

$$1. \quad p(x, y) = \frac{e^{-2}}{x!(y-x)!}$$

(a) Find the msf  $M(t_1, t_2)$  of  $(X, Y)'$

$$\begin{aligned} M_{XY}(t_1, t_2) &= E(e^{t_1 X + t_2 Y}) = \sum_{y=0}^{\infty} \sum_{x=0}^y e^{t_1 x + t_2 y} \frac{e^{-2}}{x!(y-x)!} \\ &= e^{-2} \sum_{y=0}^{\infty} \sum_{x=0}^y \frac{y! (e^{t_1})^x}{x!(y-x)!} \cdot \frac{e^{t_2 y}}{y!} = e^{-2} \sum_{y=0}^{\infty} (1 + e^{t_1})^y \frac{e^{t_2 y}}{y!} \\ &= e^{-2} \sum_{y=0}^{\infty} \frac{((1 + e^{t_1}) e^{t_2})^y}{y!} = e^{-2} e^{(1 + e^{t_1}) e^{t_2}} = e^{(1 + e^{t_1}) e^{t_2} - 2} \end{aligned}$$

(b) compute the means, the variance, the correlation coefficient of  $X$  and  $Y$

$$\begin{aligned} E(X) &= \frac{\partial}{\partial t_1} M_{XY}(t_1, t_2) \Big|_{t_1=0, t_2=0} = \frac{\partial}{\partial t_1} e^{(1 + e^{t_1}) e^{t_2} - 2} \Big|_{t_1=0, t_2=0} \\ &= (e^{t_1} + 1) (e^{(1 + e^{t_1}) e^{t_2} - 2}) = 1 \end{aligned}$$

$$\begin{aligned} E(Y) &= \frac{\partial}{\partial t_2} M_{XY}(t_1, t_2) \Big|_{t_1=0, t_2=0} = \frac{\partial}{\partial t_2} e^{(1 + e^{t_1}) e^{t_2} - 2} \Big|_{t_1=0, t_2=0} \\ &= (1 + e^{t_1}) e^{t_2} \cdot e^{(1 + e^{t_1}) e^{t_2} - 2} = 2 \end{aligned}$$

$$\begin{aligned} E(X^2) &= \frac{\partial^2}{\partial t_1^2} M_{XY}(t_1, t_2) \Big|_{(0,0)} = \frac{\partial^2}{\partial t_1^2} e^{(1 + e^{t_1}) e^{t_2} - 2} \Big|_{(0,0)} \\ &= \frac{\partial}{\partial t_1} (e^{t_1} + 1) \cdot e^{(1 + e^{t_1}) e^{t_2} - 2} \Big|_{(0,0)} \\ &= e^{t_1 + t_2} e^{(1 + e^{t_1}) e^{t_2} - 2} + e^{2t_1 + 2t_2} \cdot e^{(1 + e^{t_1}) e^{t_2} - 2} \Big|_{(0,0)} \\ &= 2 \end{aligned}$$

$$\begin{aligned} E(Y^2) &= \frac{\partial^2}{\partial t_2^2} M_{XY}(t_1, t_2) \Big|_{(0,0)} = \frac{\partial^2}{\partial t_2^2} e^{(1 + e^{t_1}) e^{t_2} - 2} \Big|_{(0,0)} \\ &= \frac{\partial}{\partial t_2} (1 + e^{t_1}) e^{t_2} \cdot e^{(1 + e^{t_1}) e^{t_2} - 2} \Big|_{(0,0)} \\ &= (1 + e^{t_1}) e^{t_2} e^{(1 + e^{t_1}) e^{t_2} - 2} + (1 + e^{t_1})^2 e^{2t_2} e^{(1 + e^{t_1}) e^{t_2} - 2} \Big|_{(0,0)} \\ &= 6 \end{aligned}$$

$$V(X) = E(X^2) - E(X)^2 = 1, \quad V(Y) = E(Y^2) - E(Y)^2 = 6 - 4 = 2$$

$$\begin{aligned} E(XY) &= \frac{\partial^2}{\partial t_1 \partial t_2} M_{XY}(t_1, t_2) \Big|_{(0,0)} = \frac{\partial^2}{\partial t_1 \partial t_2} e^{(1 + e^{t_1}) e^{t_2} - 2} \Big|_{(0,0)} \\ &= \frac{\partial}{\partial t_2} e^{t_1 + t_2} e^{(1 + e^{t_1}) e^{t_2} - 2} \Big|_{(0,0)} = e^{t_1 + t_2} e^{(1 + e^{t_1}) e^{t_2} - 2} \\ &\quad + e^{t_1} (1 + e^{t_1}) e^{2t_2} e^{(1 + e^{t_1}) e^{t_2} - 2} \Big|_{(0,0)} = 1 + 2 = 3 \end{aligned}$$

$$\rho = \frac{\text{COV}(X, Y)}{\sigma_X \sigma_Y} = \frac{(E(XY) - E(X)E(Y))}{\sigma_X \sigma_Y} = \frac{1}{\sqrt{2}} = 0.71$$

