



Solution to Exam Stochastic Differential Equations (Mastermath)  
08-06-2015; 13:30 – 16:30.

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- (5) 1. a. Let  $X \in L^1$  be a random variable and let  $\mathcal{F}$  and  $\mathcal{G}$  be  $\sigma$ -algebras such that  $\mathcal{F} \subseteq \mathcal{G}$ . Show that  $\mathbb{E}(\mathbb{E}(X|\mathcal{G})|\mathcal{F}) = \mathbb{E}(X|\mathcal{F})$ .
- (5) b. Let  $\{X_n : n \in \mathbb{N}\}$  be a sequence of random variables. Assume there is a constant  $C$  such that for all integers  $n \in \mathbb{N}$ ,  $\mathbb{E}(|X_n|^2) \leq C$ . Show that  $X$  is uniformly integrable.
- (5) c. Let  $(X_t)_{t \geq 0}$  be a nonnegative local martingale such that  $\mathbb{E}(X_0) < \infty$ . Show that  $X$  is a supermartingale.
- (2) d. Let  $0 < a < b$  and assume  $\xi : \Omega \rightarrow \mathbb{R}$  is  $\mathcal{F}_a$ -measurable. Use the definition to show that  $\mathbf{1}_{(a,b)}\xi$  is predictable.
- (4) e. Use the previous exercise and an approximation argument to show that any left-continuous adapted process  $Z : \mathbb{R} \times \Omega \rightarrow \mathbb{R}$  is predictable.

**[Soln]**

- a. Let  $Z = \mathbb{E}(\mathbb{E}(X|\mathcal{G})|\mathcal{F})$ . For every  $F \in \mathcal{F}$  we also have that  $F \in \mathcal{G}$  and thus by the definition of the conditional expectation we find that

$$\int_F Z d\mathbb{P} = \int_F \mathbb{E}(X|\mathcal{G}) d\mathbb{P} = \int_F X d\mathbb{P}.$$

Since  $Z$  is  $\mathcal{F}$ -measurable the required identity follows.

- b. For each  $r > 0$  we have

$$\int_{\{|X_n|>r\}} |X_n| d\mathbb{P} \leq r^{-1} \int_{\{|X_n|>r\}} |X_n|^2 d\mathbb{P} \leq r^{-1} \int_{\Omega} |X_n|^2 d\mathbb{P} \leq r^{-1}C.$$

Therefore,

$$\limsup_{r \rightarrow \infty} \sup_{n \geq 1} \int_{\{|X_n|>r\}} |X_n| d\mathbb{P} \leq \lim_{r \rightarrow \infty} r^{-1}C = 0.$$

- c. See one of the exercises of chapter 3.
- d. The definition of the predictability can be found in the lecture notes. There are at least two possible solutions:
1. Approximate  $\xi$  by  $\mathcal{F}_a$ -measurable simple functions  $\xi_n$ , then  $\mathbf{1}_{(a,b)}\xi_n$  can be written as a linear combination of predictable rectangles and hence is predictable. Then also the pointwise limit  $\mathbf{1}_{(a,b)}\xi$  is predictable.
  2. Check that  $\mathbf{1}_{(a,b)}\xi \in B$  is in the predictable  $\sigma$ -algebra. If  $0 \notin B$  this is simple. If  $0 \in B$ , then some more rewriting is required.
- e. Use d and approximation. See the lecture notes for details.

END

2. Let  $B$  be a standard Brownian motion and let  $\mathcal{F}_t = \sigma(B_s : s \leq t)$ .

- (6) a. Let  $\alpha \geq 0$ . Using the properties of conditional expectations and the independent increments of Brownian motion show that  $X_t = \cosh(\alpha|B_t|) \exp(-\alpha^2 t/2)$  is a martingale.

