

**Exercise 3.1**

- (i) First, assume that the random variables  $|X_n|$  are uniformly bounded by a finite constant  $c$ . Then, for  $\varepsilon > 0$ ,

$$\begin{aligned} \mathbb{E}[|X_n - X|] &= \mathbb{E}[|X_n - X| \mathbb{1}_{\{|X_n - X| > \frac{\varepsilon}{2}\}}] + \mathbb{E}[|X_n - X| \mathbb{1}_{\{|X_n - X| \leq \frac{\varepsilon}{2}\}}] \\ &\leq 2c\mathbb{P}(|X_n - X| > \frac{\varepsilon}{2}) + \frac{\varepsilon}{2}. \end{aligned}$$

Here, we used that  $|X| \leq c$   $\mathbb{P}$ -a.s since  $|X_n| \leq c$  and  $(X_n)_{n \in \mathbb{N}}$  converges  $\mathbb{P}$ -a.s. to  $X$ . Further, recall that  $\mathbb{P}$ -a.s. convergence implies convergence in probability. So there exists  $N = N(\varepsilon)$  such that the right-hand side is smaller than  $\varepsilon$  for all  $n \geq N$ . Therefore,

$$\mathbb{E}[|X_n - X|] \rightarrow 0 \text{ as } n \rightarrow \infty. \tag{1}$$

Now, for  $c \in (0, \infty)$  we consider the function  $\phi_c : \mathbb{R} \rightarrow \mathbb{R}$  defined by

$$\phi_c(x) := (x \wedge c) \vee (-c) = \begin{cases} c & \text{if } x \geq c \\ x & \text{if } x \in (-c, c), \\ -c & \text{if } x \leq -c \end{cases} \quad x \in \mathbb{R}.$$

Note that the function  $\phi_c$  is a contraction, i.e.  $|\phi_c(X_n) - \phi_c(X)| \leq |X_n - X|$ , and

$$|\phi_c(X_n)| \leq c, \quad n \in \mathbb{N}.$$

With  $X_n \rightarrow X$   $\mathbb{P}$ -a.s., it follows that  $\phi_c(X_n) \rightarrow \phi_c(X)$   $\mathbb{P}$ -a.s. (continuous mapping theorem). Hence, for any  $c$ , by (1)

$$\mathbb{E}[|\phi_c(X_n) - \phi_c(X)|] \rightarrow 0 \text{ as } n \rightarrow \infty.$$

We want to show  $\mathbb{E}[|X_n - X|] \rightarrow 0$ .

Note that since  $(X_n)_{n \in \mathbb{N}}$  is uniformly integrable, for  $\varepsilon = 1$  there exists a constant  $L > 0$  such that

$$\begin{aligned} \mathbb{E}[|X_n|] &= \mathbb{E}[|X_n| \mathbb{1}_{\{|X_n| < L\}}] + \mathbb{E}[|X_n| \mathbb{1}_{\{|X_n| \geq L\}}] \\ &\leq L + \mathbb{E}[|X_n| \mathbb{1}_{\{|X_n| \geq L\}}] \\ &\leq L + 1, \quad n \in \mathbb{N}. \end{aligned}$$

So with  $X_n \rightarrow X$   $\mathbb{P}$ -a.s., we get

$$\mathbb{E}[|X|] \leq \liminf_{n \rightarrow \infty} \mathbb{E}[|X_n|] < \infty$$

by Fatou's lemma. It follows

$$\begin{aligned} \mathbb{E}[|X_n - X|] &\leq \mathbb{E}[|X_n - \phi_c(X_n)|] + \mathbb{E}[|\phi_c(X_n) - \phi_c(X)|] + \mathbb{E}[|\phi_c(X) - X|] \\ &\leq \mathbb{E}[|X_n| \mathbb{1}_{\{|X_n| \geq c\}}] + \mathbb{E}[|\phi_c(X_n) - \phi_c(X)|] + \mathbb{E}[|X| \mathbb{1}_{\{|X| \geq c\}}] \end{aligned}$$

