

## ECE32 Assignment #1 - DEVIN BRESSER

- a) The *decision boundary* is the set of  $\{x_1, x_2\}$  that satisfy  $x_1 a_1 + x_2 a_2 = b$ . Thus, we may assign the label to the data using the sign of  $y = x_1 a_1 + x_2 a_2 - b$  since label 1 is decided if  $y > 0$  and label -1 is decided if  $y < 0$ . That is, the label may be obtained as  $\text{sign}\{y\}$ . Express  $y$  as an inner product of a vector  $\mathbf{x}$  containing the features and  $\mathbf{w}$  containing weights, that is, write  $y = \mathbf{x}^T \mathbf{w}$ .

$$y = \underline{\mathbf{x}}^T \underline{\mathbf{w}} = \begin{bmatrix} x_1 & x_2 & 1 \end{bmatrix} \begin{bmatrix} a_1 \\ a_2 \\ -b \end{bmatrix} = x_1 a_1 + x_2 a_2 - b$$

$$\rightarrow y = x_1 a_1 + x_2 a_2 - b$$

- b) Let  $x_2$  be the vertical axis and  $x_1$  be the horizontal axis. Show that the decision boundary  $y = 0$  is a straight line. Find the slope and intercept with the vertical axis as a function of  $a_1, a_2, b$ .

$$y = 0$$

$$\rightarrow x_1 a_1 + x_2 a_2 - b = 0$$

$$\rightarrow x_2 a_2 = -x_1 a_1 + b$$

$$\rightarrow x_2 = -\frac{a_1}{a_2} x_1 + \frac{b}{a_2}$$

$$\rightarrow \text{Slope: } -\frac{a_1}{a_2}$$

$$\rightarrow y\text{-int: } \frac{b}{a_2}$$

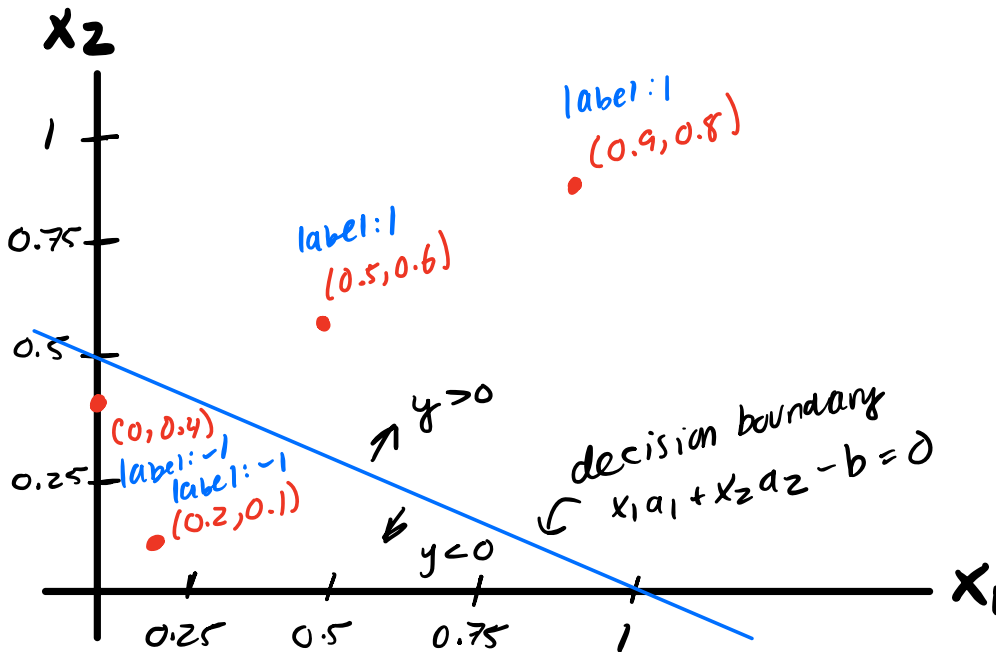
c) You classify  $n$  data samples using  $\text{sign}\{y\}$  where  $y = \begin{bmatrix} y_1 \\ y_2 \\ \vdots \\ y_n \end{bmatrix} = Xw$ . Suppose

