Perceptron

April 29, 2023

```
[10]: import numpy as np
      import matplotlib.pyplot as plt
      from tabulate import tabulate
[11]: x1 = np.array([[1], [-1], [0], [0.1], [0.2], [0.9]])
      x2 = np.array([[1], [-1], [0.5], [0.5], [0.2], [0.5]])
      print("*"*100)
      print("\n GIVEN INPUT : \n")
      print("*"*100)
      print("\n X1 : \n", x1)
      print("\n X2 : \n", x2)
      data = np.hstack((x1, x2)).tolist()
      y = np.array([[1], [-1], [-1], [-1], [1])
      print("\n class : \n", y)
      data = np.hstack((x1, x2)).tolist()
      final_data = np.hstack((data, y)).tolist()
      headers = ["x1", "x2", "Class"]
      tablefmt = "fancy_grid"
      print("*"*100)
      print("\n SAMPLE DATA : \n")
      print("*"*100)
      print(tabulate(final_data, headers=headers, tablefmt=tablefmt))
      print(np.mean(data,axis=0))
      data = data - np.mean(data,axis=0)
      final data = np.hstack((data, y)).tolist()
      print("*"*100)
      print("\n MEAN SQUARED DATA : \n")
      print("*"*100)
      print(tabulate(final_data, headers=headers, tablefmt=tablefmt))
```

GIVEN INPUT :

```
*******
X1 :
[[ 1. ]
[-1.]
[ 0. ]
[ 0.1]
[0.2]
[ 0.9]]
X2 :
[[ 1. ]
[-1.]
[ 0.5]
[ 0.5]
[ 0.2]
[ 0.5]]
class :
[[ 1]
[-1]
[-1]
[-1]
[ 1]
[ 1]]
***********************************
******
SAMPLE DATA :
************************************
******
  x1
      x2
           Class
 1
     1
              1
-1
     -1
             -1
 0
     0.5
             -1
 0.1
     0.5
             -1
 0.2
      0.2
             1
 0.9
      0.5
```

[0.2

0.28333333]

MEAN SQUARED DATA:

```
x1
                      x2
                             Class
0.8
               0.716667
                                 1
-1.2
              -1.28333
                                -1
-0.2
              0.216667
-0.1
               0.216667
                                -1
-2.77556e-17 -0.0833333
0.7
               0.216667
                                 1
```

```
[12]: # Create two arrays for the x and y coordinates of the points
      x_coords = data[:,0]
      y_coords = data[:,1]
      y = y.ravel()
      # Loop through the two classes
      for i, class_label in enumerate(np.unique(y)):
          # Get the data points for the current class
          class_data = data[y == class_label]
          # Get the x and y coordinates for the current class
          class_x = class_data[:,0]
          class_y = class_data[:,1]
          # Plot the data points with different symbols and colors for each class
          if class label == 1:
             plt.scatter(class_x, class_y, marker='+', color='green', label='Class_u
       else:
             plt.scatter(class_x, class_y, marker='*', color='red', label='Class =_u
      # Set the axis labels and legend
      plt.xlabel('x1')
      plt.ylabel('x2')
      plt.legend()
      # Show the plot
```

