## INSA Rennes, 4GM–AROM Random Models of Dynamical Systems Introduction to SDE's

### TD 2 : Some applications of the Itô formula

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**Exercise 1 [Integration by parts]** Let B(t) be a *d*-dimensional standard Brownian motion, and let  $X_1(t)$  and  $X_2(t)$  be two one-dimensional Itô processes, with

$$X_i(t) = X_i(0) + \int_0^t \psi_i(s) \, ds + \int_0^t \phi_i(s) \, dB(s) \qquad \text{for } i = 1, 2.$$

Here  $\phi_1(s)$  and  $\phi_2(s)$  are two  $1 \times d$  matrices (row vectors).

#### (i) Write the Itô formula for the one-dimensional process $X_1(t) X_2(t)$ .

#### \_\_\_\_\_ Solution \_\_\_\_\_

The two-dimensional process  $X(t) = (X_1(t), X_2(t))$  is an Itô process, with

$$\psi(s) = \begin{pmatrix} \psi_1(s) \\ \psi_2(s) \end{pmatrix} \quad \text{and} \quad \phi(s) = \begin{pmatrix} \phi_1(s) \\ \phi_2(s) \end{pmatrix}$$

The Itô formula for the Itô process X(t) and for the function  $f(x_1, x_2) = x_1 x_2$ , with

$$f'(x_1, x_2) = (x_2 \quad x_1)$$
 and  $f''(x_1, x_2) = \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix}$ ,

yields

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$$\begin{aligned} X_1(t) X_2(t) &= X_1(0) X_2(0) + \int_0^t (X_2(s) \quad X_1(s)) \left[ \begin{pmatrix} \psi_1(s) \\ \psi_2(s) \end{pmatrix} ds + \begin{pmatrix} \phi_1(s) \\ \phi_2(s) \end{pmatrix} dB(s) \right] \\ &+ \frac{1}{2} \int_0^t \operatorname{trace} \left[ \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix} \begin{pmatrix} \phi_1(s) \phi_1^*(s) & \phi_1(s) \phi_2^*(s) \\ \phi_2(s) \phi_1^*(s) & \phi_2(s) \phi_2^*(s) \end{pmatrix} \right] ds \\ &= X_1(0) X_2(0) + \int_0^t (X_2(s) \psi_1(s) + X_1(s) \psi_2(s)) ds \\ &+ \int_0^t (X_2(s) \phi_1(s) + X_1(s) \phi_2(s)) dB(s) \\ &+ \frac{1}{2} \int_0^t (\phi_1(s) \phi_2^*(s) + \phi_2(s) \phi_1^*(s)) ds , \end{aligned}$$

or in other words

$$X_1(t) X_2(t) = X_1(0) X_2(0) + \int_0^t X_2(s) dX_1(s) + \int_0^t X_1(s) dX_2(s) + \frac{1}{2} \int_0^t (\phi_1(s) \phi_2^*(s) + \phi_2(s) \phi_1^*(s)) ds .$$

Multi-dimensional version: Let B(t) be a *d*-dimensional standard Brownian motion, and let  $X_1(t)$  and  $X_2(t)$  be two *m*-dimensional Itô processes, with

$$X_i(t) = X_i(0) + \int_0^t \psi_i(s) \, ds + \int_0^t \phi_i(s) \, dB(s) \qquad \text{for } i = 1, 2.$$

Here  $\psi_1(s)$  and  $\psi_2(s)$  are two *m*-dimensional vectors, and  $\phi_1(s)$  and  $\phi_2(s)$  are two *m*×*d* matrices.

(ii) Write the Itô formula for the one-dimensional process  $X_1^*(t) X_2(t)$  and for the  $m \times m$  matrix-valued process  $X_1(t) X_2^*(t)$ .

[Hint: for the second part, apply the result obtained at question (i) to the two one-dimensional Itô processes  $u_1^* X_1(t)$  and  $u_2^* X_2(t)$  where  $u_1$  and  $u_2$  are two arbitrary vectors in  $\mathbb{R}^m$ .]

\_ Solution \_\_\_\_\_

First part (scalar product): The 2m-dimensional process  $X(t) = (X_1(t), X_2(t))$  is an Itô process, with

$$\psi(s) = \begin{pmatrix} \psi_1(s) \\ \psi_2(s) \end{pmatrix} \quad \text{and} \quad \phi(s) = \begin{pmatrix} \phi_1(s) \\ \phi_2(s) \end{pmatrix}$$