$n = 4$  and the features for the 4 data samples are 1 : (0, 0.4), 2 : (0.2, 0.1), 3 : (0.5, 0.6), 4 : (0.9, 0.8). Write out the matrix  $X$ .

$$\underset{4 \times 1}{y} = \underset{4 \times 3}{\underbrace{\begin{bmatrix} 0 & 0.4 & 1 \\ 0.2 & 0.1 & 1 \\ 0.5 & 0.6 & 1 \\ 0.9 & 0.8 & 1 \end{bmatrix}}_{\underline{X}}} \underset{3 \times 1}{\begin{bmatrix} a_1 \\ a_2 \\ -b \end{bmatrix}} \quad n \times 1 \quad \underset{n \times 1}{y} = \underset{n \times 3}{\underline{X}} \underset{3 \times 1}{\underline{w}}$$

d) Suppose  $a_1 = 1$ ,  $a_2 = 2$ , and  $b = 1$ . Sketch the decision boundary in the  $x_1$ - $x_2$  plane assuming  $x_2$  is the vertical axis and  $x_1$  is the horizontal axis. Graph the four data samples from the previous part and classify them.

$$x_2 = -\frac{a_1}{a_2} x_1 + \frac{b}{a_2} \rightarrow x_2 = -\frac{1}{2} x_1 + \frac{1}{2}$$



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In [16]: %matplotlib inline
import numpy as np
import matplotlib.pyplot as plt

# number of features
p = 2

# number of examples
n = 5000

# generate matrix of n (random) examples of p features with a column of all ones
X0 = np.random.rand(n,p)
onevec = np.ones(shape = (n,1))
X = np.concatenate((X0,onevec),axis=1)

# Classifier weights
w = [[-1], [2], [-0.4]]

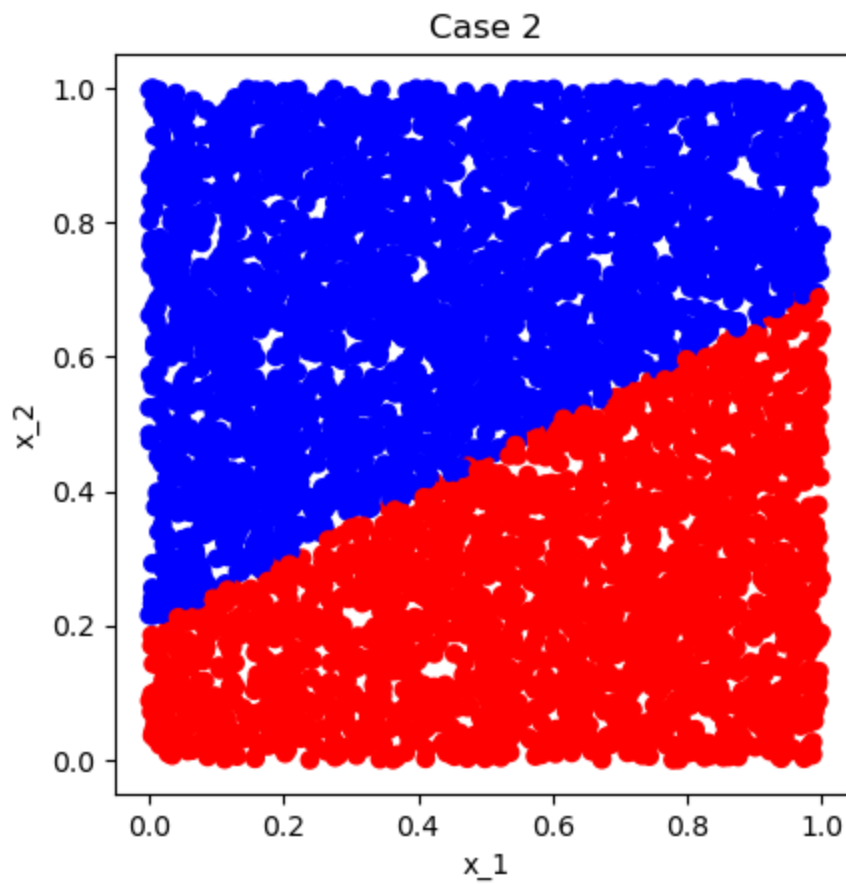
# Multiply feature matrix with weights yhat = X*w
yhat = X@w

