

## B5440 – Exercise 2, Processes and Martingales – Solutions

1. Answer and prove the following:

a. Let  $X_1, X_2, \dots$  be independent with mean 0. Is  $M_n = X_1 + \dots + X_n$  a martingale?

**Solution:** We check the definition:

$$\begin{aligned} E(M_n | \mathcal{F}_{n-1}) &= E(X_1 + \dots + X_n | X_1 + \dots + X_{n-1}) = \\ &E(X_n) + X_1 + \dots + X_{n-1} = 0 + M_n. \end{aligned}$$

Yes, it is a martingale.

b. Let  $X_1, X_2, \dots$  be independent with mean  $\mu$ . Is  $S_n = X_1 + \dots + X_n$  a martingale?

**Solution:** We check the definition:

$$\begin{aligned} E(M_n | \mathcal{F}_{n-1}) &= E(X_1 + \dots + X_n | X_1 + \dots + X_{n-1}) = \\ &E(X_n) + X_1 + \dots + X_{n-1} = \mu + M_n. \end{aligned}$$

So if  $\mu \neq 0$ , then it is **not** a martingale.

c. Let  $X_1, X_2, \dots$  be independent with mean 1. Is  $M_n = X_1 \cdot X_2 \cdots X_n$  a martingale?

**Solution:** We check the definition:

$$\begin{aligned} E(M_n | \mathcal{F}_{n-1}) &= E(X_1 \cdots X_n | X_1 \cdots X_{n-1}) = \\ &E(X_n) \cdot (X_1 \cdots X_{n-1}) = 1 \cdot M_n. \end{aligned}$$

Yes, it is a martingale.

2. Show that if  $M$  is a martingale then so is the stopped process  $M^T$ .

**Solution:** First in the discrete time case, if  $M_n$  is a martingale then the stopped process  $M_n^T = M_n$  if  $n \leq T$  and  $= M_T$  if  $n > T$ . We will use the fact that transformations of martingales by predictable processes are also martingales. So we need to find a transformation  $H_n$  that is predictable at time  $n - 1$  such that  $M^T = H \bullet M$ . It is  $H_n = 1$  if  $n \leq T$  and  $H_n = 0$  if  $n > T$ . This is predictable at time  $n - 1$  because “today” we know whether we have reached  $T$  or not, thus “tomorrow’s” value of  $H$  is known.

In continuous time we have

$$M^T(t) = \int_0^{t \wedge T} dM(u) = \int_0^t 1\{u \leq T\} dM(u)$$

which is a stochastic integral. The transformation  $H(t) = 1\{t \leq T\}$  is predictable with respect to  $\mathcal{F}_{t-}$  because it is adapted and it is left continuous. Therefore the stopped process is also a martingale.

### 3. Poisson process compensator:

Let  $N(t)$  be the number of events in  $[0, t]$  where  $N(t) - N(s) \sim \text{Poisson}((t - s)\lambda)$  for  $s < t$  and  $N$  has independent increments.

Does the Doob-Meyer decomposition apply to  $N(t)$ ? If so, identify the compensator of  $N(t)$ . Also find the predictable variation process.

### Solution

Problem statement: Find the unique predictable process  $X(t)$  such that  $N(t) - X(t)$  is a martingale.

### Steps (WAG method):

1. I think  $X(t) = \lambda t$ .
2. Is  $\lambda t$  predictable? Yes, it is left-continuous because

$$\lim_{s \uparrow t} \lambda s = \lambda t$$

and all left-continuous processes are predictable.

3. Is  $N(t) - \lambda t$  a martingale? Let's check the definition: for  $s < t$ , we have

$$\begin{aligned} E(N(t) - \lambda t | \mathcal{F}_s) &= E(N(t) - N(s) + N(s) - \lambda t | \mathcal{F}_s) = \\ &= E(N(t) - N(s) | \mathcal{F}_s) + E(N(s) | \mathcal{F}_s) - \lambda t = \\ &= E(N(t) - N(s)) + N(s) - \lambda t = \\ &= (t - s)\lambda + N(s) - \lambda t = N(s) - \lambda s. \end{aligned}$$

