

---

**Optimization with applications I**  
**série 1**

---

**Exercise 1**

Let us consider the parametric model

$$Y_n \stackrel{\text{i.i.d}}{\sim} N(\mu, \sigma^2), \quad n = 1, \dots, N,$$

with expectation  $\mu$  and standard deviation  $\sigma > 0$ .

1. Write the likelihood function and the negative log-likelihood.
2. Find the maximum likelihood estimator of  $(\mu, \sigma^2)$ . Is the solution always in the feasible set, for any values of  $y_1, \dots, y_N$ ?
3. Find the maximum likelihood estimator of  $(\mu, \sigma)$ .
4. Compute the Hessian matrix at the point of maximum likelihood. Is it positive definite?

**Exercise 2**

Let us consider the parametric model

$$Y_n \stackrel{\text{i.i.d}}{\sim} \text{Poisson}(\lambda), \quad n = 1, \dots, N,$$

with expectation  $\lambda > 0$ . Find the maximum likelihood estimator of  $\lambda$ .

**Exercise 3**

Let us consider the following linear regression model :

$$Y_n = \alpha_0 + \alpha_1 x_{n1} + \dots + \alpha_P x_{nP} + \epsilon_n, \quad n = 1, \dots, N,$$

where  $\epsilon_1, \dots, \epsilon_N \stackrel{\text{i.i.d}}{\sim} N(0, \sigma^2)$ . We denote  $\mathbf{Y} = (Y_1, \dots, Y_N)^T$  and

$$X = \begin{pmatrix} 1 & x_{11} & \dots & x_{1P} \\ 1 & x_{21} & \dots & x_{2P} \\ \vdots & \vdots & \dots & \vdots \\ 1 & x_{N1} & \dots & x_{NP} \end{pmatrix}.$$

- a) What is the joint distribution of the  $Z_n = Y_n - (\alpha_0 + \alpha_1 x_{n1} + \dots + \alpha_P x_{nP})$ ?
- b) Write the log-likelihood function of  $\boldsymbol{\alpha} = (\alpha_1, \dots, \alpha_P)^T$  and  $\sigma^2$ .
- c) Find the maximum likelihood estimator of  $\boldsymbol{\alpha}$  and  $\sigma^2$ . Is it always well-defined?
- d) Repeat a), b) and c) when the noise (i.e. the  $\epsilon_n$ ) follow a Laplace distribution with parameter  $\lambda > 0$ , defined by the following probability density function :

$$f(x \mid \lambda) = \frac{\lambda}{2} e^{-\lambda|x|}$$