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2.25 | Let X have covariance matrix $\Sigma = \begin{bmatrix} 25 & -2 & 4 \\ -2 & 4 & 1 \\ 4 & 1 & 9 \end{bmatrix}$

$$\begin{array}{c|c|c} 1 & 4 & 4 \\ 2 & 6 & 6 \\ 4 & 14 & 14 \\ \hline & 24 & 24 \end{array}$$

a) Determine S and $V^{1/2}$

b) Multiply your matrices to check the relation $V^{1/2} S V^{1/2} = \Sigma$

Solu. given a covariance matrix $\Sigma = \begin{bmatrix} 25 & -2 & 4 \\ -2 & 4 & 1 \\ 4 & 1 & 9 \end{bmatrix}$

a) We know that, $V^{1/2}$ is a standard deviation matrix and is given by

$$V^{1/2} = \begin{bmatrix} \sqrt{\sigma_{11}} & 0 & 0 \\ 0 & \sqrt{\sigma_{22}} & 0 \\ 0 & 0 & \sqrt{\sigma_{33}} \end{bmatrix} = \begin{bmatrix} 5 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 3 \end{bmatrix}$$

Now since we have Σ and $V^{1/2}$, we can find correlation matrix S using the relation

$$S = V^{-1/2} \Sigma V^{-1/2}$$

$$\Rightarrow S = (V^{1/2})^{-1} \Sigma (V^{1/2})^{-1}$$

The inverse of diagonal matrix $V^{1/2}$ is given as

$$(V^{1/2})^{-1} = \begin{bmatrix} \frac{1}{5} & 0 & 0 \\ 0 & \frac{1}{2} & 0 \\ 0 & 0 & \frac{1}{3} \end{bmatrix}$$

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$$S = \begin{bmatrix} \frac{1}{5} & 0 & 0 \\ 0 & \frac{1}{2} & 0 \\ 0 & 0 & \frac{1}{3} \end{bmatrix} \begin{bmatrix} 25 & -2 & 4 \\ -2 & 4 & 1 \\ 4 & 1 & 9 \end{bmatrix} \begin{bmatrix} \frac{1}{5} & 0 & 0 \\ 0 & \frac{1}{2} & 0 \\ 0 & 0 & \frac{1}{3} \end{bmatrix}$$

$$= \begin{bmatrix} \frac{25}{5} & \frac{-2}{5} & \frac{4}{5} \\ \frac{-2}{2} & \frac{4}{2} & \frac{1}{2} \\ \frac{4}{3} & \frac{1}{3} & \frac{9}{3} \end{bmatrix} \begin{bmatrix} \frac{1}{5} & 0 & 0 \\ 0 & \frac{1}{2} & 0 \\ 0 & 0 & \frac{1}{3} \end{bmatrix}$$

$$= \begin{bmatrix} \frac{25}{5 \cdot 5} & \frac{-2}{5 \cdot 2} & \frac{4}{5 \cdot 3} \\ \frac{-2}{2 \cdot 5} & \frac{4}{2 \cdot 2} & \frac{1}{2 \cdot 3} \\ \frac{4}{3 \cdot 5} & \frac{1}{3 \cdot 2} & \frac{9}{3 \cdot 3} \end{bmatrix}$$

$$= \begin{bmatrix} 1 & -\frac{1}{5} & \frac{4}{15} \\ -\frac{1}{5} & 1 & \frac{1}{6} \\ \frac{4}{15} & \frac{1}{6} & 1 \end{bmatrix}$$



