

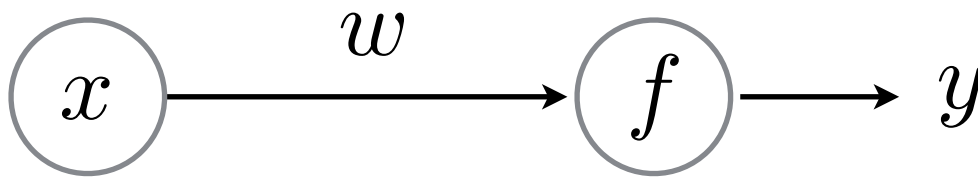


How to Train Your Perceptron

16-385 Computer Vision (Kris Kitani)
Carnegie Mellon University

Let's start easy

world's smallest perceptron!



$$y = wx$$

(a.k.a. line equation, linear regression)

Learning a Perceptron

Given a set of samples and a Perceptron

$$\{x_i, y_i\}$$

$$y = f_{\text{PER}}(x; w)$$

Estimate the parameters of the Perceptron

$$w$$

Given training data:

x	y
10	10.1
2	1.9
3.5	3.4
1	1.1

What do you think the weight parameter is?

$$y = wx$$

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not so obvious as the network gets more complicated so we use ...

An Incremental Learning Strategy

(gradient descent)

Given several examples

$$\{(x_1, y_1), (x_2, y_2), \dots, (x_N, y_N)\}$$

and a perceptron

$$\hat{y} = wx$$

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perceptron
parameter

perceptron
output

true
label

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*what does
this mean?*

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Before diving into gradient descent, we need to understand ...

Loss Function

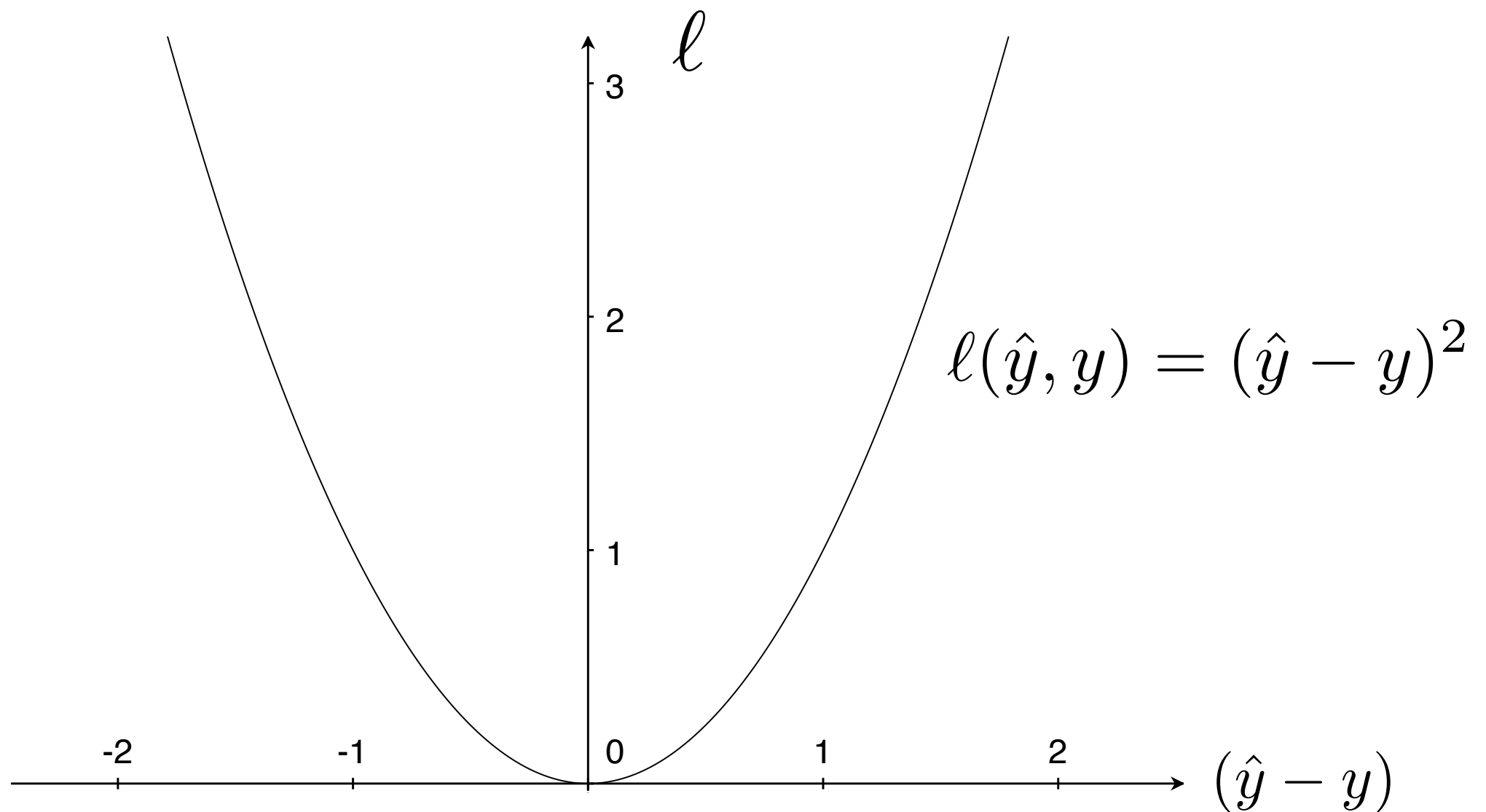
defines what it means to be
close to the true solution

