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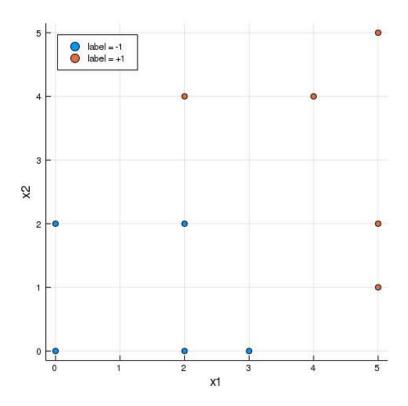
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## **Problem 1**

Midterm due Nov 10, 2020 05:29 IST Completed

#### **Problem 1. Linear Classification**

Consider a labeled training set shown in figure below:



## 1. (1)

#### 1.0/2 points (graded)

We initialize the parameters to all zero values and run the **linear perceptron algorithm** through these points in a particular order until convergence. The number of mistakes made on each point are shown in the table below. (These points correspond to the data point in the plot above)

**Note:** You should be able to arrive at the answer without programming.

What is the resulting offset parameter  $\theta_0$ ?

Enter the numerical value for  $\theta_0$ :

$$\theta_0 =$$
 -17 **\* Answer:** -18

What is the resulting parameter  $\theta$ ?

(Enter heta as a vector, e.g. type [0,1] if  $heta = \begin{bmatrix} 0 & 1 \end{bmatrix}^T$  .)

STANDARD NOTATION

#### Solution:

Let  $lpha_i$  be the number of mistakes that perceptron makes on the point  $x^{(i)}$  with label  $y^{(i)}$ . The resulting offset parameter is

$$\theta_0 = \sum_{i=1}^{10} \alpha_i y^{(i)} = -18 \tag{7.1}$$

The resulting parameter  $\theta$  is

$$heta = \sum_{i=1}^{10} lpha_i y^{(i)} x^{(i)} = egin{bmatrix} 4 & 4\end{bmatrix}^T.$$

Note that the answer does not depend on the order of data points used in the algorithm. (For reference, the sequence in the perceptron algorithm used here is (4,4),(0,0),(2,0),(3,0),(5,5),(2,4),(0,2),(2,2),(5,1),(5,2).)

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You have used 1 of 3 attempts

• Answers are displayed within the problem

## 1. (2)

0/1 point (graded)

**Setup as above:** We initialize the parameters to all zero values and run the **linear perceptron algorithm** through these points in a particular order until convergence. The number of mistakes made on each point are shown in the table below. (These points correspond to the data points in the plot above.)

Label -1 -1 -1 -1 -1 +1 +1 +1 +1 +1 Coordinates (0,0) (2,0) (3,0) (0,2) (2,2) (5,1) (5,2) (2,4) (4,4) (5,5) Perceptron mistakes 1 9 10 5 9 11 0 3 1 1

The mistakes that the algorithm makes often depend on the order in which the points were considered. Could the point (5,2) labeled +1 have been the first one considered?

yes		
Ono ✔		





#### **Solution:**

When perceptron is initialized to all zeros, the first point considered is always a mistake. Since no mistakes were made on the point (5,2) labeled +1, it could not have been the first point considered.

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You have used 1 of 3 attempts

Answers are displayed within the problem

## 1. (3)

2.0/2 points (graded)

Suppose that we now find the linear separator that **maximizes** the margin instead of running the perceptron algorithm.

What are the parameters  $\theta_0$  and  $\theta$  corresponding to the **maximum margin separator**?

(Enter  $\theta_0$  accurate to at least 3 decimal places.)

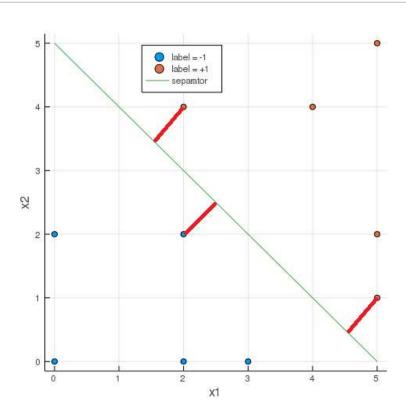
(Enter  $\theta$  as a vector, enclosed in square brackets, and components separated by commas, e.g. type [0,1] for  $\begin{bmatrix} 0 & 1 \end{bmatrix}^T$ .)

$$heta=$$
 [1,1]  $limes$  Answer: [1,1]

**STANDARD NOTATION** 

#### **Solution:**

A maximum margin separator is in particular a decision boundary whose distance to the closest data point(s) are maximal. In this problem, the data points are arranged so that the maximum margin separator can be found by inspection:



The black line is the maximum linear separater with the data points. The points (2,2), (2,4), (5,1) are support vectors. The margin, i.e. the length of the red segments, is  $1/\sqrt{2}$ .

In general, there are infinitely many equations that describes the same hyperplane, or line, in this case. However, the parameters  $\theta$  and  $\theta_0$  of the maximum margin separator satisifes one more constraint: the positive and /negative margins are defined by  $\theta \cdot x + \theta_0 = \pm 1$ . This gives the margin, i.e. the distance between either margin boundary and the separator, to be  $\frac{1}{||\theta||}$ .