Another approach

Since we know that $S = [s_{ij}]$

$$s_{ij} = \frac{\sigma_i \sigma_j}{\sqrt{\sigma_i^2} \sqrt{\sigma_j^2}} \text{ for } i \neq j$$

$$= 1 \text{ for } i=j$$

$$S = \begin{bmatrix} 1 & \frac{-2}{5 \cdot 2} & \frac{4}{5 \cdot 3} \\ \frac{-2}{2 \cdot 5} & 1 & \frac{1}{2 \cdot 3} \\ \frac{4}{3 \cdot 5} & \frac{1}{3 \cdot 2} & 1 \end{bmatrix}$$

$$= \begin{bmatrix} 1 & -\frac{1}{5} & \frac{4}{15} \\ -\frac{1}{5} & 1 & \frac{1}{6} \\ \frac{4}{15} & \frac{1}{6} & 1 \end{bmatrix}$$

b. > Checking for the relation $V^{1/2} S V^{1/2} = \Sigma$

LHS = $V^{1/2} S V^{1/2}$

$$= \begin{bmatrix} 5 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 3 \end{bmatrix} \begin{bmatrix} 1 & -\frac{1}{5} & \frac{4}{15} \\ -\frac{1}{5} & 1 & \frac{1}{6} \\ \frac{4}{15} & \frac{1}{6} & 1 \end{bmatrix} \begin{bmatrix} 5 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 3 \end{bmatrix}$$

$$= \begin{bmatrix} 5 & -1 & \frac{4}{3} \\ -\frac{2}{5} & 2 & \frac{1}{3} \\ \frac{4}{5} & \frac{1}{2} & 3 \end{bmatrix} \begin{bmatrix} 5 & 0 & 0 \\ 0 & 2 & 0 \\ 0 & 0 & 3 \end{bmatrix}$$

$$= \begin{bmatrix} 25 & -2 & 4 \\ -2 & 4 & 1 \\ 4 & 1 & 9 \end{bmatrix} = \Sigma = \text{RHS. } \square$$

2.26] Use Σ as given in exercise 2.25

2

a) Find S_{13}

b) Find the correlation between X_1 and $\frac{1}{2}X_2 + \frac{1}{2}X_3$.

solve we are given $\Sigma = \begin{bmatrix} 25 & -2 & 4 \\ -2 & 4 & 1 \\ 4 & 1 & 9 \end{bmatrix}$

$x_1 \quad x_2 \quad x_3$

a) The standard deviation matrix is

$$V^{1/2} = \begin{bmatrix} 5 & 0 & 0 \\ 0 & 2 & 1 \\ 0 & 0 & 3 \end{bmatrix}$$

so, correlation matrix is $\rho = \begin{bmatrix} 1 & -1/5 & 4/15 \\ -1/5 & 1 & 1/6 \\ 4/15 & 1/6 & 1 \end{bmatrix}$

Hence $S_{13} = 4/15$. \checkmark \square

b) To find correlation between two items, we need to first find co-variance between them and divide that by their individual variances.

so $\text{Var}(X_1) = \text{Var}(X_1) = \sigma_{11} = 25$ \checkmark

$$\begin{aligned} \text{Var}\left(\frac{1}{2}X_2 + \frac{1}{2}X_3\right) &= \left(\frac{1}{2}\right)^2 \text{Var}(X_2) + \left(\frac{1}{2}\right)^2 \text{Var}(X_3) + 2 \cdot \frac{1}{2} \cdot \frac{1}{2} \text{Cov}(X_2, X_3) \\ &= \frac{1}{4} \text{Var}(X_2) + \frac{1}{4} \text{Var}(X_3) + \frac{1}{2} \text{Cov}(X_2, X_3) \end{aligned}$$

$$= \frac{1}{4} \sigma_{22} + \frac{1}{4} \sigma_{33} + \frac{1}{2} \sigma_{23}$$

$$= \frac{1}{4} [\sigma_{22} + \sigma_{33}] + \frac{1}{2} \sigma_{23}$$

$$= \frac{1}{4} [4 + 9] + \frac{1}{2} \cdot 1$$

$$= \frac{13}{4} + \frac{1}{2}$$

$$= \frac{15}{4} \checkmark$$

#Note

$$\text{Var}(aX_1 + bX_2) = a^2 \text{Var}(X_1) + b^2 \text{Var}(X_2) + 2ab \text{Cov}(X_1, X_2)$$

