

# INM431: Machine Learning

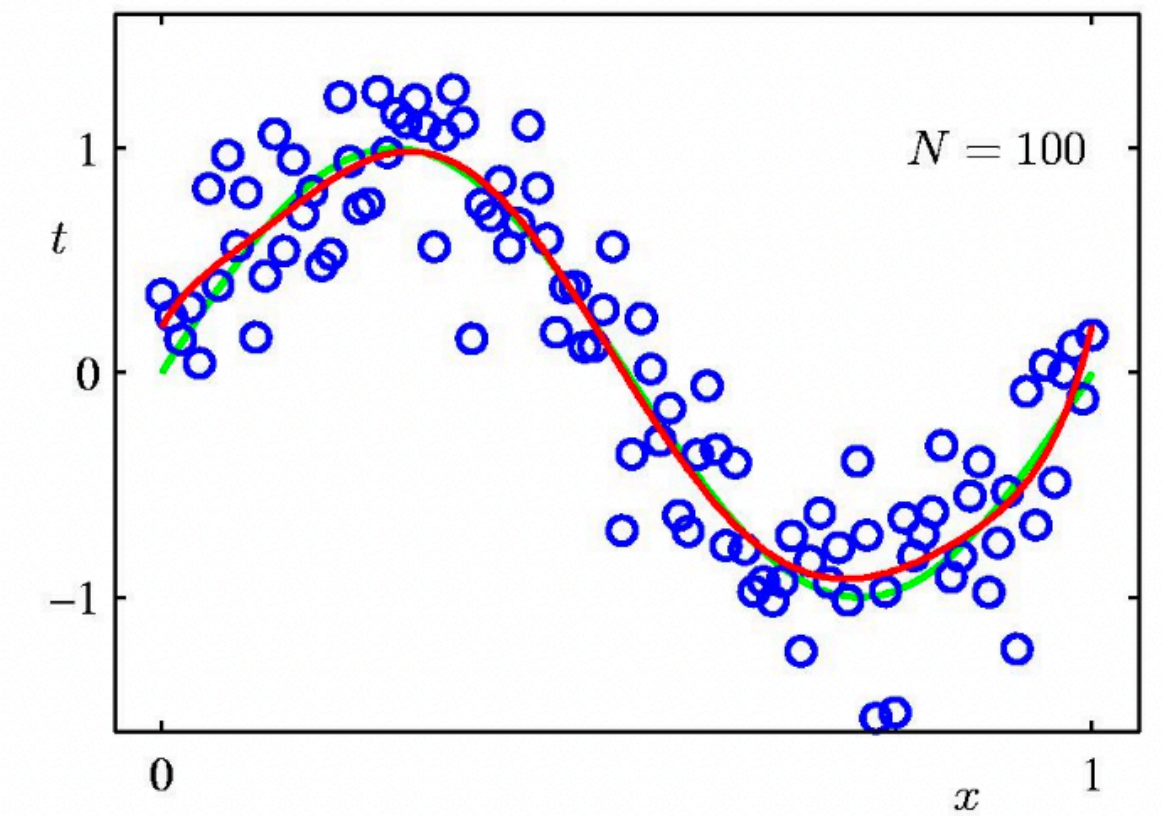