*Hint:* Recall that  $\cosh(|x|) = \cosh(x) = (e^x + e^{-x})/2$ . You may also use the identity:  $\mathbb{E}(\exp(\xi)) = \exp(\sigma^2/2)$  for  $\xi \sim N(0, \sigma^2)$ .

Fix  $A > 0$  and let  $\tau = \inf\{t \geq 0 : |B_t| = A\}$ .

- (3) b. Prove that  $\tau$  is a stopping time.
- (3) c. Show that  $\mathbb{E}(X_{t \wedge \tau}) = 1$  for all  $t \geq 0$ .
- (5) d. Show that  $\mathbb{E}(X_\tau) = 1$  and use this to find a formula for  $\mathbb{E}(e^{-\lambda\tau})$  where  $\lambda \geq 0$ .

*Hint:* You may use the known fact that  $\tau < \infty$  almost surely.

[Soln]

- a. Note that for  $s < t$  by linearity of the conditional expectation

$$\mathbb{E}(X_t|\mathcal{F}_s) = \frac{1}{2} \exp(-\alpha^2 t/2) \left( \mathbb{E}(\exp(\alpha B_t)|\mathcal{F}_s) + \mathbb{E}(\exp(-\alpha B_t)|\mathcal{F}_s) \right).$$

We calculate both conditional expectations.

$$\begin{aligned} \mathbb{E}(\exp(\alpha B_t)|\mathcal{F}_s) &= \mathbb{E}(\exp(\alpha(B_t - B_s)) \exp(\alpha B_s)|\mathcal{F}_s) \\ &= \exp(\alpha B_s) \mathbb{E}(\exp(\alpha(B_t - B_s))|\mathcal{F}_s) \quad (\text{taking out what is known}) \\ &= \exp(\alpha B_s) \mathbb{E}(\exp(\alpha(B_t - B_s))) \quad (\text{independence}) \\ &= \exp(\alpha B_s) \exp(\alpha^2(t-s)/2) \quad (\alpha(B_t - B_s) \sim N(0, \alpha^2(t-s))) \end{aligned}$$

The other conditional expectation can be calculated in the same way and is

$$\mathbb{E}(\exp(-\alpha B_t)|\mathcal{F}_s) = \exp(-\alpha B_s) \exp(\alpha^2(t-s)/2).$$

Putting everything together we find

$$\mathbb{E}(X_t|\mathcal{F}_s) = \exp(-\alpha^2 t/2) \cosh(\alpha B_s) \exp(\alpha^2(t-s)/2) = X_s.$$

- b. For every  $t \geq 0$  we can write

$$\{\tau > t\} = \{\forall s \in [0, t] B_s < A\} = \{\forall s \in [0, t] \cap \mathbb{Q} B_s < A\} = \bigcap_{s \in [0, t] \cap \mathbb{Q}} \{B_s < A\} \in \mathcal{F}_t.$$

where in the last step we used the adaptedness of  $B$  and the fact that the *countable union* of sets in  $\mathcal{F}_t$  is in  $\mathcal{F}_t$  again.

You may also use a theorem from the lecture notes. Here it is important to observe the continuity of  $B$  and the closedness of the set  $\{A, -A\}$ .

- c. Since  $\tau$  is a stopping time and  $(X_t)_{t \geq 0}$  a martingale, it follows from the stopping time theorem that  $(X_{t \wedge \tau})_{t \geq 0}$  is a martingale again. Therefore, by the properties of the conditional expectation:

$$\mathbb{E}(X_{t \wedge \tau}) = \mathbb{E}(\mathbb{E}(X_{t \wedge \tau} | \mathcal{F}_0)) = \mathbb{E}(X_0) = 0.$$

- d. First note that  $\lim_{t \rightarrow \infty} X_{t \wedge \tau} = X_\tau$  pointwise on the set  $\{\tau < \infty\}$ . Observe that  $|B_{t \wedge \tau}| \leq A$ . Therefore,

$$|X_{t \wedge \tau}| \leq \cosh(\alpha A) \exp(-\alpha^2(t \wedge \tau)/2) \leq \cosh(\alpha A).$$