(c) Determine  $E(X|Y=4)$

$$E(X|Y=4) = \sum_{x=0}^4 \frac{1(x,4)}{P(4)}$$

$$P_Y(4) = \sum_{x=0}^{\infty} \frac{e^{-2}}{x!(4-x)!} = \frac{e^{-2}}{4!} \sum_{x=0}^{\infty} \frac{4!}{x!(4-x)!} = \frac{e^{-2} 2^4}{4!}$$

$$= \sum_{x=0}^4 \frac{\frac{e^{-2}}{x!(4-x)!}}{\frac{e^{-2} 2^4}{4!}} = \sum_{x=0}^4 \frac{x 4!}{x!(4-x)! 2^4} = \sum_{x=0}^4 \frac{4!}{(x-1)!(4-x)!} 2^{-4}$$

$$= \frac{4}{2^4} \sum_{x=0}^4 \frac{(4-1)!}{(x-1)!(4-x)!} = \frac{4}{2^4} \times 2^{4-1} = \frac{4}{2} \therefore \frac{4}{2}$$

Problem 2. Let  $Y = X_1 + X_2$ , if  $Y \sim \chi^2(t)$  and  $X_1 \sim \chi^2(t_1)$

then  $X_2 \sim \chi^2(t - t_1)$

$$M_Y(t) = \chi^2(t) = \prod_{i=1}^n M_{X_i}(t) = (1-2t)^{-\frac{t_1}{2}} \cdot (1-2t)^{-\frac{k}{2}}, \dots$$

$$= (1-2t)^{-\frac{k}{2}}$$

$$t = t_1 + t_2 + \dots + t_n$$

so if  $Y = X_1 + X_2 \rightarrow t = t_1 + t_2$

$$t_2 = t - t_1$$

$\therefore$  If  $Y \sim \chi^2(t)$ ,  $X_1 \sim \chi^2(t_1)$ ,  $X_2 \sim \chi^2(t - t_1)$

Problem 3. (a) show that  $h(x) = f(x) / (1 - F(x))$

$$h(x) = \lim_{\Delta \rightarrow 0} \frac{P(x \leq X \leq x+\Delta | X \geq x)}{\Delta} = \lim_{\Delta \rightarrow 0} \frac{P(x \leq X \leq x+\Delta)}{P(X \geq x) \Delta} = \frac{1}{1-F(x)} \lim_{\Delta \rightarrow 0} \frac{P(x \leq X \leq x+\Delta)}{\Delta}$$

$$= \frac{F'(x)}{1-F(x)} = \frac{f(x)}{1-F(x)}$$

(b) when  $h(x) = Cx^b$ , find the pdf of  $X$ .

$$h(x) = \frac{f(x)}{1-F(x)} \rightarrow 1-F(x) = e^{-\int h(x) dx + c}, F(0) = 0.$$

$$1-F(x) = e^{-\frac{C}{b+1} x^{b+1} + C_k}$$

$$C_k = 0 (\because F(0) = 0)$$

$$F(x) = 1 - e^{-\frac{C}{b+1} x^{b+1}}$$

$$F'(x) = f(x) = Cx^b e^{-\frac{C}{b+1} x^{b+1}}$$



Problem 4)  $X_1 \sim N(6, 1)$ ,  $X_2 \sim N(7, 1)$  Find  $P(X_1 > X_2)$

$$P(X_1 - X_2 > 0)$$

$$E(X_1 - X_2) = E(X_1) - E(X_2) = -1$$

$$\text{Var}(X_1 - X_2) = \text{Var}(X_1) + \text{Var}(X_2) = 2$$

$$P(X_1 - X_2 > 0) = P\left(\frac{X_1 - X_2 - (-1)}{\sqrt{2}} > \frac{0 - (-1)}{\sqrt{2}}\right) = P(Z > \frac{1}{\sqrt{2}})$$

$$= 1 - \Phi\left(\frac{1}{\sqrt{2}}\right) = 1 - \int_0^{\frac{1}{\sqrt{2}}} \frac{1}{\sqrt{2\pi}} e^{-t^2/2} dt = 1 - 0.7603$$

$$\therefore 0.2397$$

Problem 5.  $X, Y \sim N(0, 1)$ , find the mgf of the random variable.

$$M_W(t) = E(e^{xyt}) = \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} e^{xyt} \frac{1}{\sqrt{2\pi}} e^{-\frac{x^2}{2}} \frac{1}{\sqrt{2\pi}} e^{-\frac{y^2}{2}} dx dy$$

$$= \frac{1}{2\pi} \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} e^{(xy)t} e^{-\frac{x^2}{2}} e^{-\frac{y^2}{2}} dx dy = \frac{1}{2\pi} \int_{-\infty}^{\infty} \int_{-\infty}^{\infty} e^{-\frac{(x-yt)^2}{2}} e^{+\frac{y^2 t^2}{2}} e^{-\frac{y^2}{2}} dx dy$$

$$= \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} e^{-y^2 \frac{(1-t^2)}{2}} dy = \frac{1}{\sqrt{2\pi}} \times \frac{\sqrt{2\pi}}{\sqrt{1-t^2}} = \frac{1}{\sqrt{1-t^2}}$$

$$\therefore \frac{1}{\sqrt{1-t^2}}$$

Problem 6.  $X = (X_1, X_2)' \sim N_2(\mu, \Sigma)$ . let  $Y_1 = X_1 + X_2$ ,  $Y_2 = X_1 - X_2$ .

