

1. [Soln]

- a. Let  $(\tau_n)_{n \geq 0}$  be a localizing sequence for the local martingale  $M$ , i.e. for all  $s, t \in \mathbb{R}_+, s < t$  we have

$$\mathbb{E}(M_{\tau_n \wedge t} | \mathcal{F}_s) = M_{\tau_n \wedge s}. \quad (1)$$

Since  $\sup_{s \geq 0} |M_s| \in L^1$ , we have that for all  $n, t$ ,  $|M_{\tau_n \wedge t}| \leq \sup_{s \geq 0} |M_s|$  is bounded by an integrable random variable. By dominated convergence we can pull the limit in (1) out and get the result. [Now taking the limit as  $n \rightarrow \infty$  in (1) the result follows from the dominated convergence theorem.]

- b. For the definition see the book.

Now call  $A_k = \{|\mathbb{E}(X|\mathcal{G})| > k\}$ . We have to show that given  $\epsilon > 0$ ,  $\exists K$  such that

$$\sup_{\mathcal{G} \subset \mathcal{F}} \mathbb{E}(1_{A_k} |\mathbb{E}(X|\mathcal{G})|) < \epsilon \quad \forall k \geq K.$$

Since  $X \in L^1$ , we have that for a given  $\epsilon > 0$  there is a  $\delta > 0$  such that if  $\mathbb{P}(A) \leq \delta$  we have  $\mathbb{E}(|X|1_A) \leq \epsilon$ . Choose  $K$  large enough such that  $K \geq \mathbb{E}(|X|)/\delta$ . By Chebyshev's and Jensen inequality for  $k \geq K$

$$\mathbb{P}(A_k) \leq \frac{1}{k} \mathbb{E}(|\mathbb{E}(X|\mathcal{G})|) \leq \frac{1}{k} \mathbb{E}(|X|) \leq \delta.$$

Since  $A_k \in \mathcal{G}$  the result follows from

$$\mathbb{E}(|\mathbb{E}(X|\mathcal{G})|1_{A_k}) \leq \mathbb{E}(\mathbb{E}(|X|\mathcal{G})1_{A_k}) = \mathbb{E}(|X|1_{A_k}) \leq \epsilon.$$

- c. i) From Itô-isometry

$$\begin{aligned} (H \bullet (K \bullet M))_t &= \int_0^t H_s d \left( \int_0^s K_u dM_u \right) \\ &= \int_0^t H_s (K_s dM_s - K_0 dM_0) = ((HK) \bullet M)_t \end{aligned}$$

- ii) Pick  $N = \int_0^\cdot H_s dM_s$ , then

$$\left\langle \int_0^\cdot H_s dM_s, \int_0^\cdot H_s dM_s \right\rangle_t = \int_0^t H_s d \left\langle M, \int_0^\cdot H_u dM_u \right\rangle_s = \int_0^t H_s^2 d\langle M \rangle_s.$$

2. [Soln]

First note that, for all  $t \geq 0$ ,

$$\int_0^t (\sin(H_s))^2 ds \leq t \quad \text{and} \quad \int_0^t (\cos(H_s))^2 ds \leq t.$$

Hence the stochastic integrals w.r.t. BM with  $\sin(H_s)$  and  $\cos(H_s)$  as integrands, i.e.,

$$\int_0^t \cos(H_s) dX_s, \quad \int_0^t \sin(H_s) dY_s, \quad \int_0^t \sin(H_s) dX_s \quad \text{and} \quad \int_0^t \cos(H_s) dY_s$$

are all local martingales. In fact, they are all  $L^2$ -martingales. Also, clearly they are continuous. Hence

$$B_t = \int_0^t \cos(H_s) dX_s + \int_0^t \sin(H_s) dY_s$$

$$\hat{B}_t = \int_0^t \sin(H_s) dX_s - \int_0^t \cos(H_s) dY_s.$$

are also continuous local martingales.