The Itô formula for the Itô process X(t) and for the function  $f(x_1, x_2) = x_1^* x_2$ , with

$$f'(x_1, x_2) = (x_2^* \quad x_1^*)$$
 and  $f''(x_1, x_2) = \begin{pmatrix} 0 & I \\ I & 0 \end{pmatrix}$ ,

yields

$$\begin{split} X_1(t) X_2^*(t) &= X_1^*(0) X_2(0) + \int_0^t (X_2^*(s) - X_1^*(s)) \left[ \begin{pmatrix} \psi_1(s) \\ \psi_2(s) \end{pmatrix} ds + \begin{pmatrix} \phi_1(s) \\ \phi_2(s) \end{pmatrix} dB(s) \right] \\ &+ \frac{1}{2} \int_0^t \operatorname{trace} \left[ \begin{pmatrix} 0 & I \\ I & 0 \end{pmatrix} \begin{pmatrix} \phi_1(s) \phi_1^*(s) - \phi_1(s) \phi_2^*(s) \\ \phi_2(s) \phi_1^*(s) - \phi_2(s) \phi_2^*(s) \end{pmatrix} \right] ds \\ &= X_1^*(0) X_2(0) + \int_0^t (X_2^*(s) \psi_1(s) + X_1^*(s) \psi_2(s)) ds \\ &+ \int_0^t (X_2^*(s) \phi_1(s) + X_1^*(s) \phi_2(s)) dB(s) \\ &+ \frac{1}{2} \int_0^t \operatorname{trace} \left[ \phi_1(s) \phi_2^*(s) + \phi_2(s) \phi_1^*(s) \right] ds \;, \end{split}$$

or in other words

$$X_1^*(t) X_2(t) = X_1^*(0) X_2(0) + \int_0^t X_2^*(s) dX_1(s) + \int_0^t X_1^*(s) dX_2(s) + \frac{1}{2} \int_0^t \operatorname{trace}[\phi_1(s) \phi_2^*(s) + \phi_2(s) \phi_1^*(s)] ds .$$

Second part: If  $u_1$  and  $u_2$  are *m*-dimensional vectors, then  $u_1^* X_1(t)$  and  $u_2^* X_2(t)$  are two onedimensional Itô processes, with

$$u_i^* X_i(t) = u_i^* X_i(0) + \int_0^t u_i^* \psi_i(s) \, ds + \int_0^t u_i^* \phi_i(s) \, dB(s) \qquad \text{for } i = 1, 2.$$

Here  $u_1^* \phi_1(s)$  and  $u_1^* \phi_2(s)$  are two  $1 \times d$  matrices (row vectors). Applying the result obtained at question (i) yields

$$u_1^* X_1(t) u_2^* X_2(t) = u_1^* X_1(0) u_2^* X_2(0) + \int_0^t u_2^* X_2(s) d[u_1^* X_1(s)] + \int_0^t u_1^* X_1(s) d[u_2^* X_2(s)] + \frac{1}{2} \int_0^t [u_1^* \phi_1(s) u_2^* \phi_2^*(s) + u_2^* \phi_2(s) u_1^* \phi_1^*(s)] ds ,$$

or equivalently, after rearranging terms

$$u_1^* X_1(t) X_2^*(t) u_2 = u_1^* X_1(0) X_2^*(0) u_2 + \int_0^t u_1^* dX_1(s) X_2^*(s) u_2 + \int_0^t u_1^* X_1(s) dX_2^*(s) u_2 + \int_0^t u_1^* \phi_1(s) \phi_2^*(s) u_2 ds ,$$

or in other words

$$\begin{aligned} X_1(t) X_2^*(t) \ &= \ X_1(0) X_2^*(0) + \int_0^t dX_1(s) X_2^*(s) + \int_0^t X_1(s) \ dX_2^*(s) \\ &+ \int_0^t \phi_1(s) \ \phi_2^*(s) \ ds \ , \end{aligned}$$

since the vectors  $u_1$  and  $u_2$  are arbitrary.

**Problem 2** [Burkholder–Davis–Gundy inequalities] Let *B* be a one–dimensional standard Brownian motion, and for any  $\phi \in M^2([0,T])$  define

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$$M(t) = \int_0^t \phi(s) \, dB(s) \,, \qquad M_*(t) = \max_{0 \le s \le t} |M(s)| \,, \qquad A(t) = \int_0^t |\phi(s)|^2 \, ds \,.$$

The objective is to show that for any  $p \ge 2$ , there exist positive constants  $0 < c_p \le C_p < \infty$ such that for any  $0 \le t \le T$ 

$$c_p \mathbb{E}|A(t)|^{p/2} \le \mathbb{E}|M_*(t)|^p \le C_p \mathbb{E}|A(t)|^{p/2}$$

i.e. the moments of a martingale can be controlled in terms of the moments of its increasing process.

#### (i) Show that the upper bound holds for p = 2.

[Hint: use the Doob inequality.]

Solution \_\_\_\_\_

The Doob inequality for p = 2 provides a uniform control in terms of the terminal value

$$\mathbb{E}[\max_{0 \le s \le t} |M(s)|^2] \le 4 \mathbb{E}|M(t)|^2 ,$$

and the Itô isometry yields

$$\mathbb{E}|M(t)|^2 = \mathbb{E}\int_0^t |\phi(s)|^2 \, ds = \mathbb{E}[A(t)] \, .$$

Combining the two estimates provide a uniform control in terms of the increasing process, i.e.

$$\mathbb{E}[\max_{0 \le s \le t} |M(s)|^2] \le 4 \mathbb{E}[A(t)] .$$

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The following boundedness assumption will be used:

there exists a positive K > 0 such that  $A(T) \le K^2$  and  $|M(t)| \le K$  for any  $0 \le t \le T$ .

# (ii) Assume that the result holds under the boundedness assumption. Show that the result can be extended to the general case.

[Hint: for any  $n \ge 1$ , consider the stopping time

$$\tau_n = \inf\{0 \le t \le T : |M(t)| \ge n \text{ or } A(t) \ge n\} \quad \text{or} \quad \tau_n = T,$$

and the stopped martingale defined by  $M^n(t) = M(t \wedge \tau_n)$  for any  $0 \le t \le T$ .]

From now on, the boundedness assumption is made.

Upper bound:

#### (iii) Show that the *p*-th order moment $\mathbb{E}|M_*(t)|^p$ can be bounded in terms of $\mathbb{E}|M(t)|^p$ .

[Hint: use the Doob inequality.]

\_\_\_\_\_ Solution \_\_

It follows from the Doob inequality that

$$\{\mathbb{E}[\max_{0 \le s \le t} |M(s)|^p]\}^{1/p} \le \frac{p}{p-1} \{\mathbb{E}|M(t)|^p\}^{1/p},\$$

or in other words

$$\mathbb{E}|M_*(t)|^p = \mathbb{E}\left[\max_{0 \le s \le t} |M(s)|^p\right] \le \left(\frac{p}{p-1}\right)^p \mathbb{E}|M(t)|^p$$

(iv) Write the Itô formula for  $|M(t)|^p$ . Show that the *p*-th order moment  $\mathbb{E}|M(t)|^p$  can be bounded in terms of  $\mathbb{E}[|M_*(t)|^{p-2} A(t)]$ , and further bounded in terms of  $\mathbb{E}|M_*(t)|^p$  and  $\mathbb{E}|A(t)|^{p/2}$ .

[Hint: for the last part, use the Hölder inequality.]

\_\_\_\_\_ Solution \_

The Itô formula for the Itô process

$$M(t) = \int_0^t \phi(s) \, dB(s) \, ,$$

and for the function  $g(x) = |x|^p$ , with

$$f'x) = p |x|^{p-1} \operatorname{sign}(x)$$
 and  $f''(x) = p (p-1) |x|^{p-2}$ ,

yields

$$|M(t)|^{p} = p \int_{0}^{t} |M(s)|^{p-1} \operatorname{sign}(M(s)) \phi(s) \, dB(s) + \frac{1}{2} p \left(p-1\right) \int_{0}^{t} |M(s)|^{p-2} \, |\phi(s)|^{2} \, ds$$

Under the boundedness assumption, the integrand  $s \mapsto |M(s)|^{p-1} \operatorname{sign}(M(s)) \phi(s)$  belongs to  $M^2([0,T])$ , since

$$\mathbb{E} \int_0^T ||M(s)|^{p-1} \operatorname{sign}(M(s)) \phi(s)|^2 \, ds \le K^{2p-2} \, \mathbb{E} \int_0^T |\phi(s)|^2 \, ds \le K^{2p} < \infty \,,$$

and therefore the stochastic integral is a martingale and its expectation is zero. Taking expectation yields

$$\begin{split} \mathbb{E}|M(t)|^{p} &= \frac{1}{2} p \left(p-1\right) \mathbb{E}\left[\int_{0}^{t} |M(s)|^{p-2} |\phi(s)|^{2} ds\right] \\ &\leq \frac{1}{2} p \left(p-1\right) \mathbb{E}\left[\max_{0 \leq s \leq t} |M(s)|^{p-2} \int_{0}^{t} |\phi(s)|^{2} ds\right] \\ &= \frac{1}{2} p \left(p-1\right) \mathbb{E}\left[|M_{*}(t)|^{p-2} A(t)\right]. \end{split}$$

The Hölder inequality for conjugate exponents q = p/(p-2) and  $q' = \frac{1}{2}p$  yields

$$\mathbb{E}[|M_*(t)|^{p-2} A(t)] \leq \{\mathbb{E}|M_*(t)|^{q(p-2)}\}^{1/q} \{\mathbb{E}|A(t)|^{q'}\}^{1/q'}$$

$$= \{\mathbb{E}|M_*(t)|^p\}^{1-2/p} \{\mathbb{E}|A(t)|^{p/2}\}^{2/p}$$

hence

$$\mathbb{E}|M(t)|^{p} \leq \frac{1}{2} p (p-1) \left\{ \mathbb{E}|M_{*}(t)|^{p} \right\}^{1-2/p} \left\{ \mathbb{E}|A(t)|^{p/2} \right\}^{2/p}$$

Reporting this inequality in the inequality obtained in question (iii) yields

$$\mathbb{E}|M_*(t)|^p \leq \left(\frac{p}{p-1}\right)^p \mathbb{E}|M(t)|^p$$
  
$$\leq \left(\frac{p}{p-1}\right)^p \frac{1}{2} p\left(p-1\right) \left\{\mathbb{E}|M_*(t)|^p\right\}^{1-2/p} \left\{\mathbb{E}|A(t)|^{p/2}\right\}^{2/p},$$

and collecting all  $\mathbb{E}|M_*(t)|^p$  terms in the left-hand side yields

$$\{\mathbb{E}|M_*(t)|^p\}^{2/p} \le \left(\frac{p}{p-1}\right)^p \frac{1}{2} p\left(p-1\right) \{\mathbb{E}|A(t)|^{p/2}\}^{2/p} ,$$

or equivalently

$$\mathbb{E}|M_*(t)|^p \le \left[\left(\frac{p}{p-1}\right)^p \frac{1}{2}p(p-1)\right]^{p/2} \mathbb{E}|A(t)|^{p/2}$$

Lower bound: Define

$$Y(t) = \int_0^t |A(s)|^{(p-1)/2} \phi(s) \, dB(s) \, .$$

(v) Show that 
$$E|Y(t)|^2 = \frac{1}{p} \mathbb{E}|A(t)|^p$$
.

#### \_\_\_\_\_ Solution \_\_\_\_\_

Under the boundedness assumption, the integrand  $s \mapsto |A(s)|^{(p-1)/2} \phi(s)$  belongs to  $M^2([0,T])$ , since

$$\mathbb{E}\int_0^T |A(s)|^{p-1} |\phi(s)|^2 \, ds \le K^{2p-2} \, \mathbb{E}\int_0^T |\phi(s)|^2 \, ds \le K^{2p} < \infty \, ,$$

and the Itô isometry yields

$$\mathbb{E}|Y(t)|^{2} = \mathbb{E}\int_{0}^{T} |A(s)|^{p-1} |\phi(s)|^{2} ds .$$

On the other hand, the usual chain rule yields

$$\frac{d}{dt}|A(t)|^{p} = p |A(t)|^{p-1} |\phi(t)|^{2} ,$$

or, in integrated form

$$|A(t)|^{p} = p \int_{0}^{t} |A(s)|^{p-1} |\phi(s)|^{2} ds .$$

Taking expectation yields

$$\mathbb{E}|A(t)|^{p} = p \mathbb{E} \int_{0}^{t} |A(t)|^{p-1} |\phi(s)|^{2} ds = p \mathbb{E}|Y(t)|^{2} . \Box$$

(vi) Using the Itô formula for  $M(t) |A(t)|^{(p-1)/2}$ , get an alternate expression for Y(t)and show the bound  $|Y(t)| \leq 2 M_*(t) |A(t)|^{(p-1)/2}$ . Show that the *p*-th order moment  $\mathbb{E}|A(t)|^p$  can be bounded in terms of  $\mathbb{E}|M_*(t)|^{2p}$  and  $\mathbb{E}|A(t)|^p$ . [Hint: for the last part, use the Hölder inequality.]

\_\_\_\_\_ Solution \_\_\_\_

Note that the usual chain rule yields

$$\frac{d}{dt} |A(t)|^{(p-1)/2} = \frac{1}{2} (p-1) |A(t)|^{(p-3)/2} |\phi(t)|^2 ,$$

or, in integrated form

$$|A(t)|^{(p-1)/2} = \frac{1}{2} (p-1) \int_0^t |A(s)|^{(p-3)/2} |\phi(s)|^2 ds .$$

The Itô formula (integration by parts) for the two component Itô process

$$\begin{pmatrix} M(t) \\ |A(t)|^{(p-1)/2} \end{pmatrix} = \frac{1}{2} (p-1) \int_0^t \begin{pmatrix} 0 \\ |A(s)|^{(p-3)/2} |\phi(s)|^2 \end{pmatrix} ds + \int_0^t \begin{pmatrix} \phi(s) \\ 0 \end{pmatrix} dB(s) ,$$

yields

$$M(t) |A(t)|^{(p-1)/2} = \frac{1}{2} (p-1) \int_0^t M(s) |A(s)|^{(p-3)/2} |\phi(s)|^2 ds + \int_0^t |A(s)|^{(p-1)/2} \phi(s) dB(s) ,$$

hence

$$Y(t) = \int_0^t |A(s)|^{(p-1)/2} \phi(s) dB(s)$$
  
=  $M(t) |A(t)|^{(p-1)/2} - \frac{1}{2} (p-1) \int_0^t M(s) |A(s)|^{(p-3)/2} |\phi(s)|^2 ds$ .