plt.show()

```
0.75
                Class = -1
                Class = +1
    0.50
    0.25
    0.00
g −0.25
   -0.50
   -0.75
   -1.00
   -1.25
                                     -0.25
        -1.25 -1.00 -0.75 -0.50
                                              0.00
                                                     0.25
                                                             0.50
                                                                    0.75
                                         x1
```

```
[13]: def graph(data,w):
          # Create boolean mask for each class
          mask_plus = y.ravel() == 1
          mask_minus = y.ravel() == -1
          # Plot points for each class with different marker symbols and colors
          plt.scatter(data[mask_plus,0], data[mask_plus,1], marker='+',__

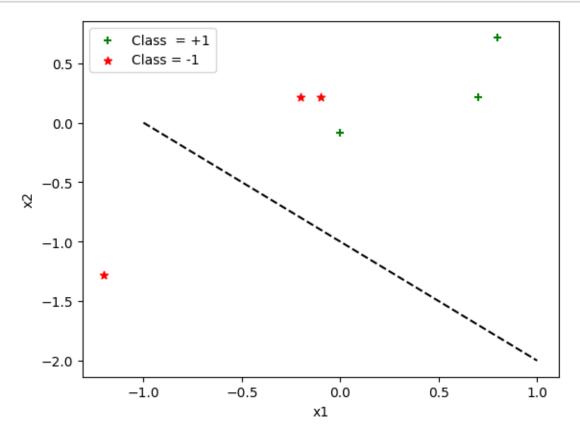
color='green', label='Class = +1')
          plt.scatter(data[mask_minus,0], data[mask_minus,1], marker='*',__

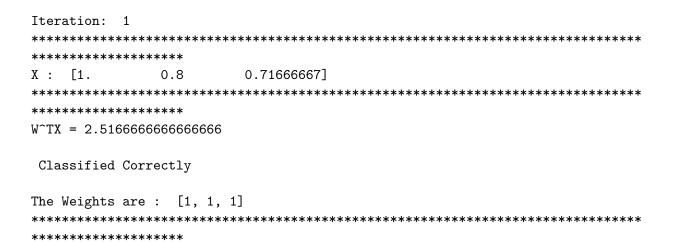
color='red', label='Class = -1')
          # Plot decision boundary
          slope = -w[1] / w[2]
          intercept = -w[0] / w[2]
          x_{vals} = np.array([-1, 1])
          y_vals = intercept + slope * x_vals
          plt.plot(x_vals, y_vals, '--', color='black')
          # Set labels and legend
```

```
plt.xlabel('x1')
plt.ylabel('x2')
plt.legend()

# Show the plot
plt.show()
```

```
[14]: correctClassified = 0
      w = [1, 1, 1]
      graph(data, w)
      iteration = 0
      while (correctClassified != len(data)): #Until everything is classified
          iteration += 1
          print("Iteration: ", iteration)
          for sample in range(len(data)):
              x = np.append(1, data[sample,0:2])
              print("*"*100)
              print("X : ",x)
              print("*"*100)
              if y[sample] == 1:
                  print("W^TX =", np.dot(np.transpose(w), x))
                  if np.dot(np.transpose(w), x) >= 0:
                      correctClassified += 1
                      print("\n","Classified Correctly","\n")
                      print("The Weights are : ", w)
                  else: #orange is classified as apple
                      x + w = w
                      print("\n","MisClassified", "\n")
                      print("w = w + x", "\n")
                      print("The Updated Weights are : ", w)
                      graph(data, w)
                      break
              else:
                  print("W^TX =", np.dot(np.transpose(w), x))
                  if np.dot(np.transpose(w), x) < 0:</pre>
                      correctClassified += 1
                      print("\n","Classified Correctly","\n")
                      print("The Weights are : ", w)
                  else:
                      M = M - X
                      print("\n", "MisClassified", "\n")
                      print("w = w - x", "\n")
                      print("The Updated Weights are : ", w)
```





 $W^TX = -1.48333333333333333$

Classified Correctly

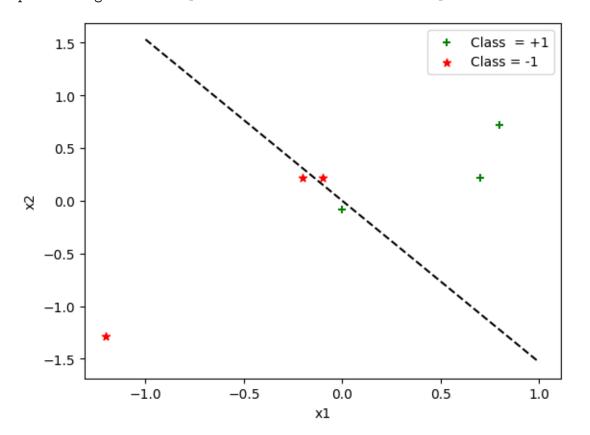
The Weights are : [1, 1, 1]

