

Return your written solutions either in person or by email  
to veska.kaarnioja@fu-berlin.de by Monday 13 June, 2022, 12:15

*Note that there is no lecture or exercise session on Monday 6 June (public holiday).*

1. Consider the following multiplicative noise model

$$y_j = a_j x_j, \quad 1 \leq j \leq n,$$

where  $y, x, a \in \mathbb{R}^n$ , and assume that  $a$  is a log-normally distributed multiplicative noise vector with independent components, that is,  $\log a_j \sim \mathcal{N}(\log a_0, \sigma^2)$ . Furthermore,  $a$  is assumed to be independent of  $x$ . By taking the logarithm, the noise model becomes additive. Using this observation, derive the likelihood density  $\mathbb{P}(y|x)$  for such  $x \in \mathbb{R}^n$  that  $x_j > 0$  for all  $j = 1, \dots, n$ .

2. Suppose our inverse problem is given by

$$y = Ax + \eta,$$

where  $y \in \mathbb{R}^k$  is the observation,  $\eta \in \mathbb{R}^k$  is additive measurement noise, and  $A \in \mathbb{R}^{k \times d}$  is the matrix modeling the measurement. Moreover, suppose that the noise distribution is given by  $\eta \sim \mathcal{N}(0, I)$  and the prior distribution by  $x \sim \mathcal{N}(0, C)$ , where  $C \in \mathbb{R}^{d \times d}$  is a symmetric and positive definite matrix.

- (a) Form the posterior density  $\pi^y(x)$ .
- (b) Notice that the MAP estimator is precisely the minimizer of  $-\log(\pi^y(x))$ . Using this observation, solve the MAP estimator explicitly.

*Hint:* In part (b), it is helpful to use the Cholesky decomposition  $C^{-1} = L^T L$ .

3. The Laplace distribution is characterized by a location parameter  $t \in \mathbb{R}$  and a scale parameter  $b > 0$ . It has the probability density

$$\pi_{t,b}(x) = \frac{1}{2b} \exp\left(-\frac{|x-t|}{b}\right).$$

Compute the Hellinger distance

$$d_{\text{H}}(\pi_{t,b}, \pi_{t,c}),$$

where  $b, c > 0$ .