SSP

# Exam

You can answer the questions in French or in English. If you get stuck in a particular part of a question, do not hesitate to try to answer the rest of the question or start the next question. Results that are available in the course notes can be quoted and do not need to be rederived.

#### **Bayesian Parameter Estimation**

#### 1. Bayesian Level Estimation in Gaussian Noise with Laplacian Prior

Assume that we obtain the following noisy measurements

$$y_i = \theta + v_i \,, \quad i = 1, \dots, n \tag{1}$$

of the unknown level  $\theta$  with Laplacian prior distribution

$$f_{\theta}(x) = \mu e^{-\lambda |x|}, -\infty < x < \infty$$
 (2)

and the measurement noise is i.i.d. and Gaussian:  $v_i \sim \mathcal{N}(0, \sigma_v^2)$ , and independent of  $\theta$ . The quantities  $\lambda$  and  $\sigma_v^2$  are known.

- (a) Given  $\lambda$ , determine  $\mu$ .
- (b) Determine the prior mean  $m_{\theta}$ .
- (c) Determine the prior variance  $\sigma_{\theta}^2$ .
- (d) Compute the MAP estimator of  $\theta$  given  $Y = [y_1 \cdots y_n]^T$ . Express  $\widehat{\theta}_{MAP}$  in terms of  $\overline{y} = \frac{1}{n} \sum_{k=1}^n y_k$ .

Hints: express  $\overline{y} = |\overline{y}| \operatorname{sign}(\overline{y})$ ,  $\theta = |\theta| \operatorname{sign}(\theta)$ . First find  $\operatorname{sign}(\theta)$  and then  $|\theta|$ , for which two cases can occur.

(e) Is  $\hat{\theta}_{MAP}$  unbiased (in a Bayesian sense)?

#### **Deterministic Parameter Estimation**

## 2. Deterministic Level Estimation in Laplacian Noise

Assume that we obtain the following noisy measurements

$$y_i = \theta + v_i \,, \quad i = 1, \dots, n \tag{3}$$

of the unknown level  $\theta$  and the measurement noise is i.i.d. with Laplacian distribution

$$f_v(x) = \frac{\delta}{2} e^{-\delta |x|} . {4}$$

- (a) Determine  $\sigma_v^2$  as a function of  $\delta$ .
- (b) Show that the maximum likelihood estimator is  $\hat{\theta}_{ML} = \arg\min_{\theta} \sum_{i=1}^{n} |y_i \theta|$ .
- (c) We shall find  $\hat{\theta}_{ML}$  explicitly as a function of  $Y = [y_1 \ y_2 \cdots y_n]^T$  by proceeding in small steps:
  - (i) By drawing  $g(\theta) = \sum_{i=1}^{n} |y_i \theta|$ , find  $\hat{\theta}_{ML}$  for n = 1, 2, 3, 4. Hint: let  $\{\tilde{y}_i, i = 1, \dots, n\}$  be a reordering of  $\{y_i, i = 1, \dots, n\}$  so that  $\tilde{y}_1 \leq \tilde{y}_2 \leq \dots < \tilde{y}_n$ .
  - (ii) Extrapolate now your results from (i) to give  $\widehat{\theta}_{ML}$  for general n. Distinguish the cases n even and n odd. In particular for n odd, what is the name of the function  $\widehat{\theta}_{ML}(Y)$ ?
- (d) We'll consider now the BLUE estimator.
  - (i) Give  $\widehat{\theta}_{BLUE}(Y)$ .
  - (ii) Give E  $\hat{\theta}_{BLUE}(Y)$ . Is  $\hat{\theta}_{BLUE}(Y)$  biased?
  - (iii) Compute  $R_{\widetilde{\theta}_{BLUE}\widetilde{\theta}_{BLUE}}$  and express it in terms of  $\sigma_v^2$ .
- (e) Compute the Fisher information and the CRB and express both in terms of  $\sigma_v^2$ .
- (f) Is the BLUE estimator efficient?
- (g) Asymptotically, as  $n \to \infty$ , give  $R_{\widetilde{\theta}_{ML}\widetilde{\theta}_{ML}}$  as a function of  $\sigma_v^2$ . Explain (remember the asymptotic properties of ML). Compare to  $R_{\widetilde{\theta}_{BLUE}\widetilde{\theta}_{BLUE}}$ .

#### Non-Parametric Spectrum Estimation

## 3. Spectral Resolution Issues

We consider the use of the Blackman-Tukey (BT) spectral estimator and more specifically with a Bartlett window of length 2M + 1. We want a spectral resolution in normalized frequency of at least 0.02 (assume the common rule of thumb for the resolution of two equiamplitude sinusoids). We furthermore want the BT spectral estimator to have a variance that is at least nine times lower than the variance of the periodogram. What is the minimum number N of samples that we need to have to reach these specifications?