Since the constant function  $\cosh(\alpha A)$  is in  $L^1$ , we may apply the dominated convergence theorem to conclude

$$1 = \lim_{t \rightarrow \infty} \mathbb{E}(X_{t \wedge \tau}) = \mathbb{E}(\lim_{t \rightarrow \infty} X_{t \wedge \tau}) = \mathbb{E}(X_\tau).$$

Next we calculate  $\mathbb{E}(X_\tau)$  in a different way. Observe that on the set  $\{\tau < \infty\}$  we have  $|B_\tau| = A$  and hence

$$1 = \mathbb{E}(X_\tau) = \mathbb{E}(\cosh(\alpha A) \exp(-\alpha^2 \tau/2)) = \cosh(\alpha A) \mathbb{E}(\exp(-\alpha^2 \tau/2)).$$

Taking  $\alpha^2/2 = \lambda$  we find that  $\mathbb{E}(\exp(-\alpha^2 \tau/2)) = 1/\cosh(\sqrt{2\lambda})$ .

END

3. Let  $M$  be a continuous  $L^2$ -martingale. We will show below, in steps (a) through (c), that  $M^2 - [M]$  is a martingale. Let  $t > s \geq 0$ . Consider a partition of  $[0, t]$  given by

$$0 = t_0 < t_1 < \cdots < t_k = s < \cdots < t_n = t.$$

- (4) a. Show that for  $t_i \geq s$ ,

$$\mathbb{E} \left[ M_{t_{i+1}}^2 - M_{t_i}^2 - (M_{t_{i+1}} - M_{t_i})^2 \middle| \mathcal{F}_s \right] = 0.$$

- (2) b. Derive that

$$\mathbb{E} \left[ M_t^2 - \sum_{i=0}^{n-1} (M_{t_{i+1}} - M_{t_i})^2 \middle| \mathcal{F}_s \right] = M_s^2 - \sum_{i=0}^{k-1} (M_{t_{i+1}} - M_{t_i})^2.$$

- (3) c. Using the previous identity and the existence theorem for quadratic variations of continuous  $L^2$ -martingales show that  $M^2 - [M]$  is a martingale.
- (6) d. Let  $M$  and  $N$  be continuous  $L^2$ -martingales with quadratic covariation  $[M, N]_t = 2t - 1$ . Let  $X : \mathbb{R}_+ \times \Omega \rightarrow \mathbb{R}$  be given by  $X_t = M_t + \cos(t)$ . Let  $U : \mathbb{R}_+ \rightarrow \mathbb{R}$  and  $V : \mathbb{R}_+ \rightarrow \mathbb{R}$  be given by  $U_t = t$  and  $V_t = t^2$ . Use the properties of quadratic (co)variations to calculate the  $[U \cdot X, V \cdot N]_t$  for  $t \geq 0$ .

[Soln]

- a. The tower property for  $s \leq t_i$  gives that  $\mathbb{E}(\cdot|\mathcal{F}_s) = \mathbb{E}(\mathbb{E}(\cdot|\mathcal{F}_{t_i})|\mathcal{F}_s)$ . Thus by the definition of a martingale we can write

$$\begin{aligned} \mathbb{E}(M_{t_{i+1}}^2 - M_{t_i}^2 - (M_{t_{i+1}} - M_{t_i})^2|\mathcal{F}_s) &= \mathbb{E}(2M_{t_i}(M_{t_{i+1}} - M_{t_i})|\mathcal{F}_s) \\ &= \mathbb{E}(\mathbb{E}(2M_{t_i}(M_{t_{i+1}} - M_{t_i})|\mathcal{F}_{t_i})|\mathcal{F}_s) \\ &= \mathbb{E}(2M_{t_i}\mathbb{E}((M_{t_{i+1}} - M_{t_i})|\mathcal{F}_{t_i})|\mathcal{F}_s) \\ &= \mathbb{E}(2M_{t_i}0|\mathcal{F}_s) = 0. \end{aligned}$$