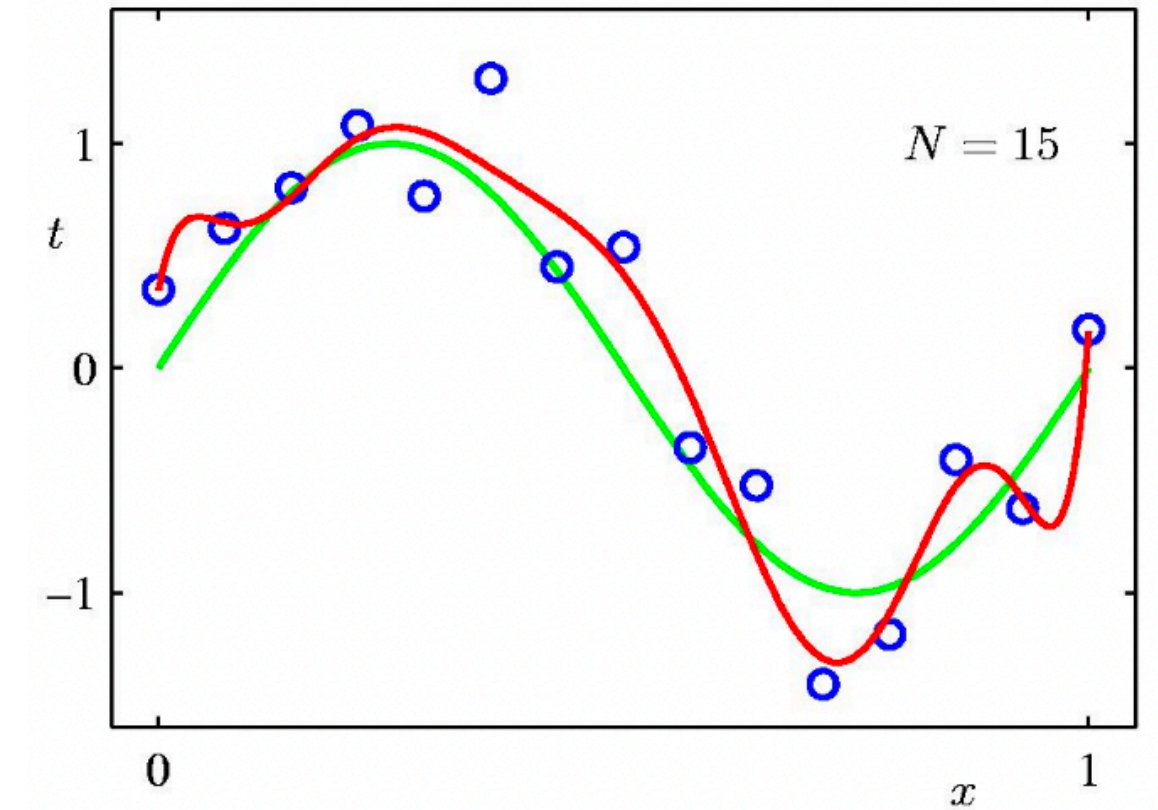
On loss functions

