

# MVE136 Random Signals Analysis

## Written exam Monday 19 August 2019 2-6 pm

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AIDS: Beta or 2 sheets (=4 pages) of hand-written notes (computer print-outs and/or xerox-copies are not allowed), but not both these aids.

GRADES: 12 (40%), 18 (60%) and 24 (80%) points for grade 3, 4 and 5, respectively.

MOTIVATIONS: All answers/solutions must be motivated. GOOD LUCK!

**Task 1.** Find the probability  $Pr(X(1)+X(2)+X(3) > 6)$  for a continuous time WSS Gaussian process  $X(t)$  with mean  $\mu_X = 1$  and autocorrelation function  $R_{XX}(\tau) = e^{-|\tau|} + 1$  for  $\tau \in \mathbb{R}$ . **(5 poäng)**

**Task 2.** Find the probability  $Pr(X(1) + X(2) > 3)$  for a Poisson process with rate 1. **(5 poäng)**

**Task 3.** Give an example of a WSS random process that is not strict sense stationary. **(5 poäng)**

**Task 4.** Calculate  $E(X[n]X[n+1])$  for non-negative integers  $n$  when  $\{X[n], n \geq 0\}$  is a Markov chain with state space (/possible values)  $S$ , initial distribution  $\pi(0)$  and transition probability matrix  $P$ , respectively, given by

$$S = \{0, 1\}, \quad \pi(0) = [1/2 \ 1/2] \quad \text{and} \quad P = \begin{bmatrix} 1/2 & 1/2 \\ 1/2 & 1/2 \end{bmatrix}. \quad \text{(5 points)}$$

**Task 5.** The sum  $X(t) = S(t) + N(t)$  of a continuous time signal process  $S(t)$  with PSD  $S_{SS}(f) = 2/(2 + (2\pi f)^2)$  and an independent (of  $S(t)$ ) continuous time noise process  $N(t)$  with PSD  $S_{NN}(f) = 1$  is input to an LTI filter. Determine the impulse response of the filter that produces an output  $Y(t)$  from the filter that minimizes the mean-square signal-recovering error  $E[(S(t) - Y(t))^2]$ . Also, find the PSD of the output  $Y(t)$  when the filter has this impulse response. **(5 points)**

**Task 6.** Prediction is an important tool in many contexts, including decision theory, planning and control. Suppose that we wish to predict  $x[n+1]$  given  $x[n]$  and  $x[n-1]$ , where  $x[n]$  is an MA(2)-process

$$x[n] = e[n] + 0.5e[n-1] + 0.2e[n-2] \quad \text{for } n \in \mathbb{N}.$$

The input noise  $e[n]$  is assumed to be a wide sense stationary zero mean white noise process with variance  $E[e[n]^2] = 1$  and autocorrelation function  $r_e[k] = E[e[n]e[n+k]] = 0$  for  $k \neq 0$ . We will further assume that we seek a linear estimator

$$\hat{x}[n+1] = h_0 x[n] + h_1 x[n-1].$$

Find the coefficients  $h_0$  and  $h_1$  that minimize  $E[(x[n+1] - \hat{x}[n+1])^2]$ . **(5 poäng)**

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### Solutions to written exam 19 August 2019

**Task 1.** As  $X(1)+X(2)+X(3)$  is  $N(m, \sigma^2)$ -distributed we have  $Pr(X(1)+X(2)+X(3) > 6) = Pr(N(m, \sigma^2) > 6) = 1 - \Phi((6-m)/\sigma)$ , where  $m = E[X(1)+X(2)+X(3)] = 3$  and  $\sigma^2 = \text{Var}(X(1)+X(2)+X(3)) = 3C_{XX}(0) + 4C_{XX}(1) + 2C_{XX}(2) = 3 + 4e^{-1} + 2e^{-2}$  [using that  $C_{XX}(\tau) = R_{XX}(\tau) - \mu_X^2 = e^{-|\tau|}$ ].

**Task 2.** As  $X(1) + X(2) = (X(2) - X(1)) + 2X(1)$  where  $X(2) - X(1)$  and  $X(1)$  are independent Po(1)-distributed we have  $Pr(X(1) + X(2) > 3) = Pr((X(2) - X(1)) + 2X(1) > 3) = Pr(X(1) \geq 2) + Pr(X(1) = 1, X(2) - X(1) > 1) + Pr(X(1) = 0, X(2) - X(1) > 3) = (1 - e^{-1} - e^{-1}) + e^{-1}(1 - e^{-1} - e^{-1}) + e^{-1}(1 - e^{-1} - e^{-1} - \frac{1}{2}e^{-1} - \frac{1}{6}e^{-1})$ .

