

# EE7401 Probability and Random Processes

## RA 3 Solutions

**Note that there can be many possible solutions. This is just one possible approach.**

**1.**

(a) Let  $N = \sum_{i=1}^{1000} X_i$ . We want to find  $\mathbb{E}[N]$ . We have

$$\begin{aligned}\mathbb{E}[N] &= \sum_{i=1}^{1000} \mathbb{E}[X_i] \\ &= \sum_{i=1}^{1000} 10^{-8} \\ &= 10^{-5}.\end{aligned}$$

(b) By the Markov inequality,

$$\begin{aligned}\mathbb{P}(N \geq 10) &\leq \frac{1}{10} \mathbb{E}[N] \\ &= 10^{-6}.\end{aligned}$$

**2.**

(a) We have

$$\begin{aligned}p_{Y|X}(y | x) &= p_{Z|X}(y - x | x) \\ &= \begin{cases} \frac{1}{4} & -2 \leq y - x \leq 2, \\ 0 & \text{otherwise} \end{cases}\end{aligned}\tag{1}$$

and

$$\begin{aligned}p_Y(y) &= \frac{1}{2} p_{Y|X}(y | 1) + \frac{1}{2} p_{Y|X}(y | -1) \\ &= \begin{cases} \frac{1}{4} & -1 \leq y \leq 1, \\ \frac{1}{8} & -3 \leq y < -1, \ 1 < y \leq 3, \\ 0 & \text{otherwise.} \end{cases}\end{aligned}$$

Using

$$p_{X|Y}(x | y) = \frac{p_{Y|X}(y | x)p_X(x)}{p_Y(y)},$$

we have

$$\begin{aligned} p_{X|Y}(1 | y) &= \begin{cases} 0 & -3 \leq y < -1, \\ \frac{1}{2} & -1 \leq y \leq 1, \\ 1 & 1 < y \leq 3, \end{cases} \\ p_{X|Y}(-1 | y) &= \begin{cases} 1 & -3 \leq y < -1, \\ \frac{1}{2} & -1 \leq y \leq 1, \\ 0 & 1 < y \leq 3. \end{cases} \end{aligned} \tag{2}$$

The MMSE estimate of  $X$  given  $Y$  is

$$\begin{aligned} \mathbb{E}[X | Y = y] &= 1 \cdot p_{X|Y}(1 | y) + (-1) \cdot p_{X|Y}(-1 | y) \\ &= \begin{cases} -1 & -3 \leq y < -1, \\ 0 & -1 \leq y \leq 1, \\ 1 & 1 < y \leq 3, \end{cases} \end{aligned}$$

and its MSE is

$$\begin{aligned} \mathbb{E}[\text{var}(X | Y)] &= \mathbb{E}[\mathbb{E}[X^2 | Y] - \mathbb{E}[X | Y]^2] \\ &= 1 - \mathbb{E}[\mathbb{E}[X | Y]^2] \\ &= 1 - \int_{-3}^{-1} \frac{1}{8} dy - \int_1^3 \frac{1}{8} dy \\ &= \frac{1}{2}. \end{aligned}$$

- (b) Let the MAP decoder be  $D(y)$ . It takes value 1 if  $p_{X|Y}(1 | y) \geq p_{X|Y}(-1 | y)$  and  $-1$  otherwise. From (2), we have

$$D(y) = \begin{cases} 1 & -1 \leq y \leq 3, \\ -1 & -3 \leq y < -1. \end{cases}$$

The error probability is

$$\begin{aligned} \mathbb{P}(D(Y) \neq X) &= \mathbb{P}(-3 \leq Y < -1 | X = 1)p_X(1) + \mathbb{P}(-1 \leq Y \leq 3 | X = -1)p_X(-1) \\ &= \frac{1}{4}, \end{aligned}$$

where we have made use of (1). Considering the MAP decoder as an estimator, its MSE is

$$\begin{aligned}
& \mathbb{E}[(D(Y) - X)^2] \\
&= \mathbb{E}[(D(Y) - X)^2 \mid D(Y) = X] \mathbb{P}(D(Y) = X) \\
&\quad + \mathbb{E}[(D(Y) - X)^2 \mid D(Y) \neq X] \mathbb{P}(D(Y) \neq X) \\
&= 4 \cdot \frac{1}{4} \\
&= 1.
\end{aligned}$$

As expected, the MSE of the MAP decoder is greater than the MMSE found in (a).

**Remark 1.** Note that when defining the MAP decoder  $D(y)$ , it does not matter which value we set  $D(y)$  to be in the case where  $p_{X|Y}(1 \mid y) = p_{X|Y}(-1 \mid y)$ . To see this, recall that we want to minimize the probability of error, which is equivalent to maximizing

$$\mathbb{P}(D(Y) = X) = \int_{-\infty}^{\infty} p_{X|Y}(D(y) \mid y) p_Y(y) \, dy. \quad (3)$$

For each  $y$ , we choose  $D(y)$  such that  $p_{X|Y}(D(y) \mid y)$  is maximized. From (3), the maximum probability is

$$\mathbb{P}(D(Y) = X) = \int_A p_{X|Y}(1 \mid y) p_Y(y) \, dy + \int_B p_{X|Y}(-1 \mid y) p_Y(y) \, dy + \int_C p_{X|Y}(D(y) \mid y) p_Y(y) \, dy,$$

where

- $A = \{y : p_{X|Y}(1 \mid y) > p_{X|Y}(-1 \mid y)\}$ ,
- $B = \{y : p_{X|Y}(1 \mid y) < p_{X|Y}(-1 \mid y)\}$  and
- $C = \{y : p_{X|Y}(1 \mid y) = p_{X|Y}(-1 \mid y)\}$ . The error probability does not change regardless of how we choose  $D(y)$  on  $C$ .

This is true in general for any Bayesian inference problem: Find a function  $D(\cdot)$  of an observed  $Y$  to infer  $X$  based on minimizing  $\mathbb{E}[\ell(D(Y), X)]$ , where  $\ell(\cdot, \cdot)$  is a non-negative risk function. From

$$\mathbb{E}[\ell(D(Y), X)] = \mathbb{E}[\mathbb{E}[\ell(D(Y), X) \mid Y]],$$

this is equivalent to minimizing  $\mathbb{E}[\ell(D(y), X) \mid Y = y]$  for each  $y$ . If  $\arg \min_{D(\cdot)} \mathbb{E}[\ell(D(y), X) \mid Y = y]$  is non-unique, we choose any of the minimizers as they all give the same  $\mathbb{E}[\ell(D(y), X) \mid Y = y]$  value.