## **Introduction to Machine Learning Course**

## Dry 6 - Deep learning

Submitted individually by Thursday, 01.07, at 23:59. Each day of delay costs 5 points.

You may answer in Hebrew or English and write on a computer or by hand (but be clear).

Please submit a PDF file named like your ID number, e.g., 123456789.pdf.

Bonus (maximal grade is 100): Writing on a computer (using LyX/LaTeX, Word + Equation tool, etc.) = 3 pts.

1. Consider a trained fully connected neural network with *L* linear layers.

Denote the function of the network by 
$$F_{\Theta}: \mathbb{R}^d \to \mathbb{R}$$
, where  $\Theta = \left(\underbrace{\mathbf{W}^{(1)}}_{\in \mathbb{R}^{d \times p}}, \underbrace{\mathbf{W}^{(2)}}_{\in \mathbb{R}^{p \times p}}, \dots, \underbrace{\mathbf{W}^{(L-1)}}_{\in \mathbb{R}^{p \times p}}, \underbrace{\mathbf{w}^{(L)}}_{\in \mathbb{R}^{p}}\right)$  is the

set of all weights and <u>no</u> biases (we follow the notations from Tutorial 11).

As an activation function, we use the ReLU function  $\sigma(z) = \max\{0, z\}$ .

The network's output is given by:

$$F_{\Theta}(x) = \boldsymbol{w}^{(L)^{\mathsf{T}}} \boldsymbol{h}^{(L-1)}(x)$$

where we recursively define the hidden layers:

$$\boldsymbol{h}^{(1)}(x) = \sigma\left(\boldsymbol{W}^{(1)^{\mathsf{T}}}x\right), \qquad h^{(\ell)}(x) = \sigma\left(\boldsymbol{W}^{(\ell)^{\mathsf{T}}}\boldsymbol{h}^{(\ell-1)}(x)\right)$$

We now scale all weights in  $\Theta$  by a factor of  $\alpha \in \mathbb{R}_{>0}$ .

Notice: the ReLU function is positive-homogeneous in the sense that  $\sigma(\alpha \cdot z) = \alpha \cdot \sigma(z)$ .

1.1. Show that the new output function holds  $F_{\alpha \cdot \Theta}(x) = c \cdot F_{\Theta}(x)$  for some scalar  $c \in \mathbb{R}$ . In your answer, find an appropriate c value for which this statement holds.

We wish the model to output a probability. As seen in Lecture 09 (for logistic regression), we can apply the sigmoid function to  $F_{\Theta}$ . That is, the model's output will be:  $\frac{1}{1+\exp\{-F_{\alpha}\cdot\Theta(x)\}}$ .

- 1.2. For  $\alpha \to 0$ , to which probability does the output converge?
- 1.3. For  $\alpha \to \infty$ , to which probability does the output converge?