Now $\text{Cov}(X_1, \frac{1}{2}X_2 + \frac{1}{2}X_3)$

$$= \frac{1}{2} \text{Cov}(X_1, X_2) + \frac{1}{2} \text{Cov}(X_1, X_3)$$

$$= \frac{1}{2} \sigma_{12} + \frac{1}{2} \sigma_{13}$$

$$= \frac{1}{2} (-2) + \frac{1}{2} \cdot (4)$$

$$= -1 + 2$$

$$= 1 \checkmark$$

finally, correlation between X_1 & $\frac{1}{2}X_2 + \frac{1}{2}X_3$ is given by

$$\rho_{X_1, \frac{1}{2}X_2 + \frac{1}{2}X_3} = \frac{\text{Cov}(X_1, \frac{1}{2}X_2 + \frac{1}{2}X_3)}{\sqrt{\text{Var}(X_1)} \sqrt{\text{Var}(\frac{1}{2}X_2 + \frac{1}{2}X_3)}}$$

$$= \frac{1}{\sqrt{25} \sqrt{15/4}}$$

$$= \frac{2}{5\sqrt{15}} \checkmark$$

$$= 0.1032$$

And from this result we can say there is weak correlation between X_1 and $\frac{1}{2}X_2 + \frac{1}{2}X_3$.

$$\left(\frac{6}{6} \right)$$

Q.27] Derive expression for the mean and variance of the following linear combinations in terms of the means and covariances of the random variable X_1, X_2 , and X_3 .

a) $X_1 - 2X_2$

b) $-X_1 + 3X_2$

c) $X_1 + 2X_2 + X_3$

d) $X_1 + 2X_2 - X_3$

e) $X_1 + 2X_2 - X_3$

f) $3X_1 - 4X_2$ if X_1 and X_2 are independent Random variables

ASK HERE: Is this X_1 and X_2 independent given only for part f or in general?

I have solved assuming only for last part but if it is in general then $\text{cov}(X_1, X_2) = 0$ in all parts.

Solve
a) $X_1 - 2X_2$

For mean: $E(X_1 - 2X_2) = E(X_1) - 2E(X_2)$

For Variance: $\text{Var}(X_1 - 2X_2) = 1^2 \text{Var}(X_1) + (-2)^2 \text{Var}(X_2) + 2 \cdot 1 \cdot (-2) \text{Cov}(X_1, X_2)$

$$\therefore \text{Var}(aX_1 + bX_2) = a^2 \text{Var}(X_1) + b^2 \text{Var}(X_2) + 2ab \text{Cov}(X_1, X_2)$$

$$= \text{Var}(X_1) + 4\text{Var}(X_2) - 4\text{Cov}(X_1, X_2)$$

b) $-X_1 + 3X_2$

For mean: $E(-X_1 + 3X_2) = -E(X_1) + 3E(X_2)$

For Variance : $\text{Var}(-X_1 + 3X_2)$

$$= (-1)^2 \text{Var}(X_1) + (3)^2 \text{Var}(X_2) + 2 \cdot (-1)(3) \text{Cov}(X_1, X_2)$$

$$= \text{Var}(X_1) + 9 \text{Var}(X_2) - 6 \text{Cov}(X_1, X_2)$$

© $X_1 + X_2 + X_3$

For mean : $E(X_1 + X_2 + X_3) = E(X_1) + E(X_2) + E(X_3)$

For Variance : $\text{Var}[X_1 + (X_2 + X_3)]$

$$= (1)^2 \text{Var}(X_1) + (1)^2 \text{Var}(X_2 + X_3) + 2 \cdot (1)(1) \text{Cov}(X_1, X_2 + X_3)$$

