

# problem set 3 (solutions)

①:  $X$  is uniform over  $\{-4, -3, \dots, 3, 4\}$ ,  $Y \triangleq \sin^2\left(\frac{\pi X}{4}\right)$

$$\Rightarrow \Pr\{X=i\} = \frac{1}{9} \text{ for } -4 \leq i \leq 4$$

$$EY = E\left\{\sin^2\left(\frac{\pi X}{4}\right)\right\} = \sum_{i=-4}^{+4} \sin^2\left(\frac{\pi i}{4}\right) \times \Pr\{X=i\} = \frac{1}{9} \sum_{i=-4}^{+4} \sin^2\left(\frac{\pi i}{4}\right)$$

$$= \frac{2}{9} \sum_{i=1}^4 \sin^2\left(\frac{\pi i}{4}\right) = \frac{2}{9} \left( \sin^2\left(\frac{\pi}{4}\right) + \sin^2\left(\frac{2\pi}{4}\right) + \sin^2\left(\frac{3\pi}{4}\right) + \sin^2\left(\frac{4\pi}{4}\right) \right)$$

$$= \frac{2}{9} \left( \frac{1}{2} + 1 + \frac{1}{2} + 0 \right) = \frac{4}{9}$$

$$EY^2 = E\left\{\sin^4\left(\frac{\pi X}{4}\right)\right\} = \sum_{i=-4}^{+4} \sin^4\left(\frac{\pi i}{4}\right) \times \Pr\{X=i\} = \frac{2}{9} \sum_{i=1}^4 \sin^4\left(\frac{\pi i}{4}\right)$$

$$= \frac{2}{9} \left( \sin^4\left(\frac{\pi}{4}\right) + \sin^4\left(\frac{2\pi}{4}\right) + \sin^4\left(\frac{3\pi}{4}\right) + \sin^4\left(\frac{4\pi}{4}\right) \right) =$$

$$= \frac{2}{9} \left( \frac{1}{4} + 1 + \frac{1}{4} \right) = \frac{2}{9} \times \frac{3}{2} = \frac{1}{3}$$

$$\Rightarrow \text{Var}(Y) = EY^2 - (EY)^2 = \frac{1}{3} - \left(\frac{4}{9}\right)^2 = \frac{11}{81}$$

②  $M$  is a geometric random variable  $\Pr(M=k) = p^k(1-p); k=0,1,2,\dots$

$$\begin{aligned} \Pr\{M \geq k+j / M \geq j\} &= \frac{\Pr\{M \geq k+j, M \geq j\}}{\Pr\{M \geq j\}} = \frac{\Pr\{M \geq k+j\}}{\Pr\{M \geq j\}} \\ &= \frac{\sum_{l=k+j}^{+\infty} p^l(1-p)}{\sum_{l=j}^{+\infty} p^l(1-p)} = \frac{(1-p) \times \frac{p^{k+j}}{1-p}}{(1-p) \times \frac{p^j}{1-p}} = p^k = \sum_{l=k}^{+\infty} p^l(1-p) = \Pr\{M \geq k\} \end{aligned}$$

③:  $f(x; \alpha, x_m) = \frac{\alpha x_m^\alpha}{x^{\alpha+1}}; x \geq x_m$  and zero elsewhere.

$$\left. \begin{matrix} \alpha=1 \\ x_m=1 \end{matrix} \right\} \Rightarrow f(x; 1, 1) = \frac{1}{x^2}; x \geq 1$$

$$\text{mean} = \int_1^{+\infty} x f(x; 1, 1) dx = \int_1^{+\infty} x \times \frac{1}{x^2} dx = \left[ \ln(x) \right]_1^{+\infty} = +\infty$$

$$(5): f_X(x) = \frac{\alpha}{\pi(x^2 + \alpha^2)}, \quad Y \triangleq \frac{1}{X} \Rightarrow g(x) = \frac{1}{x} \quad g'(x) = \frac{-1}{x^2}$$

$$y = \frac{1}{x} \Rightarrow \boxed{x = \frac{1}{y}} \quad f_Y(y) = \frac{f_X\left(\frac{1}{y}\right)}{\left|g'(x)\right|_{x=\frac{1}{y}}}$$

