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3. Margin Boundary

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3. Margin Boundary

Margin Boundary



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The **decision boundary** is the set of points x which satisfy

$$\theta \cdot x + \theta_0 = 0.$$

The **Margin Boundary** is the set of points x which satisfy

$$\theta \cdot x + \theta_0 = \pm 1.$$

So, the distance from the decision boundary to the margin boundary is $\frac{1}{\|\theta\|}$.

Margin Boundary 1

1/1 point (graded)

As explained in the lecture video, margin boundary is the set of points (x, y) at which the distance from the decision boundary to (x, y) is $\frac{1}{\|\theta\|}$. Now, what is the value of $y^{(i)} (\theta \cdot x^{(i)} + \theta_0)$ for a correctly classified point $(x^{(i)}, y^{(i)})$ on the margin boundary?

✓ Answer: 1

Solution:

From the previous problem, we know that the distance from a line $L : \theta x + \theta_0 = 0$ to $P = (x_0)$ is given by $\frac{\|\theta x_0 + \theta_0\|}{\|\theta\|}$. Because we know that the distance from the decision boundary to (x, y) is $\frac{1}{\|\theta\|}$,

$$\|\theta x_0 + \theta_0\| = 1$$

. Thus,

$$\|\theta x_0 + \theta_0\| = y^{(i)} (\theta \cdot x^{(i)} + \theta_0) = 1$$

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i Answers are displayed within the problem

Margin Boundary 2

1/1 point (graded)

What happens to the margin boundaries as we increase $\|\theta\|$?

- ☒ The margin boundaries move closer to the decision boundary
- ☐ The margin boundaries move further away from the decision boundary
- ☐ The margin boundaries converge to a certain location no matter what

**Solution:**

As we increase $\|\theta\|$, $\frac{1}{\|\theta\|}$ decreases. For now, acknowledge that $\frac{1}{\|\theta\|}$ is the distance from the decision boundary to the margin boundary (which we will closely examine in the next set of problems.) Thus, the distance from the point $(x^{(i)}, y^{(i)})$ that satisfy

$$y^{(i)} (\theta \cdot x^{(i)} + \theta_0) = 1$$

to the decision boundary will decrease. Thus the margin moves closer to the decision boundary.

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?	<u>still can not understand why set $\theta \cdot x + \theta_0 = \pm 1$.</u> <u>still can not understand why set $\theta \cdot x + \theta_0 = \pm 1$.</u>	14
?	<u>The norm of the normal</u> <u>Could someone give extra thoughts on how the norm controls decision-boundary--margin-b...</u>	1
?	<u>Question regarding Margin Boundary 1</u> <u>Dear TA or anyone who can solve my problem, The way I do this question is that it is given th...</u>	2
💬	<u>Possibly incomplete answer to Problem ?</u> <u>It seems to me that the answer to the problem is not quite correct or at least incomplete. Wh...</u>	4
✓	<u>Degree of Freedom</u> <u>The statement that the magnitude of theta is a "free parameter that we have not yet used" s...</u>	5
💬	<u>Free parameter Q</u>	1
💬	<u>Signed distance or sine distance?</u> <u>@6:35, the subtitle says "sine distance", but I think it should be "signed distance"?</u>	2

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