$$= \text{Var}(X_1) + \text{Var}(X_2 + X_3) + 2 \text{Cov}(X_1, X_2 + X_3)$$

$$= \text{Var}(X_1) + \text{Var}(X_2) + \text{Var}(X_3) + 2 \text{Cov}(X_2, X_3) + 2 \text{Cov}(X_1, X_2 + X_3)$$

$$= \text{Var}(X_1) + \text{Var}(X_2) + \text{Var}(X_3) + 2 \text{Cov}(X_2, X_3) + 2 \text{Cov}(X_1, X_2) + 2 \text{Cov}(X_1, X_3)$$

d) $X_1 + 2X_2 - X_3$

For mean : $E(X_1 + 2X_2 - X_3) = E(X_1) + 2E(X_2) - E(X_3)$

For Variance : $\text{Var}[X_1 + (2X_2 - X_3)]$

$$= (1)^2 \text{Var}(X_1) + (1)^2 \text{Var}(2X_2 - X_3) + 2 \cdot (1)(1) \text{Cov}(X_1, 2X_2 - X_3)$$

$$\therefore \text{Var}(aX_1 + bX_2) = a^2 \text{Var}(X_1) + b^2 \text{Var}(X_2) + 2ab \text{Cov}(X_1, X_2)$$

$$= \text{Var}(X_1) + \text{Var}(2X_2 - X_3) + 2 \text{Cov}(X_1, 2X_2 - X_3)$$

$$= \text{Var}(X_1) + 4 \text{Var}(X_2) + \text{Var}(X_3) - 4 \text{Cov}(X_2, X_3) + 2 \text{Cov}(X_1, 2X_2 - X_3)$$

$$= \text{Var}(X_1) + 4 \text{Var}(X_2) + \text{Var}(X_3) - 4 \text{Cov}(X_2, X_3) + 4 \text{Cov}(X_1, X_2) - 2 \text{Cov}(X_1, X_3)$$

④ $3X_1 - 4X_2$. If X_1 and X_2 are independent Random Variable.

For mean : $E(3X_1 - 4X_2) = 3E(X_1) - 4E(X_2)$

For variance : $\text{Var}(3X_1 - 4X_2)$

$$= (3)^2 \text{Var}(X_1) + (-4)^2 \text{Var}(X_2) + 2 \cdot (3) \cdot (-4) \text{Cov}(X_1, X_2)$$

$$= 9 \text{Var}(X_1) + 16 \text{Var}(X_2) - 24 \text{Cov}(X_1, X_2) \rightarrow 0$$

$\text{Cov}(X_1, X_2) = 0$ because X_1 and X_2 are independent given

$$= 9 \text{Var}(X_1) + 16 \text{Var}(X_2) . \quad \square$$

41. You are given the random vector $X' = [X_1, X_2, X_3, X_4]$ with mean vector $\mu'_X = [3, 2, -2, 0]$ and variance-covariance matrix

$$\Sigma_X = \begin{bmatrix} 3 & 0 & 0 & 0 \\ 0 & 3 & 0 & 0 \\ 0 & 0 & 3 & 0 \\ 0 & 0 & 0 & 3 \end{bmatrix}$$

Let $A = \begin{bmatrix} 1 & -1 & 0 & 0 \\ 1 & 1 & -2 & 0 \\ 1 & 1 & 1 & 3 \end{bmatrix}$

- Find $E(AX)$, the mean of AX
- Find $\text{COV}(AX)$, the variance and covariance of AX
- Which pairs of linear combinations have zero covariances?

