

4.24.

(a) $A = \begin{pmatrix} 1 & -1 & 1 \\ 2 & 1 & -1 \end{pmatrix}$ 에 대해 $Y = AX$ 이다.

$$EY = AE\bar{X} = \begin{pmatrix} 1 & -1 & 1 \\ 2 & 1 & -1 \end{pmatrix} \begin{pmatrix} 1 \\ 2 \\ 3 \end{pmatrix} = \begin{pmatrix} 2 \\ 1 \end{pmatrix}$$

$$\text{Var}(Y) = \text{Var}(AX) = A \text{Var}(X) A^T$$

$$= A \Sigma A^T = \begin{pmatrix} 1 & -1 & 1 \\ 2 & 1 & -1 \end{pmatrix} \begin{pmatrix} 1 & -1 & 0 \\ -1 & 2 & 1 \\ 0 & 1 & 2 \end{pmatrix} \begin{pmatrix} 1 & 2 \\ -1 & 1 \\ 1 & -1 \end{pmatrix}$$

$$= \begin{pmatrix} 5 & 1 \\ 1 & 2 \end{pmatrix}. Y \sim N\left(\begin{pmatrix} 2 \\ 1 \end{pmatrix}, \begin{pmatrix} 5 & 1 \\ 1 & 2 \end{pmatrix}\right)$$

(b) $\mu_1 := \begin{pmatrix} 1 \\ 2 \end{pmatrix}, \mu_2 := 3$

$$\Sigma := \begin{pmatrix} \Sigma_{11} & \Sigma_{12} \\ \Sigma_{21} & \Sigma_{22} \end{pmatrix} \text{ where } \Sigma_{22} = (2)$$

$$\mu_1 + \Sigma_{12} \Sigma_{22}^{-1} (x_3 - \mu_2) = \begin{pmatrix} 1 \\ 2 \end{pmatrix} + \begin{pmatrix} 0 \\ \frac{1}{2}(x_3 - 3) \end{pmatrix}$$

$$\Sigma_{11} - \Sigma_{12} \Sigma_{22}^{-1} \Sigma_{21} = \begin{pmatrix} 1 & -1 \\ -1 & 2 \end{pmatrix} - \begin{pmatrix} 0 \\ 1 \end{pmatrix} \frac{1}{2} \begin{pmatrix} 0 & 1 \end{pmatrix}$$

Thm 4.29(b)에 의해

$$X_1, X_2 | X_3 = x_3 \sim N\left(\begin{pmatrix} 1 \\ \frac{x_3}{2} + \frac{1}{2} \end{pmatrix}, \begin{pmatrix} 1 & -1 \\ -1 & \frac{3}{2} \end{pmatrix}\right)$$

5.10. V: or \wedge : and

(a) $|X_n + Y_n| > R \Rightarrow |X_n| > \frac{R}{2} \vee |Y_n| > \frac{R}{2}$

$$P(|X_n + Y_n| > R) \leq P(|X_n| > \frac{R}{2}) + P(|Y_n| > \frac{R}{2})$$

$$\limsup_{n \rightarrow \infty} P_n \leq 0 \text{ 이므로 } \lim_{n \rightarrow \infty} P(|X_n + Y_n| > R) = 0$$

$$\therefore X_n + Y_n = O_p(1)$$

(b) $R > 0$ 인 범위에서 R^2 과 R 은 one-to-one.

$$|X_n Y_n| > R^2 \Rightarrow |X_n| > \sqrt{R} \text{ or } |Y_n| > \sqrt{R}$$

$$P(|X_n Y_n| > R^2) \leq P(|X_n| > \sqrt{R}) + P(|Y_n| > \sqrt{R})$$

$$\limsup_{n \rightarrow \infty} P_n \leq 0 \text{ 이므로 } \lim_{n \rightarrow \infty} P(|X_n Y_n| > R^2) = 0$$

$$\therefore X_n Y_n = O_p(1)$$

(c) $\epsilon > 0, \alpha > 0$ 에 대해

$$(|Z_n X_n| > \epsilon) \Rightarrow$$

$$(|X_n| \leq \alpha \wedge |Z_n X_n| > \epsilon) \vee (|X_n| > \alpha \wedge |Z_n X_n| > \epsilon)$$

$$\Rightarrow (|Z_n| > \frac{\epsilon}{\alpha}) \vee (|X_n| > \alpha)$$

$$P(|Z_n X_n| > \epsilon) \leq P(|Z_n| > \frac{\epsilon}{\alpha}) + \sup_n P(|X_n| > \alpha)$$

$$\lim_{n \rightarrow \infty} P(|Z_n X_n| > \epsilon) \leq 0 + \lim_{n \rightarrow \infty} \sup_{X_n = O_p(1)} P(|X_n| > \alpha) = 0$$

$$\therefore Z_n X_n \xrightarrow{p} 1$$

(d) ϵ 가 임의의 ϵ 이하면 δ 는 n_1, \dots, n_2 를 갖기.

