

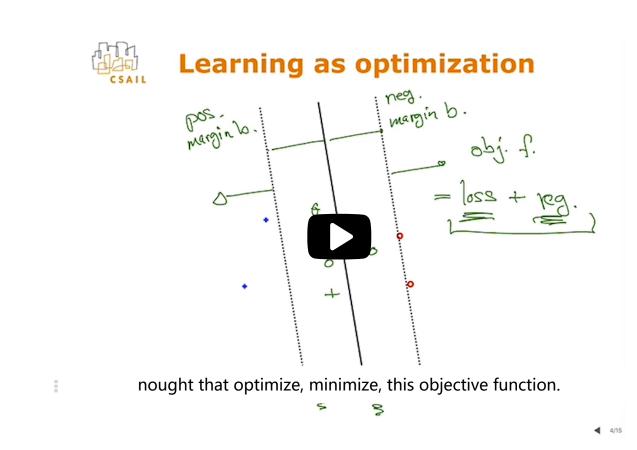
<u>Unit 1 Linear Classifiers and</u>
<u>Course</u> > <u>Generalizations (2 weeks)</u>

<u>Lecture 3 Hinge loss, Margin</u>

> boundaries and Regularization

> 2. Introduction

2. Introduction Introduction



naught.

That's a balance between the loss, how examples

fit within this ideal notion, and regularization,

our preference towards large mounting solutions.

So we will find--

we will formalize-- the objective function and then find parameters theta and theta nought that optimize, minimize, this objective function.

6:19 / 6:19

→ 1.25x

1) 🔀

CC

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Review: Distance from a Line to a Point

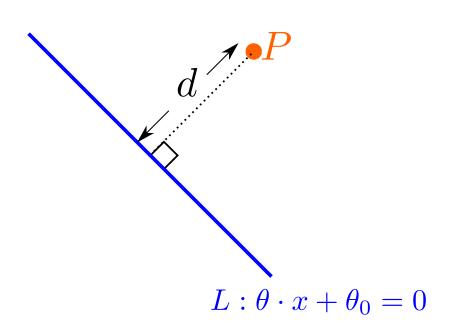
1/1 point (graded)

Consider a line L in \mathbb{R}^2 given by the equation

 $L: heta\cdot x+ heta_0=0$

where θ is a vector normal to the line L. Let the point P be the endpoint of a vector x_0 (so the coordinates of P equal the components of x_0).

What is the the shortest distance d between the line L and the point P? Express d in terms of θ, θ_0, x, x_0 .



d =

θ垂直于L,因此将x0投射到θ上,再除以norm(θ),就是没有offset的情况下的最短距离

有offset就加上offset就行(平移)
$$\frac{|\theta \cdot x + \theta_0|}{||\theta||}$$

$$ullet \left | rac{| heta \cdot x_0 + heta_0|}{|| heta||}
ight.$$

$$\frac{|\theta \cdot \theta_0 + \theta_0|}{|\theta| |\theta|}$$

$$ullet | heta \cdot x_0 + heta_0|$$

Solution:

If there is no offset θ_0 , The distance d is the projection from x_0 to θ , which is $\frac{|x_0\cdot\theta|}{||\theta||}$ (definition of projection). With the offset θ_0 added, d is $\frac{|x_0\cdot\theta+\theta_0|}{||\theta||}$. Thus the distance from a $L:\theta\cdot x+\theta_0=0$ to the point $P=x_0$ is given by $\frac{|\theta\cdot x_0+\theta_0|}{||\theta||}$.

Submit

You have used 1 of 3 attempts

1 Answers are displayed within the problem

Discussion

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