- b. Note that  $M_t^2 = M_s^2 + \sum_{i=k}^{n-1}(M_{t_{i+1}}^2 - M_{t_i}^2)$ . Therefore, we find

$$\begin{aligned} \mathbb{E} \left[ M_t^2 - \sum_{i=0}^{n-1} (M_{t_{i+1}} - M_{t_i})^2 \middle| \mathcal{F}_s \right] &- \left( M_s^2 - \sum_{i=0}^{k-1} (M_{t_{i+1}} - M_{t_i})^2 \right) \\ &= \mathbb{E} \left[ M_t^2 - M_s^2 - \sum_{i=k}^{n-1} (M_{t_{i+1}} - M_{t_i})^2 \middle| \mathcal{F}_s \right] \\ &= \sum_{i=k}^{n-1} \mathbb{E} \left[ M_{t_{i+1}}^2 - M_{t_i}^2 - (M_{t_{i+1}} - M_{t_i})^2 \middle| \mathcal{F}_s \right] = 0 \end{aligned}$$

where in the last step we used part a.

- c. The existence theorem for quadratic variations yields that for a sequence of partitions  $(\pi^m)_{m \geq 1}$  of  $[0, t]$  with  $\text{mesh}(\pi^m) \rightarrow 0$  we have with  $\pi^m = \{t_0^m, \dots, t_{n_m}^m\}$  with  $t_0^m = 0$  and  $t_{n_m}^m = t$  that

$$\sum_{i=0}^{n_m-1} (M_{t_{i+1}^m} - M_{t_i^m})^2 \rightarrow [M]_t \text{ in } L^1.$$

Without loss of generality we can assume  $s \in \pi^m$  for all  $m$ . By the contractivity of the conditional expectation (which means  $\|\mathbb{E}(X|\mathcal{F}_s)\|_{L^1} \leq \|X\|_{L^1}$ ) we also find that

$$\mathbb{E} \left[ M_t^2 - \sum_{i=0}^{n-1} (M_{t_{i+1}} - M_{t_i})^2 \middle| \mathcal{F}_s \right] \rightarrow M_t^2 - [M]_t \text{ in } L^1.$$

Similarly, letting  $k_m$  be such that  $t_{k_m} = s$ , we find

$$M_s^2 - \sum_{i=0}^{k_m-1} (M_{t_{i+1}^m} - M_{t_i^m})^2 \rightarrow M_s^2 - [M]_s \text{ in } L^1.$$

- d. First note that  $[U \cdot X, V \dot{N}] = UV \cdot [X, N]$ . Also  $[\cos(\cdot), N] = 0$  because  $\cos(\cdot)$  is of bounded variation. Thus  $[X, N] = [M, N] + [\cos(\cdot), N] = [M, N]$  and we find

$$[U \cdot X, V \dot{N}]_t = (UV \cdot [M, N])_t = \int_0^t s^3(2s-1) ds = \int_0^t 2s^4 - s^3 ds = \frac{2t^5}{5} - \frac{t^4}{4}.$$

END

4. Let  $(B_t)$  be a standard Brownian motion defined on a (filtered) probability space  $(\Omega, \mathcal{F} (\mathcal{F}_t), P)$ . Suppose  $X_t$  is a process satisfying the SDE

$$dX_t = X_t dB_t, \quad X_0 = 1. \quad (*)$$

Define the process  $(Z_t)_{t \in [0,1]}$  by

$$Z_t = X_t e^{-\int_0^t B_s^2 ds}.$$

- (6) a. Apply Itô's formula to show that

$$dZ_t = Z_t(dB_t - B_t^2 dt), \quad Z_0 = 1. \quad (**)$$

- (5) b. Find the solution  $X_t$  satisfying the SDE (\*).