Let  $\varepsilon > 0$  be given. Choosing  $c$  large enough, by uniform integrability of  $(X_n)_{n \in \mathbb{N}}$  and integrability of  $X$  the first and last summand on the right-hand side are smaller than  $\frac{\varepsilon}{3}$  for all  $n \in \mathbb{N}$ . And there exists  $N = N(c)$  such that the second summand on the right-hand side is smaller than  $\frac{\varepsilon}{3}$  for all  $n \geq N$ . Consequently,  $\mathbb{E}[|X_n - X|] < \varepsilon$  for  $n \geq N$ , thus  $X_n \rightarrow X$  in  $L^1$ .

(ii) To show: For every  $\varepsilon > 0$  there exists an  $L > 0$  such that

$$\sup_{\substack{\mathcal{G} \subseteq \mathcal{F}, \\ \mathcal{G} \text{ } \sigma\text{-algebra}}} \mathbb{E} \left[ |\mathbb{E}[X|\mathcal{G}]| \mathbb{1}_{\{|\mathbb{E}[X|\mathcal{G}]| \geq L\}} \right] < \varepsilon.$$

For  $\mathcal{G} \subseteq \mathcal{F}$ ,  $\mathcal{G}$   $\sigma$ -algebra, we have that

$$\begin{aligned} \mathbb{E} \left[ |\mathbb{E}[X|\mathcal{G}]| \mathbb{1}_{\{|\mathbb{E}[X|\mathcal{G}]| \geq L\}} \right] &\stackrel{\text{Jensen's ineq.}}{\leq} \mathbb{E} \left[ \mathbb{E}[|X| |\mathcal{G}] \mathbb{1}_{\{|\mathbb{E}[X|\mathcal{G}]| \geq L\}} \right] \\ &= \mathbb{E} \left[ \mathbb{E}[|X| \mathbb{1}_{\{|\mathbb{E}[X|\mathcal{G}]| \geq L\}} | \mathcal{G}] \right] \\ &= \mathbb{E} \left[ |X| \mathbb{1}_{\{|\mathbb{E}[X|\mathcal{G}]| \geq L\}} \right] \\ &= \mathbb{E} \left[ |X| \left( \mathbb{1}_{\{|X| > \sqrt{L}\}} + \mathbb{1}_{\{|X| \leq \sqrt{L}\}} \right) \mathbb{1}_{\{|\mathbb{E}[X|\mathcal{G}]| \geq L\}} \right] \\ &\leq \mathbb{E}[|X| \mathbb{1}_{\{|X| > \sqrt{L}\}}] + \sqrt{L} \underbrace{\mathbb{P}(\mathbb{E}[|X| |\mathcal{G}] \geq L)}_{\leq \frac{\mathbb{E}[|X|]}{L}, \text{ Markov's ineq.}} \\ &\leq \mathbb{E}[|X| \mathbb{1}_{\{|X| > \sqrt{L}\}}] + L^{-\frac{1}{2}} \mathbb{E}[|X|]. \end{aligned}$$

Let  $\varepsilon > 0$  be given. Choosing  $L$  large enough, the summands on the right-hand side are each smaller than  $\frac{\varepsilon}{2}$ . Consequently,  $\mathbb{E} \left[ |\mathbb{E}[X|\mathcal{G}]| \mathbb{1}_{\{|\mathbb{E}[X|\mathcal{G}]| \geq L\}} \right] < \varepsilon$ . Note that the upper bound holds uniformly over all  $\sigma$ -algebras  $\mathcal{G} \subset \mathcal{F}$  which proves the claim.

### Exercise 3.2

(i) By Exercise 2.3 (ii), we know that every continuous local martingale  $(M_t)_{t \in [0, T]}$ , which is bounded from below, is also a super-martingale and, in particular,

$$\mathbb{E}[M_t | \mathcal{F}_s] \leq M_s \quad \text{a.s. for all } s, t \in [0, T] \text{ with } s \leq t.$$

Hence, it is sufficient to show that

$$\mathbb{E}[M_t | \mathcal{F}_s] = M_s \quad \text{a.s. for all } s, t \in [0, T] \text{ with } s \leq t.$$

Let  $s, t \in [0, T]$  with  $s \leq t$ . We observe that

$$0 \leq M_s - \mathbb{E}[M_t | \mathcal{F}_s] \quad \text{a.s.}$$

and, using the assumption  $\mathbb{E}[M_t] = \mathbb{E}[M_0]$  for all  $t \in [0, T]$ ,

$$\mathbb{E}[M_s - \mathbb{E}[M_t | \mathcal{F}_s]] = \mathbb{E}[M_s] - \mathbb{E}[M_t] = 0.$$

Hence,  $M_s - \mathbb{E}[M_t | \mathcal{F}_s]$  is an a.s. non-negative random variable with mean zero. This implies

$$\mathbb{E}[M_t | \mathcal{F}_s] - M_s = 0 \quad \text{a.s.},$$

which shows the claim.

(ii) (a) Let  $(M_t)_{t \in [0, T]}$  be a continuous martingale in  $L^2$ .

We want to apply Theorem 2.10 (ii)  $\Rightarrow$  (i): *If  $(X_t)_{t \in [0, T]}$  is right-continuous and adapted and if  $X_\tau \in L^1$  and  $\mathbb{E}[X_\tau] = \mathbb{E}[X_0]$  for any bounded stopping time  $\tau$ , then  $X$  is a martingale.*

By Theorem 2.20,  $(M_t^2 - \langle M \rangle_t)_{t \in [0, T]}$  is a continuous local martingale. Then  $(M_t^2 - \langle M \rangle_t)_{t \in [0, T]}$  is right-continuous and adapted.

Let  $\mathcal{T} > 0$  and  $\tau \leq \mathcal{T}$  be a bounded stopping time.

We first show that  $\mathbb{E}[M_\tau^2 - \langle M \rangle_\tau] = \mathbb{E}[M_0^2 - \langle M \rangle_0] = 0$  or, equivalently,  $\mathbb{E}[M_\tau^2] = \mathbb{E}[\langle M \rangle_\tau]$ .

Again, by Theorem 2.20,  $(M_t^2 - \langle M \rangle_t)_{t \in [0, T]}$  is a continuous local martingale. Let  $(\tau_n)_{n \in \mathbb{N}}$  be a localizing sequence for  $(M_t^2 - \langle M \rangle_t)_{t \in [0, T]}$ .

Note that, since  $\varphi(u) = u^2$  is convex and  $M$  is assumed to be square-integrable,  $M^2$  is a nonnegative sub-martingale which implies

$$M_{\tau_n \wedge \tau}^2 \leq \mathbb{E}[M_{\mathcal{F}_{\tau_n \wedge \tau}}^2 | \mathcal{F}_{\tau_n \wedge \tau}].$$

Hence,  $(M_{\tau_n \wedge \tau}^2)_{n \in \mathbb{N}}$  is uniformly integrable. Further, as  $(\tau_n)_{n \in \mathbb{N}}$  is a localizing sequence for  $(M_t^2 - \langle M \rangle_t)_{t \in [0, T]}$ , for  $t \in [0, T]$ ,

$$\begin{aligned} \mathbb{E}[M_{\tau_n \wedge t}^2 - \langle M \rangle_{\tau_n \wedge t}] &= \mathbb{E}[M_{\tau_n \wedge t}^2 - \langle M \rangle_{\tau_n \wedge t} - (M_0^2 - \langle M \rangle_0)] \\ &= \mathbb{E}[M_0^2 - \langle M \rangle_0 - (M_0^2 - \langle M \rangle_0)] \\ &= 0 \end{aligned}$$

which implies  $\mathbb{E}[M_{\tau_n \wedge t}^2] = \mathbb{E}[\langle M \rangle_{\tau_n \wedge t}]$ ,  $n \in \mathbb{N}$ . Hence,  $\mathbb{E}[M_{\tau_n \wedge \tau}^2] = \mathbb{E}[\langle M \rangle_{\tau_n \wedge \tau}]$ . Thus,

$$\mathbb{E}[M_\tau^2] = \lim_{n \rightarrow \infty} \mathbb{E}[M_{\tau_n \wedge \tau}^2] = \lim_{n \rightarrow \infty} \mathbb{E}[\langle M \rangle_{\tau_n \wedge \tau}] = \mathbb{E}[\langle M \rangle_\tau]$$

where we applied Exercise 3.1.(i) in the first step and monotone convergence ( $\langle M \rangle$  non-decreasing) in the last step.

This also gives  $\mathbb{E}[M_\tau^2 - \langle M \rangle_\tau] = 0 < \infty$ , i.e.  $M_\tau^2 - \langle M \rangle_\tau \in L^1$ .

We can now conclude that  $(M_t^2 - \langle M \rangle_t)_{t \in [0, T]}$  is a martingale.

- (b) Again suppose that  $(M_t)_{t \in [0, T]}$  is a continuous martingale in  $L^2$ , in particular,  $\mathbb{E}[M_t^2] < \infty$  for all  $t \in [0, T]$ . Doob's  $L^p$ -inequality ( $p = 2$ ) implies

$$\mathbb{E} \left[ \sup_{t \in [0, T]} |M_t|^2 \right] \leq 4\mathbb{E}[M_T^2] = 4\mathbb{E}[M_T^2 - \langle M \rangle_T] = 4\mathbb{E}[M_0^2 - \langle M \rangle_0] = 0,$$

where we used that  $(M_t^2 - \langle M \rangle_t)_{t \in [0, T]}$  is a martingale by (a) and by assumption  $M_0 = 0$  and  $\langle M \rangle_t = 0$  for all  $t \in [0, T]$ .

This gives  $M_t = 0$  for all  $t \in [0, T]$ , a.s.

Alternative proof: Suppose that  $M$  is a continuous martingale in  $L^2$ . By (a)  $(M_t^2 - \langle M \rangle_t)_{t \in [0, T]}$  is a martingale. Since  $\langle M \rangle_t = 0$  for all  $t \in [0, T]$  it follows that  $M^2$  is also a martingale. Then for all  $t \in [0, T]$  it holds

$$\mathbb{E}[M_t^2] = \mathbb{E}[M_0^2] \stackrel{M_0=0}{=} 0.$$

This implies  $M_t^2 = 0$   $\mathbb{P}$ -a.s. for all  $t \in [0, T]$ , i.e.  $M_t = 0$   $\mathbb{P}$ -a.s. for all  $t \in [0, T]$ , consequently, since  $M$  is continuous,  $\mathbb{P}(M_t = 0 \forall t) = 1$ , by Exercise 2.1.(i).

- (c) Let  $(M_t)_{t \in [0, T]}$  be a continuous local martingale, not necessarily in  $L^2$ . Consider now the localizing sequence  $(\tau_n)_{n \in \mathbb{N}}$ , defined by

$$\tau_n := \inf\{t \in [0, T] : |M_t| \geq n\} \wedge T, \quad n \in \mathbb{N}.$$

Then, by Exercise 2.3.(i),  $(M_{\tau_n \wedge t} - M_0)_{t \in [0, T]} \stackrel{M_0=0}{=} (M_{\tau_n \wedge t})_{t \in [0, T]} =: (M_t^{\tau_n})_{t \in [0, T]} =: M^{\tau_n}$  is a continuous martingale and in  $L^2$  and we know, from Remark 2.22, that

$$\langle M^{\tau_n} \rangle_t = \langle M \rangle_{\tau_n \wedge t} = 0, \quad t \in [0, T],$$

for all  $n \in \mathbb{N}$ . So  $(M_t^{\tau_n})_{t \in [0, T]}$  satisfies the conditions to apply (a) and (b).

Hence, by monotone convergence we get

$$\mathbb{E} \left[ \sup_{t \in [0, T]} |M_t|^2 \right] = \lim_{n \rightarrow \infty} \mathbb{E} \left[ \sup_{t \in [0, T]} |M_{\tau_n \wedge t}|^2 \right] \stackrel{(b)}{=} 0.$$

As in (b) it follows  $M_t = 0$  for all  $t \in [0, T]$ , a.s.

### Exercise 3.3

We want to show

$$\int_0^t B_s dB_s = \frac{1}{2}(B_t^2 - t), \quad t \in [0, T]$$

Let  $T > t$ . First note that  $f = B \in \mathcal{H}^2([0, T])$  because

$$\mathbb{E} \left[ \int_0^T B_s(\omega)^2 ds \right] \stackrel{\text{Fubini}}{=} \int_0^T \mathbb{E}[B_s^2] ds = \int_0^T s ds < \infty.$$

We approximate  $B$  by

$$f_n(\omega, t) = \sum_{i=0}^{n-1} B_{t_i}(\omega) \mathbb{1}_{(t_i, t_{i+1}]}(t),$$

where  $t_i = \frac{i}{n}T$ ,  $i = 0, \dots, n$ . Note that  $B_{t_i}$  is  $\mathcal{F}_{t_i}$ -measurable and  $\mathbb{E}[B_{t_i}^2] = t_i < \infty$ ,  $i = 0, \dots, n$ , hence,  $f_n \in \mathcal{H}_0^2$ . Then

$$\begin{aligned} \|f - f_n\|_{\mathcal{H}^2}^2 &= \mathbb{E} \left[ \int_0^T \sum_{i=0}^{n-1} \mathbb{1}_{(t_i, t_{i+1}]}(t) (B_t - B_{t_i})^2 dt \right] \\ &\stackrel{\text{Fubini}}{=} \sum_{i=0}^{n-1} \int_{t_i}^{t_{i+1}} \mathbb{E}[(B_t - B_{t_i})^2] dt \\ &= \sum_{i=0}^{n-1} \int_{t_i}^{t_{i+1}} (t - t_i) dt \\ &= \frac{1}{2} \sum_{i=0}^{n-1} (t_{i+1} - t_i)^2 \\ &= \frac{T^2}{2n} \rightarrow 0, \quad n \rightarrow \infty. \end{aligned}$$

Therefore, for  $k_n := \max\{i : t_i < t\}$

$$\begin{aligned} I(f \mathbb{1}_{[0, t]}) &= \lim_{n \rightarrow \infty} I(f_n \mathbb{1}_{[0, t]}) \\ &= \lim_{n \rightarrow \infty} \sum_{i \leq k_n - 1} B_{t_i} (B_{t_{i+1}} - B_{t_i}) + B_{t_{k_n}} (B_t - B_{t_{k_n}}) \\ &= \lim_{n \rightarrow \infty} \sum_{i \leq k_n - 1} \left( \frac{1}{2} (B_{t_{i+1}}^2 - B_{t_i}^2) - \frac{1}{2} (B_{t_{i+1}} - B_{t_i})^2 \right) + B_{t_{k_n}} (B_t - B_{t_{k_n}}) \\ &= \lim_{n \rightarrow \infty} \underbrace{\frac{1}{2} B_{t_{k_n}}^2}_{\rightarrow B_t^2} + \underbrace{B_{t_{k_n}} (B_t - B_{t_{k_n}})}_{\rightarrow 0} - \frac{1}{2} \sum_{i \leq k_n - 1} (B_{t_{i+1}} - B_{t_i})^2 = \frac{1}{2} B_t^2 - \frac{1}{2} t \end{aligned}$$

since  $B$  is continuous,  $B_0 = 0$  a.s. and

$$\lim_{n \rightarrow \infty} \sum_{i \leq k_n - 1} (B_{t_{i+1}} - B_{t_i})^2 = t$$

by Corollary 2.23 ( $\langle B \rangle_t = t$ ) in the lecture notes.