# Decide which class based on whether yhat is >< 0
# sign function returns +1 when yhat(i)>0 and -1 when yhat(i)<0
pred_label = np.sign(yhat);

plt.scatter(X[:,[0]],X[:,[1]], color = ['r' if i==-1 else 'b' for i in pred_label])

plt.xlabel("x_1")
plt.ylabel("x_2")
plt.title("Case 2")
plt.axis('square')
plt.show()

# Problem 1.e comment: The boundary is a line with y-intercept 0.2 and slope 0.5
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In [18]: # Classifier weights
w = [[1.6], [2], [-1.6]]

# Multiply feature matrix with weights yhat = X*w
yhat = X@w

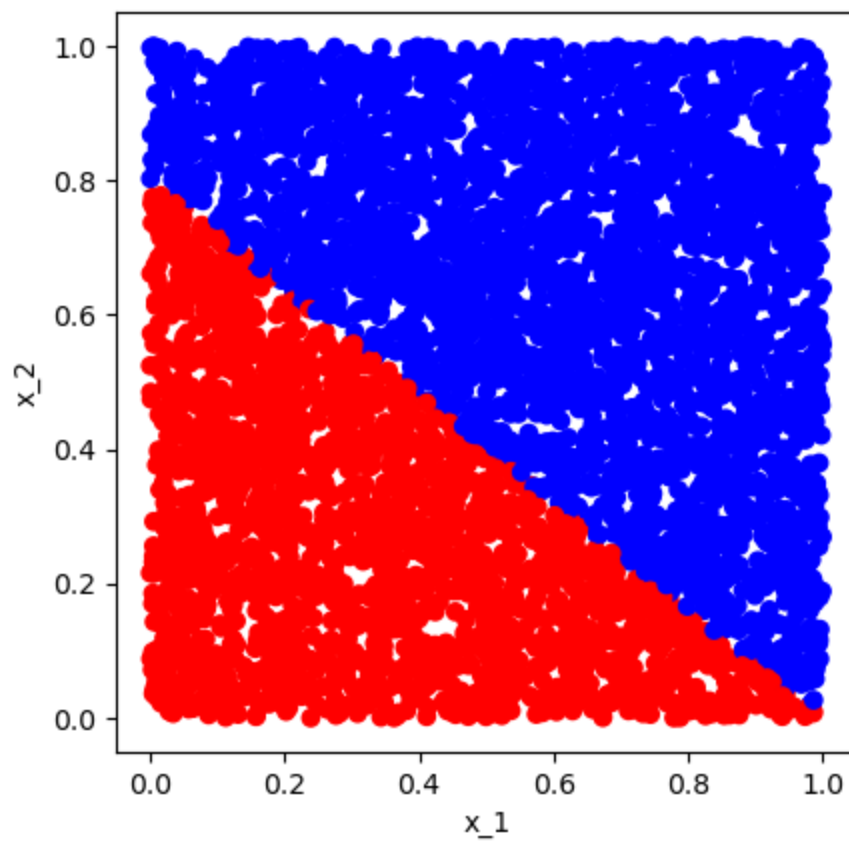
# Decide which class based on whether yhat is > 0
# sign function returns +1 when yhat(i)>0 and -1 when yhat(i)<0
pred_label = np.sign(yhat);

plt.scatter(X[:,[0]],X[:,[1]], color = ['r' if i==-1 else 'b' for i in pred_label])

plt.xlabel("x_1")
plt.ylabel("x_2")
plt.title("Case 2")
plt.axis('square')
plt.show()

# Problem 1.f comment: The new weights adjusted the slope and y-intercept of the
# decision boundary. Now, the boundary has y-intercept 0.8 and slope -0.8
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Case 2



In [ ]: