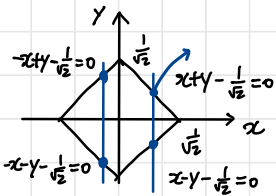
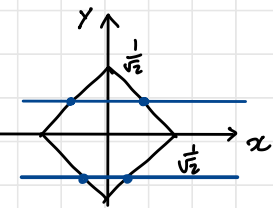


#1)
$$P(x, y) = \begin{cases} 1 & (|x| + |y| \leq \frac{1}{\sqrt{2}}) \\ 0 & (\text{others}) \end{cases}$$



$$P_X(x) = \int_{-\infty}^{\infty} p(x, y) dy = \begin{cases} \int_{x-\frac{1}{\sqrt{2}}}^{x+\frac{1}{\sqrt{2}}} 1 dy = -2x + \sqrt{2} & (0 \leq x \leq \frac{1}{\sqrt{2}}) \\ \int_{-x-\frac{1}{\sqrt{2}}}^{x+\frac{1}{\sqrt{2}}} 1 dy = 2x + \sqrt{2} & (-\frac{1}{\sqrt{2}} \leq x \leq 0) \end{cases}$$



$$P_Y(y) = \int_{-\infty}^{\infty} p(x, y) dx = \begin{cases} \int_{y-\frac{1}{\sqrt{2}}}^{-y+\frac{1}{\sqrt{2}}} 1 dx = -2y + \sqrt{2} & (0 \leq y \leq \frac{1}{\sqrt{2}}) \\ \int_{-y-\frac{1}{\sqrt{2}}}^{y+\frac{1}{\sqrt{2}}} 1 dx = 2y + \sqrt{2} & (-\frac{1}{\sqrt{2}} \leq y \leq 0) \end{cases}$$

$0 \leq x \leq \frac{1}{\sqrt{2}}, 0 \leq y \leq \frac{1}{\sqrt{2}}$ 일 때 $P_X(x)P_Y(y) = (-2x + \sqrt{2})(-2y + \sqrt{2})$
 $\neq P_{X,Y}(x, y)$

이므로 독립이 아니다!

$$\begin{aligned}
 E(X) &= \int_0^{\frac{1}{\sqrt{2}}} -2x^2 + \sqrt{2}x \, dx + \int_{-\frac{1}{\sqrt{2}}}^0 (2x^2 + \sqrt{2}x) \, dx \\
 &= \left[-\frac{2}{3}x^3 + \frac{\sqrt{2}}{2}x^2 \right]_0^{\frac{1}{\sqrt{2}}} + \left[\frac{2}{3}x^3 + \frac{\sqrt{2}}{2}x^2 \right]_{-\frac{1}{\sqrt{2}}}^0 \\
 &= \left(-\frac{1}{3\sqrt{2}} + \frac{\sqrt{2}}{4} \right) + \left(\frac{1}{3\sqrt{2}} - \frac{\sqrt{2}}{4} \right) = 0
 \end{aligned}$$

$$\begin{aligned}
 E(Y) &= \int_0^{\frac{1}{\sqrt{2}}} (-2y^2 + \sqrt{2}y) \, dy + \int_{-\frac{1}{\sqrt{2}}}^0 (2y^2 + \sqrt{2}y) \, dy \\
 &= 0
 \end{aligned}$$

$$\text{cov}(X, Y) = E((X-0)(Y-0))$$

$$= E(XY)$$

$$= \int_0^{\frac{1}{\sqrt{2}}} \int_{x-\frac{1}{\sqrt{2}}}^{-x+\frac{1}{\sqrt{2}}} xy \cdot 1 \, dy \, dx + \int_{-\frac{1}{\sqrt{2}}}^0 \int_{-x-\frac{1}{\sqrt{2}}}^{-x+\frac{1}{\sqrt{2}}} xy \cdot 1 \, dy \, dx$$

$$= \int_0^{\frac{1}{\sqrt{2}}} x \left[\frac{1}{2}y^2 \right]_{x-\frac{1}{\sqrt{2}}}^{-x+\frac{1}{\sqrt{2}}} dx + \int_{-\frac{1}{\sqrt{2}}}^0 x \left[\frac{1}{2}y^2 \right]_{-x-\frac{1}{\sqrt{2}}}^{-x+\frac{1}{\sqrt{2}}} dx$$