X: [1. -0.2 0.21666667]

MisClassified

w = w - x

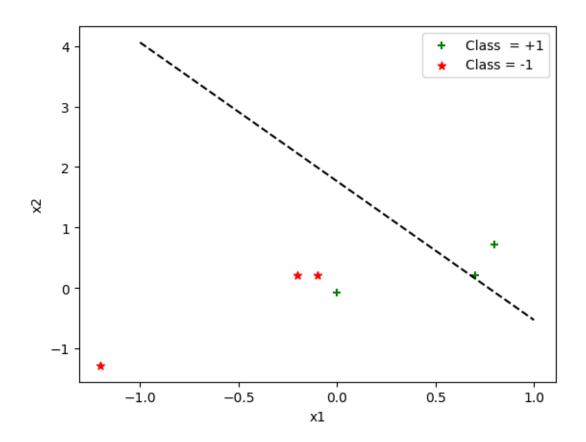
The Updated Weights are : [0. 1.2 0.78333333]



Iteration: 2

******	******	******	******	**********
******	****			
X : [1.	0.8	0.71666667]		
*******	******	******	******	*********
*******	****			
$W^TX = 1.52138888$	88888887			
Classified Corre	ectly			
m	Γο.	4.0	0.70000000	1
The Weights are		1.2		
		******	*****	**********
**************************************		1 0000000	7	
		-1.28333333		********
******		r~~~~~~~~	******	· · · · · · · · · · · · · · · · · · ·
$W^TX = -2.445277$				
W IX - 2.440211				
Classified Corre	ectlv			
0140011104 0011				
The Weights are	: [0.	1.2	0.78333333	
-				***********
*******	****			
X : [1.	-0.2	0.21666667]	
******	******	******	*****	*********
*******	****			
$W^TX = -0.070277$	7777777783			
Classified Corre	ectly			
	_		_	
The Weights are		1.2		
		******	******	*********
*********			-	
X : [1.		0.21666667	_	
		******	******	**********

$W^TX = 0.04972222$	222222218			
MisClassified				
MISCIASSIFIED				
w = w - x				
w A				
The Updated Weigh	hts are : [-	-1.	1.3	0.56666667]



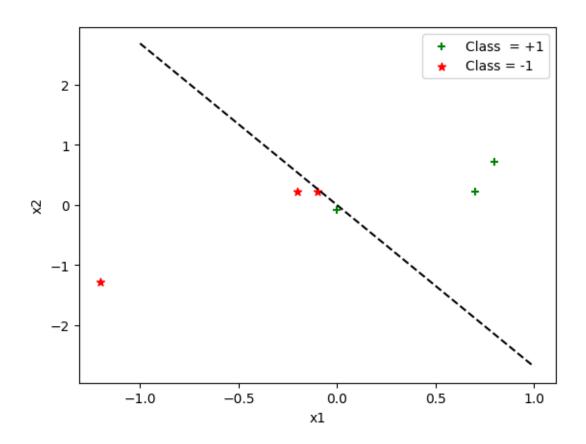
Iteration: 3 ********************************* ****** 0.71666667] X : [1.0.8 ************************* ****** $W^TX = 0.446111111111111111$ Classified Correctly The Weights are : [-1. 0.56666667] 1.3 ********************************** ****** -1.2 -1.28333333] ********************************* ****** $W^TX = -3.28722222222222$ Classified Correctly

0.56666667]

1.3

The Weights are : [-1.