From Levy characterization theorem, it will follow that continuous local martingales  $B$  and  $\hat{B}$  are two independent Brownian motions, if we show that the quadratic variation processes are given by

$$\langle B \rangle_t = \langle \hat{B} \rangle_t = t, \quad \langle B, \hat{B} \rangle_t = 0.$$

But, since  $X$  and  $Y$  are independent Brownian motions, we have

$$\begin{aligned} \langle B \rangle_t &= \left\langle \int_0^{(\cdot)} \cos(H_s) dX_s \right\rangle_t + \left\langle \int_0^{(\cdot)} \sin(H_s) dY_s \right\rangle_t + \left\langle \int_0^{(\cdot)} \cos(H_s) dX_s, \int_0^{(\cdot)} \sin(H_s) dY_s \right\rangle_t \\ &= \int_0^t (\cos(H_s))^2 d\langle X \rangle_s + \int_0^t (\sin(H_s))^2 d\langle Y \rangle_s + \int_0^t \cos(H_s) \sin(H_s) d\langle X, Y \rangle_s \\ &= \int_0^t (\cos(H_s))^2 ds + \int_0^t (\sin(H_s))^2 ds + 0 = \int_0^t ds \\ &= t. \end{aligned}$$

Similarly, we can show that  $\langle \hat{B} \rangle_t = t$ . Finally,

$$\begin{aligned} \langle B, \hat{B} \rangle_t &= \int_0^t \cos(H_s) \sin(H_s) d\langle X \rangle_s - \int_0^t (\cos(H_s))^2 d\langle X, Y \rangle_s \\ &\quad + \int_0^t (\sin(H_s))^2 d\langle Y, X \rangle_s - \int_0^t \sin(H_s) \cos(H_s) d\langle Y \rangle_s \\ &= \int_0^t \cos(H_s) \sin(H_s) ds - 0 + 0 - \int_0^t \sin(H_s) \cos(H_s) ds \\ &= 0. \end{aligned}$$

Hence the proof is complete.

3. [Soln]

a.)  $X$  is continuous and starts at a point  $x_0 \in (a, b)$  hence the exit time of this interval can be determined by  $T_a \wedge T_b$ .

b.) We can use Itô's formula to show that  $f(X_t)$  and  $f(X_{t \wedge T})$  are semimartingales. Note since

$$\langle X \rangle_s = \left\langle x_0 + \int_0^s \sigma(X_u) dB_u \right\rangle_s = \int_0^s \sigma^2(X_u) du$$

the decompositions are equal to

$$f(X_t) = f(x_0) + \int_0^t f'(X_s) \sigma(X_s) dB_s + \frac{1}{2} \int_0^t f''(X_s) \sigma^2(X_s) ds$$

resp. analogous for  $f(X_{t \wedge T})$  which is a combination of a local martingale and stochastic integral w.r.t. the Lebesgue measure.

c.) First we show that for all  $t > 0$ ,  $\langle M \rangle_t \in L^1$  by upper bounding

$$\left\langle \int_0^{\cdot \wedge T} f'(X_s) \sigma(X_s) dB_s \right\rangle_t \leq \int_0^t \sup_{0 \leq u \leq T} (f')^2(X_u) \sigma^2(X_u) du \leq Ct$$

where the last inequality follows from the fact that  $\sigma$  is continuous and  $f \in C^2$ . Since  $M$  is a martingale, via Doob's decomposition we get that  $(M_t^2 - \langle M \rangle_t)_{t \geq 0}$  is also a martingale and  $\mathbb{E}(M_t^2) = \mathbb{E}(\langle M \rangle_t) < \infty$ , hence  $M \in L^2$ .

d.) Direct consequence.

e.) Trivial.

f.) Follows from

$$\mathbb{E}(h(X_{t \wedge T})) = h(x_0) + \frac{1}{2} \mathbb{E} \left( \int_0^{t \wedge T} h''(X_s) \sigma^2(X_s) ds \right) = h(x_0) + \frac{1}{2} \mathbb{E} \left( \int_0^{t \wedge T} 1 ds \right).$$

g.) It is trivial to show that  $t \wedge T \in L^1$  and since  $t \wedge T \rightarrow T$  almost surely, we have by monotone convergence that  $T \in L^2$

h.) Trivial.

i.) From (d) we get

$$\begin{aligned} \mathbb{E}(w(X_{t \wedge T})) &= w(x_0) + \frac{1}{2} \mathbb{E} \left( \int_0^{t \wedge T} w''(X_s) \sigma^2(X_s) ds \right) = w(x_0) + \frac{1}{2} \mathbb{E}(t \wedge T) \\ &\stackrel{(f)}{=} \mathbb{E}(h(X_{t \wedge T})) - \frac{b-x_0}{b-a} \end{aligned}$$

we can take the limit  $t \rightarrow \infty$  and get

$$\mathbb{E}(w(X_T)) = \mathbb{E}(h(X_{t \wedge T})) + \frac{b-x_0}{b-a} = \frac{b-x_0}{b-a}$$

We use that  $w(X_T) = h(X_T) + \frac{b-X_T}{b-a}$  and solve the above equation

$$\mathbb{E}(X_T) = x_0.$$

The claim follows from  $\mathbb{E}(X_T) = a\mathbb{P}(X_T = a) + b\mathbb{P}(X_T = b)$  and  $\mathbb{P}(X_T = a) + \mathbb{P}(X_T = b) = 1$ .

