

Q1

X	$\mu_1 = 1$	$\sigma_1^2 = 4$	$p = \frac{1}{2}$	(Independent r.v.)
Y	$\mu_2 = 4$	$\sigma_2^2 = 6$		

$$Z = 3X - 2Y$$

$$\begin{aligned} E[Z] &= E[3X - 2Y] = 3E[X] - 2E[Y] \\ &= 3 \cdot 1 - 2 \cdot 4 \\ &\Rightarrow -5 \end{aligned}$$

$$\begin{aligned} \text{Var}[Z] &= E[Z^2] - \mu^2 \\ &= \text{Var}[3X - 2Y] \\ &= 9 \cdot \text{Var}[X] + 4 \cdot \text{Var}[Y] \\ &= 9 \cdot 4 + 4 \cdot 6 \\ &\Rightarrow 60 \end{aligned}$$

Q2

$$P[x_i = 1] = P[x_i = -1] = \begin{cases} 1/2 \\ 0 \text{ o.w} \end{cases}$$

$$S = a_1 x_1 + \dots + a_n x_n$$

(a) $E[S] = E[a_1 x_1 + a_2 x_2 + \dots + a_n x_n]$

$$E[x_i] = 1 \cdot \frac{1}{2} + (-1) \cdot \frac{1}{2} = 0$$

$$E[S] = 0$$

(b) $\text{Var}[S] = \text{Var}[a_1 x_1 + a_2 x_2 + \dots + a_n x_n]$

$$\text{Var}[x_i] = E[x_i^2] - (\mu_{x_i})^2$$

$$= 1 \cdot \frac{1}{2} + (-1) \cdot \frac{1}{2} - 0 = 1$$

$$\text{Var}[S] = \sum_{i=1}^n a_i^2 \cdot \text{Var}[x_i] = \sum_{i=1}^n a_i^2$$

Q3

$$X = \begin{bmatrix} X_1 \\ X_2 \\ \vdots \\ X_n \end{bmatrix}$$

$$Y = a_1 X_1 + \dots + a_n X_n + c$$

$$Z = b_1 X_1 + \dots + b_n X_n + d$$

$\Sigma = \text{Cov}(X) \Rightarrow$ variance-covariance matrix

$$\text{Var}[Y] = \text{Var}[a_1 X_1 + a_2 X_2 + \dots + a_n X_n + c]$$

$$= \sum_{i=1}^n a_i^2 \text{Var}[X_i]$$

$$= \sum_{i=1}^n a_i^2 [\Sigma]_{ii} \quad \left(\begin{array}{l} \text{as } i\text{th diagonal entry in} \\ \Sigma \text{ is } \text{Var}[X_i] \end{array} \right)$$

$$\text{Cov}[Y, Z]$$

$$= \text{Cov}[a_1 X_1 + \dots + a_n X_n + c, b_1 X_1 + \dots + b_n X_n + d]$$

$$= \sum_{i,j=1}^n a_i b_j \text{Cov}(X_i, X_j)$$

$$= \sum_{i,j=1}^n a_i b_j [\Sigma]_{ij}$$

Q4

$$\begin{matrix} X_1 & \mu_1 & \sigma_1^2 \\ X_2 & \mu_2 & \sigma_2^2 \end{matrix}$$

$$\rho = \text{Cov}(X, Y)$$

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$$E[(X_1 - \mu_1) | X_2 = x_2] = a + bx_2$$

(a)

$$\Rightarrow E[X_1 | X_2 = x_2] - E[\mu_1 | X_2 = x_2]$$

$$\Rightarrow E[X_1 | X_2 = x_2] - \mu_1 = a + bx_2$$

$$E[X_1 | X_2 = x_2] = a + bx_2 + \mu_1$$

or

$$E[E[X_1 | X_2 = x_2]] = E[a + bx_2 + \mu_1]$$

$$\hookrightarrow = E[X_1] = \mu_1 = a + bE[X_2] + \mu_1$$

$$\Rightarrow a + b\mu_2 = 0$$

$$\Rightarrow a = -b\mu_2$$

(b)

$$\sigma_1^2 = E[(X_1 - \mu_1)^2]$$

$$\rho = \frac{E[(X_1 - \mu_1)(X_2 - \mu_2)]}{\sqrt{\sigma_1^2 \sigma_2^2}}$$

$$\rho = \frac{\text{Cov}(X_1, X_2)}{\sigma_1 \sigma_2}$$

$$\rho \sigma_1 \sigma_2 = \text{Cov}(X_1, X_2)$$

$$= E[X_1 X_2] - \mu_1 \mu_2$$

$$= E[X_2] \cdot E[X_1 | X_2] - \mu_1 \mu_2 + \mu_1 \mu_2$$

$$= \mu_2 \cdot a$$

$$E[X_1 X_2] = \iint x_1 x_2 \cdot f_{X_1, X_2}(x_1, x_2) dx_1 dx_2$$

$$= \iint x_1 x_2 \cdot f_{X_1 | X_2}(x_1 | x_2) \cdot f_{X_2}(x_2) dx_1 dx_2$$

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$$\rho = \frac{\text{Cov}(X_1, X_2)}{\sigma_1 \sigma_2}$$

$$\rho \sigma_1 \sigma_2 = \text{Cov}(X_1, X_2)$$

$$= E[X_1 X_2] - \mu_1 \mu_2$$

$$= E[(X_1 - \mu_1)(X_2 - \mu_2)]$$

$$= \iint (x_1 - \mu_1)(x_2 - \mu_2) \cdot f_{X_1, X_2}(x_1, x_2) dx_1 dx_2$$

$$= \iint (x_1 - \mu_1)(x_2 - \mu_2) f_{X_1|X_2}(x_1|x_2) \cdot f_{X_2}(x_2) dx_1 dx_2$$

$$= \int (x_2 - \mu_2) \cdot f_{X_2}(x_2) \cdot \underbrace{\int (x_1 - \mu_1) \cdot f_{X_1|X_2}(x_1|x_2) dx_1}_{a + bx_2} dx_2$$

$$= \int (x_2 - \mu_2) \cdot f_{X_2}(x_2) \cdot (a + bx_2) dx_2$$

$$= \int \cancel{b} \cdot (x_2 - \mu_2) \cdot (bx_2 - b\mu_2 + \underbrace{b\mu_2 + a}_0) \cdot f_{X_2}(x_2) dx_2$$

$$= \int (x_2 - \mu_2) \cdot b(x_2 - \mu_2) \cdot f_{X_2}(x_2) dx_2$$

$$= b \int (x_2 - \mu_2)^2 f_{X_2}(x_2) dx_2$$

$$= b \cdot \sigma_2^2$$