

STK4080 Survival and event history analysis, UiO 2019

# **Slides 5: Counting processes and martingales**

**SOLUTIONS TO EXERCISES**

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## EXERCISE 1

Throw a die several times. Let  $Y_i$  be result in  $i$ th throw, and let  $X_i = Y_1 + \dots + Y_i$  be the sum of the  $i$  first throws.

**(a)** Find an expression for  $f(x_1, x_2)$ .

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$$f(x_1, x_2) = P(X_1 = x_1, X_2 = x_2) = \frac{1}{36}I(1 \leq x_1 \leq 6, 1 + x_1 \leq x_2 \leq 6 + x_1)$$

**(b)** Find an expression for  $f(x_2|x_1)$ .

\*\*\*\*\*

$$f(x_2|x_1) = \frac{f(x_1, x_2)}{f(x_1)} = \frac{1}{6}I(1 + x_1 \leq x_2 \leq 6 + x_1)$$

**(c)** Find an expression for  $E(X_2|x_1)$  as a function of  $x_1$ .

\*\*\*\*\*

$$E(X_2|x_1) = x_1 + 7/2$$

- (d) Find the expectation of the random variable  $E(X_2|X_1)$ . Find  $E(X_2)$  from this. Find also  $E(X_2)$  without using the rule of double expectation.

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$$E(X_2|X_1) = X_1 + 7/2$$

$$E(X_2) = E(E(X_2|X_1)) = E(X_1) + 7/2 = 7$$

or

$$E(X_2) = E(Y_1 + Y_2) = E(Y_1) + E(Y_2) = 7/2 + 7/2 = 7$$

- (e) Find also  $E(X_3|X_2)$ ,  $E(X_3|X_1, X_2)$  and  $E(X_3|X_1)$ .

\*\*\*\*\*

$$E(X_3|X_2) = X_2 + 7/2$$

$$E(X_3|X_1, X_2) = X_2 + 7/2$$

$$E(X_3|X_1) = X_1 + 7$$

## EXERCISE 2

Let  $X_1, X_2, \dots$  be independent with  $E(X_n) = \mu$  for all  $n$ .

Let  $S_n = X_1 + \dots + X_n$ .

Find  $E(S_4|\mathcal{F}_2)$  and in general  $E(S_n|\mathcal{F}_m)$  when  $m < n$

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$$\begin{aligned} E(S_4|\mathcal{F}_2) &= E(X_1 + X_2 + X_3 + X_4|X_1, X_2) \\ &= X_1 + X_2 + E(X_3) + E(X_4) \\ &= X_1 + X_2 + 2\mu \end{aligned}$$

$$E(S_n|\mathcal{F}_m) = S_m + (n - m)\mu$$

### EXERCISE 3

Let  $X_1, X_2, \dots$  be independent with

$$P(X_i = 1) = P(X_i = -1) = 1/2$$

Think of  $X_i$  as result of a game where one flips a coin and wins 1 unit if “heads” and loses 1 unit if “tails”.

Let  $M_n = X_1 + \dots + X_n$  be the gain after  $n$  games,  $n = 1, 2, \dots$

Show that  $M_n$  is a (mean zero) martingale.

\*\*\*\*\*

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

## EXERCISE 4

**(a)** Let  $X_1, X_2, \dots$  be independent with  $E(X_n) = 0$  for all  $n$ .

Let  $M_n = X_1 + \dots + X_n$ .

Show that  $M_n$  is a (mean zero) martingale.

\*\*\*\*\*

Same proof as for Exercise 3

**(b)** Let  $X_1, X_2, \dots$  be independent with  $E(X_n) = \mu$  for all  $n$ .

Let  $S_n = X_1 + \dots + X_n$ .

Compute  $E(S_n | \mathcal{F}_{n-1})$  (see Exercise 2). Is  $\{S_n\}$  a martingale?

Can you find a simple transformation of  $S_n$  which is a martingale?

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$$E(S_n | \mathcal{F}_{n-1}) = S_{n-1} + \mu$$

This is a martingale only if  $\mu = 0$ .

Easy to see that  $M_n = S_n - n\mu$  is a martingale.