Turn the page please.

## Wiener Filtering and Equalization

## 4. FIR (U)MMSE Linear Equalization of a FIR Channel

Consider a causal FIR equalizer H with N coefficients,  $\hat{x}_k = H^T Y_k$ . For a FIR channel of length L, the received signal vector  $Y_k$  can be written as

$$\begin{bmatrix} y_k \\ y_{k-1} \\ \vdots \\ y_{k-N+1} \end{bmatrix} = \begin{bmatrix} c_0 & c_1 & \cdots & c_{L-1} & 0 & \cdots & 0 \\ 0 & c_0 & \cdots & c_{L-2} & c_{L-1} & \cdots & 0 \\ \vdots & \ddots & \ddots & & \ddots & \ddots & \vdots \\ 0 & \cdots & 0 & c_0 & \cdots & \cdots & c_{L-1} \end{bmatrix} \begin{bmatrix} s_k \\ s_{k-1} \\ \vdots \\ s_{k-N-L+2} \end{bmatrix} + \begin{bmatrix} v_k \\ v_{k-1} \\ \vdots \\ v_{k-N+1} \end{bmatrix}$$

or

$$\underbrace{Y_k}_{N\times 1} = \underbrace{C}_{N\times (N+L-1)} \underbrace{S_k}_{(N+L-1)\times 1} + \underbrace{V_k}_{N\times 1} = \left[\underline{c_0} \ \underline{c_1} \ \cdots \ \underline{c_{N+L-2}}\right] S_k + V_k = \sum_{i=0}^{N+L-2} \underbrace{\underline{c_i}}_{N\times 1} \ s_{k-i} + V_k \ .$$

The vector  $\underline{c}_i$  is column i+1 of the matrix C. The symbol sequence  $s_k$  is considered to be white noise with zero mean and variance  $\sigma_s^2$ . The additive noise  $v_k$  is independent of the symbol sequence and white Gaussian with zero mean and variance  $\sigma_v^2$ . We shall see that it may be advantageous to introduce an equalization delay d. Hence consider  $x_k^{(d)} = s_{k-d}, d \in \{0, 1, \ldots, N+L-2\}$ .

MMSE FIR equalization is a particular instance of FIR Wiener filtering. Hence the MMSE FIR equalizer coefficients  $H_{MMSE}^{(d)}$  satisfy the normal equations

$$R_{YY} H_{MMSE}^{(d)} = R_{Yx^{(d)}}$$
 or hence  $H_{MMSE}^{(d)} = R_{YY}^{-1} R_{Yx^{(d)}}$ , and the MMSE is  $\sigma_{\widetilde{x}_{MMSE}^{(d)}}^2 = R_{x^{(d)}x^{(d)}} - R_{x^{(d)}Y}R_{YY}^{-1} R_{Yx^{(d)}}$ .

- (a) Determine  $R_{YY}$  in terms of  $\sigma_s^2$ ,  $\sigma_v^2$ , the matrix C and the identity matrix  $I_N$ , and determine  $R_{Yx^{(d)}}$  in terms of  $\sigma_s^2$  and the vector  $\underline{c}_d$ .
- (b) Express the MMSE  $\sigma_{\widetilde{x}_{MMSE}^{(d)}}^2$  in terms of these same quantities.

The corresponding (naive) SNR is 
$$\text{SNR}_{MMSE}^{(d)} = \frac{\sigma_s^2}{\sigma_{\widetilde{x}_{MMSE}}^2}$$
.

In what follows, we shall consider the specific case of an equalizer with N=2 coefficients, a channel with L=2 coefficients  $c_0=1, c_1=a$ , and no noise  $\sigma_v^2=0$ . The range of possible delays is now limited to  $d \in \{0,1,2\}$ . In the absence of noise, the MSE is determined by intersymbol interference which is unavoidable here with an FIR equalizer.

- (c) Compute the MMSE  $\sigma^2_{\widetilde{x}^{(d)}_{MMSE}}$  for d=0,1,2, in terms of  $\sigma^2_s$  and a.
- (d) Compute, for d=0,1,2,  $SNR_{UMMSE}^{(d)}$  (via  $SNR_{UMMSE}^{(d)} = SNR_{MMSE}^{(d)} 1$ ) in terms of a.
- (e) Determine the optimal delay  $d \in \{0, 1, 2\}$  as a function of |a| (optimal in the sense of maximizing  $SNR_{UMMSE}^{(d)}$ ).