Solu.

a) Mean of $AX : E(AX)$

$$= A E(X)$$

$$= \begin{bmatrix} 1 & -1 & 0 & 0 \\ 1 & 1 & -2 & 0 \\ 1 & 1 & 1 & 3 \end{bmatrix}_{3 \times 4} \begin{bmatrix} 3 \\ 2 \\ -2 \\ 0 \end{bmatrix}_{4 \times 1}$$

$$= \begin{bmatrix} 1 \\ 9 \\ 3 \end{bmatrix}$$

$$b.) \text{Cov}(AX) = A \Sigma_X A'$$

#Note:

$$\text{Cov}(CX) = C \Sigma_X C'$$

$$= \begin{bmatrix} 1 & -1 & 0 & 0 \\ 1 & 1 & -2 & 0 \\ 1 & 1 & 1 & 3 \end{bmatrix} \begin{bmatrix} 3 & 0 & 0 & 0 \\ 0 & 3 & 0 & 0 \\ 0 & 0 & 3 & 0 \\ 0 & 0 & 0 & 3 \end{bmatrix} \begin{bmatrix} 1 & 1 & 1 \\ -1 & 1 & 1 \\ 0 & -2 & 1 \\ 0 & 0 & 3 \end{bmatrix}$$

$$= \begin{bmatrix} 3 & -3 & 0 & 0 \\ 3 & 3 & -6 & 0 \\ 3 & 3 & 3 & 9 \end{bmatrix} \begin{bmatrix} 1 & 1 & 1 & 1 \\ -1 & 1 & 1 & 1 \\ 0 & -2 & 1 & 1 \\ 0 & 0 & 3 & 1 \end{bmatrix}$$

$$= \begin{bmatrix} 6 & 0 & 0 \\ 0 & 18 & 0 \\ 0 & 0 & 36 \end{bmatrix}$$

Now

$$\Rightarrow AX = \begin{bmatrix} 1 & -1 & 0 & 0 \\ 1 & 1 & -2 & 0 \\ 1 & 1 & 1 & 3 \end{bmatrix} \begin{bmatrix} X_1 \\ X_2 \\ X_3 \\ X_4 \end{bmatrix}$$

$$= \begin{bmatrix} X_1 - X_2 \\ X_1 + X_2 - 2X_3 \\ X_1 + X_2 + X_3 + X_4 \end{bmatrix}$$

$$= \begin{bmatrix} Y_1 \\ Y_2 \\ Y_3 \end{bmatrix}$$

say

also let $Y = AX$

So clearly we have

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$$\text{COV}(AX) = \text{COV}(Y) = \begin{bmatrix} 6 & 0 & 0 \\ 0 & 18 & 0 \\ 0 & 0 & 36 \end{bmatrix} = \Sigma_Y$$

So

$$\text{Var}(Y_1) = \text{Var}(X_1 - X_2) = 6$$

$$\text{Var}(Y_2) = \text{Var}(X_1 + X_2 - 2X_3) = 18$$

$$\text{Var}(Y_3) = \text{Var}(X_1 + X_2 + X_3 + X_4) = 36$$

$$\text{COV}(Y_1, Y_2) = \text{COV}(X_1 - X_2, X_1 + X_2 - 2X_3) = 0 = \text{COV}(Y_2, Y_1)$$

$$\text{COV}(Y_1, Y_3) = \text{COV}(X_1 - X_2, X_1 + X_2 + X_3 + X_4) = 0 = \text{COV}(Y_3, Y_1)$$

$$\text{COV}(Y_2, Y_3) = \text{COV}(X_1 + X_2 - 2X_3, X_1 + X_2 + X_3 + X_4) = 0 = \text{COV}(Y_3, Y_2)$$

So, clearly All pairs of Linear combination have zero covariance. \square

* * The End * *

2

Q.30] You are given the random vector $X' = [X_1, X_2, X_3, X_4]$ with mean vector $\mu_X' = [4, 3, 2, 1]$ and variance-covariance matrix

$$\Sigma_X = \begin{bmatrix} 3 & 0 & 2 & 2 \\ 0 & 1 & 1 & 0 \\ 2 & 1 & 9 & -2 \\ 2 & 0 & -2 & 4 \end{bmatrix} \quad \text{partition } X \text{ as } X = \begin{bmatrix} X_1 \\ X_2 \\ \vdots \\ X_3 \\ X_4 \end{bmatrix} = \begin{bmatrix} X^{(1)} \\ \vdots \\ X^{(2)} \end{bmatrix}$$

