

## 271B - Homework 2

**Problem 1.** Let  $S$ ,  $T$ , and  $T_n$ ,  $n = 1, 2, \dots$  be stopping times (with respect to some filtration  $\{\mathcal{F}_t\}_{t \geq 0}$ ). Show that  $T \vee S$ ,  $T \wedge S$ ,  $T + S$ ,  $\sup_n T_n$  are also stopping times.

*Proof.* The pointwise minimum, maximum, sum, and supremum of measurable functions are measurable. For the minimum and maximum we have

$$\{(T \wedge S) \leq t\} = \{T \leq t\} \cup \{S \leq t\}$$

$$\{(T \vee S) \leq t\} = \{T \leq t\} \cap \{S \leq t\}.$$

Unions and intersections of measurable sets are measurable, so both of these sets live in  $\mathcal{F}_t$ . Thus,  $T \wedge S$  and  $T \vee S$  are stopping times. For the sum, we can write the set  $\{T + S \leq t\}$  as a countable union:

$$\{T + S \leq t\} = \bigcup_{\alpha, \beta \in \mathbb{Q}, \alpha + \beta \leq t} \{T \leq \alpha\} \cap \{S \leq \beta\}.$$

As  $\mathcal{F}_t$ -measurability is closed under countable union and intersection, the sum is a stopping time. Finally, we have

$$\{\sup_n T_n \leq t\} = \bigcap_{n=1}^{\infty} \{T_n \leq t\},$$

which is measurable, so the supremum is also a stopping time. □

**Problem 2.** Let  $X_t$  be an adapted and continuous stochastic process, and define

$$T_\Gamma = \inf\{t \geq 0 : X_t \in \Gamma\}$$

for  $\Gamma$  a closed set. Show that  $T_\Gamma$  is a stopping time.

*Proof.* As  $\Gamma$  is closed, for every  $x$  there is a well-defined “distance to  $\Gamma$ ” function

$$d(x, \Gamma) = \inf_{y \in \Gamma} |x - y|.$$

In fact, this function is continuous. Since  $X_t$  has continuous paths and is  $\mathcal{F}_t$  measurable, the composition  $Y_t = d(X_t, \Gamma)$  is  $\mathcal{F}_t$  measurable.

Since  $\Gamma$  is closed,  $X_t \in \Gamma$  if and only if  $Y_t = d(X_t, \Gamma) = 0$ . From this it follows that  $T_\Gamma > t$  if and only if  $Y_s > 0$  for all  $s \leq t$ . Intuitively, if  $T_\Gamma > t$ , then  $X_t$  arrives in  $\Gamma$  at some time strictly later than  $t$ . In order for this to happen,  $X_t$  must be outside of  $\Gamma$  at all times  $s \leq t$ , in which case  $Y_s = d(X_s, \Gamma) > 0$ . This set is ostensibly an uncountable intersection, but we can write it as a union of countable intersections by approximating by rational points.

$$\{T_\Gamma > t\} = \bigcap_{s \leq t} \{Y_s > 0\} = \bigcup_{n \geq 1} \bigcap_{q \in \mathbb{Q} \cap [0, t]} \{Y_q > 1/n\} \in \mathcal{F}_t.$$

Hence,  $T_\Gamma$  is a stopping time. □

**Problem 3.** Show that if  $X_t$  is a martingale with respect to some filtration (say  $\mathcal{F}_t$ ) then it is also a martingale with respect to the filtration generated by itself.