******	******	******	***********
******	**		
X : [1.	-0.2	0.21666667]	
******	******	******	***********
******	**		
$W^TX = -1.13722222$	22222221		
Classified Correc	tly		
The Weights are :	[-1.	1.3	0.56666667]
******	*****	******	************
******	**		
X : [1.	-0.1	0.21666667]	
******	******	******	************
******	**		
$W^TX = -1.00722222$	2222222		
Classified Correc	tly		
The Weights are :	[-1.	1.3	0.56666667]
******	*****	******	***********
******	**		
X : [1.00000000e	+00 -2.77555	756e-17 -8.33	333333e-02]
******	******	******	************
******	**		
$W^TX = -1.04722222$	2222223		
MisClassified			
W = W + X			
The Updated Weight	s are : [0.	1.3	0.48333333]



Iteration: 4 ********* ****** X : [1. 0.71666667] 0.8 ************************* ******* $W^TX = 1.386388888888889$ Classified Correctly The Weights are : [0. 0.48333333] 1.3 ********************************** ****** X : [1.-1.2 -1.28333333] ********************************* ****** $W^TX = -2.180277777777775$ Classified Correctly

0.48333333]

1.3

The Weights are : [0.

*********	*******	*************			

X: [10.2	0.2166666	0.21666667]			
*********	*******	*************			

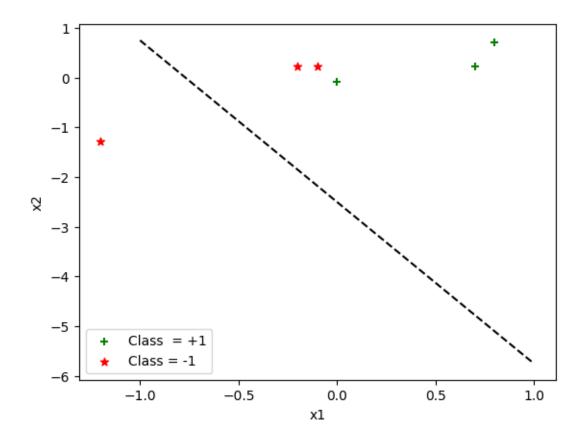
$W^TX = -0.15527777777777785$					
Classified Correctly					
The Weights are : [0.	1.3	0.48333333]			
***********	*********	*************			

X: [10.1	0.2166666	7]			
*********	********	*************			

$W^TX = -0.02527777777777783$					
Classified Correctly					
The Weights are : [0.	1.3	0.48333333]			
********	******	************			

X : [1.0000000e+00 -2.7755	55756e-17 -8	.33333333e-02]			
- ************************************					

W^TX = -0.04027777777778					
MisClassified					
W = W + X					



Classified Correctly

X: [1. -1.2 -1.28333333]

Classified Correctly

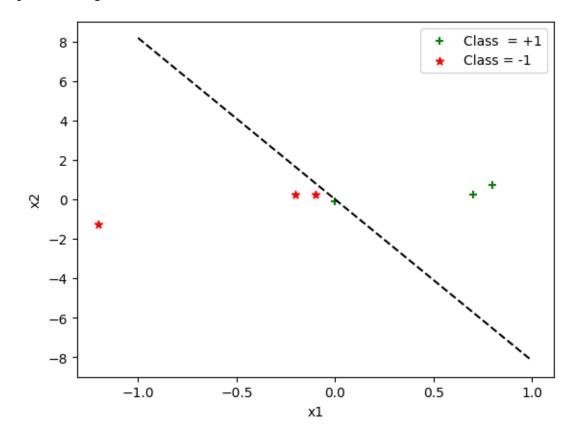
The Weights are : [1. 1.3 0.4]

X: [1. -0.2 0.21666667]

MisClassified

w = w - x

The Updated Weights are : [0. 1.5 0.18333333]

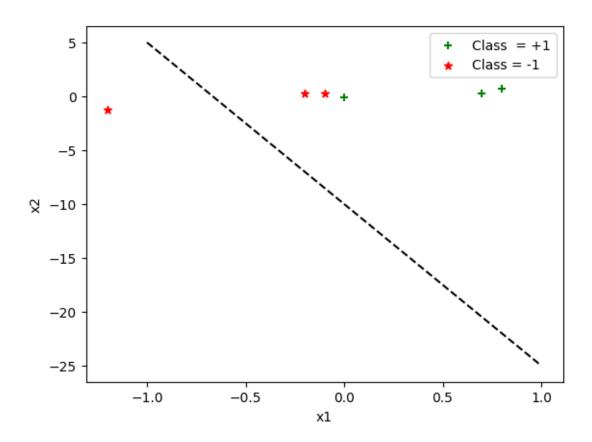


X: [1. 0.8 0.71666667]