Let $A = [1 \ 2]$ and $B = \begin{bmatrix} 1 & -2 \\ 2 & -1 \end{bmatrix}$

and consider the linear combinations $AX^{(1)}$ and $BX^{(2)}$. Find

- a) $E(X^{(1)})$
- b) $E(AX^{(1)})$
- c) $\text{Cov}(AX^{(1)})$
- d) $\text{Cov}(AX^{(1)})$
- e) $E(X^{(2)})$
- f) $E(BX^{(2)})$
- g) $\text{Cov}(X^{(2)})$
- h) $\text{Cov}(BX^{(2)})$
- i) $\text{Cov}(X^{(1)}, X^{(2)})$
- j) $\text{Cov}(AX^{(1)}, BX^{(2)})$.

Solu.

We are given that, a random vector

$X' = [X_1, X_2, X_3, X_4]$, with mean vector $\mu_X' = [4 \ 3 \ 2 \ 1]$

and variance-covariance matrix

$$\Sigma_X = \begin{bmatrix} 3 & 0 & 2 & 2 \\ 0 & 1 & 1 & 0 \\ 2 & 1 & 9 & -2 \\ 2 & 0 & -2 & 4 \end{bmatrix}$$

Consequently their partitions

$$\Sigma_{11} = \begin{bmatrix} 3 & 0 \\ 0 & 1 \end{bmatrix}, \quad \Sigma_{12} = \begin{bmatrix} 2 & 2 \\ 1 & 0 \end{bmatrix} = \Sigma_{21}^T$$

$$\text{and } \Sigma_{22} = \begin{bmatrix} 9 & -2 \\ -2 & 4 \end{bmatrix}, \quad \text{for } X = \begin{bmatrix} X_1 \\ X_2 \\ \dots \\ X_3 \\ X_4 \end{bmatrix} = \begin{bmatrix} X^{(1)} \\ \dots \\ X^{(2)} \end{bmatrix}$$

Also given $A = [1 \ 2]$ and $B = \begin{bmatrix} 1 & -2 \\ 2 & -1 \end{bmatrix}$ and their linear combination $AX^{(1)}$ and $BX^{(2)}$.

$$a) E(X^{(1)}) = E\left(\begin{bmatrix} X_1 \\ X_2 \end{bmatrix}\right) = \begin{pmatrix} E(X_1) \\ E(X_2) \end{pmatrix} = \begin{bmatrix} 4 \\ 3 \end{bmatrix} \checkmark$$

$$b) E(AX^{(1)}) = A E(X^{(1)}) = [1 \ 2] \begin{bmatrix} 4 \\ 3 \end{bmatrix}$$

$$= 4 + 6 \\ = 10 \checkmark$$

$$c) \text{Cov}(X^{(1)}) = \Sigma_{11} = \begin{bmatrix} 3 & 0 \\ 0 & 1 \end{bmatrix} \checkmark$$

$$d) \text{Cov}(AX^{(1)}) = A \text{Cov}(X^{(1)}) A'$$

#Note:

$$\text{Cov}(CX) = C \Sigma_X C'$$

$$= \begin{bmatrix} 1 & 2 \end{bmatrix}_{1 \times 2} \begin{bmatrix} 3 & 0 \\ 0 & 1 \end{bmatrix}_{2 \times 2} \begin{bmatrix} 1 \\ 2 \end{bmatrix}_{2 \times 1}$$