*Proof.* Let  $\mathcal{G}_t = \sigma(X_s : s \leq t)$  be the filtration  $X$  generates. We then have  $\mathcal{G}_t \subseteq \mathcal{F}_t$  for all  $t$  since  $\mathcal{G}_t$  is the smallest  $\sigma$ -algebra with respect to which  $X_t$  is measurable. By the law of total expectation and the martingale property of  $X_t$  with respect to  $\mathcal{F}_t$  we have for any  $s \leq t$

$$\mathbb{E}[X_t \mid \mathcal{G}_s] = \mathbb{E}[\mathbb{E}[X_t \mid \mathcal{F}_s] \mid \mathcal{G}_s] = \mathbb{E}[X_s \mid \mathcal{G}_s] = X_s.$$

Thus,  $X_t$  is a martingale with respect to  $\{\mathcal{G}_t\}$ . □

**Problem 4.** Let  $a, b$  be deterministic and  $f, g$  of class I. Show that if

$$a + \int_0^T f_s dB_s = b + \int_0^T g_s dB_s \quad (1)$$

then  $a = b$  and  $f = g$  a.a. for  $(t, \omega) \in (0, T) \times \Omega$ .

*Proof.* Since  $f$  and  $g$  are of class I,  $\int_0^t f_s dB_s$  and  $\int_0^t g_s dB_s$  are martingales and  $\int_0^0 f_s dB_s = 0$  a.s. (the same holds for  $g$ ). Taking the expectation of both sides of the given relation shows that  $a = b$  a.s. and

$$\int_0^T (f_s - g_s) dB_s = 0.$$

By the Itô isometry we have

$$0 = \mathbb{E} \left[ \left( \int_0^T (f_s - g_s) dB_s \right)^2 \right] = \mathbb{E} \left[ \int_0^T (f_s - g_s)^2 ds \right].$$

We conclude that  $f_t(\omega) = g_t(\omega)$  for almost all  $(t, \omega) \in (0, T) \times \Omega$ . □

**Problem 5.** Assume that  $X_t$  is of class I and continuous in mean square on  $[0, T]$ , that is for  $t \in [0, T]$

$$\mathbb{E}[X_t^2] < \infty, \quad \lim_{s \rightarrow t} \mathbb{E}[(X_t - X_s)^2] = 0.$$

Define

$$\phi_t^{(n)} = \sum_j X_{t_{j-1}^{(n)}} \chi_{[t_{j-1}^{(n)}, t_j^{(n)})}(t), \quad t_j^{(n)} = j2^{-n}.$$

Show that for  $0 \leq t \leq T$

$$\int_0^t X_s dB_s = \lim_{n \rightarrow \infty} \int_0^t \phi_s^{(n)} dB_s,$$

where the limit is in  $L^2(\mathbb{P})$ .

*Proof.* For any  $n$  we have by the Itô isometry

$$\mathbb{E} \left[ \left( \int_0^t (X_s - \phi_s^{(n)}) dB_s \right)^2 \right] = \mathbb{E} \left[ \int_0^t (X_s - \phi_s^{(n)})^2 ds \right] = \mathbb{E} \left[ \sum_j \int_{t_{j-1}^{(n)}}^{t_j^{(n)}} (X_s - X_{t_{j-1}^{(n)}})^2 ds \right].$$

Now we claim that continuity in mean square on the compact set  $[0, T]$  implies uniform continuity in mean square. Assuming this claim, we can choose  $n$  large enough so that  $\mathbb{E}[(X_s - X_{t_{j-1}^{(n)}})^2]$  is smaller than say  $\epsilon$  for all  $j$ . For  $n$  at least this large we have

$$\mathbb{E} \left[ \left( \int_0^t (X_s - \phi_s^{(n)}) dB_s \right)^2 \right] \leq \sum_j (t_j^{(n)} - t_{j-1}^{(n)}) \epsilon = \epsilon T.$$

Since the  $L^2$  distance between  $\int_0^t \phi_s^{(n)} dB_s$  and  $\int_0^t X_s dB_s$  can be made arbitrarily small, we conclude that  $\int_0^t \phi_s^{(n)} dB_s \rightarrow \int_0^t X_s dB_s$  in  $L^2$ .

