplotlib.pyplot as plt

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dataOR = [( 1,  1,  1,  1,  1),
           ( 1,  1, -1,  1,  1),
           ( 1, -1,  1,  1,  1),
           ( 1, -1, -1,  1,  1),
           (-1,  1,  1,  1,  1),
           (-1,  1, -1,  1,  1),
           (-1, -1,  1,  1,  1),
           (-1, -1, -1,  1, -1)]

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dataXOR = [( 1,  1,  1,  1, -1),
            ( 1,  1, -1,  1,  1),
            ( 1, -1,  1,  1,  1),
            ( 1, -1, -1,  1,  1),
            (-1,  1,  1,  1,  1),
            (-1,  1, -1,  1,  1),
            (-1, -1,  1,  1,  1),
            (-1, -1, -1,  1, -1)]

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class perceptron:
    def __init__(self, lr, iters, prt, recTrainHist):
        self.lr = lr
        self.iters = iters
        self.af = self.sign
        self.prt = prt
        #self.loss = L
        self.recTrainHist = recTrainHist
        if self.recTrainHist:
            self.WHist = list()
            self.bHist = list()
            self.XHist = list()
            self.YHist = list()
            self.YestHist = list()
            self.ErrHist = list()
            self.deltaWHist = list()
            self.updateHist = list()
            self.acchHist = list()
        self.W = None
        self.b = None
        return

    def sign(self, val):
        return np.where(val>0, 1, -1)

    def train(self, X, Y):
        self.nSamples, self.mFeatures = X.shape
        self.W = np.random.uniform(low=-.1, high=.1, size=self.mFeatures)
        self.b = 0

        XY = np.concatenate((X, Y), axis=1)
        XY_tmp = np.ndarray.copy(XY)
        for _ in range(self.iters):
            i = random.randint(0, len(XY_tmp)-1)
            Xdatum = XY_tmp[i][::-1]

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Ydatum = XY_tmp[i][-1]
XY_tmp = np.delete(XY_tmp, i, axis=0)
#pdb.set_trace()
linOutput = np.dot(Xdatum, self.W) + self.b
y_ = self.af(linOutput)
# apply the Perceptron rule
error = Ydatum - y_
update = self.lr * error
delta_W = update * Xdatum
self.W += delta_W
self.b += error
if len(XY_tmp)==0: XY_tmp = np.ndarray.copy(XY)
if self.prt==True:
    print("iter:{:<4d}".format(_), " |X:{:2}".format(Xdatum[0]), \
          "{:2}".format(Xdatum[1]), "{:2}".format(Xdatum[2]), \
          "{:2}".format(Xdatum[3]), " |Y:{:2}".format(Ydatum), \
          " |y_{:2}".format(y_), " | err:{:2}".format(error), \
          " | W:{:3.2f}".format(self.W[0]), "{:3.2f}".format(self.W[1]), \
          "{:3.2f}".format(self.W[2]), '{:3.2f}'.format(self.W[3]))
if self.recTrainHist==True:
    self.WHist.append(self.W)
    self.bHist.append(self.b)
    self.XHist.append(Xdatum)
    self.YHist.append(Ydatum)
    self.YestHist.append(y_)
    self.ErrHist.append(error)
    self.deltaWHist.append(delta_W)
    #self.updateHist.append(update)
    self.accHist.append(self.getacc(self.YHist, self.YestHist))
return

def predict(self, X):
    dotprod = np.dot(X, self.W) + self.b
    return self.sign(dotprod)

def getacc(self, groundtruth, prediction):
    correct = 0
    for i in range(len(groundtruth)):
        if groundtruth[i] == prediction[i]:
            correct += 1
    return correct / float(len(groundtruth)) * 100.0

def get_data(data):
    X = list()
    Y = list()
    for i in range(len(data)):
        X.append(np.asarray(data[i][: -1]))
        Y.append(np.asarray(data[i][ -1]))
    X = np.asarray(X)
    Y = [1 if i > 0 else -1 for i in Y]
    Y = np.asarray(Y)
    Y = np.expand_dims(Y,axis=1)
    return X, Y

if __name__ == '__main__':
    print('-->> Training and testing with OR')

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X, Y = get_data(dataOR)
pOR = perceptron(0.1, 1000, True, True)    Learning rate is passed in.
pOR.train(X,Y)
plt.style.use('ggplot')
plt.plot(range(len(pOR.acchist)), pOR.acchist, label='OR Perceptron Training Acc')
#plt.show()
#figOR = plt.figure()
#ax = figOR.add_subplot(111, projection='3d')

print("\n\n")
print('-->> Training and testing with XOR')
xorX, xorY = get_data(dataXOR)
pXOR = perceptron(0.1, 1000, True, True)
pXOR.train(xorX,xorY)
#plt.style.use('fivethirtyeight')

plt.plot(range(len(pXOR.acchist)), pXOR.acchist, label='XOR Perceptron Training Acc')

plt.xlabel('Training iterations')
plt.ylabel('Training accuracy [%]')
plt.title('OR vs. XOR training accuracy')
plt.legend()
#plt.grid(True)
#plt.tight_layout()
plt.show()

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