

EE7401 Probability and Random Processes

Practice

1.

Two discrete random variables X and Y have possible values of 4, 5 and 6. Table 1 specifies one of their probability-mass functions (PMFs): $p_{X,Y}(x,y)$, $p_{Y|X}(y|x)$ or $p_{X|Y}(x|y)$.

Table 1

$y \backslash x$	4	5	6
4	a	0.1	0.5
5	0.2	b	0.4
6	0.3	0.2	c

$$\sum_{x,y} P(x,y) = 1$$

$$a, b, c \geq 0$$

Give the conditions (values of a , b , and c), under which the Table 1 specifies PMFs, $p_{Y|X}(y|x)$, $p_{X|Y}(x|y)$ and $p_{X,Y}(x,y)$, respectively.

$$p_{Y|X}(y|x) : a=0.5, b=0.7, c=0.1$$

$$p_{X|Y}(x|y) : a=0.4, b=0.4, c=0.5$$

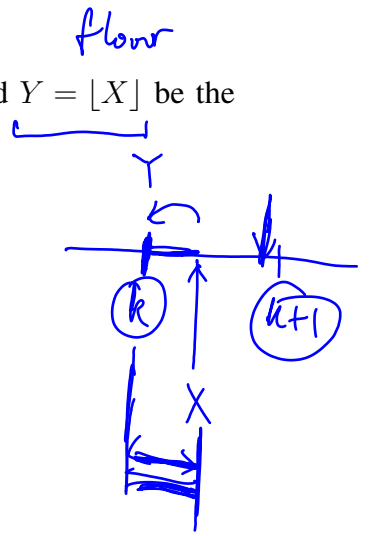
$$p_{X,Y}(x,y) : \text{not possible.}$$

$$f_X(x) = \begin{cases} \lambda e^{-\lambda x}, & x \geq 0 \\ 0, & \text{o.w.} \end{cases}$$

2. Let $X \sim \text{Exp}(\lambda)$ be an exponential random variable with parameter λ and $Y = \lfloor X \rfloor$ be the integer part of X , i.e., $Y = k$ for $k \leq X < k+1$, $k = 0, 1, \dots$

(i) Find the pmf of Y .

(ii) Let $Z = X - Y$ be the quantization error. Find the pdf of Z .



$$\begin{aligned} \text{(i)} \quad P_Y(k) &= \mathbb{P}(Y=k), \quad k=0, 1, 2, \dots \\ &= \mathbb{P}(k \leq X < k+1) \end{aligned}$$

$$= \int_k^{k+1} f_X(x) dx$$

$$= \int_k^{k+1} \lambda e^{-\lambda x} dx = -e^{-\lambda x} \Big|_k^{k+1}$$

total prob. law $\Omega = \bigcup_{k \geq 0} A_k = e^{-\lambda k} - e^{-\lambda(k+1)}$

$$\mathbb{P}(B) = \sum_{k=0}^{\infty} \mathbb{P}(B \cap A_k) = e^{-\lambda k} (1 - e^{-\lambda})$$

(ii) cdf \rightarrow pdf $f_Z(z) = \frac{d}{dz} \left(\frac{1 - e^{-\lambda z}}{1 - e^{-\lambda}} \right) = \begin{cases} \frac{\lambda e^{-\lambda z}}{1 - e^{-\lambda}}, & 0 \leq z < 1 \\ 0, & \text{o.w.} \end{cases}$

$$\mathbb{P}(Z \leq z) = \mathbb{P}(X - Y \leq z)$$

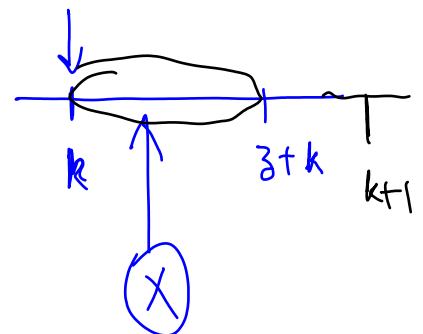
$$= \sum_{k=0}^{\infty} \mathbb{P}(X - Y \leq z, Y=k)$$

