

# Homework 2 : Problem 3.5

Consider two-classes that follow multivariate normal distribution with:

- $\text{mean1} = [1, 1]$  and  $\text{mean2} = [0, 0]$
- $\text{var} = 0.2$

Produce 50 vectors from each class. Use these vectors to design a linear classifier using the perceptron algorithm. After convergence, draw the corresponding decision line.

## Note:

To guarantee linear separability of the classes disregard:

- vectors with  $x_1 + x_2 < 1$  for the  $[1, 1]$  class.
- vectors with  $x_1 + x_2 > 1$  for the  $[0, 0]$  class.

```
# -*- coding: utf-8 -*-
"""
Homework 2 - Problem 3.4

Perceptron Algorithm
Using Punishment and Reward Method
@author: Nazneen Kotwal
"""

import os
import numpy as np
import matplotlib.pyplot as plt
from matplotlib.colors import ListedColormap

def decbound(wht):
    """
    To plot the decision boundary
    Parameters
    -----
    arg1 : wht
        Weights associated with the feature vector
        Format: [w0,w1,w2]
    """
    x = np.linspace(-2, 2.5, 50)
    y = -(wht[0] + wht[1]*x)/wht[2]
    plt.plot(x, y)

def percepalgo(trainingSet,wht,rho):
    """
    Perceptron Algorithm for Punishment and Reward Method
    -----
    arg1 : trainingSet
        Feature vector
    arg2 : wht
        Weights associated with the feature vector
        Format [w0,w1,w2]
    arg3 : rho
        Controls the rate of convergence of the algorithm
    """
    if ((np.dot(wht,trainingSet[0:3]) <= 0) and (trainingSet[-1] == 0)):
        mul = trainingSet[0:3]
        wht = wht + (rho * mul)
    elif ((np.dot(wht,trainingSet[0:3]) >= 0) and (trainingSet[-1] == 1)):
        mul = trainingSet[0:3]
```

```

        wht = wht - (rho * mul)
    else:
        wht = wht
    return(wht)

def main():

    path = "C:/Users/nazne/OneDrive/Documents/1_ECE 759 Pattern Recognition/Homework/Homework2/proble
    print ("The current working directory is", os.getcwd())
    os.chdir(path)

    # Creating a Data Set for Training and Testing
    mean1 = (1, 1)
    cov = ((0.2, 0),(0, 0.2))
    sample1 = np.random.multivariate_normal(mean1, cov, 50)
    mean2 = (0, 0)
    sample2 = np.random.multivariate_normal(mean2, cov, 50)
    sample11=[]
    sample22=[]
    for x in range(50):
        if((sample1[x][0] + sample1[x][1]) > 1):
            sample11.append(sample1[x])
        if((sample2[x][0]+sample2[x][1]) < 1):
            sample22.append(sample2[x])
    print(sample11)
    print(sample22)
    z1 = np.zeros((len(sample11),1))
    z2 = np.ones((len(sample22),1))
    z3 = np.ones((len(sample11),1))
    sample11 = np.hstack((z3,sample11))
    sample22 = np.hstack((z2,sample22))
    a = np.hstack((sample11,z1))
    b = np.hstack((sample22,z2))
    trainingSet = np.concatenate((a,b),axis=0)
    print(trainingSet)
    x1 = trainingSet[:,1]
    x2 = trainingSet[:,2]
    cmap_bold = ListedColormap(['#FF0000', '#00FF00'])
    plt.scatter(x1, x2, c=trainingSet[:,3], cmap=cmap_bold,
                edgecolor='k', s=20)

    print('No. of Train set: ' + repr(len(trainingSet)))
    iniwht = [2,2,2]
    rho = 0.1
    count = 0
    while True:
        flag = 0
        count += 1
        for x in range(len(trainingSet)):
            wht = percepalgo(trainingSet[x],iniwht,rho)
            print(wht)
            print(iniwht)
            if (np.array_equal(iniwht,wht)==False):
                print(np.array_equal(iniwht,wht))
                flag = 1
                iniwht = wht
                print(flag)
            if (flag == 0) or (count == 100):
                break
        decbound(wht)
    print('The Update Weight Vector on convergence w = [w0 w1 w2] is %s: ' %(wht))
    print('Number of Loops untill Covergence: %d' % count)

if __name__ == '__main__':
    main()

```

### Note:

- The Update Weight Vector on convergence  $w = [w_0 \ w_1 \ w_2]$  is  $[-1.3 \ 1.443 \ 1.333]$
- Number of Loops untill Covergence: 4

