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# TIME SERIES ANALYSIS

Solutions to problems in Chapter 2

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IMM

## Solution 2.1

### Question 1.

A second order moment representation of  $(H \ L)^\top$  consists of

$$E[H], \quad V[H], \quad \rho[H, L], \quad E[L], \quad V[L]$$

which are calculated as

$$E[H] = E[2X + 3Y] = 2E[X] + 3E[Y] = \underline{\underline{40}}$$

$$E[L] = E[-X + 2Y] = -E[X] + 2E[Y] = \underline{\underline{15}}$$

$$\begin{aligned} V[H] &= V[2X + 3Y] = 2^2V[X] + 3^2V[Y] + 2 \cdot 6\text{Cov}[X, Y] \\ &= 2^2 \cdot 1 + 3^2 \cdot 2^2 + 2 \cdot 6 \cdot 1 \cdot 2 \cdot \frac{1}{2} = \underline{\underline{52}} \end{aligned}$$

$$\begin{aligned} V[L] &= V[-X + 2Y] = (-1)^2V[X] + 2^2V[Y] - 2 \cdot 2 \cdot \text{Cov}[X, Y] \\ &= (-1)^2 \cdot 1 + 2^2 \cdot 2^2 - 2 \cdot 2 \cdot 1 \cdot 2 \cdot \frac{1}{2} = \underline{\underline{13}} \end{aligned}$$

$$\begin{aligned} \text{Cov}[H, L] &= \text{Cov}[2X + 3Y, -X + 2Y] \\ &= -2V[X] + 6V[Y] + 4\text{Cov}[X, Y] - 3\text{Cov}[X, Y] \\ &= -2 \cdot 1 + 6 \cdot 2^2 + 1 \cdot 1 \cdot 2 \cdot \frac{1}{2} = 23 \quad \Rightarrow \end{aligned}$$

$$\rho[H, L] = \frac{\text{Cov}[H, L]}{\sqrt{V[H]V[L]}} = \frac{23}{\underline{\underline{26}}}$$

Alternatively, we define

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

I.e.

$$\begin{bmatrix} H \\ L \end{bmatrix} = B \begin{bmatrix} X \\ Y \end{bmatrix}$$

and the second order moment representation of  $(H \ L)^\top$  can then be calcu-

lated by

$$\begin{aligned} E \begin{bmatrix} H \\ L \end{bmatrix} &= BE \begin{bmatrix} X \\ Y \end{bmatrix} = \begin{bmatrix} 2 & 3 \\ -1 & 2 \end{bmatrix} \begin{bmatrix} 5 \\ 10 \end{bmatrix} = \underline{\underline{\begin{bmatrix} 40 \\ 15 \end{bmatrix}}} \\ V \begin{bmatrix} H \\ L \end{bmatrix} &= BV \begin{bmatrix} X \\ Y \end{bmatrix} B^T = \begin{bmatrix} 2 & 3 \\ -1 & 2 \end{bmatrix} \begin{bmatrix} 1 & 1 \\ 1 & 4 \end{bmatrix} \begin{bmatrix} 2 & -1 \\ 3 & 2 \end{bmatrix} = \\ &= \underline{\underline{\begin{bmatrix} 52 & 23 \\ 23 & 13 \end{bmatrix}}} \end{aligned}$$

In a second order moment representation the connection between the two stochastic variables can either be represented by the correlation  $\rho[X, Y]$  or by the covariance  $\text{Cov}[X, Y]$ .

## Solution 2.2

*Question 1.*

$$E[Y|X] = E[\alpha + \beta X + \epsilon|X] = \alpha + \beta E[X|X] + E[\epsilon|X] = \underline{\underline{\alpha + \beta X}}$$

$$V[Y|X] = V[\alpha + \beta X + \epsilon|X] = \beta^2 V[X|X] + V[\epsilon|X] = \underline{\underline{\sigma_\epsilon^2}}$$

(Since  $V[X|X] = E[(X - E[X|X])^2|X] = 0$ )

*Question 2.*

$$E[Y] = E[E[Y|X]] = E[\alpha + \beta X] = \alpha + \beta E[X] = \underline{\underline{\alpha + \beta \mu_x}}$$

$$V[Y] = E[V[Y|X]] + V[E[Y|X]] = E[\sigma_\epsilon^2] + V[\alpha + \beta X] = \underline{\underline{\sigma_\epsilon^2 + \beta^2 \sigma_X^2}}$$

*Question 3.*

(We define  $\mu_Y = E[Y]$ ,  $\sigma_Y^2 = V[Y]$  and  $\sigma_X^2 = V[X]$ )

$$\text{Cov}[X, Y] = \text{Cov}[X, \alpha + \beta X + \epsilon] = \beta \sigma_X^2$$

And since  $\text{Cov}[X, Y] = \rho(X, Y) \sigma_X \sigma_Y = \rho \sigma_X \sigma_Y$  the moment estimate of  $\beta$  is

$$\beta = \rho \frac{\sigma_Y}{\sigma_X}$$

From the first result in question 2  $\alpha$  is found as

$$\alpha = \mu_Y - \beta \mu_x = \underline{\underline{\mu_Y - \frac{\rho \sigma_Y \mu_X}{\sigma_X}}}$$

## Solution 2.3

### Question 1.

The mean and variance of  $Y$  can be found by first determining the conditional mean and variance of  $Y$  given  $N$ . The conditional mean is

$$E[Y|N] = E[X_1 + X_2 + \cdots + X_N|N] = N \cdot E[X_i],$$

and the conditional variance is

$$V[Y|N] = V[X_1 + X_2 + \cdots + X_N|N] = N \cdot V[X_i]$$

applying for the conditional variance that  $\text{Cov}[X_r, X_s|N = j] = 0$  for  $r \neq s$ . The mean is obtained by using the property (2.43)

$$\begin{aligned} E[Y] &= E[E[Y|N]] = E[N \cdot E[X_i]] = E[N] \cdot E[X_i] \\ &= 20 \cdot 2 = \underline{40}, \end{aligned}$$

and by using the theorem (2.51), we get

$$\begin{aligned} V[Y] &= E[V[Y|N]] + V[E[Y|N]] \\ &= E[N \cdot V[X_i]] + V[N \cdot E[X_i]] \\ &= E[N] \cdot V[X_i] + E[X_i]^2 \cdot V[N] \\ &= 20 \cdot \left(\frac{1}{8}\right)^2 + 2^2 \cdot 2^2 = \underline{\underline{\frac{261}{16}}}. \end{aligned}$$

### Question 2.

By using the theorem (2.52)

$$\text{Cov}[Y, Z] = E[\text{Cov}[Y, Z|N]] + \text{Cov}[E[Y|N], E[Z|N]],$$

where  $E[\text{Cov}[Y, Z|N]] = E[N \cdot V[X_i]]$  as  $\text{Cov}[X_r, X_s|N] = 0$  for  $r \neq s$ , i.e.

$$\text{Cov}[Y, Z] = E[N \cdot V[X_i]] + \text{Cov}[N \cdot E[X_i], N \cdot \alpha E[X_i]]$$

$$= E[N] \cdot V[X_i] + E[X_i]^2 \alpha \cdot V[N]$$

$$= 20 \cdot \left(\frac{1}{8}\right)^2 + 2^2 \cdot \alpha \cdot 2^2$$

$$= \frac{5}{16} + 16\alpha = \underline{\underline{\frac{5 + 256\alpha}{16}}}.$$