**Task 3.** For example, the process  $\{X(t)\}_{t \in \mathbb{Z}}$  made up of independent random variables that are  $N(0, 1)$ -distributed for  $t$  even and that have a discrete Rademacher distribution with PMF  $P_{X(t)}(-1) = P_{X(t)}(1) = 1/2$  for  $t$  odd, as this process is zero-mean with autocorrelation function  $R_{XX}(\tau) = \delta(\tau)$ , but clearly is not strict sense stationary.

**Task 4.** As the only possible values of the random variable  $X[n]X[n+1]$  are 0 and 1, we have  $E(X[n]X[n+1]) = 0 \cdot P(X[n]X[n+1] = 0) + 1 \cdot P(X[n]X[n+1] = 1) = P(X[n]X[n+1] = 1) = P(X[n] = 1, X[n+1] = 1) = \pi(n)_1 P_{11} = \pi(n)_1 (1/2)$ . Noting that  $\pi(0) = \pi$  is in fact a stationary distribution for the chain we have  $\pi(n) = \pi(0) = \pi$ , so that  $\pi(n)_1 = \pi(0)_1 = 1/2$  and  $E(X[n]X[n+1]) = 1/4$ . (The latter result is in fact also more or less obviously true already from the beginning without calculations from symmetry considerations ... .)

**Task 5.** We are looking for the impulse response of the Wiener filter with transfer function  $H(f) = S_{SS}(f)/(S_{SS}(f) + S_{NN}(f)) = 2/(4 + (2\pi f)^2)$  so that the impulse response (see Table E.1 in the book of Miller and Childers) is  $h(t) = 2e^{-2|t|}$ . Further, we have

$$S_{YY}(f) = |H(f)|^2 S_{XX}(f) = \frac{S_{SS}(f)^2}{S_{SS}(f) + S_{NN}(f)} = \frac{4}{(4 + (2\pi f)^2)(2 + (2\pi f)^2)}.$$

**Task 6.** Let us first formulate this problem in terms of our standard notation for Wiener filtering: Our quantity of interest is usually denoted  $d[n]$  whereas our estimator is denoted  $\hat{d}[n]$ . In this problem, we have  $d[n] = x[n+1]$  and  $\hat{d}[n] = h_0 x[n] + h_1 x[n-1]$ . In terms of this notation, the Wiener-Hopf equations can be expressed as

$$\begin{cases} h_0 r_x[0] + h_1 r_x[1] = r_{dx}[0] \\ h_0 r_x[1] + h_1 r_x[0] = r_{dx}[1] \end{cases},$$

which can also be written on matrix form

$$\begin{bmatrix} r_x[0] & r_x[1] \\ r_x[1] & r_x[0] \end{bmatrix} \begin{bmatrix} h_0 \\ h_1 \end{bmatrix} = \begin{bmatrix} r_{dx}[0] \\ r_{dx}[1] \end{bmatrix}. \quad (1)$$

For this problem, we note that  $r_{dx}[0] = E[d[n]x[n]] = E[x[n+1]x[n]] = r_x[1]$  and  $r_{dx}[1] = E[d[n]x[n-1]] = E[x[n+1]x[n-1]] = r_x[2]$  so that it is therefore sufficient to compute  $r_x[k]$  for  $k = 0, 1$  and  $2$  before we can solve for  $h_0$  and  $h_1$ .

Let us try to derive a general expression for  $r_x[k]$ : Given our expression for  $x[n]$  it holds that

$$\begin{aligned} r_x[k] &= E[x[n]x[n-k]] \\ &= E[(e[n] + 0.5e[n-1] + 0.2e[n-2])(e[n-k] + 0.5e[n-1-k] + 0.2e[n-2-k])] \\ &= \begin{cases} 0 & \text{if } |k| > 2 \\ 0.2 & \text{if } |k| = 2 \\ 0.5 + 0.2 \cdot 0.5 = 0.51 & \text{if } |k| = 1 \\ 1 + 0.5 + 0.2 = 1.7 & \text{if } k = 0 \end{cases}. \end{aligned} \quad (2)$$

By combining (1) with (2), we obtain the matrix equation

$$\begin{bmatrix} 1.7 & 0.51 \\ 0.51 & 1.7 \end{bmatrix} \begin{bmatrix} h_0 \\ h_1 \end{bmatrix} = \begin{bmatrix} 0.51 \\ 0.2 \end{bmatrix},$$

from which we can find the solutions  $h_0 \approx 0.29$  and  $h_1 \approx 0.03$ .