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# Jumping back to linear regression

$$L(w) = \frac{1}{2} \sum_{n=1}^N (f(x_n, w) - t_n)^2$$

Least squares as a loss function?



# Revisiting MLE

$$\log P(D | \theta) = \sum_c \log P(d_c | \theta)$$

Loss function:

$$-\log P(\theta | D) = -\log P(D | \theta) - \log P(\theta) + \log P(D)$$

Where, we obtain  $\theta$  by optimising the log-likelihood.

# Squared loss

$$L(w) = \frac{1}{2} \sum_{n=1}^N (f(x_n, w) - t_n)^2$$

Given a collection of input features, outputs, and a hypothesis function  $f$  (or the machine learning function of choice), we want to find the parameters  $w$  that minimises the sum of the losses.

# Recipe

$$L(w) = \frac{1}{2} \sum_{n=1}^N (f(x_n, w) - t_n)^2$$

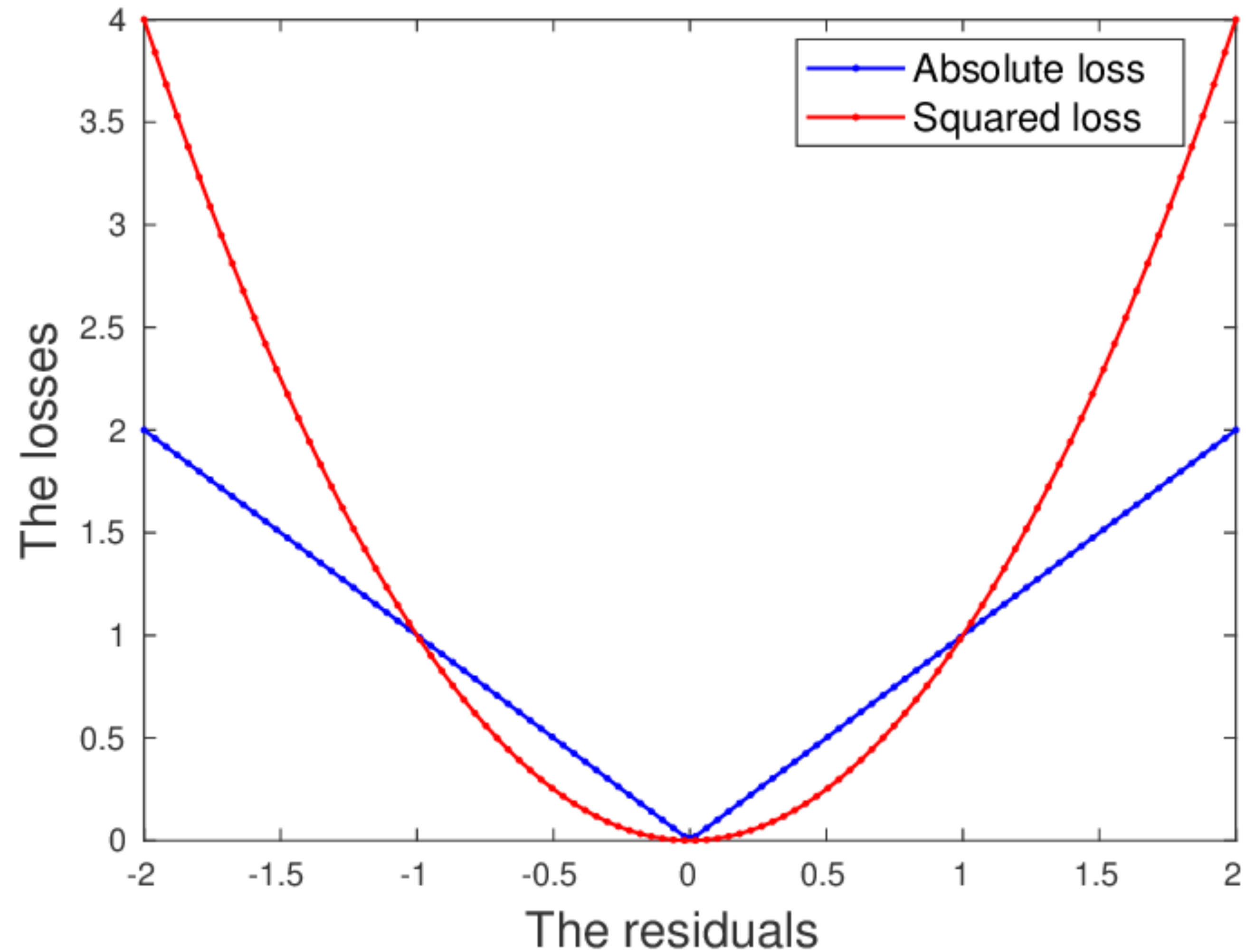
Minimising the loss:

$$\min_w \frac{1}{2} \sum_{n=1}^N (f(x_n, w) - t_n)^2 = 0$$

Approach 1: Gradient descent - take first order derivative

Approach 2: Analytic solution, solve for  $w$

# Visualising the losses

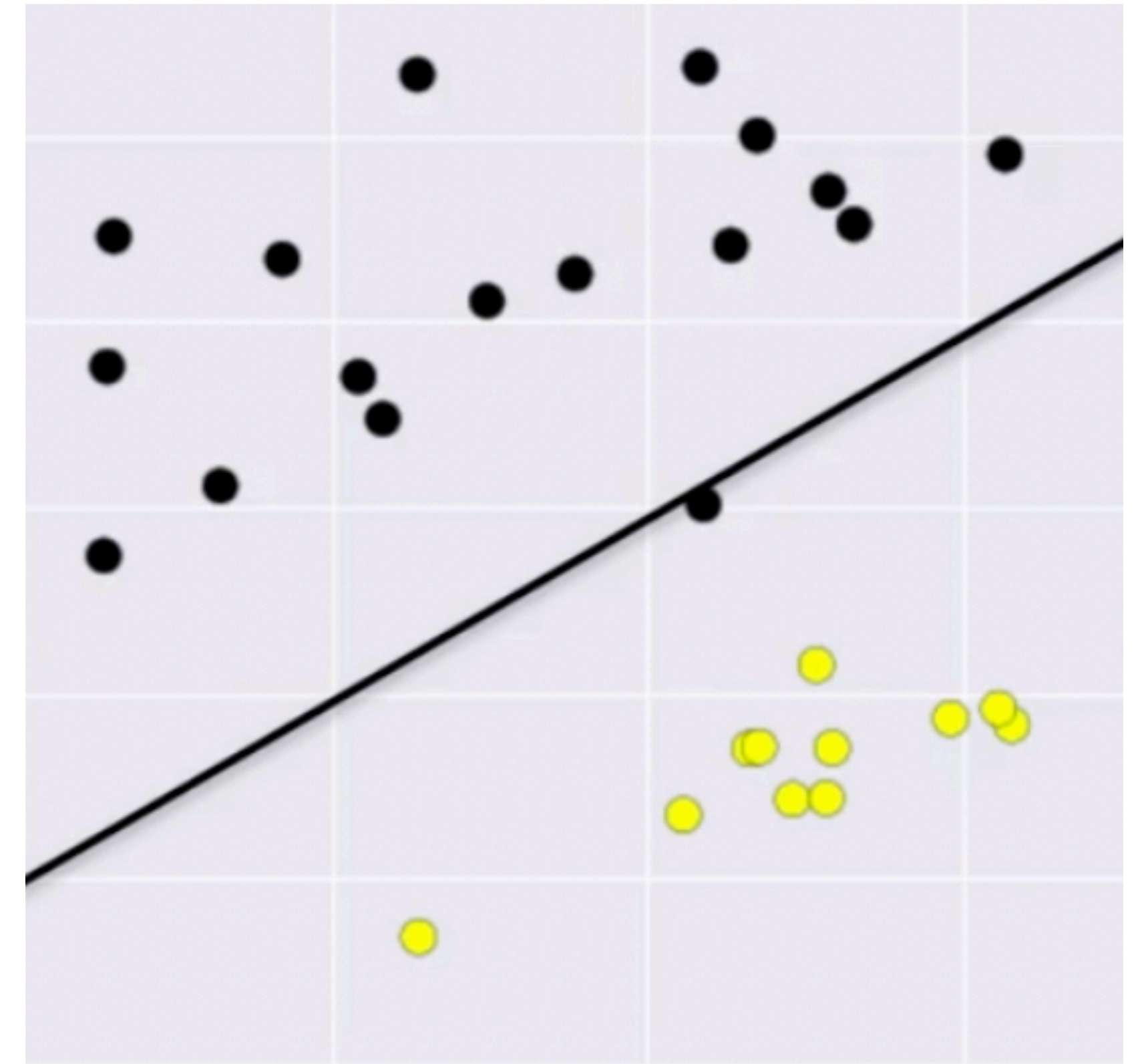


# Linear classification

A linear classifier computes the linear score:

$$f(w) = w^{\top} x_i + b$$

Predict label of  $x_i$  based on linear score  $s_i$





# **A few more loss functions for classification**

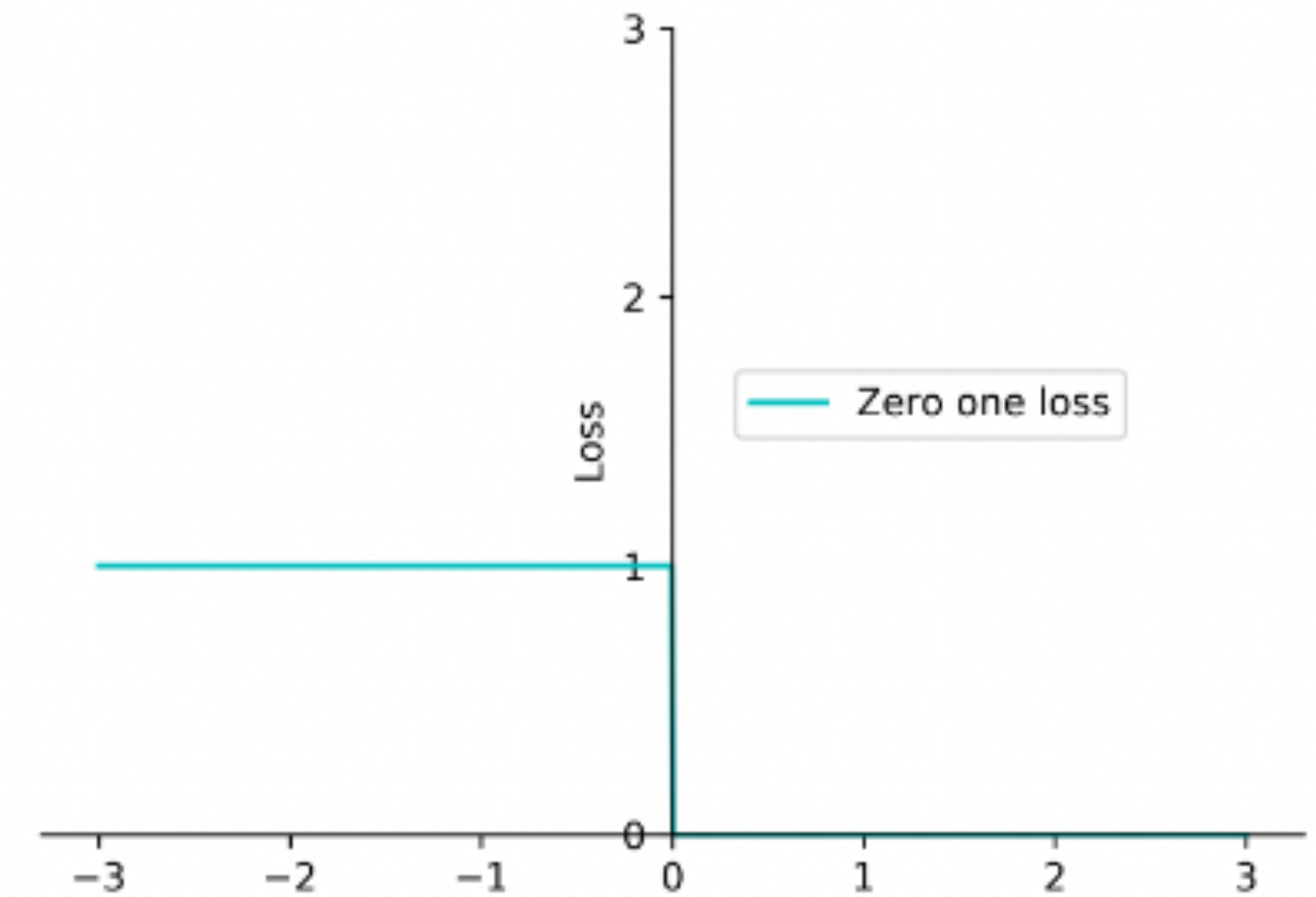


# Zero-one loss

Measure the total number of prediction errors for the classifier.

$$L_{0/1} = \begin{cases} 1, & \text{if } y_i f(w, x_i) < 0 \\ 0, & \text{otherwise} \end{cases}$$

when,  $y_i = \pm 1$  (binary case, with labels +1/-1)