Now we show uniform mean square continuity. Suppose for the sake of contradiction that for some  $\epsilon$  there is no  $\delta$  such that  $|s - t| < \delta$  implies that  $\|X_s - X_t\|_{L^2} < \epsilon$ . Then we can find a sequence  $s_n, t_n$  so that  $|s_n - t_n| < 1/n$  but  $\|X_{s_n} - X_{t_n}\|_{L^2} > \epsilon$ . By the compactness of  $[0, T]$ , we can assume that  $s_n \rightarrow s^* \in [0, T]$ . We then have  $\|X_{s^*} - X_{t_n}\|_{L^2} > \epsilon$ , but this contradicts the mean square continuity of  $X$  at  $s^*$ .  $\square$

**Problem 6.** Let  $X_t$  be a deterministic continuous function and

$$Y_t = \int_0^t X_s dB_s.$$

Deduce the law of the process  $Y$ .

*Solution.* We assume  $t \in [0, T]$  for some  $T < \infty$ . Since  $X_t$  is continuous, it is bounded and  $\int_0^t X_s ds < \infty$  for all  $t$  and  $\omega$ . In particular, the family  $\{X_t\}_{t \in [0, T]}$  is uniformly integrable in  $\omega$ , so we have

$$\int_0^t X_s dB_s = \lim_{n \rightarrow \infty} \sum_{j=1}^{t/\Delta t} X_{t_{j-1}} \Delta B_{t_j},$$

where  $\Delta B_{t_j} = B_{t_j} - B_{t_{j-1}}$  and the limit is in  $L^2$ . (Alternatively,  $X$  satisfies the hypotheses of problem 5, so we could have used the result from that problem to get this limit.) Since the Brownian increments on the right-hand side are disjoint, they are independent normal random variables, so the whole sum on the right is a normal random variable with distribution

$$\mathcal{N} \left( 0, \sum_{j=1}^{t/\Delta t} X_{t_{j-1}}^2 \Delta t \right).$$

Since  $X$  is continuous and deterministic, we recognize the above sum as a Riemann sum as  $\Delta t \rightarrow 0$ . Since the  $L^2$  limit of normal random variables is normal when the mean and variance converge as sequences of real numbers, we have

$$Y_t = \int_0^t X_s dB_s \sim \mathcal{N} \left( 0, \int_0^t X_s^2 ds \right).$$

Now the process  $Y$  is Gaussian if and only if for every  $t_1 < \dots < t_k \in T$ , any linear combination of the  $Y_{t_j}$ 's has univariate normal distribution. Since  $\int_a^b X_s dB_s = \int_a^c X_s dB_s + \int_c^b X_s dB_s$  for any  $a < c < b$ , we have

$$c_1 Y_{t_1} + c_2 Y_{t_2} + \dots + c_k Y_{t_k} = (c_1 + \dots + c_k) Y_{t_1} + (c_2 + \dots + c_k)(Y_{t_2} - Y_{t_1}) + \dots + c_k(Y_{t_k} - Y_{t_{k-1}}).$$

The variables  $Y_{t_j} - Y_{t_{j-1}}$  are themselves Itô integrals over disjoint intervals, so they are independent normal random variables. We conclude that the above linear combination is normally distributed, so the process  $Y$  is Gaussian.

A Gaussian process, and therefore its law, is determined by its mean and covariance. Since  $Y_t \sim \mathcal{N}(0, \int_0^t X_s^2 ds)$ , the process  $Y$  has zero mean. As for the covariance, we have for any  $s, t$

$$\begin{aligned} \text{Cov}(Y_s, Y_t) &= \mathbb{E} \left[ \int_0^s X_u dB_u \cdot \int_0^t X_u dB_u \right] \\ &= \mathbb{E} \left[ \left( \int_0^{s \wedge t} X_u dB_u \right)^2 \right] + \mathbb{E} \left[ \int_0^{s \wedge t} X_u dB_u \cdot \int_{s \wedge t}^{s \vee t} X_u dB_u \right] \\ &= \mathbb{E} \left[ \int_0^{s \wedge t} X_u^2 du \right] = \int_0^{s \wedge t} X_u^2 du. \end{aligned}$$

The last line follows from the Itô isometry and the independence of Itô integrals over disjoint intervals. □