Note that

$$|M(t)|A(t)|^{(p-1)/2} \le M_*(t) |A(t)|^{(p-1)/2}$$
,

and

$$\begin{aligned} |\frac{1}{2}(p-1) \int_0^t M(s) |A(s)|^{(p-3)/2} |\phi(s)|^2 \, ds| &\leq M_*(t) \frac{1}{2}(p-1) \int_0^t |A(s)|^{(p-3)/2} |\phi(s)|^2 \, ds\\ &\leq M_*(t) |A(t)|^{(p-1)/2} \, .\end{aligned}$$

and the triangle inequality yields

$$|Y(t)| \le 2 M_*(t) |A(t)|^{(p-1)/2}$$
.

Therefore

$$\mathbb{E}|A(t)|^{p} = p \mathbb{E}|Y(t)|^{2} \le 4p \mathbb{E}[|M_{*}(t)|^{2} |A(t)|^{p-1}].$$

The Hölder inequality for conjugate exponents q = p/(p-1) and q' = p yields

$$\mathbb{E}[|M_*(t)|^2 |A(t)|^{p-1}] \leq \{\mathbb{E}|M_*(t)|^{2q'}\}^{1/q'} \{\mathbb{E}|A(t)|^{q(p-1)}\}^{1/q}$$
$$= \{\mathbb{E}|M_*(t)|^{2p}\}^{1/p} \{\mathbb{E}|A(t)|^p\}^{1-1/p},$$

hence

$$\mathbb{E}|A(t)|^{p} \leq 4p \ \{\mathbb{E}|M_{*}(t)|^{2p}\}^{1/p} \ \{\mathbb{E}|A(t)|^{p}\}^{1-1/p} \ .$$

and collecting all  $\mathbb{E}|A(t)|^p$  terms in the left-hand side yields

$$\{\mathbb{E}|A(t)|^p\}^{1/p} \le 4p \{\mathbb{E}|M_*(t)|^{2p}\}^{1/p}$$

or equivalently

$$\mathbb{E}|A(t)|^p \le (4p)^p \mathbb{E}|M_*(t)|^{2p}$$
.

**Exercice 3 [Exponential bound]** Let B be a one-dimensional standard Brownian motion, and for any  $\phi \in M^2([0,T])$  define

$$M(t) = \int_0^t \phi(s) \, dB(s) \, , \qquad A(t) = \int_0^t |\phi(s)|^2 \, ds \, .$$

For any positive  $\lambda > 0$  define

$$Z^{\lambda}(t) = \exp\{\lambda M(t) - \frac{1}{2}\lambda^2 A(t)\}.$$

In the special case where  $\phi \equiv 1$ , the process M is a Brownian motion, and the process  $Z^{\lambda}$  is a martingale. This was shown using the expression of the Laplace transform of a Gaussian r.v. and this trick cannot be used in the general case.

(i) Write the Itô formula for  $Z^{\lambda}(t)$ , show that it is a (local) martingale, hence  $\mathbb{E}[Z^{\lambda}(t)] \leq 1$  for any  $0 \leq t \leq T$ .

[Hint: a nonnegative local martingale is a (true) supermartingale.]

SOLUTION \_\_\_

The Itô formula for the Itô process

$$X(t) = \lambda M(t) - \frac{1}{2} \lambda^2 A(t) = \int_0^t (-\frac{1}{2} \lambda^2 |\phi(s)|^2) \, ds + \int_0^t \lambda \, \phi(s) \, dB(s) \, ,$$

and for the function  $f(x) = \exp\{x\}$ , with

$$f'(x) = f''(x) = \exp\{x\}$$
,

yields

$$Z^{\lambda}(t) = 1 + \int_{0}^{t} Z^{\lambda}(s) \left[ -\frac{1}{2} \lambda^{2} |\phi(s)|^{2} ds + \lambda \phi(s) dB(s) \right] + \frac{1}{2} \int_{0}^{t} Z^{\lambda}(s) \lambda^{2} |\phi(s)|^{2} ds$$

$$= 1 + \lambda \int_{0}^{t} Z^{\lambda}(s) \phi(s) dB(s) .$$
(\*)

The integrand  $s\mapsto Z^\lambda(s)\,\phi(s)$  belongs to  $M^2_{\rm loc}$  only, since

$$\int_0^T |Z^{\lambda}(t) \phi(t)|^2 dt \le \max_{0 \le t \le T} |Z^{\lambda}(t)|^2 \int_0^T |\phi(t)|^2 dt < \infty$$

almost surely, for any  $T \ge 0$ , and therefore the stochastic integral is only a (nonnegative) local martingale, and its expectation is not necessarily zero.