YOU get to choose the loss function!

(some are better than others depending on what you want to do)

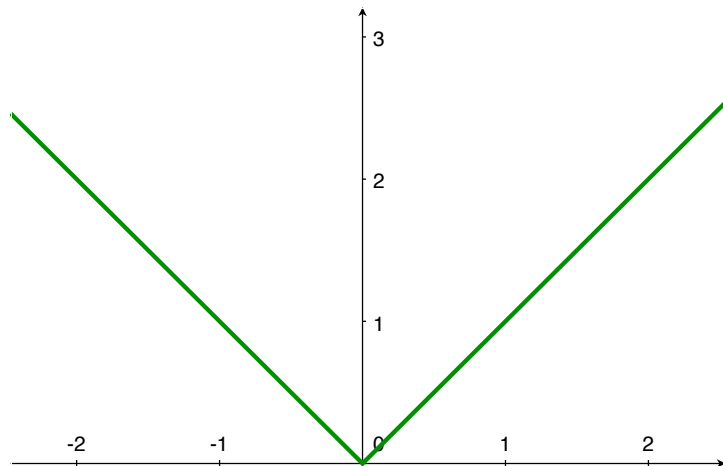
Squared Error (L2)

(a popular loss function)



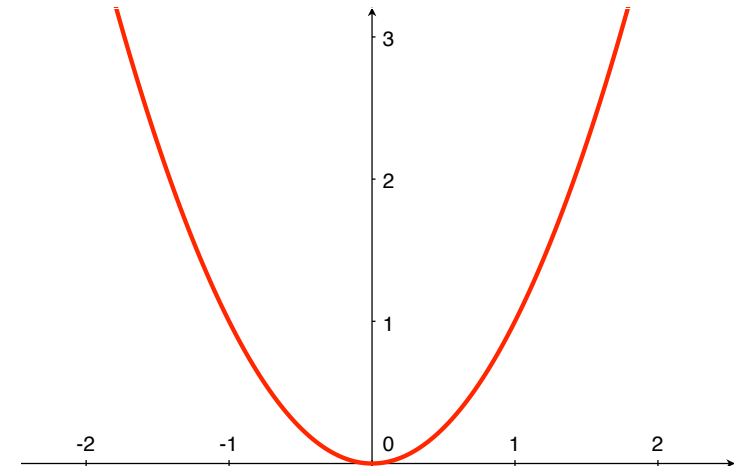
L1 Loss

$$\ell(\hat{y}, y) = |\hat{y} - y|$$



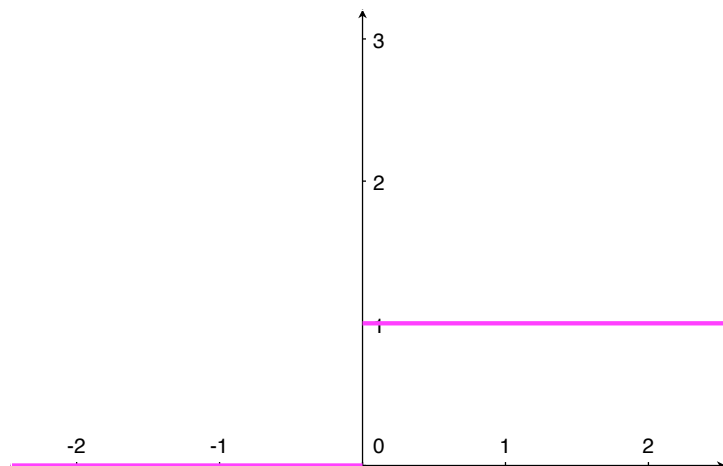
L2 Loss

$$\ell(\hat{y}, y) = (\hat{y} - y)^2$$



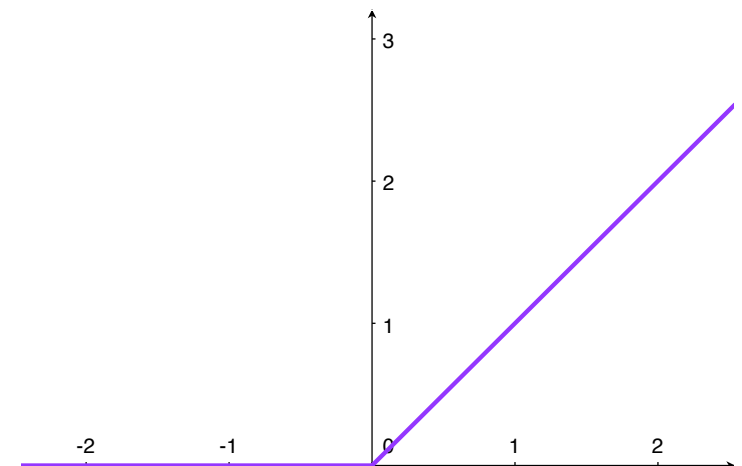
Zero-One Loss

$$\ell(\hat{y}, y) = \mathbf{1}[\hat{y} \neq y]$$



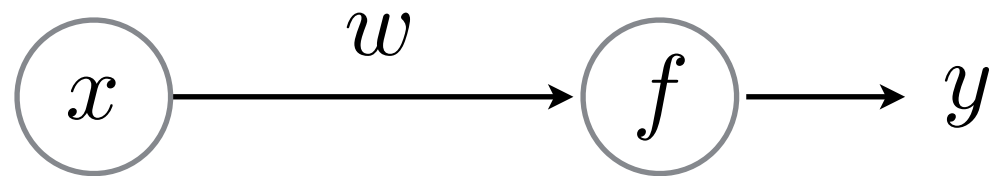
Hinge Loss

$$\ell(\hat{y}, y) = \max(0, 1 - y \cdot \hat{y})$$



back to the...

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function of **ONE** parameter!

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linear function! $f(x) = wx$

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Let's demystify this process first...

Code to train your perceptron:

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for $n = 1 \dots N$

$w = w + (y_n - \hat{y})x_i;$

just one line of code!

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Now where does this come from?