(c) Let  $X_1, X_2, \dots$  be independent with  $E(X_n) = 1$  for all  $n$ .

Let  $M_n = X_1 \cdot X_2 \cdot \dots \cdot X_n$ .

Show that  $M_n$  is a martingale. What is  $E(M_n)$ ?

\*\*\*\*\*

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

$$E(M_n) = 1$$

## EXERCISE 5

Show that for a martingale  $\{M_n\}$  is

**(a)**  $E(M_n|\mathcal{F}_m) = M_m$  for all  $m < n$ .

This is essentially Exercise 2.1 in book.

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We use induction on  $k$  to prove that

$$(*) \quad E(M_n|\mathcal{F}_{n-k}) = M_{n-k} \text{ for } k = 1, 2, \dots, n-1$$

$k = 1$  is definition of martingale.

Suppose  $(*)$  holds for  $k$ . Then

$$E(M_n|\mathcal{F}_{n-k-1}) = E\{E(M_n|\mathcal{F}_{n-k})|\mathcal{F}_{n-k-1}\} = E(M_{n-k}|\mathcal{F}_{n-k-1}) = M_{n-k-1}$$

Hence  $(*)$  holds for all the required  $k$ .

**(b)**  $E(M_m|\mathcal{F}_n) = M_m$  for all  $m < n$

\*\*\*\*\*

This holds trivially since  $M_m \in \mathcal{F}_n$

**(c)** Give verbal (intuitive) interpretations of each of (a) and (b).



## EXERCISE 6

Define the *martingale differences* by

$$\Delta M_n = M_n - M_{n-1}$$

- (a) Show that the definition of martingale,  $E(M_n|\mathcal{F}_{n-1}) = M_{n-1}$ , is equivalent to

$$E(M_n - M_{n-1}|\mathcal{F}_{n-1}) = 0, \text{ i.e. } E(\Delta M_n|\mathcal{F}_{n-1}) = 0 \quad (1)$$

(Hint: Why is  $E(M_{n-1}|\mathcal{F}_{n-1}) = M_{n-1}$ ?)

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The following are clearly equivalent by 'Hint' which is obvious:

$$\begin{aligned} E(M_n|\mathcal{F}_{n-1}) &= M_{n-1} \\ E(M_n|\mathcal{F}_{n-1}) - M_{n-1} &= 0 \\ E(M_n|\mathcal{F}_{n-1}) - E(M_{n-1}|\mathcal{F}_{n-1}) &= 0 \\ E(M_n - M_{n-1}|\mathcal{F}_{n-1}) &= 0 \end{aligned}$$

**(b)** Show that for a martingale we have

$$\text{Cov}(M_{n-1} - M_{n-2}, M_n - M_{n-1}) = 0 \text{ for all } n \quad (2)$$

i.e.

$$\text{Cov}(\Delta M_{n-1}, \Delta M_n) = 0$$

Explain in word what this means.

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$$\begin{aligned} \text{Cov}(M_{n-1} - M_{n-2}, M_n - M_{n-1}) &= E\{(M_{n-1} - M_{n-2})(M_n - M_{n-1})\} \\ &= E[E\{(M_{n-1} - M_{n-2})(M_n - M_{n-1})|\mathcal{F}_{n-1}\}] \\ &= E[(M_{n-1} - M_{n-2})E\{(M_n - M_{n-1})|\mathcal{F}_{n-1}\}] \\ &= 0 \end{aligned}$$

**(c)** Show that (1) and (2) automatically hold when  $M_n = X_1 + \dots + X_n$  for independent  $X_1, X_2, \dots$

Note that in this case the differences  $\Delta M_n = M_n - M_{n-1} = X_n$  are independent.

Thus (2) shows that martingale differences correspond to a weakening of the independent increments property

## EXERCISE 7

Suppose that you use the martingale betting strategy, but that you in any case must stop after  $n$  games.

Calculate the expected gain. Can you “beat the game”?

*(Hint: You either win 1 or lose  $2^n - 1$  units.)*

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You lose all  $n$  times with probability  $1/2^n$ . The loss is then  $1 + 2 + 4 + \dots + 2^{n-1} = 2^n - 1$ .

You otherwise win 1, which must have probability  $1 - 1/2^n$ . Hence expected gain is:

$$-(2^n - 1) \cdot (1/2^n) + 1 \cdot (1 - 1/2^n) = -1 + 1/2^n + 1 - 1/2^n = 0$$

so you do not “beat the game”

## EXERCISE 8

Do Exercise 2.6 in book:

Show that the stopped process  $M^T$  is a martingale. (*Hint: Find a predictable process  $H$  such that  $M^T = H \bullet M$* ).

\*\*\*\*\*

We have

$$M_n^T = M_{n \wedge T} \equiv \begin{cases} M_n & \text{if } n \leq T \\ M_T & \text{if } n > T \end{cases}$$

Want to write

$$(*) \quad M_n^T = H_1(M_1 - M_0) + H_2(M_2 - M_1) + \dots + H_n(M_n - M_{n-1})$$

Let

$$H_i = \begin{cases} 1 & \text{if } T \geq i \\ 0 & \text{otherwise} \end{cases}$$

Then  $H_i$  is known at time  $i - 1$  and we may check that (\*) does the job.

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This suggests that  $E(M_T) \equiv E(M_T^T) = E(M_0^T) = E(M_0) = 0$ .

But this does obviously not hold for the martingale strategy case (where indeed the process is a martingale, stopped at a stopping time).

What is the clue here? *We can conclude from the above that  $E(M_n^T) = 0$  for all  $n$ , but not that this  $n$  can be replaced by a random  $T$ . But see next slide in lecture!*

## EXERCISE 9

Suppose  $X_n = U_1 + \dots + U_n$  where  $U_1, \dots, U_n$  are independent with  $E(U_i) = \mu$ .

Find the Doob decomposition of the process  $X$  and identify the predictable part and the innovation part.

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Recall the Doob decomposition:

$$X_n = E(X_n | \mathcal{F}_{n-1}) + (X_n - E(X_n | \mathcal{F}_{n-1})) = \text{predictable} + \text{innovation}$$

Here the predictable part is:

$$E(X_n | \mathcal{F}_{n-1}) = E(X_{n-1} + U_n | \mathcal{F}_{n-1}) = X_{n-1} + \mu$$

and the innovation is hence

$$X_n - (X_{n-1} + \mu) = X_n - X_{n-1} - \mu = U_n - \mu$$

## EXERCISE 10

Prove the results on the covariances (see Exercise 2.2 in book)

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Let  $0 \leq s < t < u < v \leq \tau$

$$\begin{aligned} \text{Cov}(M(t) - M(s), M(v) - M(u)) &= E\{(M(t) - M(s))(M(v) - M(u))\} \\ &= E[E\{(M(t) - M(s))(M(v) - M(u)) | \mathcal{F}_u\}] \\ &= E[(M(t) - M(s))E\{M(v) - M(u) | \mathcal{F}_u\}] \\ &= 0 \end{aligned}$$

## POISSON PROCESSES

$N(t) = \#$  events in  $[0, t]$

Characterizing properties:

- $N(t) - N(s)$  is Poisson-distributed with parameter  $\lambda(t - s)$
- $N(t)$  has independent increments, i.e. number of events in disjoint intervals are independent.

### EXERCISE 11

Show that

$$M(t) = N(t) - \lambda t$$

is a martingale. Identify the compensator of  $N(t)$ .

*Hint:*

For  $t > s$  is  $E(N(t)|\mathcal{F}_s) = N(s) + \lambda(t - s)$

\*\*\*\*\*

The compensator must be  $\lambda t$  by Doob-Meyer (and  $\lambda t$  is increasing, predictable).

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



## EXERCISE 12

This is Exercise 2.3 in book, plus a new question (d). (a) was done as our Exercise 3(a)

Thus let  $X_1, X_2, \dots$  be independent with  $E(X_n) = 0$  and  $\text{Var}(X_n) = \sigma^2$  for all  $n$ . Let  $M_n = X_1 + \dots + X_n$ .