(1) Find distribution of  $Y$

$$\text{If } Y = AX \Rightarrow \begin{bmatrix} Y_1 \\ Y_2 \end{bmatrix} = \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix} \begin{bmatrix} X_1 \\ X_2 \end{bmatrix} \quad A = \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix}$$

$$\text{mean} = A\mu, \text{ Variance} = A\Sigma A'$$

$$\therefore \text{mean} = \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix} \mu, \text{ variance} = \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix} \Sigma \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix}'$$

(2) Find condition that  $Y_1$  and  $Y_2$  are independent.

$$\text{Var}(X_1) = \sigma_1^2, \text{Var}(X_2) = \sigma_2^2, \text{COV}(X_1, X_2) = \rho \sigma_1 \sigma_2$$

$$\Sigma = \begin{bmatrix} \sigma_1^2 & \rho \sigma_1 \sigma_2 \\ \rho \sigma_1 \sigma_2 & \sigma_2^2 \end{bmatrix} \quad A \Sigma A' = \begin{bmatrix} \sigma_1^2 + \rho \sigma_1 \sigma_2 & \sigma_2^2 + \rho \sigma_1 \sigma_2 \\ \sigma_1^2 - \rho \sigma_1 \sigma_2 & -\sigma_2^2 + \rho \sigma_1 \sigma_2 \end{bmatrix} \begin{bmatrix} 1 & 1 \\ 1 & -1 \end{bmatrix}$$

$$= \begin{bmatrix} \sigma_1^2 + \sigma_2^2 & \sigma_1^2 - \sigma_2^2 \\ +2\rho \sigma_1 \sigma_2 & \sigma_1^2 + \sigma_2^2 \\ \sigma_1^2 - \sigma_2^2 & 2\rho \sigma_1 \sigma_2 \end{bmatrix}$$

$$\text{If } \text{Var}(X_1) = \text{Var}(X_2) = \sigma_1 = \sigma_2, \quad A \Sigma A'_{12} = 0$$

then,  $Y_1, Y_2$  are independent

$$\therefore \text{Var}(X_1) = \text{Var}(X_2)$$

Problem 7.  $\mu_x = 1, \mu_y = 4, \sigma_x^2 = 4, \sigma_y^2 = 6, \rho_{xy} = 1/2$

find the mean and variance of  $z = 3x - 2y$

$$Z = (3 \ -2) \begin{pmatrix} X \\ Y \end{pmatrix} \quad A = (3 \ -2)$$

$$\text{mean} = (3 \ -2) \begin{pmatrix} \mu_x \\ \mu_y \end{pmatrix} = 3\mu_x - 2\mu_y = -5$$

$$\Sigma = \begin{pmatrix} \sigma_x^2 & \rho\sigma_x\sigma_y \\ \rho\sigma_x\sigma_y & \sigma_y^2 \end{pmatrix} = \begin{pmatrix} 4 & \sqrt{6} \\ \sqrt{6} & 6 \end{pmatrix}$$

$$\text{Var}(Z) = A \Sigma A' = (3 \ -2) \begin{pmatrix} 4 & \sqrt{6} \\ \sqrt{6} & 6 \end{pmatrix} \begin{pmatrix} 3 \\ -2 \end{pmatrix} = (12 - 2\sqrt{6} \ 3\sqrt{6} - 6) \begin{pmatrix} 3 \\ -2 \end{pmatrix} = \begin{pmatrix} 36 - 6\sqrt{6} \\ -6\sqrt{6} + 12 \end{pmatrix} = 48 - 12\sqrt{6}$$

$$\therefore \text{mean}(Z) = -5$$

$$\text{Var}(Z) = 48 - 12\sqrt{6}$$

Problem 8. Is  $S$  be an unbiased estimator of sigma?

$$E(X_i) = \mu, \text{Var}(X_i) = \sigma^2, E(X_i^2) = \mu^2 + \sigma^2, E(\bar{X}) = \frac{\sigma^2}{n} + \mu^2$$

$$E(\sum (X_i - \bar{X})^2) = E(\sum X_i^2 - 2\bar{X} \sum X_i + n\bar{X}^2) = \sum E(X_i^2) - E(n\bar{X}^2) \\ = n\mu^2 + n\sigma^2 - \sigma^2 - n\mu^2 = (n-1)\sigma^2.$$

$$E(S^2) = E\left(\frac{\sum (X_i - \bar{X})^2}{n-1}\right) = \frac{1}{n-1} E(\sum (X_i - \bar{X})^2) = \frac{n-1}{n-1} \sigma^2 = \sigma^2.$$

$\therefore S$  be an unbiased estimator of  $\sigma$ .