

## ET 4386 Estimation and Detection

Autumn 2017

### Exercises - Estimation

**Problem 1:** We observe the data  $x[n] = Ar^n + w[n]$  for  $n = 0, \dots, N - 1$ . The noise process  $w[n]$  is uncorrelated Gaussian with variance  $\sigma^2$ , and  $r > 0$  is known.

**Problem 1a:** Assume that  $A$  is deterministic. Find the CRLB for  $A$ .

**Problem 1b:** Determine the MVU estimator for  $A$  and find its variance.

**Problem 1c:** Determine the BLUE estimator for  $A$ .

**Problem 1d:** What happens with the variance of the MVU estimator if  $N \rightarrow \infty$ ?

**Problem 1e:** Now assume that  $A$  is a zero-mean random variable that is Gaussian distributed with variance  $\sigma_A^2$ , independent of  $w[n]$ . Calculate the MMSE estimator of  $A$ .

**Problem 1f:** Determine the minimum Bayesian MSE.

**Problem 1g:** Find the LMMSE estimator.

**Problem 2:** We observe the data  $x[n]$  for  $n = 0, \dots, N - 1$ . Each sample has the conditional pdf

$$p(x[n]|\theta) = \begin{cases} \frac{1}{\theta}, & 0 \leq x[n] \leq \theta \\ 0, & \text{otherwise.} \end{cases}$$

Conditioned on  $\theta$ , the observations are independent. The distribution of  $\theta$  is  $\theta \sim U(0, \beta)$ .

**Problem 2a:** Find the MMSE estimator of  $\theta$ .

**Problem 2b:** Consider the situation with little prior knowledge. For what value of  $\beta$  is the prior non-informative?

**Problem 2c:** How does the estimator change in the case of little prior knowledge.

**Problem 2d:** Determine the Bayesian ML estimator.

**Problem 2e:** Determine the MAP estimator.

**Problem 3:** We observe data  $x[n] = s[n] + w[n]$  for  $n = 0, \dots, N - 1$ . Both  $s[n]$  and  $w[n]$  are zero-mean WSS processes, mutually uncorrelated. Their autocorrelation functions are given by  $r_{ss}[k] = \sigma_s^2[k]$  and  $r_{ww}[k] = \sigma^2[k]$

**Problem 3a:** Determine the LMMSE estimator of  $\mathbf{s} = [s[0], s[1], \dots, s[N - 1]]^T$ .

**Problem 3b:** Determine the minimum Bayesian MSE  $Bmse(\hat{\theta}_i)$  for this linear estimator.

**Problem 4:** Given is the data model

$$x[n] = A + w[n]$$

for  $n = 0, \dots, N - 1$ . The unknown parameter  $A$  is assumed to have the distribution

$$p(A) = \begin{cases} \lambda \exp[-\lambda A], & A > 0 \\ 0, & A < 0, \end{cases}$$

with  $\lambda > 0$  and  $w[n]$  WGN independent of  $A$  with variance  $\sigma^2$ .

**Problem 4a:** Determine the MAP estimator of  $A$ .

**Problem 4b:** Determine the LMMSE estimator of  $A$ .

**Problem 5:** Given is the a posteriori pdf

$$p(\theta|x) = \begin{cases} \exp[-(\theta - x)], & \theta > x \\ 0, & \theta < x. \end{cases}$$

**Problem 5a:** Calculate the MMSE and MAP estimator of  $\theta$ .

**Problem 6:** Given is the distribution

$$p(x[n]) = \begin{cases} \theta \exp[-\theta x[n]], & x[n] > 0 \\ 0, & x[n] < 0, \end{cases}$$

for  $n = 0, \dots, N - 1$ . The samples  $x[n]$  are independent over time.

**Problem 6a:** Give the CRLB for  $\hat{\theta}$ .

**Problem 6b:** Does an unbiased estimator exist that achieves the CRL bound?

**Problem 6c:** Determine the MLE estimator.

The a priori distribution  $p(\theta)$  is given by

$$p(\theta) = \begin{cases} \lambda \exp[-\lambda\theta], & \theta > 0 \\ 0, & \theta < 0. \end{cases}$$

**Problem 6d:** Calculate the MAP estimator.

**Problem 6e:** Calculate the MMSE estimator.