

# SOLUTIONS TO EXERCISES WEEK 37

## Exercise 2.5

$$M_1 = \{M_{10}, M_{20}, \dots\} \quad \text{and} \quad M_2 = \{M_{20}, M_{21}, \dots\}$$

are martingales with  $M_{10} = M_{20} = 0$  w.v.t.

The history  $\{\mathcal{F}_n\}$ .

Predictable covariation process:  $\langle M_1, M_2 \rangle$

$$\langle M_1, M_2 \rangle_n = \sum_{i=1}^n E\{\Delta M_{1i} \Delta M_{2i} \mid \mathcal{F}_{i-1}\}$$

$$= \sum_{i=1}^n \text{Cov}(\Delta M_{1i}, \Delta M_{2i} \mid \mathcal{F}_{i-1})$$

Optional covariation process:  $[M_1, M_2]$

$$[M_1, M_2] = \sum_{i=1}^n \Delta M_{1i} \Delta M_{2i}$$

a) We immediately have that

$$M_{10} M_{20} - \langle M_1, M_2 \rangle_0 = 0 \quad \text{and}$$

$$M_{10} M_{20} - [M_1, M_2]_0 = 0.$$

Further we have that (cf. exercise 2.4):

(2)

$$E(M_{1n} M_{2n} - \langle M_1, M_2 \rangle_n | F_{n-1})$$

$$= E\{(M_{1,n-1} + \Delta M_{1n})(M_{2,n-1} + \Delta M_{2n}) | F_{n-1}\} - \langle M_1, M_2 \rangle_n$$

$$= E\{M_{1,n-1} M_{2,n-1} + M_{1,n-1} \Delta M_{2n} + M_{2,n-1} \Delta M_{1n} + \Delta M_{1n} \Delta M_{2n} | F_{n-1}\}$$

$$= \langle M_1, M_2 \rangle_{n-1} - E(\Delta M_{1n} \Delta M_{2n} | F_{n-1})$$

$$= M_{1,n-1} M_{2,n-1} + O + O + E(\Delta M_{1n} \Delta M_{2n} | F_{n-1})$$

$$- \langle M_1, M_2 \rangle_{n-1} - E(\Delta M_{1n} \Delta M_{2n} | F_{n-1})$$

$$= M_{1,n-1} M_{2,n-1} - \langle M_1, M_2 \rangle_{n-1}$$

Thus  $M_1 M_2 - \langle M_1, M_2 \rangle$  is a mean zero martingale.

Further we have (cf. page 45 in ABG)

$$E(M_{1n} M_{2n} - [M_1, M_2]_n | F_{n-1})$$

$$= E(M_{1,n-1} M_{2,n-1} + M_{1,n-1} \Delta M_{2n} + M_{2,n-1} \Delta M_{1n} - [M_1, M_2]_{n-1} | F_{n-1})$$

(3)

$$= M_{1,n-1} M_{2,n-1} + 0 + 0 - \{m_1, m_2\}_{n-1}$$

$$= M_{1,n-1} M_{2,n-1} - \{m_1, m_2\}_{n-1}$$

Thus  $M_1 M_2 - \{m_1, m_2\}$  is a mean zero martingale.

b) Since  $M_1$  and  $M_2$  are mean zero martingales, we have

$$\text{Cov}(M_{1n}, M_{2n}) = E(M_{1n} M_{2n})$$

Further by the results in a) we have that

$$E(M_{1n} M_{2n} - \langle m_1, m_2 \rangle_n) = 0$$

$$E(M_1 M_2 - \{m_1, m_2\}_n) = 0$$

This gives

$$\text{Cov}(M_{1n}, M_{2n}) = E \langle m_1, m_2 \rangle_n = E \{m_1, m_2\}_n$$

(4)

Exercise 2.8.

Let  $M_1$  and  $M_2$  be mean zero martingales as in exercise 2.5 and let  $H_1 = \{H_{10}, H_{11}, \dots\}$  and  $H_2 = \{H_{20}, H_{21}, \dots\}$  be predictable processes. Then we have

$$\langle H_1 \bullet M_1, H_2 \bullet M_2 \rangle_n$$

$$= \sum_{s=1}^n E \{ \Delta(H_1 \bullet M_1)_s \cdot \Delta(H_2 \bullet M_2)_s \mid \bar{\mathcal{F}}_{s-1} \}$$

$$= \sum_{s=1}^n E \{ H_{1s} \Delta M_{1s} \cdot H_{2s} \Delta M_{2s} \mid \bar{\mathcal{F}}_{s-1} \}$$

$$= \sum_{s=1}^n H_{1s} H_{2s} E \{ \Delta M_{1s} \Delta M_{2s} \mid \bar{\mathcal{F}}_{s-1} \}$$

$$= \sum_{s=1}^n H_{1s} H_{2s} \Delta \langle M_1, M_2 \rangle_s$$

(5)

In a similar manner we have that

$$[H_1 \circ M_1, H_2 \circ M_2]_n$$

$$= \sum_{s=1}^n \Delta(H_1 \circ M_1)_s \circ \Delta(H_2 \circ M_2)_s$$

$$= \sum_{s=1}^n H_{1s} \Delta M_{1s} \circ H_{2s} \circ \Delta M_{2s}$$

$$= \sum_{s=1}^n H_{1s} H_{2s} \Delta [M_1, M_2]_s$$

### Exercise 2.10

For two martingales  $M_1$  and  $M_2$   
we have

$$(*) \quad \langle M_1 + M_2 \rangle = \langle M_1 \rangle + \langle M_2 \rangle + 2 \langle M_1, M_2 \rangle$$

and

$$(**) \quad \{M_1 + M_2\} = \{M_1\} + \{M_2\} + 2 \{M_1, M_2\}$$

These results may be proved from

the definition of predictable and optional variation processes, cf (2.19) and (2.20) and the discussion on the top of page 50.