# Log loss/cross-entropy loss/max-entropy loss

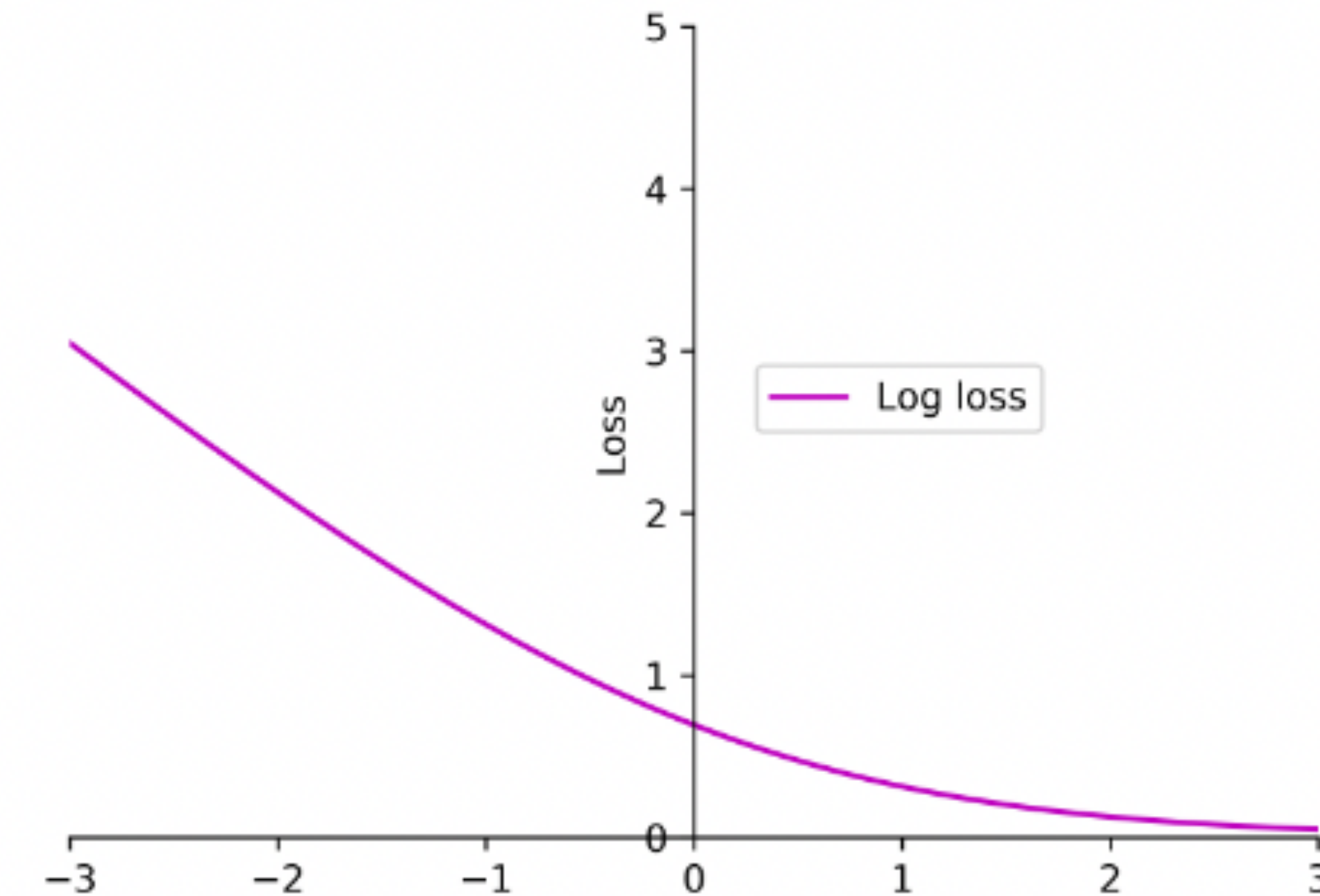
Log loss is one of the most popular loss functions for training:

$$\frac{1}{n} \sum_{i=1}^n L_{\log}(y_i, f(w, x_i))$$

where,  $L_{\log}$  is defined as:

$$L_{\log}(y_i, f(w, x_i)) = \log(1 + e^{-y_i f(w, x_i)})$$

Can be used for multi-class discrete classification tasks, i.e. one-hot target vectors.



# Exercise

Suppose your ML model output is the following in comparison with the respective target values as shown in the table

Calculate:

- (1) the classification error,
- (2) the mean square error,
- (3) least squares and
- (4) cross-entropy

ML output				Target		
0.1	0.3	0.6		0	0	1
0.2	0.6	0.2		0	1	0
0.3	0.4	0.3		1	0	0