$$= \sum_{k=0}^{\infty} \mathbb{P}(X \leq z+k, Y=k)$$

$$= \sum_{k=0}^{\infty} \mathbb{P}(k \leq X \leq z+k)$$

$$\sum_{k=0}^{\infty} r^k = \frac{1}{1-r}$$

$$\int_k^{z+k} \lambda e^{-\lambda x} dx = e^{-\lambda k} (1 - e^{-\lambda z})$$



$$= \left(\sum_{k=0}^{\infty} e^{-\lambda k} \right) \underbrace{(1 - e^{-\lambda})}_{\text{MMSE}} = \frac{1}{1 - e^{-\lambda}} (1 - e^{-\lambda})$$

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3.

Two continuous random variables X and Y have joint probability density function (pdf)

$$p_{X,Y}(x,y) = \lambda y^2 e^{-xy}, \quad x \geq 0, 1 \leq y \leq 5,$$

where λ is a constant.

- (a) Express the MMSE estimation of X as a function of the random variable Y . If we observed that $Y=2$, what is the estimate of X that has the minimal mean square error?

- (b) For unknown value of X , find the estimate of Y that has the minimal mean square error. What is the mean square error of the estimate?

(a) $E[X|Y]$

$x \geq 0, 1 \leq y \leq 5$

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

$= \frac{\lambda y^2 e^{-xy}}{\lambda y} = y e^{-xy}$

$E[(Y - \hat{Y})^2]$

minimizer = $E[Y]$

$1 \leq y \leq 5$

$p_Y(y) = \int_{-\infty}^{\infty} p_{X,Y}(x,y) dx$

$= \int_0^{\infty} \lambda y^2 e^{-xy} dx$
 $= \lambda y^2 \left[-\frac{e^{-xy}}{y} \right]_0^{\infty}$

$= \lambda y$

$E[X|Y=y] = \int_0^{\infty} x p_{X|Y}(x|y) dx$

$= \int_0^{\infty} x y e^{-xy} dx$

$= y \left[\int_0^{\infty} x e^{-xy} dx \right]$

$= -\frac{1}{y^2} e^{-xy} \Big|_{x=0}^{x=\infty}$

integration by parts

$= y \left(\left[-\frac{x e^{-xy}}{y} \right]_{x=0}^{x=\infty} + \int_0^{\infty} \frac{e^{-xy}}{y} dx \right)$

$= y \cdot \frac{1}{y} = \frac{1}{y}$

$Y=2$

$\hat{X}_{\text{MMSE}} = E[X|Y=2]$

$= \frac{1}{2}$

$$1 \leq y \leq 5, \quad p_Y(y) = \lambda y \quad \int_1^5 \lambda y \, dy = 1 \Rightarrow \lambda \left. \frac{y^2}{2} \right|_1^5 = 1$$

$$\mathbb{E}[Y] = \int_1^5 y \cdot \lambda y \, dy = \lambda \left(\frac{y^3}{3} \right) \Big|_1^5 = \frac{1}{12} \cdot 0 = \frac{31}{9}$$

$$\begin{aligned} \text{MSE} = \text{var}(Y) &= \mathbb{E}[Y^2] - (\mathbb{E}[Y])^2 \\ &= \lambda \int_1^5 y^3 \, dy - \left(\frac{31}{9} \right)^2 \\ &\approx 1.136 \end{aligned}$$

$$\text{MSE} : \text{mean squared error} = \mathbb{E}[(\hat{X} - X)^2]$$

$$\text{MMSE} : \text{minimum mean squared error} = \text{minimum MSE}$$

$$\text{when } \hat{X} = \mathbb{E}[X|Y]$$

MMSE estimator

MMSE : minimum MSE estimate