### Steps (start at the end method):

1.  $N(t) - X(t)$  is a martingale iff  $E(N(t) - X(t) | \mathcal{F}_s) = N(s) - X(s)$  for  $s < t$  and a predictable process  $X$ . Hence,

$$\begin{aligned} E(N(t) - X(t) - N(s) + X(s) | \mathcal{F}_s) &= 0 \Leftrightarrow \\ E((N(t) - N(s)) + (X(s) - X(t)) | \mathcal{F}_s) &= 0 \Leftrightarrow \\ E(N(t) - N(s) | \mathcal{F}_s) = E(X(t) - X(s) | \mathcal{F}_s) &\Leftrightarrow \\ (t - s)\lambda = E(X(t) - X(s) | \mathcal{F}_s). \end{aligned}$$

Since  $X(t)$  is predictable, we must have  $E(X(t) - X(s) | \mathcal{F}_s) = X(t) - X(s)$ . If not, then we could find a  $t^*$  such that  $E(X(t^*) | \mathcal{F}_{t^*-}) \neq X(t^*)$ , which would violate the definition of predictability. The result follows.

## Predictable and optional variation processes

We just showed that  $M(t) = N(t) - \lambda t$  is a martingale. Since the Poisson distribution has the same mean and variance, we can guess that  $\langle M \rangle = \lambda t$ .

By the Doob Meyer, and the fact that  $\lambda t$  is predictable, we have to show that  $M^2(t) - \lambda t$  is a martingale. We do this by checking the definition:

$$\begin{aligned} E((N(t) - \lambda t)^2 - \lambda t | \mathcal{F}_s) &= \\ E(N^2(t) - 2N(t)\lambda t + \lambda^2 t^2 - \lambda t | \mathcal{F}_s). \end{aligned}$$

Above we showed that  $E(N(t) | \mathcal{F}_s) = N(s) + \lambda(t - s)$  by adding and subtracting  $N(s)$ , so we have now

$$\begin{aligned} E(N^2(t) | \mathcal{F}_s) - 2(N(s) + \lambda(t - s))\lambda t + \lambda^2 t^2 - \lambda t &= \\ E(N^2(t) | \mathcal{F}_s) - 2N(s)\lambda t - 2\lambda^2 t^2 + 2\lambda^2 st + \lambda^2 t^2 - \lambda t &= \\ E(N^2(t) | \mathcal{F}_s) - 2N(s)\lambda t - \lambda^2 t^2 + 2\lambda^2 st - \lambda t. \end{aligned}$$

Looking now at the first term, it equals

$$\begin{aligned} E((N(s) + N(t) - N(s))^2 | \mathcal{F}_s) &= \\ N(s)^2 + 2N(s)E(N(t) - N(s)) + E((N(t) - N(s))^2) &= \\ N(s)^2 + 2N(s)\lambda(t - s) + Var(N(t) - N(s)) + (E(N(t) - N(s)))^2 &= \\ N(s)^2 + 2N(s)\lambda(t - s) + \lambda(t - s) + \lambda^2(t - s)^2. \end{aligned}$$

Plugging back into the last line of the previous display:

$$\begin{aligned} N(s)^2 + 2N(s)\lambda(t - s) + \lambda(t - s) + \lambda^2(t - s)^2 - 2N(s)\lambda t - \lambda^2 t^2 + 2\lambda^2 st - \lambda t &= \\ N(s)^2 - 2N(s)\lambda s - \lambda s + \lambda^2(t - s)^2 - \lambda^2 t^2 + 2\lambda^2 st &= \\ N(s)^2 - 2N(s)\lambda s - \lambda s - \lambda^2 s^2 &= \\ (N(s) - \lambda s)^2 - \lambda s, \end{aligned}$$

which shows that the martingale property is satisfied.

What about  $[M](t)$ ? For this we use the definition of the optional variation process, and observe that it equals

$$\begin{aligned} \sum_{s \leq t} (M(s) - M(s-))^2 &= \sum_{s \leq t} (N(s) - N(s-) + \lambda s - \lambda s)^2 = \\ &= \sum_{s \leq t} (N(s) - N(s-))^2. \end{aligned}$$

Now if there is a jump at  $s$  the term in the sum equals 1, and if not, it equals 0. So this sum counts the number of events that occurred up to  $t$ , in other words it equals  $N(t)$  itself. This is true for any counting process.