$$\lim_{n \rightarrow \infty} P(|Z_n| \geq \delta) = 0 \Leftrightarrow n \geq n_0 \Rightarrow P(|Z_n| \geq \delta_0) < \epsilon$$

아니 $n \leq n_0$ 인 n 에 대해 $\lim_{n \rightarrow \infty} P(|Z_n| > R) = 0$ 이므로

$$1 \leq i \leq n_0 \text{에 대해 } \forall \epsilon > 0, \exists \delta_i \text{ s.t. } P(|Z_i| > \delta_i) < \epsilon$$

$$\delta := \max_i \{\delta_i : 0 \leq i \leq n_0\}$$

$$(i \leq n_0) \max_i P(|X_i| > \delta) \leq \max_i P(|X_i| > \delta_i) < \epsilon$$

$$(i > n_0) \sup P(|X_i| > \delta) \leq \sup P(|X_i| > \delta_0) < \epsilon$$

$$\therefore X > \delta \Rightarrow \sup P(|X_i| > \delta) \leq \epsilon, Z_n = O_p(1)$$

4.26.

(a)

(d) (a)에 의해 $\frac{1}{p} \sum_{i=1}^p Z_i^2 = S_{ZZ} \sim \chi^2(n-1)$

$$\frac{r}{\sqrt{1-r^2}} = \frac{\frac{1}{\sqrt{1-p^2}} S_{ZZ} + S_{ZW}}{\sqrt{S_{ZZ} S_{WW} - S_{ZW}^2}} = \frac{\frac{p}{\sqrt{1-p^2}} \frac{S_{ZZ}}{\sqrt{S_{ZZ}}} + \frac{S_{ZW}}{\sqrt{S_{ZZ}}}}{\sqrt{S_{WW} - \frac{S_{ZW}^2}{S_{ZZ}}}}$$

이때 (C-i)에서 $\sqrt{S_{WW} - S_{ZW}^2/S_{ZZ}} \stackrel{d}{=} V_2 \sim \chi^2(n-2)$

(C-ii)에서 $S_{ZW}/S_{ZZ} \stackrel{d}{=} U \sim N(0, 1^2)$

$$S_{ZZ} = \sum_{i=1}^n (Z_i - \bar{Z})^2 \sim \chi^2(n-1) = V_1$$

$$\therefore \frac{r}{\sqrt{1-r^2}} \stackrel{d}{=} \frac{\frac{p}{\sqrt{1-p^2}} \sqrt{V_1} + U}{\sqrt{V_2}} \quad (\because \text{(iii)} - \frac{S_{ZW}}{S_{ZZ}})$$

2020-15709 4등재

2. $\text{pdf}_{X,Y}(x,y) = \frac{1}{2\pi\sqrt{1-\rho^2}} e^{-\frac{x^2 - 2\rho xy + y^2}{2(1-\rho^2)}} \quad (x,y \in \mathbb{R})$

$z = \frac{y}{\sqrt{1-\rho^2}}$ $P(X < 0, Y < 0) =$
 $P(X < 0 | Y = X < 0)$
 $= P(X < 0) P(Z < \frac{-\rho X}{\sqrt{1-\rho^2}})$
 $P(X < 0 | X < 0) = P(X < 0, Z < -\frac{\rho X}{\sqrt{1-\rho^2}})$
 $z = -\frac{\rho}{\sqrt{1-\rho^2}} x = P(X > 0, Z > -\frac{\rho X}{\sqrt{1-\rho^2}})$

$= \int_0^\infty \int_{\arctan(\frac{\rho}{\sqrt{1-\rho^2}})}^{\frac{\pi}{2}} \frac{1}{2\pi} e^{-\frac{r^2}{2}} dr d\theta$ (3차원 변환)
 $= \int_0^\infty \int_{\arctan(\frac{\rho}{\sqrt{1-\rho^2}})}^{\frac{\pi}{2}} \frac{r}{2\pi} e^{-\frac{r^2}{2}} dr d\theta$
 $= \int_0^\infty \frac{1}{2\pi} \left(\frac{\pi}{2} - \arctan\left(-\frac{\rho}{\sqrt{1-\rho^2}}\right) \right) r e^{-\frac{r^2}{2}} dr$
 $= \left(\frac{1}{4} + \arctan\left(\frac{\rho}{\sqrt{1-\rho^2}}\right) \cdot \frac{1}{2\pi} \right) [-e^{-\frac{r^2}{2}}]_0^\infty$
 $= \frac{1}{4} + \frac{1}{2\pi} \cdot \arctan\left(\frac{\rho}{\sqrt{1-\rho^2}}\right)$