**(a)** Show that  $M_n$  is a (mean zero) martingale.

**(b)** Compute  $\langle M \rangle_n$

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$$\langle M \rangle_n = \sum_{i=1}^n \text{Var}(\Delta M_i | \mathcal{F}_{i-1})$$

Here  $\Delta M_i = M_i - M_{i-1} = X_i$ , so we get

$$\langle M \rangle_n = \sum_{i=1}^n \text{Var}(\Delta M_i | \mathcal{F}_{i-1}) = \sum_{i=1}^n \text{Var}(X_i | \mathcal{F}_{i-1}) = \sum_{i=1}^n \text{Var}(X_i) = n\sigma^2$$

since  $X_i$  is independent of  $\mathcal{F}_{i-1}$ . (Why?)

(c) Compute  $[M]_n$

\*\*\*\*\*

$$[M]_n = \sum_{i=1}^n (\Delta M_i)^2 = \sum_{i=1}^n X_i^2$$

(d) Compute  $E \langle M \rangle_n$ ,  $E [M]_n$  and  $\text{Var}(M_n)$ .

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$$E \langle M \rangle_n = E [M]_n = \text{Var}(M_n) = n\sigma^2$$

(and these are equal in general, see next exercise)

## EXERCISE 13

Consider a general discrete time martingale  $M$

**(a)**  $M_n^2 - \langle M \rangle_n$  is a mean zero martingale - Exercise 2.4 in book

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Write

$$\begin{aligned} M_n^2 &= (M_{n-1} + M_n - M_{n-1})^2 \\ &= M_{n-1}^2 + 2M_{n-1}(M_n - M_{n-1}) + (M_n - M_{n-1})^2 \end{aligned}$$

Thus

$$\begin{aligned} &E\{M_n^2 - \langle M \rangle_n | \mathcal{F}_{n-1}\} \\ &= E\{M_{n-1}^2 + 2M_{n-1}(M_n - M_{n-1}) + (M_n - M_{n-1})^2 - \langle M \rangle_n | \mathcal{F}_{n-1}\} \\ &= M_{n-1}^2 + 2M_{n-1}E\{(M_n - M_{n-1}) | \mathcal{F}_{n-1}\} + E\{(M_n - M_{n-1})^2 | \mathcal{F}_{n-1}\} \\ &\quad - \sum_{i=1}^n E\{(M_i - M_{i-1})^2 | \mathcal{F}_{i-1}\} \\ &= M_{n-1}^2 + 2 \cdot 0 - \sum_{i=1}^{n-1} E\{(M_i - M_{i-1})^2 | \mathcal{F}_{i-1}\} \\ &= M_{n-1}^2 - \langle M \rangle_{n-1} \end{aligned}$$

**(b)**  $M_n^2 - [M]_n$  is a mean zero martingale - See book p. 45

**(c)** Use (a) and (b) to prove that

$$\text{Var}(M_n) = E \langle M \rangle_n = E [M]_n$$

What is the intuitive essence of this result?

## EXERCISE 14

Consider again the Poisson process, where we have shown that

$$M(t) = N(t) - \lambda t$$

is a martingale.

Prove now that:

$$M^2(t) - \lambda t$$

is a martingale (Exercise 2.10 in book)

Then prove that  $\langle M \rangle (t) = \lambda t$

What is  $[M] (t)$ ?

*Hint:*

For  $t > s$  is  $E(N(t)|\mathcal{F}_s) = N(s) + \lambda(t - s)$

For  $t > s$  is  $E(N^2(t)|\mathcal{F}_s) = N(s)^2 + 2N(s)\lambda(t - s) + \lambda(t - s) + (\lambda(t - s))^2$

## EXERCISE 15

Do Exercise 2.13.

## EXERCISE 16

Do Exercise 2.5:

$M_1M_2 - \langle M_1, M_2 \rangle$  is a mean zero martingale

## EXERCISE 17

Do Exercise 2.5:

$M_1M_2 - [M_1, M_2]$  is a mean zero martingale

(Hint: We know that  $M^2 - \langle M \rangle$  is a mean zero martingale. Use that  $M_1M_2 = (1/4) [(M_1 + M_2)^2 - (M_1 - M_2)^2]$ )

## EXERCISE 18

Prove also (2.28) in book.

Do Exercise 2.11 in book.