Classified Correctly

The Weights are: [0. 1.5 0.18333333] ****** X : [1. -1.2 -1.28333333] ************************************* ******* $W^TX = -2.035277777777775$ Classified Correctly 1.5 0.18333333] The Weights are : [0. ************************* ****** X: [1. -0.2 0.21666667] ******* $W^TX = -0.2602777777777783$ Classified Correctly 1.5 0.18333333] The Weights are : [0. ****** X : [1. -0.1 0.21666667] ******* $W^TX = -0.1102777777777782$ Classified Correctly The Weights are : [0. 1.5 0.18333333] ************************************ ******* X : [1.00000000e+00 -2.77555756e-17 -8.33333333e-02] ******* $W^TX = -0.01527777777777817$ MisClassified w = w + x

The Updated Weights are: [1. 1.5 0.1]



Iteration: 7 ********* ****** X : [1. 0.71666667] 0.8 ************************* ******* $W^TX = 2.271666666666667$ Classified Correctly The Weights are : [1. 1.5 0.1] *********************************** ****** -1.2-1.28333333] ********************************* ****** $W^TX = -0.92833333333333333$

Classified Correctly

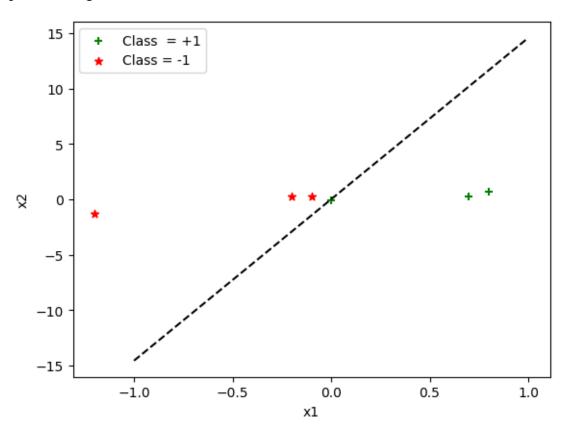
The Weights are : [1. 1.5 0.1]

X: [1. -0.2 0.21666667]

MisClassified

w = w - x

The Updated Weights are : [0. 1.7 -0.11666667]



Iteration:	8
*******	**************************
******	*****

X: [1. 0.8 0.71666667]

W^TX = 1.2763888888888888

Classified Correctly

The Weights are	: [0.	1.7	-0.11666667]
******	******	*******	*************
******	****		
X : [1.	-1.2	-1.28333333]	
******	******	******	*************
******	****		
$W^TX = -1.890277$	77777778		
Classified Corre	ectly		
The Weights are	: [0.	1.7	-0.11666667]
*********	*******	**********	************

X : [1.	-0.2	0.21666667]	
******	*******	*******	************
******	****		
$W^TX = -0.365277$	7777777787		
Classified Corre	ectly		
The Weights are	: [0.	1.7	-0.11666667]
******	******	*******	*************
******	****		
X : [1.	-0.1	0.21666667]	
******	******	*******	************
******	****		
$W^TX = -0.195277$	7777777783		
Classified Corre	ectly		
The Weights are	: [0.	1.7	-0.11666667]
*********	*******	*********	************
*******	****		
X : [1.00000000	0e+00 -2.77	555756e-17 -8.33	3333333e-02]
******	******	*********	************
*********	****		
$W^TX = 0.00972222$	222222217		
Classified Corre	ectly		
The Weights are	: [0.	1.7	-0.11666667]
*********	*******	*******	************

X : [1.	0.7	0.21666667]	
******	*******	*******	************
******	****		
$W^TX = 1.16472222$	22222222		

Classified Correctly