However, a nonnegative local martingale is a supermartingale. Indeed, let L be a nonnegative local martingale, i.e. there exists a non-decreasing sequence of stopping times, such that  $\tau_n \uparrow \infty$  almost surely and such that the stopped process defined by  $L(t \land \tau_n)$  for any  $t \ge 0$  is a martingale. Then, for any  $0 \le s \le t$  and any  $A \in \mathcal{F}(s)$  it holds

$$\mathbb{E}[1_{A} \ 1_{(\tau_{n} \ge s)} \ L(s)] = \mathbb{E}[1_{A} \ 1_{(\tau_{n} \ge s)} \ L(s \land \tau_{n})] = \mathbb{E}[1_{A} \ 1_{(\tau_{n} \ge s)} \ L(t \land \tau_{n})] ,$$

since  $A \cap \{\tau_n \ge s\} \in \mathcal{F}(s)$ . The Lebesgue dominated convergence theorem yields

$$\lim_{n\uparrow\infty} \mathbb{E}[\mathbf{1}_A \ \mathbf{1}_{\left(\tau_n \geq s\right)} \ L(s)] = \mathbb{E}[\mathbf{1}_A \ L(s)] \ ,$$

and the Fatou lemma yields

$$\lim_{n\uparrow\infty} \mathbb{E}[1_A \ 1_{(\tau_n \ge s)} \ L(t \land \tau_n)] \ge \mathbb{E}[1_A \ \liminf_{n\uparrow\infty} [1_{(\tau_n \ge s)} \ L(t \land \tau_n)]] = \mathbb{E}[1_A \ L(t)] ,$$

hence

$$\mathbb{E}[\mathbf{1}_A \ L(s)] \ge \mathbb{E}[\mathbf{1}_A \ L(t)] \ ,$$

for any  $A \in \mathcal{F}(s)$ , i.e.

$$L(s) \ge \mathbb{E}[L(t) \mid \mathcal{F}(s)]$$
.

Therefore, the stochastic integral in  $(\star)$  is a supermartingale and its expectation is smaller than zero, hence

$$\mathbb{E}[Z^{\lambda}(t)] \geq 1 \; .$$

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(ii) Assume that  $A(t) \leq Kt$  for any  $0 \leq t \leq T$ . Show the following exponential bound: for any positive c > 0

$$\mathbb{P}[\max_{0 \le t \le T} |M(t)| > c] \le 2 \exp\{-\frac{c^2}{2 K T}\}.$$

[Hint: use the Chernoff approach to large deviations estimates, and the (easy) inequality

$$\mu \mathbb{P}[\max_{0 \le t \le T} X(t) > \mu] \le \mathbb{E}[X(0)] \ ,$$

valid for any nonnegative supermartingale.]

\_\_\_\_ Solution \_\_\_\_

The proof of the inequality for a nonnegative supermartingale follows the same lines as in the discrete times case. Indeed, let L be a nonnegative supermartingale, and introduce the bounded stopping time

$$\tau = \inf\{0 \le s \le t : L(s) \ge \mu\},\$$

if such time exists, and  $\tau = t$  otherwise.

Note that if  $\max_{0 \le s \le t} L(s) \ge \mu$ , then  $L(s) \ge \mu$  for some  $0 \le s \le t$ , hence  $L(\tau) \ge \mu$ . It follows from the entired compliant theorem that

It follows from the optional sampling theorem that

$$\mathbb{E}[L(0)] \geq \mathbb{E}[L(\tau)]$$
  
=  $\mathbb{E}[1(\max_{0 \leq s \leq t} L(s) \geq \mu) L(\tau)] + \mathbb{E}[1(\max_{0 \leq s \leq t} L(s) < \mu) L(\tau)]$   
 $\geq \mu \mathbb{P}[\max_{0 \leq s \leq t} L(s) \geq \mu],$ 

hence

$$\mu \mathbb{P}[\max_{0 \le s \le t} L(s) \ge \mu] \le \mathbb{E}[L(0)] ,$$

and the claim is proved.

