EE60039 – Autumn Semester 2023-24

Exercise 5: Probability and Random Processes for Signals and Systems

Prof. Ashish Ranjan Hota Department of Electrical Engineering, IIT Kharagpur

# Q 5.1: Estimation Task

Let X be a random variable with Exponential distribution with parameter  $\Lambda$ , and  $\Lambda$  itself is a discrete random variable with  $\mathbb{P}(\Lambda = 1) = \mathbb{P}(\Lambda = 2) = 0.5$ . Find  $\mathbb{E}[X]$ . Suppose you observe that X = 1. Find the maximum likelihood and maximum a-posteriori estimates of  $\Lambda$ .

Recall: the density of an exponential random variable X with parameter  $\Lambda$  is  $f_X(x) = \Lambda e^{-\Lambda x}, x \ge 0$ .

# Q 5.2: ML Estimation Task

Suppose  $\hat{x}_1, \hat{x}_2, \ldots, \hat{x}_N$  be N i.i.d. realizations of a random variable with density given by

$$f(x,\theta) = \begin{cases} \theta x^{\theta-1}, & x \in (0,1), \\ 0, & \text{otherwise.} \end{cases}$$

Find the maximum likelihood estimator of  $\theta$ .

# Q 5.3: Simple Estimation Task

Let X and Y be a random variables with joint pdf:

$$f_{X,Y}(x,y) = \begin{cases} x+y, & 0 \le x \le 1, 0 \le y \le 1, \\ 0, & \text{otherwise.} \end{cases}$$

- 1. Find the MAP estimator of X given Y = y.
- 2. Find the ML estimator of X given Y = y.
- 3. Find the MMSE estimator of X given Y = y.
- 4. Find the LMSE estimator of X given Y = y.

# Q 5.4: Another Simple Estimation Task

Let X be a random variable (r.v.) that denotes the temperature of a room. X is a discrete r.v. that can takes values in the finite set  $\{30, 32, 34\}$  where the value corresponds to the temperature in degree Celsius. Let Y be a random variable that indicates the comfort level of an occupant. Y = 1 is the occupant is comfortable, Y = -1 otherwise. The following is known about the conditional probability of Y for different values of X.

$$\begin{split} p_{Y|X}(Y=1|X=30) &= 0.8, \quad p_{Y|X}(Y=-1|X=30) = 1-0.8 = 0.2, \\ p_{Y|X}(Y=1|X=32) &= 0.6, \quad p_{Y|X}(Y=-1|X=32) = 1-0.6 = 0.4, \\ p_{Y|X}(Y=1|X=34) = 0.2, \quad p_{Y|X}(Y=-1|X=34) = 1-0.8 = 0.8. \end{split}$$

1. Suppose an occupant feels comfortable (Y = 1). Then, what the maximum likelihood estimate of the room temperature X?

Suppose X has a prior distribution (p.m.f.) given by  $p_X(X = 30) = p_X(X = 32) = 0.5$ , and  $p_X(X = 34) = 0$ . Then,

- 2. What is the posterior distribution of X if Y = 1?
- 3. What is the MAP estimate of X if Y = 1?
- 4. What is the MMSE estimate of X if Y = 1?
- 5. Repeat the above questions when Y = -1.

# Q 5.5: Yet Another Simple Estimation Task

Consider a variant of the above problem where the level of comfort Y is a continuous r.v. with conditional distributions are given by

$$\begin{split} f_{Y|X}(y|X=30) &\sim U[-1,0], \\ f_{Y|X}(y|X=32) &\sim U[-0.5,0.5], \\ f_{Y|X}(y|X=34) &\sim U[0,1], \end{split}$$

where U[a, b] denotes the uniform distribution between a and b.

1. Suppose we observe that Y = 0.2. Then, what the maximum likelihood estimate of the room temperature X?

Suppose X has a prior distribution (p.m.f.) given by  $p_X(X = 30) = p_X(X = 32) = 0.5$ , and  $p_X(X = 34) = 0$  as before. Then,

- 2. What is the posterior distribution of X if Y = 0.2?
- 3. What is the MAP estimate of X if Y = 0.2?
- 4. What is the MMSE estimate of X if Y = 0.2?
- 5. Repeat the above questions when the prior distribution of X is given by  $p_X(X = 30) = p_X(X = 32) = 0.4$ , and  $p_X(X = 34) = 0.2$ .

### **Q 5.6:** Mean Square Estimation

Let  $Y_i = X + V_i$ , where X is a random variable we want to estimate,  $Y_i$  is the *i*-th observation, and  $V_i$  is a random variable which acts as measurement noise. Let X and  $V_i$ 's be zero-mean and independent random variables with variance of X being  $\sigma_X^2 = 1$  and variance of  $V_i$  be  $\sigma_V^2 = 1$ .

- (a) Suppose we make two noisy observations  $Y_1$  and  $Y_2$ . Find the linear mean-square estimator for X in terms of  $Y_1$  and  $Y_2$ .
- (b) (**Bonus**) Suppose we make N noisy measurements  $Y_1, Y_2, \ldots, Y_N$ . Find the linear meansquare estimator for X in terms of  $Y_1, Y_2, \ldots, Y_N$ . Use the fact that  $(I + aa^T)^{-1} = I - \frac{1}{N+1}aa^T$ , where I is the identity matrix of dimension N and  $a = \begin{bmatrix} 1 & 1 & \ldots & 1 \end{bmatrix}^T$  is the column vector of length N with all entires 1.