We then consider two counting processes  $N_1$  and  $N_2$  that have no simultaneous jump. The corresponding martingales are  $M_1$  and  $M_2$ . Then  $N_1 + N_2$  is a counting process (since all jumps have size +1) and it follows from (2.42) that

$$\begin{aligned} [M_1 + M_2]_t &= N_1(t) + N_2(t) \\ &= [M_1]_t + [M_2]_t. \end{aligned}$$

Using (\*\*) we then obtain

$$[M_1, M_2]_t = 0$$

⑦

Next the counting process

$N_1 + N_2$  has intensity process

$\lambda_1(t) + \lambda_2(t)$ , where  $\lambda_j(t)$  is the intensity process of  $N_j(t)$ ;  $j=1,2$ .

Hence

$$\langle M_1 + M_2 \rangle(t) = \int_0^t (\lambda_1(u) + \lambda_2(u)) du,$$

$$= \int_0^t \lambda_1(u) du + \int_0^t \lambda_2(u) du$$

$$= \langle M_1 \rangle(t) + \langle M_2 \rangle(t)$$

Using (\*) we then obtain

$$\langle M_1, M_2 \rangle(t) = 0$$

(8)

## Exercise 2.11

This exercise generalizes the results (2.36) and (2.37) to inhomogeneous Poisson processes.

$N(t)$  is an inhomogeneous Poisson process with intensity  $\lambda(t)$  (a deterministic function). Then  $N(t) - N(s)$  is Poisson distributed with parameter  $\int_s^t \lambda(u) du$ , and  $N(t) - N(s)$  is independent of  $N(s)$ . We introduce

$$M(t) = N(t) - \int_0^t \lambda(u) du$$

a) We have that ( $t > s$ )

$$\begin{aligned} E(M(t) | \mathcal{F}_s) &= E(N(t) - \int_0^t \lambda(u) du | N(s)) \\ &= E(N(s) + N(t) - N(s) | N(s)) - \int_0^t \lambda(u) du \\ &= N(s) + E(N(t) - N(s)) - \int_0^t \lambda(u) du \\ &= N(s) + \int_s^t \lambda(u) du - \int_0^t \lambda(u) du \\ &= N(s) - \int_0^s \lambda(u) du = M(s) \end{aligned}$$

(9)

$$\begin{aligned}
 f) & E \left\{ M^2(t) - \int_0^t \lambda(u) du \mid F_s \right\} \\
 &= E \left\{ (N(t) - \int_0^t \lambda(u) du)^2 \mid N(s) \right\} - \int_0^t \lambda(u) du \\
 &= E \left\{ N^2(t) - 2N(t) \int_0^t \lambda(u) du + (\int_0^t \lambda(u) du)^2 \mid N(s) \right\} \\
 &\quad - \int_0^t \lambda(u) du \\
 &= E \left\{ (N(s) + N(t) - N(s))^2 \mid N(s) \right\} \\
 &\quad - 2 \int_0^t \lambda(u) du \cdot E \left\{ N(s) + N(t) - N(s) \mid N(s) \right\} \\
 &\quad + (\int_0^t \lambda(u) du)^2 - \int_0^t \lambda(u) du \\
 &= N^2(s) + 2N(s) E \left\{ N(t) - N(s) \right\} + E \left\{ (N(t) - N(s))^2 \right\} \\
 &\quad - 2 \int_0^t \lambda(u) du \cdot N(s) - 2 \int_0^t \lambda(u) du \cdot E \left\{ N(t) - N(s) \right\} \\
 &\quad + (\int_0^t \lambda(u) du)^2 - \int_0^t \lambda(u) du \\
 &= N^2(s) + 2N(s) \int_s^t \lambda(u) du + \text{Var}(N(t) - N(s)) + (E(N(t) - N(s)))^2 \\
 &\quad - 2N(s) \int_0^t \lambda(u) du - 2 \int_0^t \lambda(u) du \cdot \int_s^t \lambda(u) du \\
 &\quad + (\int_0^t \lambda(u) du)^2 - \int_0^t \lambda(u) du
 \end{aligned}$$

⑩

$$= N^2(s) + 2N(s) \int_s^t \lambda(u) du + \int_s^t \lambda(u) du + (\int_s^t \lambda(u) du)^2$$

$$- 2N(s) \int_0^t \lambda(u) du - 2 \int_0^t \lambda(u) du \int_s^t \lambda(u) du + (\int_0^t \lambda(u) du)^2 - \int_0^t \lambda(u) du$$

$$= N^2(s) - 2N(s) \int_0^s \lambda(u) du + (\int_0^t \lambda(u) du - \int_s^t \lambda(u) du)^2 - \int_0^s \lambda(u) du$$

$$= N^2(s) - 2N(s) \int_0^s \lambda(u) du + (\int_0^s \lambda(u) du)^2 - \int_0^s \lambda(u) du$$

$$= M^2(s) - \int_0^s \lambda(u) du$$

Thus  $M^2(t) - \int_0^t \lambda(u) du$  is a martingale