

ELEC3810 Homework 2

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Q1

Question 1 (15 points)

The input X to a communication channel is +1 or -1 with probability p and $1 - p$, respectively. The received signal Y is the sum of X and noise N which has a Gaussian distribution with zero mean and variance $\sigma^2 = 0.16$.

- (a) Find the joint probability $P[X = j, Y \leq y]$.
- (b) Find the marginal pmf of X and the marginal pdf of Y .
- (c) Design a MAP decoder and write the decoding rule.

$$Y = X + N$$

(a)

When $j = 1$:

$$\begin{aligned} P[X = j, Y \leq y] &= P[X = 1, 1 + N \leq y] = P[X = 1, N \leq y - 1] \\ &= p * \frac{1}{0.5 * \sqrt{2\pi}} * \int_{-\infty}^{y-1} e^{-\frac{x^2}{2 * 0.25}} dx \\ &= p * \frac{1}{0.5 * \sqrt{2\pi}} * \int_{-\infty}^{y-1} e^{-2x^2} dx \end{aligned}$$

When $j = -1$:

$$\begin{aligned} P[X = j, Y \leq y] &= P[X = -1, -1 + N \leq y] = P[X = -1, N \leq y + 1] \\ &= (1 - p) * \frac{1}{0.5 * \sqrt{2\pi}} * \int_{-\infty}^{y+1} e^{-\frac{x^2}{2 * 0.25}} dx \\ &= (1 - p) * \frac{1}{0.5 * \sqrt{2\pi}} * \int_{-\infty}^{y+1} e^{-2x^2} dx \end{aligned}$$

(b)

Marginal pmf of X:

$$p_x(1) = p, p_x(-1) = 1 - p$$

Marginal cdf of Y:

$$\begin{aligned} F_y(y) &= P[Y \leq y | x = -1]P[x = -1] + P[Y \leq y | x = 1]P[x = 1] \\ &= P[N - 1 \leq y | x = -1] * (1 - p) + P[N + 1 \leq y | x = 1] * p \\ &= (1 - p) * \frac{1}{0.5 * \sqrt{2\pi}} * \int_{-\infty}^y e^{-2(y+1)^2} dy + (p) * \frac{1}{0.5 * \sqrt{2\pi}} * \int_{-\infty}^y e^{-2(y-1)^2} dy \end{aligned}$$

Marginal pdf of Y:

$$f_y(y) = \frac{d}{dy} F_y(y)$$

$$= \frac{(1-p)}{0.5\sqrt{2\pi}} * e^{-2(y+1)^2} dy + \frac{p}{0.5\sqrt{2\pi}} * e^{-2(y-1)^2} dy$$

(c)

MAP decoder rule:

$$\hat{X} = +1 \text{ if } p_{X|Y}(X = +1|Y) > p_{X|Y}(X = -1|Y)$$

$$\hat{X} = -1 \text{ otherwise}$$

Q2

Question 2 (20 points)

Let X and Y be discrete random variables with joint pmf's:

X/Y	-1	0	1
-1	1/6	1/6	0
0	0	0	1/3
1	1/6	1/6	0

- Find the minimum mean square error linear estimator for Y given X.
- Find the minimum mean square error estimator for Y given X.
- Find the MAP and ML estimators for Y given X.
- Compare the mean square error of the estimators in parts a, b and c.

$$E[X] = 0, E[Y] = 0, E[XY] = 0, E[X^2] = 2/3, E[Y^2] = 2/3$$

$$Var[X] = E[X^2] - E[X]^2 = 2/3$$

$$Var[Y] = 2/3$$

$$Cov(X, Y) = E[XY] - E[X]E[Y] = 0$$

(a)

$$\hat{Y}_L = \frac{Cov(X, Y)}{Var(X)}(X - E[X]) + E[Y]$$

$$= 0$$

(b)

$$\hat{Y}(-1) = E[Y|X = -1] = -1 * 1/2 + 0 * 1/2 + 0$$

$$= -1/2$$

$$\hat{Y}(0) = E[Y|X = 0] = 0 + 0 + 1 * 1$$

$$= 1$$

$$\hat{Y}(1) = E[Y|X = 1] = -1 * 1/2 + 0 * 1/2 + 0$$

$$= -1/2$$

(c)

MAP:

value of Y that maximizes: $P_Y(Y|X) = \frac{P(X|Y)P(Y)}{P(X)}$

$$\hat{Y}(-1) = 0 \text{ or } -1$$

$$\hat{Y}(-1) = 1$$

$$\hat{Y}(-1) = 0 \text{ or } -1$$

ML:

value of Y that maximizes: $P_X(X|Y) = \frac{P(Y|X)P(X)}{P(Y)}$

$$\hat{Y}(-1) = 0 \text{ or } -1$$

$$\hat{Y}(-1) = 1$$

$$\hat{Y}(-1) = 0 \text{ or } -1$$

(d)