3. (a) $\begin{pmatrix} W_1 \\ W_2 \end{pmatrix} := \begin{pmatrix} X+Y \\ X-Y \end{pmatrix} = \begin{pmatrix} 1 & 1 \\ 1 & -1 \end{pmatrix} \begin{pmatrix} X \\ Y \end{pmatrix} \text{이라}$
 $\frac{Z}{\sim} \begin{pmatrix} W_1 \\ W_2 \end{pmatrix} \sim N_2(A\mu, A\Sigma A^T)$ 에서
 $A\mu = \begin{pmatrix} 1 & 1 \\ 1 & -1 \end{pmatrix} \begin{pmatrix} \mu_1 \\ \mu_2 \end{pmatrix} = \begin{pmatrix} \mu_1 + \mu_2 \\ \mu_1 - \mu_2 \end{pmatrix}$
 $A\Sigma A^T = \begin{pmatrix} 1 & 1 \\ 1 & -1 \end{pmatrix} \begin{pmatrix} \sigma_1^2 & \rho\sigma_1\sigma_2 \\ \rho\sigma_1\sigma_2 & \sigma_2^2 \end{pmatrix} \begin{pmatrix} 1 & 1 \\ 1 & -1 \end{pmatrix}$
 $= \begin{pmatrix} \sigma_1^2 + 2\rho\sigma_1\sigma_2 + \sigma_2^2 & \sigma_1^2 - \sigma_2^2 \\ \sigma_1^2 - \sigma_2^2 & \sigma_1^2 - 2\rho\sigma_1\sigma_2 + \sigma_2^2 \end{pmatrix}$

Prop 4.28 (b)에서 $W_1 \perp W_2 \Leftrightarrow \sigma_1^2 - \sigma_2^2 = 0$
 $\therefore (X+Y) \perp (X-Y) \Leftrightarrow \sigma_1^2 = \sigma_2^2$

(b) $\begin{pmatrix} V_1 \\ V_2 \end{pmatrix} = \begin{pmatrix} aX+bY \\ cX+dY \end{pmatrix} = \begin{pmatrix} a & b \\ c & d \end{pmatrix} \begin{pmatrix} X \\ Y \end{pmatrix}$
 따라서 $\begin{pmatrix} V_1 \\ V_2 \end{pmatrix} \sim N\left(B \begin{pmatrix} \mu_1 \\ \mu_2 \end{pmatrix}, B \begin{pmatrix} \sigma_1^2 & \rho\sigma_1\sigma_2 \\ \rho\sigma_1\sigma_2 & \sigma_2^2 \end{pmatrix} B^T\right)$

Thm 4.29 (b)에서 $V_1/V_2 = t$ 는 이변량
 정규분포 ~~가~~ 다르다.

4. 우선 $U_{(1)} := \min U_i$, $U_{(n)} := \max U_i$

$$\text{pdf}_{U_{(1)}}(x) = \frac{n}{a} \left(1 - \frac{x}{a}\right)^{n-1} I_{(0,a)}(x)$$

$$E(U_{(1)}) = \int_0^a x \cdot \frac{n}{a} \left(1 - \frac{x}{a}\right)^{n-1} dx = \frac{a}{n+1} \quad E(U_i)^2$$

$$\begin{aligned} \text{Var}(U_{(1)}) &= \int_0^a n a \left(\frac{a}{x}\right)^{n-1} \left(1 - \frac{a}{x}\right)^{n-1} dx - \frac{E(U_i)^2}{P(n+3)} \\ &= \frac{2a^2}{(n+1)(n+2)} - \frac{a^2}{(n+1)^2} \end{aligned}$$

$$\text{pdf}_{U_{(n)}}(x) = \frac{n}{a} \left(\frac{x}{a}\right)^{n-1} I_{(0,a)}(x)$$

$$E(U_{(n)}) = \int_0^a \frac{n}{a^n} x^n dx = \frac{na}{n+1}$$

$$\text{Var}(U_{(n)}) = \int_0^a \frac{n}{a^n} x^{n+1} dx - \frac{n^2 a^2}{(n+1)^2}$$

$$= \frac{a^2 n}{(n+1)^2 (n+2)}$$

$$\lim_{n \rightarrow \infty} E(U_{(1)}) = 0, \lim_{n \rightarrow \infty} \text{Var}(U_{(1)}) = 0$$

$$\lim_{n \rightarrow \infty} E(U_{(n)}) = a, \lim_{n \rightarrow \infty} \text{Var}(U_{(n)}) = 0$$

$$p\lim_{n \rightarrow \infty} U_{(1)} = 0, p\lim_{n \rightarrow \infty} U_{(n)} = a$$

$$p\lim_{n \rightarrow \infty} X_n = p\lim_{n \rightarrow \infty} U_{(1)} + p\lim_{n \rightarrow \infty} U_{(n)} = a$$