Note that for any  $0 \le s \le t$ 

$$\lambda M(s) = \lambda M(s) - \frac{1}{2} \lambda^2 A(s) + \frac{1}{2} \lambda^2 A(s) \le \lambda M(s) - \frac{1}{2} \lambda^2 A(s) + \frac{1}{2} \lambda^2 K t ,$$

hence for any positive  $\lambda > 0$ 

$$\begin{aligned} \max_{0 \le s \le t} M(s) \ge c \ \Rightarrow \ \max_{0 \le s \le t} [\lambda \, M(s) - \frac{1}{2} \, \lambda^2 \, A(s)] + \frac{1}{2} \, \lambda^2 \, K \, t \ge \lambda \, c \\ \Rightarrow \ \max_{0 \le s \le t} [\lambda \, M(s) - \frac{1}{2} \, \lambda^2 \, A(s)] \ge \lambda \, c - \frac{1}{2} \, \lambda^2 \, K \, t \\ \Rightarrow \ \max_{0 \le s \le t} Z^{\lambda}(s) \ge \exp\{\lambda \, c - \frac{1}{2} \, \lambda^2 \, K \, t\} \; .\end{aligned}$$

Therefore

$$\mathbb{P}[\max_{0 \le s \le t} M(s) \ge c] \le \mathbb{P}[\max_{0 \le s \le t} Z^{\lambda}(s) \ge \exp\{\lambda c - \frac{1}{2}\lambda^2 K t\}]$$
$$\le \exp\{-\lambda c + \frac{1}{2}\lambda^2 K t\},$$

since  $Z^{\lambda}$  is a nonnegative supermartingale with  $Z^{\lambda}(0) = 1$ .

Similarly, note that for any  $0 \le s \le t$ 

$$\begin{split} \lambda\left(-M(s)\right) &= -\lambda M(s) - \frac{1}{2}\,\lambda^2 A(s) + \frac{1}{2}\,\lambda^2 A(s) \le -\lambda M(s) - \frac{1}{2}\,\lambda^2 A(s) + \frac{1}{2}\,\lambda^2 K t \ , \end{split}$$
 for any positive  $\lambda > 0$ 

$$\begin{aligned} \max_{0 \le s \le t} (-M(s)) \ge c \ \Rightarrow \ \max_{0 \le s \le t} [-\lambda M(s) - \frac{1}{2} \lambda^2 A(s)] + \frac{1}{2} \lambda^2 K t \ge \lambda c \\ \Rightarrow \ \max_{0 \le s \le t} [-\lambda M(s) - \frac{1}{2} \lambda^2 A(s)] \ge \lambda c - \frac{1}{2} \lambda^2 K t \\ \Rightarrow \ \max_{0 \le s \le t} Z^{-\lambda}(s) \ge \exp\{\lambda c - \frac{1}{2} \lambda^2 K t\} \;. \end{aligned}$$

Therefore

hence

$$\mathbb{P}[\max_{0 \le s \le t} (-M(s)) \ge c] \le \mathbb{P}[\max_{0 \le s \le t} Z^{-\lambda}(s) \ge \exp\{\lambda c - \frac{1}{2}\lambda^2 K t\}]$$
$$\le \exp\{-\lambda c + \frac{1}{2}\lambda^2 K t\} ,$$

since  $Z^{-\lambda}$  is a nonnegative supermartingale with  $Z^{-\lambda}(0) = 1$ . Combining the two estimates yields

$$\begin{split} \mathbb{P}[\max_{0 \le s \le t} |M(s)| \ge c] &\leq \mathbb{P}[\max_{0 \le s \le t} M(s) \ge c] + \mathbb{P}[\min_{0 \le s \le t} M(s) \le -c] \\ &= \mathbb{P}[\max_{0 \le s \le t} M(s) \ge c] + \mathbb{P}[\max_{0 \le s \le t} (-M(s)) \ge c] \\ &\leq 2 \exp\{-\lambda \, c + \frac{1}{2} \, \lambda^2 \, K \, t\} \; . \end{split}$$

The bound holds for any positive  $\lambda > 0$ , hence it holds also for the minimum over all possible values of  $\lambda > 0$ . The minimum is achieved for  $\lambda = c/(Kt) > 0$  and the minimum value is  $2 \exp\{-\frac{1}{2}c^2/(Kt)\}$ , hence

$$\mathbb{P}[\max_{0 \le s \le t} |M(s)| \ge c] \le 2 \min_{\lambda > 0} \exp\{-\lambda c + \frac{1}{2}\lambda^2 K\} = 2 \exp\{-\frac{1}{2}\frac{c^2}{Kt}\}.$$

(iii) In the general case, show that for any positive c, K > 0

$$\mathbb{P}[\max_{0 \le t \le T} |M(t)| > c] \le 2 \exp\{-\frac{c^2}{2 K T}\} + \mathbb{P}[A(T) > K T] .$$

\_\_\_\_\_ Solution \_\_\_\_\_

Simply

$$\begin{split} \mathbb{P}[\max_{0 \le t \le T} |M(t)| > c] &= \mathbb{P}[\max_{0 \le t \le T} |M(t)| > c, A(T) \le KT] + \mathbb{P}[\max_{0 \le t \le T} |M(t)| > c, A(T) > KT] \\ &\le 2 \exp\{-\frac{c^2}{2 \, KT}\} + \mathbb{P}[A(T) > KT] \;. \end{split}$$

