

## Midterm Exam (10\_2)

1. Suppose that  $z = s + v$ , where  $s$  and  $v$  are independent and  $v \sim N(0, \sigma_v^2)$ . Further assume that we have a knowledge on  $s$ , viz.,  $s \sim N(\bar{s}, \sigma_s^2)$ .
  - (1) Find the likelihood function and the maximum likelihood (ML) estimate  $\hat{s}_{ML}$ . (5 points)
  - (2) Find the maximum *a posteriori* (MAP) estimate  $\hat{s}_{MAP}$ . (10 points)
  - (3) Find the mean square (MS) estimate  $\hat{s}_{MS}$ . (15 points)

2. A stochastic unknown  $X$  and the data  $Z$  are jointly distributed according to

$$f_{XZ}(x, z) = \frac{6}{7}(x + z)^2; \quad 0 \leq x \leq 1, 0 \leq z \leq 1.$$

We want to find the best linear mean square (LMS) estimate  $\hat{X}_{LMS}(z)$ .

- (1) First, calculate the best estimate for  $X$  and its error covariance in the absence of data  $Z$  to compare with the LMS estimate. (5 points)
- (2) Show that  $\hat{X}_{LMS}(z)$  is obtained from

$$\hat{X}_{LMS}(z) = \bar{X} + P_{XZ}P_Z^{-1}(z - \bar{z}). \quad (10 \text{ points})$$

- (3) Find  $\hat{X}_{LMS}(z)$  using the equation in (2). (10 points)
- (4) Calculate the error covariance of  $\hat{X}_{LMS}(z)$  and compare with the result in (1). (5 points)

3. Suppose a signal  $s(n)$  is generated by the stochastic difference equation

$$s(n) = 0.75s(n-1) + w(n)$$

where  $w(n)$  is white noise  $w(n) \sim N(0, 4.9)$ . The signal is measured in the presence of additive white noise  $v(n) \sim N(0, 8)$  that is uncorrelated with the signal.

- (1) Show that the power spectral density of  $s(n)$  is given by

$$S_s(z) = \frac{4.9}{(1 - 0.75z^{-1})(1 - 0.75z)}. \quad (5 \text{ points})$$

- (2) Find the causal Wiener filter for estimating  $\hat{s}(n)$  from the noisy

measurements. Give the filter's transfer function  $H(z)$  and impulse response  $h(n)$ . (15 points)

(3) Show that the mean square estimation error (MSE) can be written by

$$\text{MSE} = R_s(0) - \sum_{i=0}^{\infty} h(i)R_{sz}(i). \quad (5 \text{ points})$$

(4) Using the equation in (3), compute the MSE associated with  $\hat{s}(n)$ . (5 points)

(Solutions)

1.

(1) (Kamen Example 3.3)

The likelihood function is

$$f_{z|s}(z | s) = f_v(z - s) = \frac{1}{\sqrt{2\pi}\sigma_v} e^{-(z-s)^2 / 2\sigma_v^2}.$$

$$\hat{s}_{ML} = z.$$

(2) (Kamen Example 3.6)

$$\hat{s}_{MAP} = \bar{s} + \frac{\sigma_s^2}{\sigma_v^2 + \sigma_s^2}(z - \bar{s}) \quad \text{or} \quad \hat{s}_{MAP} = \frac{\sigma_v^2 \bar{s} + \sigma_s^2 z}{\sigma_v^2 + \sigma_s^2}.$$

(3) (Kamen Example 3.8)

$$\hat{s}_{MS} = \hat{s}_{MAP}.$$

2.

(1)

$$f_X(x) = \int_{-\infty}^{\infty} f_{XZ}(x, z) dz = \int_0^1 \frac{6}{7}(x+z)^2 dz = \frac{6}{7} \left( x^2 + x + \frac{1}{3} \right)$$

$$\hat{X}_{ap} = \bar{X} = \frac{9}{14} = 0.643$$

$$P_{\hat{X}_{ap}} = E \{ \tilde{X}_{ap}^2 \} = E \left\{ \left( X - \frac{9}{14} \right)^2 \right\} = 0.0677.$$

(2) (ETNote10\_2.doc)

