TIME SERIES ANALYSIS

Solutions to problems in Chapter 2



Solution 2.1

Question 1.

A second order momement representation of $(H \ L)^{\top}$ consists of

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

which are calculated as

$$\begin{split} E[H] &= E[2X + 3Y] = 2E[X] + 3E[Y] = \underline{40} \\ E[L] &= E[-X + 2Y] = -E[X] + 2E[Y] = \underline{15} \\ V[H] &= V[2X + 3Y] = 2^2V[X] + 3^2V[Y] + 2 \cdot 6\text{Cov}[X, Y] \\ &= 2^2 \cdot 1 + 3^2 \cdot 2^2 + 2 \cdot 6 \cdot 1 \cdot 2 \cdot \frac{1}{2} = \underline{52} \\ V[L] &= V[-X + 2Y] = (-1)^2V[X] + 2^2V[Y] - 2 \cdot 2 \cdot \text{Cov}[X, Y] \\ &= (-1)^2 \cdot 1 + 2^2 \cdot 2^2 - 2 \cdot 2 \cdot 1 \cdot 2 \cdot \frac{1}{2} = \underline{13} \\ \text{Cov}[H, L] &= \text{Cov}[2X + 4Y, -X + 2Y] \\ &= -2V[X] + 6V[Y] + 4\text{Cov}[X, Y] - 3\text{Cov}[X, Y] \\ &= -2 \cdot 1 + 6 \cdot 2^2 + 1 \cdot 1 \cdot 2 \cdot \frac{1}{2} = 23 \quad \Rightarrow \\ \rho[H, L] &= \frac{\text{Cov}[H, L]}{\sqrt{V[X]V[Y]}} = \underline{23} \\ \end{split}$$

Alternatively, we define

$$B = \begin{bmatrix} 2 & 3\\ -1 & 2 \end{bmatrix}$$
$$\begin{bmatrix} H\\ L \end{bmatrix} = B \begin{bmatrix} X\\ Y \end{bmatrix}$$

I.e.

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

lated by

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

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

Solution 2.2

Question 1.

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

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

Question 2.

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

Question 3.

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

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

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

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

From the first result in question 2 α is found as

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

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

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

and the conditional variance is

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

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)

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

and by using the theorem (2.51), we get

$$V[Y] = E[V[Y|N]] + V[E[Y|N]]$$

= $E[N \cdot V[X_i]] + V[N \cdot [E[X_i]]]$
= $E[N] \cdot V[X_i] + E[X_i]^2 \cdot V[N]$
= $20 \cdot (\frac{1}{8})^2 + 2^2 \cdot 2^2 = \frac{261}{16}$.

Question 2.

By using the theorem (2.52)

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

where
$$E[\operatorname{Cov}[Y, Z|N]] = E[N \cdot V[X_i]]$$
 as $\operatorname{Cov}[X_r, X_s|N] = 0$ for $r \neq s$, i.e.
 $\operatorname{Cov}[Y, Z] = E[N \cdot V[X_i]] + \operatorname{Cov}[N \cdot E[X_i], N \cdot \alpha E[X_i]]$
 $= E[N] \cdot V[X_i] + E[X_i]^2 \alpha \cdot V[N]$
 $= 20 \cdot (\frac{1}{8})^2 + 2^2 \cdot \alpha \cdot 2^2$
 $= \frac{5}{16} + 16\alpha = \frac{5 + 256\alpha}{\underline{16}}$.