*Hint:* You may propose a solution and appeal to the uniqueness theorem.

- (3) c. Use the solution in (b) to show that  $\mathbb{E}(Z_t^2) \leq e^t$ .

*Hint:* You may use the identity  $\mathbb{E}(\exp(\xi)) = \exp(\sigma^2/2)$  for  $\xi \sim N(0, \sigma^2)$ .

- (6) d. In the "integrated version" of the SDE (\*\*) two relevant random variables are  $\int_0^1 Z_t dB_t$  and  $\int_0^1 Z_t B_t^2 dt$ . Use (c) to show that both of the following hold.

$$(i) \quad \mathbb{E} \left[ \left( \int_0^1 Z_t dB_t \right)^2 \right] < \infty \quad \text{and} \quad (ii) \quad \mathbb{E} \left[ \int_0^1 |Z_t B_t^2| dt \right] < \infty.$$

*Hint:* If needed, you may assume the expressions for the moments of Gaussian distribution, without proof. For example,  $\mathbb{E}(\xi^4) = 3\sigma^4$  for  $\xi \sim N(0, \sigma^2)$ .

[Soln]

a. Note that (\*) implies that  $X_t$  is a continuous process satisfying

$$(i) \int_0^t X_s^2 ds < \infty \text{ a.s.} \quad \text{and} \quad (ii) \quad X_t = 1 + \int_0^t X_s dB_s.$$

Property (i) implies that  $\int_0^t X_s dB_s$  is a (continuous) local martingale and so is  $X_t$  from (ii). Also,  $Y_t := \int_0^t B_s^2 ds$ , being an increasing function of  $t$ , is a FV process and hence is a semimartingale. Note that  $Z_t = f(X_t, Y_t)$  where  $f(x, y) = xe^{-y}$ . Clearly,  $Z_0 = X_0 = 1$  and  $f \in C^2(\mathbb{R}^2)$ , with

$$f_x(x, y) = e^{-y} = -f_{xy}(x, y); \quad f_{xx} = 0; \quad f_y(x, y) = -xe^{-y} = -f(x, y) \quad \text{and} \quad f_{yy} = f.$$

Applying (vector-valued) Itô's formula to  $Z_t = f(X_t, Y_t)$  we have

$$dZ_t = d[f(X_t, Y_t)] = f_x dX_t + f_y dY_t + \frac{1}{2} f_{xx} d[X]_t + f_{xy} d[X, Y]_t + \frac{1}{2} f_{yy} d[Y]_t.$$

Since  $Y_t$  is a FV process, the quadratic (co)variation processes  $[X, Y]_t = [Y]_t = 0$ . Noting further that  $dY_t = B_t^2 dt$ , from (\*) we have,

$$\begin{aligned} dZ_t &= f_x(X_t, Y_t) dX_t + f_y(X_t, Y_t) dY_t = e^{-Y_t} dX_t - f(X_t, Y_t) dY_t \\ &= e^{-Y_t} X_t dB_t - Z_t B_t^2 dt = Z_t (dB_t - B_t^2 dt). \end{aligned}$$

b. Integrating factor method or an educated guess (together with an application of Itô formula) shows that

$$X_t = e^{B_t - \frac{1}{2}t}$$

is **a** solution to the SDE (\*). The uniqueness theorem ensures that this is the (unique) solution. Note that the uniqueness theorem is applicable because the coefficient of the SDE is  $b(x) = x$  and it satisfies both the Lipschitz and Growth conditions:

$$|b(x) - b(y)| \leq K|x - y| \quad \text{and} \quad |b(x)|^2 \leq K(1 + |x|^2) \quad \text{with} \quad K = 1.$$

c. From nonnegativity of  $X_t$  and  $Y_t := \int_0^t B_s^2 ds$  it follows that,  $0 \leq Z_t = X_t e^{-Y_t} \leq X_t$  and as a result,