4. [Soln]

a. We are going to consider solutions of the form

$$X_t = a(t) \left\{ x_0 + \int_0^t b(s) dB_s \right\}.$$

Let

- (i)  $Y_t = x_0 + \int_0^t b(s) dB_s$  be a local martingale and
- (ii)  $f(t, x) := a(t)x \in C^{1,2}(\mathbb{R}_+ \times \mathbb{R})$ .

Note that  $X_t = f(t, Y_t)$  and

$$f_t(t, x) = a'(t)x = \frac{a'(t)}{a(t)} f(t, x), \quad f_x(t, x) = a(t), \quad f_{xx}(t, x) = 0.$$

Using Itô formula we have

$$\begin{aligned} dX_t &= d[f(t, Y_t)] = f_t(t, Y_t) dt + f_x(t, Y_t) dY_t + \frac{1}{2} f_{xx}(t, Y_t) d[Y]_t \\ &= \frac{a'(t)}{a(t)} f(t, Y_t) dt + a(t) dY_t + 0 = \frac{a'(t)}{a(t)} X_t dt + a(t) b(t) dB_t. \end{aligned}$$

So, for  $X$  to be a solution to the SDE

$$dX_t = -\alpha X_t dt + \sigma dB_t$$

it suffices to take

$$\frac{a'(t)}{a(t)} = -\alpha, \text{ i.e., } a(t) = c e^{-\alpha t}, \text{ for some constant } c, \quad \text{and} \quad a(t)b(t) = \sigma.$$

To satisfy the initial condition  $X_0 = x_0$  we need

$$x_0 = f(0, Y_0) = a(0)x_0 \Leftrightarrow a(0) = 1 \Leftrightarrow c = 1.$$

Hence we consider  $a(t) = e^{-\alpha t}$  and  $b(t) = \sigma/a(t) = \sigma e^{\alpha t}$ .

Note that with this choice of  $a(\cdot)$  and  $b(\cdot)$ , both (i) and (ii) are satisfied. In fact,  $Y$  is a  $L^2$ -martingale, because it is a stochastic integral w.r.t. the Brownian motion where the integrand satisfies

$$\int_0^t b(s)^2 ds = \int_0^t \sigma^2 e^{\alpha s} ds = \frac{\sigma^2}{\alpha} (e^{\alpha t} - 1) < \infty, \quad \forall t.$$

Hence a solution to the SDE is given by

$$X_t = a(t) \left\{ x_0 + \int_0^t b(s) dB_s \right\} = x_0 e^{-\alpha t} + \sigma \int_0^t e^{-\alpha(t-s)} dB_s.$$

From the uniqueness of solution it follows that this is the only/unique solution.

b. We now consider the process

$$Y_t = \exp \left\{ X_t - \frac{\eta}{\alpha} \right\} = g(X_t), \text{ say, where } g(x) = e^{x-\eta/\alpha}.$$

Clearly  $g \in C^2(\mathbb{R})$ , with  $g' = g'' = g$ . Using Itô formula we have

$$dY_t = d[g(X_t)] = g'(X_t) dX_t + \frac{1}{2} g''(X_t) d[X]_t = Y_t [-\alpha X_t dt + \sigma dB_t] + \frac{1}{2} Y_t d[X]_t.$$

