

*Mathematics of Learning – Worksheet 10*

- The exercise sheets will be uploaded every Monday. Solution sketches will be uploaded one week later.
- You can hand in your own solutions via StudOn and we correct them - this is not mandatory. Please hand in small groups of 2-3 students.
- For questions, please use the forum on StudOn since other students may have similar questions. If you have a more personal question about the exercises please send an email to ehsan.waiezi@fau.de or lars.weidner@fau.de respectively.

**Exercise 1 [Descent directions, gradients, niveau lines]**

Consider a function  $f: \mathbb{R}^n \rightarrow \mathbb{R}$ , to be continuously differentiable. We define a descent direction for  $f$  at a point  $\tilde{x} \in \mathbb{R}^n$  as all vectors  $\theta \in \mathbb{R}^n$ , such that  $\exists \rho > 0$  such that  $f(\tilde{x} + \lambda\theta) < f(\tilde{x})$  for all  $0 < \lambda < \rho$  ("if we make a tiny ( $\rho$ ) or an even smaller ( $\lambda$ ) step in that direction ( $\theta$ ), the value of  $f$  gets smaller"). Prove or disprove:

a) The set of descent directions for  $f$  at a point  $\tilde{x}$  with  $\nabla f(\tilde{x}) \neq 0$  is the set  $\{\theta \in \mathbb{R}^n : \langle \nabla f(\tilde{x}), \theta \rangle < 0\}$ .

b) For  $\nabla f(\tilde{x}) \neq 0$ , the minimization problem

$$\min_{\theta \in \mathbb{R}^n: \|\theta\|_2=1} \lim_{\lambda \rightarrow 0} \frac{f(\tilde{x} + \lambda\theta) - f(\tilde{x})}{\lambda}$$

is solved by  $-\frac{\nabla f(\tilde{x})}{\|\nabla f(\tilde{x})\|_2}$ .

**Exercise 2 [Derivative of logistic activation function].**

Let  $\psi: \mathbb{R} \rightarrow \mathbb{R}$  the logistic activation function for a perceptron defined as

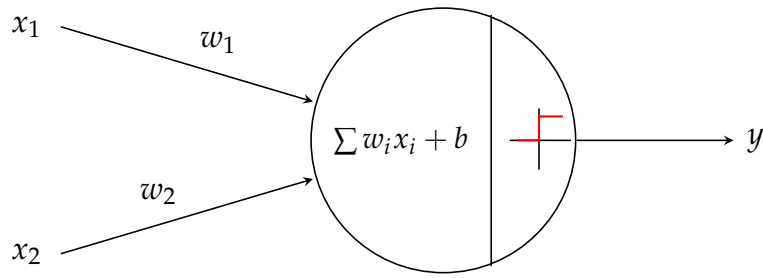
$$\psi(t) := \frac{1}{1 + e^{-t}}$$

Show that the derivative  $\psi'$  of  $\psi$  can be computed as:

$$\psi'(t) = \psi(t)(1 - \psi(t)).$$

**Exercise 3 [Implementation of perceptrons for binary logic functions].**

Let  $f_\theta: \mathbb{R}^2 \rightarrow \mathbb{R}$  be a parametrized map realized by the following binary perceptron that maps two inputs  $\vec{x} = (x_1, x_2)$  to an output  $y$ :



Here,  $\theta \in \mathbb{R}^3$  is the vector of free parameters with  $\theta := (w_1, w_2, b)$ , where  $w_1, w_2$  are the weights of the respective inputs and  $b$  is the bias of the perceptron. We assume that the activation function of the perceptron is the *Heavyside step function*  $H: \mathbb{R} \rightarrow \{0, 1\}$  defined as :

$$H(x) := \begin{cases} 0, & \text{if } x < 0, \\ 1, & \text{if } x \geq 0. \end{cases}$$

Implement a Python function `perceptron(x, theta)` that return a output  $y$ , which is either 0 or 1. Use this function to implement a family of perceptrons, which realize the following binary logic functions:

AND			OR			XOR			NAND			NOR		
$x_1$	$x_2$	$y$	$x_1$	$x_2$	$y$	$x_1$	$x_2$	$y$	$x_1$	$x_2$	$y$	$x_1$	$x_2$	$y$
0	0	0	0	0	0	0	0	0	0	0	1	0	0	1
1	0	0	1	0	1	1	0	1	1	0	1	1	0	0
0	1	0	0	1	1	0	1	1	0	1	1	0	1	0
1	1	1	1	1	1	1	1	0	1	1	0	1	1	0

**Hint:** One of the binary logic functions cannot be realized by a simple perceptron. Explain the reasons for this and suggest an alternative realization.