

## Statistical Parameter Estimation 2024

### Exercises 2

*Deadline 24 January 2024, 23:59 Lappeenranta time*

Let us consider a simple exponential decay

$$y(t) = \theta_1 + (A_0 - \theta_1) \exp(-\theta_2 t) + \varepsilon(t) \quad (1)$$

Let the data be

$$\begin{aligned} y(t) &= (0.487, 0.572, 0.369, 0.179, 0.119, 0.0809, 0.104, 0.091, 0.047, 0.051)^\top, \\ t &= (1, 2, 3, 4, 5, 6, 7, 8, 9, 10)^\top, \\ \varepsilon(t) &\sim \mathcal{N}(0, \sigma^2 I). \end{aligned}$$

1. Assume first that  $A_0 = 1$ .
  - Write the likelihood density of  $\theta$  given  $y(t)$ .
  - Derive the negative log-likelihood.
2. Maximum likelihood estimation: use the methods in the optimisation toolbox e.g. `fminsearch` in Matlab, to obtain a numerical optimizer for the maximum likelihood estimator of  $\theta$ .
3. Compute conditional mean estimate with
  - Metropolis-Hastings,
  - Adaptive Metropolis,
  - Delayed rejection, and,
  - Delayed rejection with adaptation DRAM.
4. Priors
  - Choose a Gaussian prior for  $\theta$  and define a posterior distribution. Calculate the negative log posterior. Compute the MAP and CM estimators as above.
  - Do the same as above, but choose a prior distribution that is uniformly distributed.
5. Visualisation and MCMC diagnostics
  - Plot chains, ACFs, densities – compute ESS and OES.
  - Study the chain convergence. How long chains do you need?

- Write down parameter estimates and the Monte Carlo errors of the estimates of posterior mean and posterior covariance.
- Consider different strategies to handle uncertainty in observation,  $\sigma$ .
- Generate suitable predictive envelopes around the best fit model.
- How would handle the fact that observations are constrained to positive?
- Add  $A_0$  as an extra parameter.