From the given SDE it follows that  $X_t$  satisfies

$$X_t = x_0 - \alpha \int_0^t X_s ds + \sigma B_t.$$

Since the second term in the above expression is of finite variation (a.s.), the quadratic variation of  $X$  is given by  $[X]_t = \sigma^2 [B]_t = \sigma^2 t$ . We then get the SDE satisfied by  $Y_t$ :

$$\begin{aligned}
dY_t &= -\alpha Y_t X_t dt + \sigma Y_t dB_t + \frac{1}{2} Y_t \sigma^2 dt \\
&= Y_t \left[ -\alpha X_t + \frac{1}{2} \sigma^2 \right] dt + \sigma Y_t dB_t \\
&= Y_t \left[ -\alpha \left( \ln Y_t + \frac{\eta}{\alpha} \right) + \frac{1}{2} \sigma^2 \right] dt + \sigma Y_t dB_t \\
&= Y_t \left( \frac{1}{2} \sigma^2 - \eta - \alpha \ln Y_t \right) dt + \sigma Y_t dB_t, \quad t > 0
\end{aligned}$$

with  $Y_0 = e^{x_0 - \eta/\alpha}$ .

c. Now consider the SDE satisfied by a geometric mean reverting process:

$$dr_t = r_t (\theta - \alpha \ln r_t) dt + \sigma r_t dB_t, \quad t > 0$$

with  $r_0 = 1$ .

Comparing this with the SDE in (b), satisfied by  $Y_t$ , we realize that it suffices to take

$$\frac{1}{2} \sigma^2 - \eta = \theta, \quad \text{i.e.,} \quad \eta = \frac{1}{2} \sigma^2 - \theta, \quad \text{and} \quad x_0 = \frac{\eta}{\alpha} = \frac{\sigma^2/2 - \theta}{\alpha}.$$

Hence from (a) and (b) it follows that the required solution is given by

$$r_t = \exp \left\{ \tilde{r}_t - \frac{\sigma^2/2 - \theta}{\alpha} \right\}$$

where

$$\tilde{r}_t = \frac{\sigma^2/2 - \theta}{\alpha} e^{-\alpha t} + \sigma \int_0^t e^{-\alpha(t-s)} dB_s.$$

5. **[Soln]**

Note that formally we can rewrite  $X_t$  under  $\mathbb{P}$  as follows.

$$\begin{aligned} X_t &= x + \int_0^t \mu_s ds + \int_0^t \sigma_s dB_s \\ &= x + \int_0^t \nu_s ds + \int_0^t (\mu_s - \nu_s) ds + \int_0^t \sigma_s dB_s \\ &= x + \int_0^t \nu_s ds + \int_0^t \sigma_s \left[ dB_s + \frac{\mu_s - \nu_s}{\sigma_s} ds \right] \end{aligned}$$

Hence by defining

$$\tilde{B}_t = B_t + \int_0^t \frac{\mu_s - \nu_s}{\sigma_s} ds = B_t - \int_0^t H_s ds \quad \text{with } H_t = \frac{\nu_t - \mu_t}{\sigma_t},$$

we have

$$(\star) \quad X_t = x + \int_0^t \nu_s ds + \int_0^t \sigma_s d\tilde{B}_s \quad \text{a.s. } [\mathbb{P}].$$

To obtain a new measure  $\mathbb{Q}$ , under which  $\tilde{B}$  is a Brownian motion, we are going to apply Girsanov theorem. Now assume that  $\sigma_t \neq 0 \forall t \in \mathbb{R}_+$ , so that  $H_t = \frac{\nu_t - \mu_t}{\sigma_t}$  is well defined. Clearly,  $H$  is an adapted process. To apply Girsanov theorem we need furthermore

$$(\dagger) \quad \int_0^T H_s^2 ds < \infty \quad \text{a.s. } [\mathbb{P}],$$

and  $Z_T \equiv Z_T(H)$  to be a  $\mathbb{P}$ -martingale, where

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

The martingale condition is satisfied if, for example,  $(\nu_t)$ ,  $(\mu_t)$  and  $(\sigma_t)$  are deterministic/non-random processes satisfying  $\sigma_t \neq 0$  and  $(\dagger)$ .

In this case,  $\tilde{B}$  is a BM under  $\mathbb{Q}$ , which is defined as  $d\mathbb{Q} = Z_T d\mathbb{P}$ . Since  $Z_T > 0$ , it follows that  $\mathbb{Q} \equiv \mathbb{P}$ , i.e., the null sets of both the measures are the same. From  $(\star)$  it then follows that

$$X_t = x + \int_0^t \nu_s ds + \int_0^t \sigma_s d\tilde{B}_s \quad \text{a.s. } [\mathbb{Q}]$$

where  $\tilde{B}$  is a  $\mathbb{Q}$ -BM.