

# Stochastic analysis, 8. exercises

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**Exercise 1** Let  $\tau(\omega) \in [0, +\infty]$  be a random time,  $F(t) = P(\tau \leq t)$  for  $t \in [0, \infty)$ .

Consider the single jump counting process  $N_t = \mathbf{1}(\tau(\omega) \leq t)$  which generates the filtration  $\mathbb{F} = (\mathcal{F}_t)$  with  $\mathcal{F}_t^N = \sigma(N_s : s \leq t)$ .

(a) Show that  $\tau$  is a stopping time in the filtration  $\mathbb{F}$ .

(b) Show first that for every Borel function  $f(x)$ , the random variable

$$f(\tau(\omega))\mathbf{1}(\tau(\omega) \leq s)$$

is  $\mathcal{F}_s$ -measurable.

(c) Define the *cumulative hazard function*

$$\Lambda(t) = \int_0^t \frac{1}{1 - F(s-)} F(ds),$$

where  $F(-s) = P(\tau < s)$  denotes the limit from the left. Show that

$$M_t = N_t - \Lambda_{t \wedge \tau}$$

is an  $\mathbb{F}$ -martingale.

(d) Assume that  $t \mapsto F(t)$  and therefore also  $t \mapsto \Lambda(t)$  are continuous, which means  $P(\tau = t) = 0$  for all  $t \in \mathbb{R}^+$ . Show that  $\Lambda_\tau$  has 1-exponential distribution:

$$P(\Lambda_\tau > x) = \exp(-x), \quad x \geq 0$$

(e) Show that the martingale  $M_t$  is uniformly integrable, what is  $M_\infty$ ?

**Solution 1** (a) We have  $\{\tau(\omega) \leq s\} = N_s^{-1}\{1\} \in \mathcal{F}_s^N$ , so  $\tau$  is a stopping time.

(b) We notice that when  $f$  is a simple function of the form  $f(x) = \sum_{k=1}^n c_k \mathbf{1}(a_k, b_k)$ , the claim holds, since

$$f(\tau(\omega))\mathbf{1}(\tau(\omega) \leq s) = \sum_{k=1}^n c_k \mathbf{1}(\tau(\omega) \in (a_k, b_k) \cap [0, s])$$

and every set of the form  $\{\tau \in (a, b) \cap [0, s]\}$  is  $\mathcal{F}_s$  measurable. The rest follows by a standard limiting argument.

(c) Notice first that  $M_t$  is integrable, since

$$\int_{\Omega} \Lambda_{t \wedge \tau} dP = \int_{\Omega} \int_0^{t \wedge \tau} \frac{1}{1 - F(s-)} F(ds) dP \leq \int_0^{\infty} \int_{\{\tau \geq s\}} \frac{1}{1 - F(s-)} dP F(ds) = \int_0^{\infty} \mathbf{1} F(ds) = 1.$$

We have

$$E(M_t - M_s | \mathcal{F}_s) = E(N_t - \int_0^{t \wedge \tau} \frac{1}{1 - F(u-)} F(du) - N_s + \int_0^{s \wedge \tau} \frac{1}{1 - F(u-)} F(du) | \mathcal{F}_s) = 0.$$

Thus it is enough to show that for any  $A \in \mathcal{F}_s$  we have

$$\int_A \mathbf{1}(s < \tau \leq t) dP = \int_A \int_{(s \wedge \tau, t \wedge \tau]} \frac{1}{1 - F(u-)} F(du) dP.$$

Notice that if  $\omega \in A$  is such that  $\tau(\omega) \leq s$ , then

$$\mathbf{1}(s < \tau \leq t) = 0 = \int_{(s \wedge \tau(\omega), t \wedge \tau(\omega)]} \frac{1}{1 - F(u-)} F(du).$$

Therefore we may assume that  $\tau(\omega) > s$  for all  $\omega \in A$ , but this implies that  $A = \{\omega \in \Omega : \tau(\omega) > s\}$  (or the empty set, but that is a trivial case). Now

$$\begin{aligned} \int_A \int_{(s \wedge \tau, t \wedge \tau]} \frac{1}{1 - F(u-)} F(du) dP &= \int_{\Omega} \int_{(s, t \wedge \tau]} \frac{\mathbf{1}(\tau(\omega) > s)}{1 - F(u-)} F(du) dP(\omega) \\ &= \int_{(s, t]} \int_{\{\tau \geq u\}} \frac{1}{1 - F(u-)} dP F(du) \\ &= \int_{(s, t]} \frac{P(\tau \geq u)}{1 - P(\tau < u)} F(du) \\ &= \int_{(s, t]} F(du) = F(t) - F(s+) = P(\tau \leq u) - P(\tau \leq s) \\ &= P(s < \tau \leq u) = \int_A \mathbf{1}(s < \tau \leq t). \end{aligned}$$