$$= 0$$

관산이 0이므로, 즉 두 확률변수가 상관없어도 독립은 아닐수 있다

#2. $d \times d$ 행렬 A , 확률변수 $X = (X_1, X_2, \dots, X_d)^T$ 에 대해

#2.1 $E(AX) = AE(X)$

솔리) 나만의 풀이

$$E(AX) = E \left(\begin{pmatrix} \sum_{j=1}^d a_{1j} X_j \\ \sum_{j=1}^d a_{2j} X_j \\ \vdots \\ \sum_{j=1}^d a_{dj} X_j \end{pmatrix} \right) = \begin{pmatrix} E\left(\sum_{j=1}^d a_{1j} X_j\right) \\ E\left(\sum_{j=1}^d a_{2j} X_j\right) \\ \vdots \\ E\left(\sum_{j=1}^d a_{dj} X_j\right) \end{pmatrix}$$

$$= \begin{pmatrix} \sum_{j=1}^d a_{1j} E(X_j) \\ \sum_{j=1}^d a_{2j} E(X_j) \\ \vdots \\ \sum_{j=1}^d a_{dj} E(X_j) \end{pmatrix} = \begin{pmatrix} a_{11} & a_{12} & \dots & a_{1d} \\ a_{21} & a_{22} & \dots & a_{2d} \\ \vdots & \vdots & \ddots & \vdots \\ a_{d1} & a_{d2} & \dots & a_{dd} \end{pmatrix} \begin{pmatrix} E(X_1) \\ E(X_2) \\ \vdots \\ E(X_d) \end{pmatrix} = AE(X)$$

전개

#2.1 $E(AX) = AE(X)$

so 2) 교수님 언급하신 풀이

$$E(AX) = E \left(\begin{pmatrix} \sum_{j=1}^d a_{1j} x_j \\ \sum_{j=1}^d a_{2j} x_j \\ \vdots \\ \sum_{j=1}^d a_{dj} x_j \end{pmatrix} \right) = \begin{pmatrix} E\left(\sum_{j=1}^d a_{1j} x_j\right) \\ E\left(\sum_{j=1}^d a_{2j} x_j\right) \\ \vdots \\ E\left(\sum_{j=1}^d a_{dj} x_j\right) \end{pmatrix}$$

$$= \begin{pmatrix} \int \dots \int \sum_{j=1}^d a_{1j} x_j p(x_1, x_2, \dots, x_d) dx_1 \dots dx_d \\ \int \dots \int \sum_{j=1}^d a_{2j} x_j p(x_1, x_2, \dots, x_d) dx_1 \dots dx_d \\ \vdots \\ \int \dots \int \sum_{j=1}^d a_{dj} x_j p(x_1, x_2, \dots, x_d) dx_1 \dots dx_d \end{pmatrix}$$

$$= \begin{pmatrix} \sum_{j=1}^d a_{1j} \int \dots \int x_j p(x_1, x_2, \dots, x_d) dx_1 \dots dx_n \\ \sum_{j=1}^d a_{2j} \int \dots \int x_j p(x_1, x_2, \dots, x_d) dx_1 \dots dx_n \\ \vdots \\ \sum_{j=1}^d a_{dj} \int \dots \int x_j p(x_1, x_2, \dots, x_d) dx_1 \dots dx_n \end{pmatrix}$$

$$= \begin{pmatrix} a_{1j} \\ a_{2j} \\ \vdots \\ a_{dj} \end{pmatrix} \begin{pmatrix} E(x_1) \\ E(x_2) \\ \vdots \\ E(x_d) \end{pmatrix} = AE(X)$$

cf)

$$\begin{aligned} E_{x_1}(x_1) &= \int \dots \int x_1 p(x_1, x_2, \dots, x_d) dx_2 \dots dx_d \\ &= \int x_1 \int \dots \int p(x_1, x_2, \dots, x_d) dx_2 \dots dx_d dx_1 \\ &= \int x_1 p_{x_1}(x_1) dx_1 \end{aligned}$$

$$\#2.2 \quad V(AX) = AV(X)A^T$$

$$V(AX) = E \left((AX - E(AX))(AX - E(AX))^T \right)$$

$$= E \left((AX - AE(X))(AX - AE(X))^T \right)$$

$$= E \left(A(X - E(X))(X - E(X))^T A^T \right)$$

$$= A E \left((X - E(X))(X - E(X))^T \right) A^T$$

$$= AV(X)A^T$$