(3)

$$P_{XZ} = E \{ XZ \} - \bar{X}\bar{Z} = \int_0^1 \int_0^1 xz \frac{6}{7}(x+z)^2 dx dz - \bar{X}\bar{Z} = -0.0085$$

$$P_Z = E \{ Z^2 \} - \bar{Z}^2 = \int_0^1 z^2 \frac{6}{7} \left( z^2 + z + \frac{1}{3} \right) dz - \bar{Z}^2 = 0.068$$

$$\hat{X}_{LMS}(z) = 0.643 - \frac{0.0085}{0.068}(z - 0.643) = 0.724 - 0.126z.$$

(4)

$$\begin{aligned} P_{\hat{X}_{LMS}} &= E \left\{ \left[ (X - \bar{X}) - P_{XZ} P_Z^{-1} (Z - \bar{Z}) \right] \left[ (X - \bar{X}) - P_{XZ} P_Z^{-1} (Z - \bar{Z}) \right]^T \right\} \\ &= E \left\{ (X - \bar{X})(X - \bar{X})^T - P_{XZ} P_Z^{-1} (Z - \bar{Z})(X - \bar{X})^T \right. \\ &\quad \left. - (X - \bar{X})(Z - \bar{Z})^T P_Z^{-1} P_{ZX} + P_{XZ} P_Z^{-1} (Z - \bar{Z})(Z - \bar{Z})^T P_Z^{-1} P_{ZX} \right\} \\ &= P_{\hat{X}} - P_{XZ} P_Z^{-1} P_{ZX} = P_{\hat{X}} - P_{XZ} P_Z^{-1} P_{XZ}^T \leq P_{\hat{X}} \end{aligned}$$

In this problem,

$$P_{\hat{X}} \approx 0.0677$$

$$P_{\hat{X}_{LMS}} \approx 0.0666.$$

3. (Kamen Problem 4.11)

(1)

$$s(n) = 0.75s(n-1) + w(n)$$

$$S_s(z) = 0.75z^{-1}S_s(z) + S_w(z); S_w(z) = 4.9$$

$$H_s(z) = \frac{S_s(z)}{S_w(z)} = \frac{1}{1 - 0.75z^{-1}}$$

$$S_s(z) = H_s(z)H_s(z^{-1})S_w(z) = \frac{4.9}{(1 - 0.75z^{-1})(1 - 0.75z)}$$

$$R_s(n) = 11.2(0.75)^{|n|}$$

(2)

$$S_z(z) = S_s(z) + S_v(z) = \frac{4.9}{(1 - 0.75z^{-1})(1 - 0.75z)} + 8$$

$$= \frac{15(1 - 0.4z)(1 - 0.4z^{-1})}{(1 - 0.75z^{-1})(1 - 0.75z)}$$

$$S_z^+(z) = \frac{3.873(1 - 0.4z^{-1})}{(1 - 0.75z^{-1})}; \quad S_z^-(z) = \frac{3.873(1 - 0.4z)}{(1 - 0.75z)}$$

$$\frac{S_{sz}(z)}{S_z^-(z)} = \frac{(1 - 0.75z)}{3.873(1 - 0.4z)} \cdot \frac{4.9}{(1 - 0.75z^{-1})(1 - 0.75z)}$$

$$= \frac{1.265}{(1 - 0.4z)(1 - 0.75z^{-1})} = \frac{-3.16z^{-1}}{(1 - 0.75z^{-1})(1 - 2.5z^{-1})}$$

$$= \frac{1.81}{1 - 0.75z^{-1}} - \frac{1.81}{1 - 2.88z^{-1}}$$

$$\left[ \frac{S_{sz}(z)}{S_z^-(z)} \right]_+ = \frac{1.81}{1 - 0.75z^{-1}}$$

$$H(z) = \frac{1}{S_z^+(z)} \left[ \frac{S_{sz}(z)}{S_z^-(z)} \right]_+$$

$$= \frac{(1 - 0.75z^{-1})}{3.873(1 - 0.4z^{-1})} \cdot \frac{1.81}{(1 - 0.75z^{-1})} = \frac{0.4667}{(1 - 0.4z^{-1})}$$

$$h(n) = 0.4667(0.4)^n 1(n).$$

(3) (ETNote10\_3.doc)

(4)

$$\begin{aligned}\text{MSE} &= R_s(0) - \sum_{i=0}^{\infty} h(i)R_{sz}(i) \\ &= 11.2 - \sum_{i=0}^{\infty} 0.4667 \times 0.4^i \times 11.2 \times 0.75^i \\ &= 11.2 - 5.227 \sum_{i=0}^{\infty} 0.4^i \times 0.75^i \\ &= 11.2 - 5.227 \times \frac{1}{1 - 0.4 \times 0.75} \approx 3.733\end{aligned}$$