Mean Square error of (a) = 1/3

Mean Square error of (b) = 1/6

Mean Square error of (c) (ML and MAP same) = 1/3

Q3

Question 3 (10 points)

Let U_1, U_2, U_3 be independent random variables with zero mean and variance 1. Find the linear MMSE estimator of S in terms of Z_1 and Z_2 , and the corresponding MSE.

$$\begin{bmatrix} S \\ Z_1 \\ Z_2 \end{bmatrix} = \begin{bmatrix} 1 & 1 & 1 \\ 1 & 2 & 0 \\ 0 & 1 & 2 \end{bmatrix} \begin{bmatrix} U_1 \\ U_2 \\ U_3 \end{bmatrix}$$

Q4

Question 4 (10 points)

Let X be uniformly distributed in the interval (-1,1) and let $Y = X^3$. Find the best linear estimator for Y in terms of X. Compare its performance to the best estimator.

$f(x) = 1/2$ in interval (-1,1), 0 otherwise

$$E[X] = \int_{-1}^1 x * f(x) dx = \int_{-1}^1 \frac{x}{2} dx = 0$$

$$E[Y] = E[X^3] = \int_{-1}^1 \frac{x^3}{2} dx = 0$$

$$E[XY] = E[X * X^3] = \int_{-1}^1 \frac{x^4}{2} dx = 0$$

$$\text{Cov}(X, Y) = E[XY] - E[X]E[Y] = 0$$

The best linear estimator for Y is:

$$\hat{Y} = \frac{\text{Cov}(X, Y)}{\text{Var}(X)} + E[Y]$$

$$= E[Y] = 0$$

the mean square error is $\text{Var}[Y]$

$$\text{Var}[Y] = E[Y^2] - E[Y]^2$$

$$E[Y^2] = E[X^6] = \int_{-1}^1 \frac{x^6}{2} dx = 1/7$$

$\text{Var}[Y] = 1/7$ which is the mean square error for the best linear estimator

The best estimator for Y is:

$$E[Y|X = x] = E[X^3|X = x] = x^3$$

with mean square error of 0, meaning that the best estimator performs better than best linear estimator.

Q5

Question 5 (15 points)

Suppose $X \sim U[1, 2]$, and given $X = x$, $Y \sim \text{Exp}(\lambda)$ with parameter $\lambda = \frac{1}{x}$.

- Find the linear MMSE estimate of X given Y.
- Find the MSE of this estimator.
- Check that $E[\tilde{X}Y] = 0$. ($\tilde{X} = X - \hat{X}$)

$$E[X] = \int_1^2 x * 1 dx = 3/2$$

$$E[Y] = E[E[Y|X]] = E[X] = 3/2$$

$$E[Y^2] = E[E[Y^2|X]] = E[2X^2] = \int_1^2 2x^2 * 1 dx = 14/3$$

$$\text{Var}[Y] = E[Y^2] - E[Y]^2 = 29/12$$

$$E[XY] = \int_1^2 x^2 * 1 dx = 7/3$$

$$\text{Cov}(X, Y) = E[XY] - E[X]E[Y] = 1/12$$

(a)

$$\begin{aligned}\hat{X}_L &= \frac{\text{Cov}(X,Y)}{\text{Var}(Y)}(Y - E[Y]) + E[X]. \\ &= \frac{1}{29} \left(Y - \frac{3}{2} \right) + \frac{3}{2} \\ &= \frac{Y}{29} + \frac{42}{29}\end{aligned}$$

(b)

$$\begin{aligned}\rho^2 &= \frac{\text{Cov}^2(X,Y)}{\text{Var}(X)\text{Var}(Y)} = \frac{1}{29} \\ MSE &= (1 - \rho^2)\text{Var}(X) \\ \$ &= (1 - 1/29) \cdot 1/12 \$ \\ &= \frac{7}{87}\end{aligned}$$

(c)

$$\begin{aligned}\tilde{X} &= X - \hat{X}_L \\ &= X - \frac{Y}{29} - \frac{42}{29} \\ E[\tilde{X}Y] &= E \left[\left(X - \frac{Y}{29} - \frac{42}{29} \right) Y \right] \\ &= E[XY] - \frac{EY^2}{29} - \frac{42}{29} EY \\ &= 0\end{aligned}$$

Q6

Question 6 (10 points)

Let X be an unobserved random variable with $E(X) = 0, \text{Var}(X) = 4$. Assume that we have observed Y_1 and Y_2 given by

$$Y_1 = X + W_1,$$

$$Y_2 = X + W_2,$$

where $E(W_1) = E(W_2) = 0, \text{Var}(W_1) = 1$, and $\text{Var}(W_2) = 4$. Assume that W_1, W_2 , and X are independent random variables. Find the linear MMSE estimator of X , given Y_1 and Y_2

$$E[\tilde{X}] = -aE[Y_1] - bE[Y_2] - c = -a \cdot 0 - b \cdot 0 - c = -c.$$

Since $E[X] = 0, c = 0$

$$\begin{aligned}\text{Cov}(\hat{X}_L, Y_1) &= \text{Cov}(aY_1 + bY_2, Y_1) \\ &= a\text{Cov}(X + W_1, X + W_1) + b\text{Cov}(X + W_1, X + W_2) \\ &= a(\text{Var}(X) + \text{Var}(W_1)) + b\text{Var}(X) \\ &= 5a + 4b\end{aligned}$$

$$\begin{aligned}\text{Cov}(\hat{X}_L, Y_2) &= \text{Cov}(aY_1 + bY_2, Y_2) \\ &= a\text{Var}(X) + b(\text{Var}(X) + \text{Var}(W_2)) \\ &= 4a + 8b\end{aligned}$$