# Q 5.7: Linear Mean Square Estimation

Assume that a zero-mean random variable  $X = X_c + Z$  where Z is uncorrelated with the observation Y. Show that the LMSE of X given Y equals the LMSE of  $X_c$  given Y. In other words,  $\widehat{\mathbb{E}}(X|Y) = \widehat{\mathbb{E}}(X_c|Y)$ . This shows that LMSE does not give us any information about the component Z that is uncorrelated with the observation Y.

#### Q 5.8: Linear Mean Square Estimation

Let Y = X + V, where X and V are independent zero-mean Gaussian random variables with variances  $\sigma_X^2$  and  $\sigma_V^2$ , respectively. Show that the LMSE estimator of  $X^2$  given Y and  $Y^2$  is

$$\hat{X}^{2} = \sigma_{X}^{2} + \frac{\sigma_{X}^{4}}{\sigma_{X}^{4} + 2\sigma_{X}^{2}\sigma_{V}^{2} + \sigma_{V}^{4}}(Y^{2} - \sigma_{X}^{2} - \sigma_{V}^{2}).$$

Hint: Use the fact that if Z is a zero-mean Gaussian r.v. with variance  $\sigma_Z^2$ , then  $\mathbb{E}[Z^3] = 0$  and  $\mathbb{E}[Z^4] = 3\sigma_Z^4$ .

#### Q 5.9: Linear Mean Square Estimation

Let  $Y_1 = (X + W_1)^2$ ,  $Y_2 = W_2 X$ . Find the LMSE estimator of X in terms of  $Y_1$  and  $Y_2$ (treat  $Y = [Y_1 \quad Y_2]^T$  as a random vector and use the appropriate formula for  $\widehat{\mathbb{E}}[X|Y]$ ). Use the following information:  $\mathbb{E}[W_1] = 0$ ,  $\mathbb{E}[W_1^2] = 1$ ,  $\mathbb{E}[W_2] = 2$ ,  $\mathbb{E}[W_2^2] = 1$ ,  $\mathbb{E}[X] = 1$ ,  $\mathbb{E}[X^2] = 2$ ,  $\mathbb{E}[X^3] = 3$ ,  $\mathbb{E}[X^4] = 4$ .

#### **Q 5.10:** Linear Mean Square Estimation

Let  $y_k = v_k + v_{k-1}, k \ge 0$ , where  $v_j, j \ge -1$  is a zero-mean stationary white noise scalar process with unit variance. Show that

$$\hat{y}_{k+1|k} = \frac{k+1}{k+2}(y_k - \hat{y}_{k|k-1}),$$

where  $\hat{y}_{k|m} = \widehat{\mathbb{E}}[y_k|Y_M]$  is the LMSE estimator of  $y_k$  given  $Y_M = (y_1, y_2, \dots, y_m)$ .

### **Q 5.11:** Prediction and Interpolation

Consider a series of observations  $y(0), y(1), \ldots$  where

 $y(k) = a_0 w(k) + a_1 w(k-1) + \ldots + a_N w(k-N),$ 

where  $w(-N), w(-N+1), \ldots$  are independent with  $\mathbb{E}[w(l)] = 0$  and  $\mathbb{E}[w^2(l)] = \sigma_w$ . The coefficients  $a_0, \ldots, a_N$  are known. Solve the following problems.

- 1. Prediction: Find the LMSE predictor of y(k+1) based on y(k) (based only on y(k), not y(k-1), etc).
- 2. Interpolation: Find the LMSE predictor of y(k) based (only) on y(k-1) and y(k+1) for k > 1.

#### Q 5.12: Mean Square Estimation

Let s[n] be a discrete-time zero-mean wide sense stationary (WSS) signal with auto-covariance function given by

$$C_{ss}[m] = \begin{cases} 15, & \text{for } m = 0, \\ -4, & \text{for } m = 1, -1 \\ 5, & \text{for } m = 2, -2, \\ 0, & \text{for } m \ge 3. \end{cases}$$

Suppose we observe this signal with a time-delay and with some measurement noise w[n]. w[n] is a zero-mean WSS signal uncorrelated with s[n] and  $C_{ww}[m] = 3$  if m = 0 and  $C_{ww}[m] = 0$  for  $m \neq 0$ . The received signal is given by

$$r[n] = s[n-1] + w[n].$$

We would like to estimate  $\hat{s}[n]$  that minimizes the mean square error (MSE)  $\mathbb{E}\left[(s[n] - \hat{s}[n])^2\right]$ .

- 1. Find the covariance functions  $C_{rs}[m]$  and  $C_{rr}[m]$  (in terms of  $C_{ss}[m]$  and  $C_{ww}[m]$ ).
- 2. If the estimator is to be of the form  $\hat{s}[n] = ar[n]$ , then find the value of a that minimizes the MSE.
- 3. If the estimator is to be of the form  $\hat{s}[n] = a_0 r[n] + a_1 r[n-1]$ , then find the values of  $a_0$  and  $a_1$  that minimizes the MSE.
- 4. If the estimator is to be of the form  $\hat{s}[n] = \frac{1}{2}r[n-n_0]$  with the integer  $n_0 \ge 0$ , then find the value of  $n_0$  that minimizes the MSE.

### Q 5.13: Past End-Sem Exam Question

X is a three-dimensional random vector with E[X] = 0 and autocorrelation matrix  $R_X$  with elements  $r_{ij} = (-0.80)^{|i-j|}$ . Use  $X_1$  and  $X_2$  to form a linear estimate of  $X_3 : \hat{X}_3 = a_1 X_1 + a_2 X_2$ , i.e., determine  $a_1$  and  $a_2$  that minimizes mean-square error.