$$\Rightarrow f_Y(y) = \frac{\frac{\alpha}{\pi\left(\frac{1}{y^2} + \alpha^2\right)}}{\left|\frac{-1}{\left(\frac{1}{y}\right)^2}\right|} = \frac{\alpha}{\pi y^2 \left(\frac{1}{y^2} + \alpha^2\right)} = \frac{\alpha}{\pi(1 + \alpha^2 y^2)}$$

$$\Rightarrow f_Y(y) = \frac{\frac{1}{\alpha}}{\pi\left(y^2 + \left(\frac{1}{\alpha}\right)^2\right)} \rightsquigarrow \text{Cauchy random variable with parameter } \frac{1}{\alpha}$$

$$(6): U \text{ uniform over } [0, 1], \quad X \triangleq -\ln\left(1 - \frac{1}{U}\right)$$

$$0 < U < 1 \Rightarrow 1 - \frac{1}{U} < 0 \Rightarrow X = -\ln\left(\frac{1}{U} - 1\right)$$

$$F_X(x) = \Pr\{X \leq x\} = \Pr\left\{-\ln\left(\frac{1}{U} - 1\right) \leq x\right\} = \Pr\left\{U \leq \frac{1}{1 + e^{-x}}\right\}$$

$$= \frac{1}{1 + e^{-x}}$$

$$f_X(x) = \frac{d}{dx} F_X(x) = \frac{-(-1)e^{-x}}{(1 + e^{-x})^2} = \frac{e^{-x}}{(1 + e^{-x})^2}$$

$$\Rightarrow f_X(x) = \frac{e^x}{(1 + e^x)^2}$$



$$(7): f_X(x) = \frac{1}{\sqrt{2\pi}\sigma} e^{-\frac{(x-\mu)^2}{2\sigma^2}}, y \triangleq \begin{cases} 0; & x \leq 0 \\ x; & x > 0 \end{cases}$$

$$F_Y(y) = \Pr\{Y \leq y\}$$

we consider the following two cases:

$$\text{case 1 } (y < 0) \Rightarrow F_Y(y) = \Pr\{X \leq y\} = 0$$

$$\begin{aligned} \text{case 2 } (y \geq 0) &\Rightarrow F_Y(y) = \Pr\{X \leq y\} = \Pr\{Y=0\} + \Pr\{0 < Y \leq y\} \\ &= \Pr\{X \leq 0\} + \Pr\{0 < X \leq y\} = \Pr\{X \leq 0\} + \Pr\{X \leq y\} - \Pr\{X \leq 0\} \\ &= \Pr\{X \leq y\} = F_X(y) \end{aligned}$$

$$\Rightarrow F_Y(y) = \begin{cases} 0 & ; y < 0 \\ F_X(y) & ; y \geq 0 \end{cases}$$

$$\Rightarrow f_Y(y) = \frac{d}{dy} F_Y(y) = F_X(0) \delta(y) + \begin{cases} 0 & ; y < 0 \\ f_X(y) & ; y \geq 0 \end{cases}$$

⑧ Let  $X = \begin{pmatrix} x_1 \\ x_2 \\ \vdots \\ x_n \end{pmatrix}$  be a random vector

$$K_X = E\left\{(X - EX)(X - EX)^T\right\} \quad \text{— covariance matrix which is symmetric}$$

A symmetric matrix  $M$  with real entries is positive semi-definite if the real number  $Z^T M Z$  is nonnegative for every nonzero column vector  $Z$ .

$$\text{Let } Z = \begin{pmatrix} z_1 \\ z_2 \\ \vdots \\ z_n \end{pmatrix} \in \mathbb{R}^n$$

$$\Rightarrow Z^T K_X Z = Z^T E\left\{(X - EX)(X - EX)^T\right\} Z = E\left\{Z^T (X - EX) (Z^T (X - EX))^T\right\}$$

Defining the new random variable  $Y \triangleq Z^T (X - EX)$

we have:

$$Z^T K_X Z = E\{Y^2\} \geq 0$$

$\Rightarrow K_X$  is positive semi-definite

⑨:  $X$  is a  $n \times 1$  random vector.

$$\mu = EX \quad K_x = E\{(X - EX)(X - EX)^T\}$$

$$Y = AX + b$$

$$EY = AEX + b$$

$$\begin{aligned} K_y &= E\{(Y - EY)(Y - EY)^T\} = E\{(AX + b - AEX - b)(AX + b - AEX - b)^T\} \\ &= E\{A(X - EX)(X - EX)^T A^T\} = A K_x A^T \end{aligned}$$