$$\mathbb{E}(Z_t^2) \leq \mathbb{E}(X_t^2) = \mathbb{E}[e^{2B_t - t}] = e^{-t} e^{\frac{1}{2}4t} = e^t.$$

d. Note that  $Z_t \in \mathcal{L}^2(B)$ , because, from part (c),

$$\int_0^1 \mathbb{E}(Z_t^2) dt \leq \int_0^1 e^t dt \leq (e - 1) < \infty.$$

We can then apply Itô isometry and this leads to (i):

$$\mathbb{E} \left[ \left( \int_0^1 Z_t dB_t \right)^2 \right] = \int_0^1 \mathbb{E}(Z_t^2) dt \leq (e - 1) < \infty.$$

For (ii), note that using Fubini and (two times) Cauchy-Schwartz we have

$$\begin{aligned} \mathbb{E} \left[ \int_0^1 |Z_t B_t^2| dt \right] &= \int_0^1 \mathbb{E}[|Z_t B_t^2|] dt \leq \int_0^1 (E[Z_t^2])^{\frac{1}{2}} (E[B_t^4])^{\frac{1}{2}} dt \\ &\leq \left( \int_0^1 E[Z_t^2] dt \right)^{\frac{1}{2}} \left( \int_0^1 E[B_t^4] dt \right)^{\frac{1}{2}} \leq (e - 1)^{\frac{1}{2}} \left( \int_0^1 3t^2 dt \right)^{\frac{1}{2}} \\ &= \sqrt{e - 1} < \infty. \end{aligned}$$

END

5. Let  $0 < T < \infty$ . Suppose  $H$  is a predictable process satisfying

$$\int_0^T H_t^2 dt < \infty, \quad \text{a.s.} \quad (\dagger)$$

and let  $Z_t \equiv Z_t(H)$  be given by

$$Z_t(H) = \exp \left\{ \int_0^t H_s dB_s - \frac{1}{2} \int_0^t H_s^2 ds \right\}.$$

(5) a. Use Itô's formula to  $Z$  to show that, under  $(\dagger)$ ,  $Z$  is a continuous local martingale.

In the Girsanov Theorem an important assumption is that  $(Z_t)_{t \in [0, T]}$  is a martingale. Since  $Z$  is nonnegative using Question (1c) one sees that  $Z$  is a supermartingale. It can be shown that  $(Z_t)_{t \in [0, T]}$  is a martingale if and only if  $\mathbb{E}[Z_T] = 1$ . In this exercise we will check the latter in several steps under the assumption that there is a constant  $C$  such that

$$\int_0^T H_t^2 dt \leq C, \quad \text{a.s.} \quad (\dagger\dagger)$$

Let  $\sigma_n = \inf\{0 \leq t \leq T : Z_t \geq n\}$ , where we set  $\sigma_n = T$  if the set is empty. In this way  $(\sigma_n)_{n \geq 1}$  is a localizing sequence for  $Z$ . Denote by  $Z_t^{(n)}$  the stopped process  $Z_{t \wedge \sigma_n}$ . It then follows immediately that

$$\mathbb{E}(Z_t^{(n)}) = 1, \quad n \geq 1, \quad 0 \leq t \leq T.$$

(3) b. Show from the definitions that  $Z_t^{(n)} = Z_t(H^{(n)})$  a.s., where  $H^{(n)}$  is given by

$$H_s^{(n)}(\omega) = H_s(\omega) \mathbf{1}_{[0, \sigma_n(\omega)]}(s), \quad \text{for } 0 \leq s \leq T.$$

(5) c. Show that if  $(\dagger\dagger)$  holds then  $\mathbb{E}\left(\left(Z_T^{(n)}\right)^2\right) \leq e^C$ , for all  $n \geq 1$ .

