

**Machine Learning**  
Summer 2021  
**Exercise Sheet 3**

**Exercise 3-1** Linear Regression

Let  $\mathbf{X} \in \mathbb{R}^{N \times D}$  be a dataset with  $N$  samples of dimension  $D$  in which the first column contains only ones (to represent the bias),  $\mathbf{y} \in \mathbb{R}^N$  a vector with the target values, and  $\mathbf{w} \in \mathbb{R}^D$  the weight vector we want to learn. Given the scalar loss

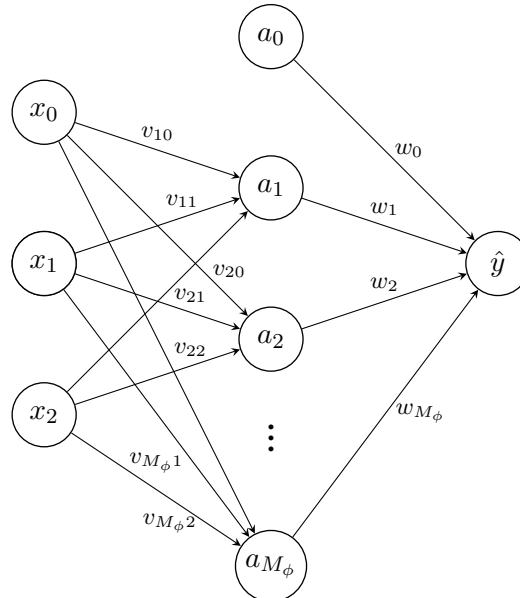
$$L = (\mathbf{y} - \mathbf{X}\mathbf{w})^T(\mathbf{y} - \mathbf{X}\mathbf{w}) + \lambda \mathbf{w}^T \mathbf{w},$$

show that the analytical solution that minimizes the loss  $L$  is  $\mathbf{w}^* = (\mathbf{X}^T \mathbf{X} + \lambda \mathbf{I})^{-1} \mathbf{X}^T \mathbf{y}$ .

*Hint:* Use the following identities:  $(\mathbf{UV})^T = \mathbf{V}^T \mathbf{U}^T$ ,  $\frac{\partial \mathbf{x}^T \mathbf{a}}{\partial \mathbf{x}} = \mathbf{a}$ ,  $\frac{\partial \mathbf{Ax}}{\partial \mathbf{x}} = \mathbf{A}^T$  and  $\frac{\partial \mathbf{x}^T \mathbf{Ax}}{\partial \mathbf{x}} = 2\mathbf{Ax}$ .

**Exercise 3-2** A simple Neural Network

The illustration below depicts a two-layered neural network with inputs  $x \in \mathbb{R}^2$  and for each input one bias term  $x_0 = 1$ , i.e.  $\mathbf{x}_i = (1, x_{i,1}, x_{i,2})^T$ . Analogously, there is a bias  $a_0 = 1$  in the hidden layer.



As activation function for the hidden neurons we employ a sigmoid, i.e.

$$a_h = \sigma(\mathbf{x}_i^T \mathbf{v}_h) = \frac{1}{1 + e^{-\mathbf{x}_i^T \mathbf{v}_h}}; \quad \forall h = 1, \dots, M_\phi.$$

The output value  $\hat{y}$  is calculated as weighted sum of the neurons in the hidden layer:  $\hat{y} = \mathbf{a}^T \mathbf{w} = \sum_{j=0}^{M_\phi} a_j w_j$ .

- (a) Prove that the following holds:  $\frac{\partial a_h}{\partial \mathbf{v}_h} = a_h (1 - a_h) \mathbf{x}_i$
- (b) Express the maximal value of  $\hat{y}$  in terms of  $\mathbf{w}$  if all weights  $w_h$  ( $h \in \{0, \dots, M_\phi\}$ ) are positive. What's the minimal value?
- (c) If  $v_{h,j} = 0$  for all  $j \in \{0, \dots, M\}$ ,  $h \in \{1, \dots, M_\phi\}$ , then what is  $\hat{y}$ ?

### **Exercise 3-3**      PyTorch - Feed Forward Neural Network

On the course website you will find an ipython notebook leading you through the implementation of a simple feed forward neural network in PyTorch for classifying handwritten digits from the MNIST dataset.