(d) We compute the Laplace transform.

$$\begin{aligned} \int_{\Omega} e^{-\theta \Lambda_{\tau}} dP &= \int_{\Omega} e^{-\theta \int_0^{\tau(\omega)} \frac{1}{1 - F(s)} F(ds)} dP(\omega) = \int_0^{\infty} e^{-\theta \int_0^t \frac{1}{1 - F(s)} F(ds)} F(dt) \\ &= \int_0^{\infty} e^{\theta \log(1 - F(t))} F(dt) = \int_0^{\infty} (1 - F(t))^{\theta} F(dt) = \frac{1}{\theta + 1}, \end{aligned}$$

which is the same as the Laplace transform of a random variable with pdf  $\exp(-x)$ ,

$$\int_0^{\infty} e^{-\theta x} \exp(-x) dx = \int_0^{\infty} e^{-(\theta+1)x} dx = \frac{1}{\theta + 1}.$$

(e) The martingale is uniformly integrable, since  $M_t \leq 1 + \Lambda_{\tau}$ , which is integrable. Moreover,  $M_t > 0$ , so  $M_t$  converges and  $E(M_{\infty}) = E(M_0) = 0$ .

**Exercise 2** Let  $(M_t : t \in \mathbb{R}^+)$  be a  $\mathbb{F}$ -martingale, and  $\mathbb{G}$  a filtration with  $\mathcal{G}_t \subset \mathcal{F}_t$ . We assume that  $(M_t)$  is also  $\mathbb{G}$ -adapted. Show that  $(M_t)$  is a martingale in the smaller filtration  $\mathbb{G}$ .

**Solution 2** We have

$$E(M_t | \mathcal{G}_s) = E((M_t | \mathcal{F}_s) | \mathcal{G}_s) = E(M_s | \mathcal{G}_s) = M_s.$$

**Exercise 3** Let  $(M_t : t \in \mathbb{R})$  be a  $\mathbb{F}$ -martingale under  $P$ , and  $\mathcal{G}_t$  a filtration such that  $\forall t \geq 0$ , the  $\sigma$ -algebrae  $\mathcal{G}_t$  and  $\sigma(M_s : s \leq t)$  are  $P$ -independent. Show that under  $P$ ,  $(M_t : t \in \mathbb{R}^+)$  is a martingale in the enlarged filtration  $(\mathcal{F}_t \vee \mathcal{G}_t : t \geq 0)$ .

**Solution 3** *Solution missing.*

Let  $(B_t : t \geq 0)$  be a Brownian motion in the filtration  $\mathbb{F}$ , which means

- $B_0(\omega) = 0$
- $t \mapsto B_t(\omega)$  is continuous
- $\forall 0 \leq s \leq t$ ,  $(B_t - B_s)$  is  $P$ -independent from  $\mathcal{F}_s$ , conditionally gaussian with conditional mean  $E(B_t - B_s | \mathcal{F}_s) = 0$  and conditional variance  $E((B_t - B_s)^2 | \mathcal{F}_s) = t - s$ .

**Exercise 4** Show that for  $a > 0$  the process  $(a^{-1/2}B_{at} : t \in \mathbb{R}^+)$  is also a Brownian motion.

**Solution 4** Notice that it is a Brownian motion w.r.t. the filtration  $\mathcal{F}_{at}$ !

Clearly at  $t = 0$  we have  $a^{-1/2}B_{at} = 0$ . The continuity is also satisfied. Finally,  $a^{-1/2}B_{at} - a^{-1/2}B_{as}$  is clearly independent from  $\mathcal{F}_{as}$  and its conditional probability density is just

$$a^{1/2} \frac{1}{\sqrt{2\pi(at - as)}} e^{-\frac{(a^{-1/2}x)^2}{2(at - as)}} = \frac{1}{\sqrt{2\pi(t - s)}} e^{-\frac{x^2}{2(t - s)}},$$

which is what we needed to prove.

**Exercise 5** The process  $W_0 = 0$ ,  $W_t = tB_{1/t}$  is also a Brownian motion.

**Solution 5** We will show that  $W_t$  is a Brownian motion with respect to its own filtration  $\mathcal{G}_t = \sigma(W_s : s \leq t)$ .