Hence, in this problem, we need to choose  $\theta$  such that  $\|\theta\|=\sqrt{2}$ , where  $1/\sqrt{2}$  is the distance between the separator and the closest data point(s). The parameters corresponding to the maximum margin separator are:

$$\theta = [1, 1]^T \text{ and } \theta_0 = -5 \tag{7.2}$$

#### **Remark:**

Although it is not required by the problem, we feel it might be helpful to also discuss how to find "maximum margin separater" generally. In fact, this is known as the hard margin SVM, i.e., to find the linear hyperplane with the maximum margin, such that all points are classified correctly. In particular, if we omit the bias term  $\theta_0$ , the goal is to maximize  $1/\|\theta\|$  (or minimize  $\theta^2$ , equivalently), given  $y^{(i)}$  ( $\theta \cdot x^{(i)}$ )  $\geq 1$  for all i. This optmization problem can be then written as the following quadratic programming formulation:

$$egin{array}{ll} \min _{ heta} & heta^2 \ & ext{s.t.} & y^{(i)}\left( heta \cdot x^{(i)}
ight) \geq 1, orall \ i \end{array}$$

Note this is different from the commonly seen soft margin SVM, where you can allow some classification errors (that's why it's called "soft" instead of "hard"). In comparison, the soft margin SVM corresponds to the following optmization problem:

$$egin{aligned} \min & heta^2 + C \sum_i \xi_i \ & ext{s.t.} \quad y^{(i)} \left( heta \cdot x^{(i)} 
ight) \geq 1 - \xi_i, orall i \end{aligned}$$

If you use SVM program directly from machine learning toolkits such as scikit-learn, it is by default the soft-margin SVM (in scikit-learn the default C equals to 1); and some regularizer term is also likely to be there by default on the objective. If you want to numerically use the soft SVM program to approximately produce a hard SVM result, you need to be very careful on the parameters ( $C\leftarrow\infty$ ,  $\lambda\leftarrow0$ ). Then you would get something very close within the numerical tolerance.

Submit

You have used 1 of 3 attempts

**1** Answers are displayed within the problem

## 1. (4)

1.0/1 point (graded)

What is the value of the margin attained?

(Enter an exact answer or decimal accurate to at least 2 decimal places.)

**Grading note:** Both reasonable answers wil be accepted. In case the definition of the margin is not clear, we have accepted both the distance between the separator and the margin, and the distance between the 2 margin boundaries, as correct answers.

#### **Solution:**

The support vectors (points closest to the max-margin separator) are (2,2),(2,4) and (5,1). The distance between any one of these points and the separator is  $\frac{\sqrt{2}}{2} = \frac{1}{\sqrt{2}}$ .

By definition, this distance equals the margin of a maximum margin separator, because the support vector(s) lie on the margin boundaries in this case.

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You have used 1 of 3 attempts

**1** Answers are displayed within the problem

## 1. (5)

1/1 point (graded)

Using the parameters  $heta_0$  and heta corresponding to the **maximum margin separator**, what is the sum of Hinge losses evaluated on each example?

Sum of hinge losses: 0



Answer: 0

#### Solution:

Since the points are linearly separated, the hinge loss is 0.

You can also verify by computing the sum of the hinge losses explicitly by:

$$\sum_{i=1}^{10} \max\{0, 1 - y^{(i)} \left(\theta \cdot x^{(i)} + \theta_0
ight)\} = 0.$$
 (7.3)

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You have used 1 of 3 attempts

**1** Answers are displayed within the problem

# 1. (6)

1/1 point (graded)

Suppose we modify the maximum margin solution a bit and divide both  $\theta$  and  $\theta_0$  by 2. What is the sum of hinge losses evaluated on each example for this new separator?

Sum of hinge losses:

1.5

**✓ Answer:** 1.5

#### Solution:

We can find the hinge loss visually. Since both  $\theta$  and  $\theta_0$  are scaled by the same constant 1/2, the decision boundary stays the same, but the margin is twice what it was before. The points that were right on the margin [i.e. (2,2),(2,4) and (5,1)] will now have a loss of 1/2 each, and all other points has loss of 0, resulting in a total loss of 1.5. The sum of the hinge losses for the new parameters is:

$$\sum_{i=1}^{10} \max\{0, 1-y^{(i)}\, (rac{1}{2} heta \cdot x^{(i)} + rac{1}{2} heta_0)\} = 1.5$$

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You have used 1 of 3 attempts

**1** Answers are displayed within the problem

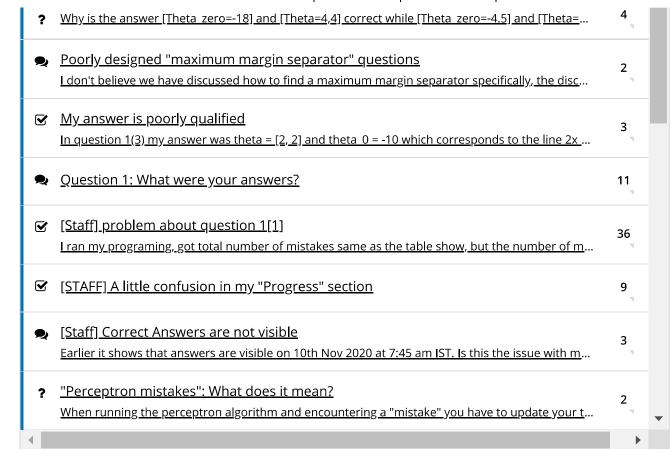
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