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## 4. Empirical Risk

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Exercises due Mar 8, 2023 08:59 -03 Completed

**the Objective: Empirical Risk****Video** [Download video file](#)**Transcripts** [Download SubRip \(.srt\) file](#) [Download Text \(.txt\) file](#)**Compute Hinge Loss**

1/1 point (graded)

The empirical risk  $R_n$  is defined as

$$R_n(\theta) = \frac{1}{n} \sum_{t=1}^n \text{Loss}(y^{(t)} - \theta \cdot x^{(t)})$$

where  $(x^{(t)}, y^{(t)})$  is the  $t$ th training example (and there are  $n$  in total), and **Loss** is some hinge loss.

Recall from a previous lecture that the definition of hinge loss:

Also, we have  $\hat{y} = 1.25$ .

Compute the value of



Submit

You have used 1 of 3 attempts

## Compute Squared Error Loss

1/1 point (graded)

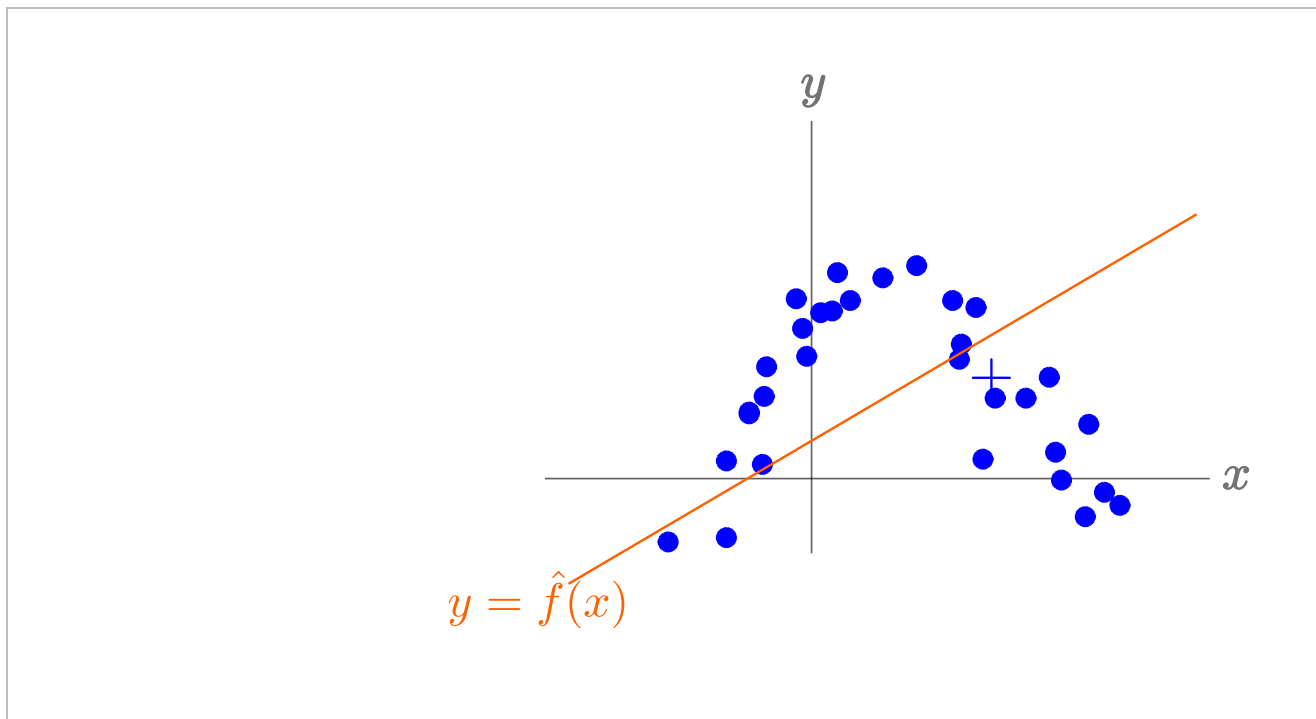
Now, we will calculate the empirical risk with the squared error loss. Remember that the loss function is given by

The 4 training examples are as in the previous problem:

## Geometrically Identifying Error

1/1 point (graded)

What type of error does the figure below depict? The blue dots are the training examples and the orange line is the predictor



☒ Structural error

☐ Estimation error



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You have used 1 of 1 attempt

## Increasing the Number of Training Examples

0/1 point (graded)

If we increase  $n$ , the number of training examples, which of the following types of error

☒ Structural error

☐ Estimation error



  
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You have used 1 of 1 attempt

**(Optional) Error decomposition and the bias-variance trade-off**

## Discussion

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? Squaring only magnifies differences if they are larger than one

? Why do we use Hinge Loss here?

Why do we penalize for deviation that is  $<1$  and do not penalize for deviation that is  $\geq 1$ ? It doesn't make sense

? Error decomposition and the bias-variance trade-off

the third term : is  $E[(\text{error})^2]$  independent of the bias-variance trade-off. if so, then why. The whole idea of

? How to get the result in optional part

💬 What is the exact definition of  $z$  ?

In the explanation of the problems they always refer to the Loss function  $L(z)$ , but nowhere is defined what  $z$

? Empirical risk

What does she mean by "projected empirical risk"?

? (Optional) Error decomposition and the bias-variance trade-off

💬 Last Problem

I believe the definition of 'large amount of training data' is quite arbitrary which, in my opinion, somewhat con

💬 Confusing explanation about Empirical Risk  $R_n$

$R_n$  is defined as:  $\text{Sum}(\text{Loss}(y(t) - \theta.X(t))) / n$  and then they mention Loss can be for instance the hinge los

💬 Confusion about the deviation

💬 Rationale of choosing square function

Imho, in addition to what the professor mentioned (tolerate small deviation, truly penalize large deviation), as



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