*Hint:* Argue that  $\left(Z_t^{(n)}\right)^2 \leq M_t^{(n)} e^C$  for some suitable martingale  $M_t^{(n)}$  with  $M_0^{(n)} = 1$ .

(4) d. Show that, under  $(\dagger\dagger)$ ,  $\mathbb{E}[Z_T] = 1$ .

*Hint:* Argue that  $\sigma_n \rightarrow T$  a.s. and  $Z_T^{(n)} \rightarrow Z_T$  a.s. Use Question 1 to conclude that  $\{Z_T^{(n)}, n \geq 1\}$  is uniformly integrable. Recall further that for  $X, X_n \in L^1$  ( $n \geq 1$ ),  $X_n \rightarrow X$  in  $L^1$  if and only if (i)  $X_n \rightarrow X$  in probability and (ii)  $\{X_n, n \geq 1\}$  is uniformly integrable.



[Soln]

- a. Note that under  $(\dagger)$ ,  $M_t := \int_0^t H_s dB_s$  is a local martingale and  $N_t := \int_0^t H_s^2 ds$  is a FV process on  $[0, T]$ . In particular, both of them are semimartingales. Since  $Z_t \equiv Z_t(H) = f(M_t, N_t)$  where  $f(x, y) = e^{x - \frac{1}{2}y}$  and both  $M$  and  $N$  are continuous, it follows that  $Z_t$  is also continuous. Noting that  $f \in C^2(\mathbb{R}^2)$ , we can apply (vector-valued) Itô formula to  $Z_t = f(M_t, N_t)$  to obtain

$$\begin{aligned} dZ_t &= f_x dM_t + f_y dN_t + \frac{1}{2} f_{xx} d[M]_t + f_{xy} d[M, N]_t + \frac{1}{2} f_{yy} d[N]_t \\ &= f_x dM_t + f_y dN_t + \frac{1}{2} f_{xx} d[M]_t, \quad \text{since } [N]_t = [M, N]_t = 0. \end{aligned}$$

Noting that  $[M]_t = \int_0^t H_s^2 ds = N_t$  and  $f_x = f_{xx} = f = -2f_y$ , we get

$$Z_t = Z_0 + \int_0^t f(M_s, N_s) dM_s = 1 + \int_0^t Z_s dM_s = 1 + \int_0^t Z_s H_s dB_s.$$

Now, almost sure continuity of  $Z$  implies that

$$\int_0^t (Z_s H_s)^2 ds \leq \left( \sup_{s \in [0, T]} Z_s \right) \int_0^t H_s^2 ds < \infty \quad (\text{a.s.})$$

This in turn implies that  $\int_0^t Z_s H_s dB_s$  and hence  $Z_t = 1 + \int_0^t Z_s H_s dB_s$  is a (continuous) local martingale.

- b. Let  $n \geq 1$ . Then

$$\begin{aligned} Z_t^{(n)} &= Z_{t \wedge \sigma_n} = e^{\int_0^{t \wedge \sigma_n} H_s dB_s - \frac{1}{2} \int_0^{t \wedge \sigma_n} H_s^2 ds} = e^{\int_0^t H_s \mathbf{1}_{[0, \sigma_n]}(s) dB_s - \frac{1}{2} \int_0^t H_s^2 \mathbf{1}_{[0, \sigma_n]}(s) ds} \\ &= e^{\int_0^t H_s^{(n)} dB_s - \frac{1}{2} \int_0^t (H_s^{(n)})^2 ds} = Z_t(H^{(n)}), \quad \text{from the definition of } Z(H). \end{aligned}$$

- c. First note that, since  $H^{(n)}$  satisfies  $(\dagger)$ , it follows from part (a) that  $Z_t^{(n)} = Z_t(H^{(n)})$  is a continuous local martingale. But it is also nonnegative and hence by Question 1(c), it is a supermartingale. The fact that  $E\left(Z_t^{(n)}\right) = 1$  then implies that  $Z_t^{(n)}$  is a martingale. The same argument will also lead to the conclusion that  $Z_t(2H^{(n)})$  is a martingale.