$$= \begin{bmatrix} 3 & 2 \end{bmatrix} \begin{bmatrix} 1 \\ 2 \end{bmatrix}$$

$$= 7 \quad \checkmark$$

$$\left\{ \begin{array}{l} \because A = \begin{bmatrix} 1 & 2 \end{bmatrix} \\ A' = \begin{bmatrix} 1 \\ 2 \end{bmatrix} \\ \& \text{cov}(X^{(1)}) = \Sigma_{11} \end{array} \right\}$$

$$e) E(X^{(2)}) = E\left(\begin{bmatrix} X_3 \\ X_4 \end{bmatrix}\right) = \begin{pmatrix} E(X_3) \\ E(X_4) \end{pmatrix} = \begin{bmatrix} 2 \\ 1 \end{bmatrix} \quad \checkmark$$

$$f) E(BX^{(2)}) = B E(X^{(2)}) = \begin{bmatrix} 1 & -2 \\ 2 & -1 \end{bmatrix}_{2 \times 2} \begin{bmatrix} 2 \\ 1 \end{bmatrix}_{2 \times 1} \\ = \begin{bmatrix} 0 \\ 3 \end{bmatrix} \quad \checkmark$$

$$g) \text{cov}(X^{(2)}) = \Sigma_{22} = \begin{bmatrix} 9 & -2 \\ -2 & 4 \end{bmatrix} \quad \checkmark$$

$$h) \text{cov}(BX^{(2)}) = B \text{cov}(X^{(2)}) B^T$$

Note

$$\text{cov}(CX) = C \Sigma_X C^T$$

$$= \begin{bmatrix} 1 & -2 \\ 2 & -1 \end{bmatrix}_{2 \times 2} \begin{bmatrix} 9 & -2 \\ -2 & 4 \end{bmatrix}_{2 \times 2} \begin{bmatrix} 1 & 2 \\ -2 & -1 \end{bmatrix}_{2 \times 2} \\ = \begin{bmatrix} 13 & -10 \\ 20 & -8 \end{bmatrix} \begin{bmatrix} 1 & 2 \\ -2 & -1 \end{bmatrix}$$

$$= \begin{bmatrix} 33 & 36 \\ 36 & 48 \end{bmatrix}$$

$$i.) \text{COV}(X^{(1)}, X^{(2)})$$

$$= \text{COV}\left(\begin{bmatrix} X_1 \\ X_2 \end{bmatrix}, \begin{bmatrix} X_3 \\ X_4 \end{bmatrix}\right)$$

$$= \begin{bmatrix} \text{COV}(X_1, X_3) & \text{COV}(X_1, X_4) \\ \text{COV}(X_2, X_3) & \text{COV}(X_2, X_4) \end{bmatrix}$$

$$= \begin{bmatrix} 2 & 2 \\ 1 & 0 \end{bmatrix} = \Sigma_{12}$$

$$j.) \text{COV}(AX^{(1)}, BX^{(2)})$$

$$= E(A X^{(1)} - \mu(A X^{(1)}))(B X^{(2)} - \mu(B X^{(2)}))^T$$

$$= A \cdot E(X^{(1)} - \mu(X^{(1)}))(X^{(2)T} B^T - \mu(X^{(2)})^T B^T)$$

$$= A \cdot \text{COV}(X^{(1)}, X^{(2)}) \cdot B^T$$

→ Ask here - The source for this relation is chatgpt, so I am not 100% sure and look for explanation on this.

$$= \begin{bmatrix} 1 & 2 \end{bmatrix}_{1 \times 2} \begin{bmatrix} 2 & 2 \\ 1 & 0 \end{bmatrix}_{2 \times 2} \begin{bmatrix} 1 & 2 \\ -2 & -1 \end{bmatrix}_{2 \times 2}$$

$$= \begin{bmatrix} 4 & 2 \end{bmatrix}_{1 \times 2} \begin{bmatrix} 1 & 2 \\ -2 & -1 \end{bmatrix}_{2 \times 2}$$

$$= [0 \ 6] \checkmark$$

$$\begin{array}{r} \overline{14} \\ 14 \end{array}$$