

# MVE136 Random Signals Analysis

Written exam Monday 17 August 2015 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  $\Pr(X(1) = 1)$  for a Poisson process  $X(t)$  with  $E[X(2)^2] = 6$ . **(5 points)**

**Task 2.** Let  $X(t)$  be a WSS Gaussian process such that  $\Pr(X(1)+X(2) \leq 3) = \Phi(1/3)$  and  $\Pr(X(1)+X(2) \leq 4) = \Phi(1/2)$ , where as usual  $\Phi(x)$  denotes the standard Gaussian CDF. Find the mean function  $\mu_X(t) = E[X(t)]$  of the process  $X(t)$ . **(5 points)**

**Task 3.** Consider a Markov chain  $X_k$  with states 0 and 1, and with  $X_0 = 0$ . Find the transition matrix  $P$  if it is known that  $E[X_1] = E[X_2] = 1/2$ . **(5 points)**

**Task 4.** Write a little essay providing the basic ideas and techniques of spectral estimation for WSS random processes. **(5 points)**

**Task 5.** A continuous time LTI system has input  $X(t) = U \cos(\omega_0 t) + V \sin(\omega_0 t)$ , where  $U$  and  $V$  are independent standard normal random variables and  $\omega_0$  is a constant. Find the impulse response  $h(t)$  if the output is  $Y(t) = U \sin(\omega_0 t) - V \cos(\omega_0 t)$ . **(5 points)**

**Task 6.** Your task here is to derive the Wiener-Hopf equations for an FIR-filter of order 2. To be more specific, suppose that the impulse response of a linear filter is

$$h[n] = h_0 \delta[n] + h_1 \delta[n-1] + h_2 \delta[n-2],$$

and that the input signal is a WSS signal  $x[n]$ . Derive a matrix equation that determines the filter coefficients  $h_0$ ,  $h_1$  and  $h_2$  so that the output from the filter  $\hat{d}[n]$  minimizes  $E[(d[n] - \hat{d}[n])^2]$ . See Figure 1 below for an illustration. You may use the orthogonality principle to derive the equation asked for if you want. **(5 points)**

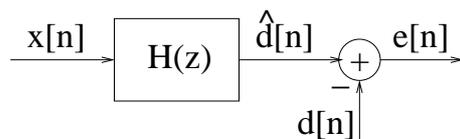


Figure 1: An illustration of the relation between  $x[n]$ ,  $h[n]$ ,  $\hat{d}[n]$ ,  $d[n]$  and the filtering error  $e[n]$  that we seek to minimize.

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### Solutions to written exam 17 August 2015

**Task 1.** As  $E[X(2)^2] = 2\lambda + (2\lambda)^2 = (2\lambda + 1/2)^2 - 1/4$  equals  $6 = 24/4$  we have  $2\lambda = -1/2 \pm \sqrt{1/4 + 24/4} = -1/2 \pm 5/2 = 2$ , so that  $\lambda = 1$  and  $\Pr(X(1) = 1) = \lambda e^{-\lambda} = 1/e$ .

**Task 2.** As  $X(t)$  is WSS the mean function is constant  $\mu_X(t) = \mu_X$ . As  $X(1)+X(2)$  is  $N(m, \sigma^2)$ -distributed with  $m = E[X(1)+X(2)] = 2\mu_X$ , we have  $\Pr(X(1)+X(2) \leq 3) = \Phi((3-2\mu_X)/\sigma)$  and  $\Pr(X(1)+X(2) \leq 4) = \Phi((4-2\mu_X)/\sigma)$ , so that  $(3-2\mu_X)/\sigma = 1/3$  and  $(4-2\mu_X)/\sigma = 1/2$ , giving  $\mu_X = 1/2$  (and  $\sigma = 6$ ).

**Task 3.** As  $E[X_1] = 0 \cdot p_{0,0} + 1 \cdot p_{0,1} = p_{0,1} = 1/2$  we have  $p_{0,0} = 1 - p_{0,1} = 1/2$ . As  $E[X_2] = 0 \cdot \Pr(X_2=0) + 1 \cdot \Pr(X_2=1) = \Pr(X_2=1) = ([1 \ 0] P^2)_1 = \dots = 1/4 + p_{1,1}/2 = 1/2$  we have  $p_{1,1} = 1/2$  and  $p_{1,0} = 1 - p_{1,1} = 1/2$ .

**Task 4.** See Section 10.4 in the book by Miller and Childers.

**Task 5.** It is easy to see mathematically or figure out more “hands-on” that the impulse response must be  $h(t) = \delta(t - \pi/(2\omega_0))$ .

**Task 6.** According to the orthogonality principle, the optimal filter must satisfy

$$E[(d[n] - \hat{d}[n])x[n]] = E[(d[n] - \hat{d}[n])x[n-1]] = E[(d[n] - \hat{d}[n])x[n-2]] = 0,$$

since the error  $d[n] - \hat{d}[n]$  should be orthogonal to the data  $x[n]$ ,  $x[n-1]$  and  $x[n-2]$ .

By substituting  $\hat{d}[n] = h_0 x[n] + h_1 x[n-1] + h_2 x[n-2]$  above we readily obtain

$$\begin{bmatrix} r_x[0] & r_x[1] & r_x[2] \\ r_x[1] & r_x[0] & r_x[1] \\ r_x[2] & r_x[1] & r_x[0] \end{bmatrix} \begin{bmatrix} h_0 \\ h_1 \\ h_2 \end{bmatrix} = \begin{bmatrix} r_{dx}[0] \\ r_{dx}[1] \\ r_{dx}[2] \end{bmatrix}.$$

These are the Wiener-Hopf equations that we were asked to derive. Instead of using the orthogonality principle, it is also possible to derive the Wiener-Hopf equations by setting the derivatives of our cost function to zero,

$$\frac{\partial}{\partial h_i} E[(d[n] - \hat{d}[n])^2] = 0 \quad \text{for } i = 0, 1, 2.$$