Since $\text{Cov}(X, Y_1) = \text{Cov}(X, Y_2) = \text{Var}(X) = 4$:

$5a + 4b = 4$ and $4a + 8b = 4$, solving both equations give $a = 2/3$ and $b = 1/6$

The linear MMSE of X :

$$\begin{aligned}\hat{X}_L &= aY_1 + bY_2 + c \\ &= \hat{X}_L = \frac{2}{3}Y_1 + \frac{1}{6}Y_2\end{aligned}$$

Q7

Question 7 (20 points)

Let X, Y, Z have joint pdf

$$f_{X,Y,Z}(x, y, z) = k(x + y + z) \text{ for } 0 \leq x \leq 1, 0 \leq y \leq 1, 0 \leq z \leq 1$$

- Find k
- Find the minimum mean square error linear estimator for Y given X and Z .
(Hint: given two observations, you can either calculate the derivative of MSE in terms of every coefficient or list the observations as a vector form.)
- Find the minimum mean square estimator for Y given X and Z .
- Find the MAP and ML estimators for Y given X and Z .

(a)

$$1 = \int_0^1 \int_0^1 \int_0^1 k(x + y + z) dx dy dz$$

$$1 = 1.5k$$

$$k = 2/3$$

(b)

$$f(x) = \frac{2}{3}(x + 1), f(x, y) = \frac{2}{3}(x + y + \frac{1}{2})$$

$$E[X] = E[Y] = E[Z] = \int_0^1 f(x) * x dx = \int_0^1 \frac{2}{3}x(x + 1)dx = \frac{5}{9}$$

$$E[XY] = \int_0^1 \frac{2}{3}(x + y + \frac{1}{2}) * x * y dx dy = \frac{11}{36}$$

$$\text{Var}[X] = \text{Var}[Y] = \text{Var}[Z] = E[X^2] - E[X]^2 = \frac{13}{162}$$

$$\text{Cov}(X, Y) = \text{Cov}(Y, Z) = \text{Cov}(X, Z) = E[XY] - E[X]E[Y] = \frac{11}{36} - \frac{5}{9} * \frac{5}{9} = \frac{-1}{324}$$

Linear Estimator:

$$a = R_{XZ}^{-1} * E[XZ]$$

$$\begin{aligned}
(a_1, a_2)^T &= \begin{pmatrix} Var(x) & Cov(X, Z) \\ Cov(X, Z) & Var(Z) \end{pmatrix}^{-1} \begin{pmatrix} Cov(Z, X) \\ Cov(Y, Z) \end{pmatrix} \\
&= \begin{pmatrix} 13/162 & -1/324 \\ -1/324 & 13/162 \end{pmatrix}^{-1} \begin{pmatrix} -1/324 \\ -1/324 \end{pmatrix} \\
&= \begin{pmatrix} -27/705 \\ -27/705 \end{pmatrix}
\end{aligned}$$

$$\begin{aligned}
\hat{Y} &= (a_1, a_2) \begin{pmatrix} X - E[X] \\ Z - E[Z] \end{pmatrix} + E[Y] \\
&= (-27/705, -27/705) \begin{pmatrix} X - 5/9 \\ Z - 5/9 \end{pmatrix} + 5/9 \\
&= \frac{-27}{705}(x + z) + \frac{217}{423}
\end{aligned}$$

Mean Square Error:

$$\begin{aligned}
&= Var[X] - a^T \begin{pmatrix} Cov(X, Y) \\ Cov(Y, Z) \end{pmatrix} \\
&= \frac{13}{162} - \frac{1}{1410} \\
&= \frac{1514}{19035}
\end{aligned}$$

(c)

$$f(y|x, z) = \frac{(x+y+z)}{(x+y+0.5)}, \quad 0 < y < 1$$

$$E[Y|X, Z] = \int_0^1 \frac{(x+y+z)*y}{x+y+0.5} dy = \frac{0.5(x+z)+\frac{1}{3}}{x+y+0.5}$$

MMSE Estimator for Y given X and Z:

$$\hat{Y} = \frac{0.5(x+z)+\frac{1}{3}}{x+y+0.5}$$

(d)

ML Estimator:

$$\hat{Y} = 1 \text{ if } x + z < 1$$

$$\hat{Y} = 0 \text{ if } x + z > 1$$