Next note that from  $(\dagger\dagger)$  it follows that

$$\begin{aligned} \left(Z_t^{(n)}\right)^2 &= e^{2 \int_0^t H_s^{(n)} dB_s - \int_0^t (H_s^{(n)})^2 ds} = e^{\int_0^t 2H_s^{(n)} dB_s - \frac{1}{2} \int_0^t (2H_s^{(n)})^2 ds + \int_0^t (H_s^{(n)})^2 ds} \\ &= Z_t(2H^{(n)}) e^{\int_0^t (H_s^{(n)})^2 ds} \leq Z_t(2H^{(n)}) e^C, \end{aligned}$$

where, as mentioned earlier,  $Z_t(2H^{(n)})$  is a martingale. Hence

$$E\left[\left(Z_T^{(n)}\right)^2\right] \leq E\left[Z_T(2H^{(n)})\right] e^C = E\left[Z_0(2H^{(n)})\right] e^C = e^C.$$

- d. Almost sure continuity of  $Z$  implies that  $\exists \Omega_0$  with  $P(\Omega_0) = 1$  such that  $Z_t(\omega)$  is continuous in  $t$  for each  $\omega \in \Omega_0$ . Then for  $\omega \in \Omega_0$ ,  $Z^*(\omega) := \sup_{t \in [0, T]} Z_t(\omega) < \infty$ . Note that for all  $n \geq Z^*(\omega)$ ,  $\sigma_n(\omega) = T$ . Hence  $\lim_{n \rightarrow \infty} \sigma_n(\omega) = T$  for all  $\omega \in \Omega_0$ . In other words,  $\sigma_n \rightarrow T$  (a.s.), as  $n \rightarrow \infty$ . Consequently,  $\lim_{n \rightarrow \infty} Z_T^{(n)} = \lim_{n \rightarrow \infty} Z_{T \wedge \sigma_n} = Z_T$  (a.s.) In particular,  $Z_T^{(n)}$  converges to  $Z_T$  in probability. The  $L^2$ -boundedness of  $\{Z_T^{(n)}, n \geq 1\}$ , obtained in part (c), together with Question 1(b), imply uniform integrability of the sequence. Uniform integrability and convergence in prob imply that  $Z_T^{(n)}$  converges to  $Z_T$  in  $L^1$ . In particular,  $E[Z_T] = \lim_{n \rightarrow \infty} E\left[Z_T^{(n)}\right]$ . But  $E\left[Z_T^{(n)}\right] = 1$  for  $n \geq 1$ . Hence,  $E[Z_T] = 1$ .

END

## Some comments on grading of Question 4 and 5

- Before applying any theorem or formula in/to a situation, you should make sure that your situation satisfies all conditions stated in the statement of the theorem/formula. For example,
  - Ito isometry: integrand must be in  $\mathcal{L}^2(B)$
  - Ito formula: function  $f \in C^{1,2}(\mathbb{R}_+ \times \mathbb{R}^n)$ .  
Note that function  $f$  should be deterministic, as the notations suggest, and should not involve  $\omega$  (randomness).  $f(t, x) = x \exp(\int_0^t H_s^2 ds)$  where  $H_s$  is a stochastic process is not an appropriate function to apply Ito formula to.  
Also, it is better to write out the full/complete formula and then fill-in the values (e.g.,  $f_{xx} = 0$  or  $[X, Y]_t = 0$ ).
  - Uniqueness theorem of solution: coefficients must satisfy growth and Lipschitz conditions.
- In (5a) one needs, as intermediate step,  $\int_0^t Z_s H_s dB_s$  to be a local martingale. This needs proof. One way is to check that  $\int_0^t Z_s^2 H_s^2 ds < \infty$  a.s. Then the integral makes sense and is a local martingale.

END