# Machine Learning Assignment 3 Neural Networks

Submission deadline: December 9, 2024

Please submit your solution in PDF format (preferably, but not necessarily, L<sup>A</sup>T<sub>E</sub>X— this .tex file can be found on iCorsi). Handwriting and scanned documents are not allowed. In case you need further help, please write on iCorsi or contact me at mikhail.andronov@idsia.ch.

## 1 Estimating the parameters of a statistical model (26 points)

You are given a data set of N measurements  $\{\mathbf{x}^{(1)}, \dots, \mathbf{x}^{(N)}\}$ , and every measurement  $\mathbf{x}^{(n)}$  contains D numbers  $(x_1^{(n)}, \dots, x_D^{(n)})$ , such as  $x_d^{(n)} \in \mathbb{N} \cup \{0\}$  for all  $n \in \{1, \dots, N\}$  and  $d \in \{1, \dots, D\}$ . You decide to model the true distribution of this dataset with an independent multivariate Poisson distribution with the parameter vector  $\lambda = (\lambda_1, \dots, \lambda_D)$ , which has the form

$$p(\mathbf{x}|\lambda) = \prod_{d=1}^{D} \frac{\lambda_d^{x_d}}{x_d!} e^{-\lambda_d}$$
 (1)

You want to estimate the optimal parameters of the model given the data.

# 1.1 Likelihood (3 points)

What is the likelihood function of  $\lambda$  given the data set of N measurements  $\{\mathbf{x}^{(1)}, \dots, \mathbf{x}^{(N)}\}$ ? (3 points)

# 1.2 Log-likelihood (3 points)

Derive the log-likelihood. Include all intermediate steps and simplify the final result.

# 1.3 MLE (10 points)

Derive the maximum likelihood estimate (MLE) of  $\lambda$ . You can assume the critical point to be the maximum, no second derivatives are required. Include all intermediate steps and simplify the final result.

#### 1.4 MAP (10 points)

You place a constraint on the parameters of the model by introducing a prior distribution on them. You assume independent exponential priors on the parameters  $\lambda_d$ 

$$p(\lambda) = \prod_{d=1}^{D} p(\lambda_d) = \prod_{d=1}^{D} \beta_d e^{-\beta_d \lambda_d}$$

where  $\beta_i > 0$ . What is the maximum a posteriori (MAP) estimate of  $\lambda$ ? Include all intermediate steps and simplify the final result.

## 2 Additional questions (7 points)

Give answers to the following questions.

# 2.1 Different prior (3 points)

What would be the MAP estimate of  $\lambda$  if we chose the uniform prior, i.e., the prior that treats all parameter values as equally likely? Explain your reasoning.

## 2.2 Choice of prior (2 points)

When would the exponential prior on  $\lambda$  be a good choice? What kind of our belief about the model parameters are we expressing in this choice of prior?

# 2.3 Prior parameters (2 points)

If we make the  $\beta$  parameters of the prior smaller and smaller, how will the shape of the prior and the MAP estimate change?