First of all notice that  $W_t - W_s = tB_{1/t} - sB_{1/s}$  is independent of  $\mathcal{G}_s = \sigma(B_{1/r} : r \leq s)$ . We can use the 1st exercise of the 2nd exercise set to see that  $B_{1/t}$  has Gaussian conditional distribution with mean  $\frac{B_{1/s}/t}{1/s} = \frac{s}{t}B_{1/s}$  and variance  $\frac{(1/t)(1/s - 1/t)}{1/s} = \frac{t-s}{t^2}$ . Thus  $tB_{1/t}$  has Gaussian conditional distribution with mean  $sB_{1/s}$  and variance  $t - s$  and finally  $W_t - W_s$  has Gaussian conditional distribution with mean 0 and variance  $t - s$ , which is what we wanted to show.

**Exercise 6** Let  $\theta \in \mathbb{R}$ , and  $i = \sqrt{-1}$  be the imaginary unit.

Show that

$$E(\exp(i\theta B_t)) = \exp\left(-\frac{1}{2}\theta^2 t\right).$$

**Solution 6** We calculate

$$\begin{aligned} E(\exp(i\theta B_t)) &= \int_{-\infty}^{\infty} e^{i\theta x} \frac{1}{\sqrt{2\pi t}} e^{-\frac{x^2}{2t}} dx \\ &= \int_{-\infty}^{\infty} e^{-\left(\frac{x}{\sqrt{2t}} + i\theta\sqrt{\frac{t}{2}}\right)^2 - \frac{\theta^2 t}{2}} \frac{dx}{\sqrt{2\pi t}} \\ &= e^{-\frac{1}{2}\theta^2 t} \frac{1}{\sqrt{\pi}} \int_{(-\infty, \theta\sqrt{\frac{t}{2}})}^{(\infty, \theta\sqrt{\frac{t}{2}})} e^{-z^2} dy. \end{aligned}$$

Thus it is enough to show that

$$\int_{(-\infty, a)}^{(\infty, a)} e^{-z^2} dz = \sqrt{\pi}$$

for all  $a$ . We do this by contour integration. The function  $e^{-z^2}$  is analytic, so its integral over the rectangular path  $(-R, 0) \rightarrow (R, 0) \rightarrow (R, a) \rightarrow (-R, a) \rightarrow (-R, 0)$  is 0. Moreover

$$\int_{(R, 0)}^{(R, a)} e^{-z^2} dz \rightarrow 0, \quad \int_{(-R, a)}^{(-R, 0)} e^{-z^2} dz \rightarrow 0$$

as  $R \rightarrow \infty$ . Thus letting  $R \rightarrow \infty$ , we get that

$$\int_{(-\infty, a)}^{(\infty, a)} e^{-z^2} dz = \int_{-\infty}^{\infty} e^{-z^2} dz = \sqrt{\pi}.$$

**Exercise 7** For  $\theta \in \mathbb{R}$ , consider now

$$M_t = \exp(i\theta B_t + \frac{1}{2}\theta^2 t) = \left\{ \exp\left(\frac{1}{2}\theta^2 t\right) \cos(\theta B_t) + \sqrt{-1} \exp\left(\frac{1}{2}\theta^2 t\right) \sin(\theta B_t) \right\} \in \mathbb{C}$$

where  $i = \sqrt{-1}$  is the imaginary unit.

Recall that  $E(\exp(i\theta G)) = \exp(-\theta^2 \sigma^2 / 2)$  when  $G(\omega) \sim \mathcal{N}(0, \sigma^2)$ .

- Show that  $M_t$  is a complex valued  $\mathbb{F}$ -martingale, which means that real and imaginary parts are  $\mathbb{F}$ -martingales.
- Show that  $\lim_{t \rightarrow \infty} |M_t(\omega)| = \infty$ .

**Solution 7** Assuming  $\theta \neq 0$ , clearly  $\lim_{t \rightarrow \infty} |M_t(\omega)| = \lim_{t \rightarrow \infty} e^{\frac{1}{2}\theta^2 t} = \infty$ .

Now,  $M_t$  is integrable since  $|M_t| = e^{\frac{1}{2}\theta^2 t} < \infty$ . Also the martingale property is satisfied:

$$E(M_t | \mathcal{F}_s) = e^{\frac{1}{2}\theta^2 t + i\theta B_s} E(e^{i\theta(B_t - B_s)} | \mathcal{F}_s) = e^{i\theta B_s + \frac{1}{2}\theta^2 t} e^{-\frac{1}{2}\theta^2(t-s)} = e^{i\theta B_s + \frac{1}{2}\theta^2 s}$$