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Exercise 4 [Feynman–Kac formula] Consider the following linear parabolic PDE

$$\frac{\partial u}{\partial t}(t,x) = \frac{1}{2} \Delta u(t,x) - c(x) u(t,x) , \quad \text{for any } (t,x) \in [0,\infty) \times \mathbb{R}^d$$

with initial condition u(0, x) = g(x) for any  $x \in \mathbb{R}^d$ . Here, the coefficient c(x) is non-negative and bounded from above, its derivative c'(x) is bounded, and the initial condition g(x) together with its derivative g'(x) are bounded. It is assumed that a solution u(t, x) exists that is  $C^{1,2}$ with a bounded derivative w.r.t. the space variable.

Fix  $x \in \mathbb{R}^d$  and let B be a standard d-dimensional Brownian motion starting from B(0) = x.

#### (i) Fix t > 0, and write the Itô formula for

$$u(t-s, B(s)) \exp\{-\int_0^s c(B(r)) dr\}$$
.

[Hint: show that the process

$$V(s) = \exp\{-\int_0^s c(B(r)) \, dr\}$$

is an Itô process, and write the Itô formula for the (d+1)-dimensional Itô process (B(s), V(s))and for the time-dependent function f(s, x, v) = u(t - s, x) v.]

The usual chain rule yields

$$\frac{d}{dt}V(t) = -c(B(t)) V(t) ,$$

or in integrated form

$$V(s) = 1 - \int_0^s c(B(r))) V(r) \, dr \; .$$

Next, the Itô formula for the (d+1)-dimensional Itô process

$$\begin{pmatrix} B(s)\\V(s) \end{pmatrix} = \begin{pmatrix} 0\\1 \end{pmatrix} + \int_0^s \begin{pmatrix} 0\\-c(B(r))V(r) \end{pmatrix} dr + \int_0^s \begin{pmatrix} I\\0 \end{pmatrix} dB(r) ,$$

and for the time–dependent function f(s, x, v) = u(t - s, x) v, with

$$\frac{\partial f}{\partial t}(s,x,v) = -\frac{\partial u}{\partial t}(t-s,x) v$$

and

$$f'(s,x,v) = \left(\frac{\partial u}{\partial x}(t-s,x) \ v \quad u(t-s,x)\right) \quad \text{and} \quad f''(s,x,v) = \begin{pmatrix} \frac{\partial^2 u}{\partial x^2}(t-s,x) \ v \quad \frac{\partial u}{\partial x}(t-s,x) \\ \frac{\partial u}{\partial x}(t-s,x) \ v \quad 0 \end{pmatrix}$$

yields

$$\begin{split} u(t-s,B(s))\,V(s) \,&=\, u(t,x) + \int_0^s \left[-\frac{\partial u}{\partial t}(t-r,B(r))\,V(r)\right]dr \\ &+ \int_0^s \left(\frac{\partial u}{\partial x}(t-r,B(r))\,V(r) - u(t-r,B(r))\right) \, \left(\begin{array}{c} 0\\ -c(B(r))\,V(r) \end{array}\right)dr \\ &+ \int_0^s \left(\frac{\partial u}{\partial x}(t-r,B(r))\,V(r) - u(t-r,B(r))\right) \, \left(\begin{array}{c} I\\ 0 \end{array}\right)dB(r) \\ &+ \int_0^s \operatorname{trace}\left[\left(\begin{array}{c} \frac{\partial^2 u}{\partial x^2}(t-r,B(r))\,V(r) - \frac{\partial u}{\partial x}(t-r,B(r)) \\ \frac{\partial u}{\partial x}(t-r,B(r)) & 0 \end{array}\right) \, \left(\begin{array}{c} I & 0\\ 0 & 0 \end{array}\right)\right]dr \\ &= ui(t,x) - \int_0^s \frac{\partial u}{\partial t}(t-r,B(r))\,V(r)\,dr \\ &- \int_0^s c(B(r))\,u(t-r,B(r))\,V(r)\,dr \\ &+ \int_0^s \frac{\partial u}{\partial x}(t-r,B(r))\,V(r)\,dB(r) \\ &+ \frac{1}{2}\int_0^s \Delta u(t-r,B(r))\,V(r)\,dr \end{split}$$

Collecting all the ordinary integral terms reduces to

$$-\int_{0}^{s} \left[\frac{\partial u}{\partial t}(t-r,B(r)) - \frac{1}{2}\Delta u(t-r,B(r)) + c(B(r))u(t-r,B(r))\right]V(r)\,dr = 0$$

since |

$$\frac{\partial u}{\partial t}(s,y) - \frac{1}{2}\Delta u(s,y) + c(y) u(s,y) = 0 ,$$

for any  $s \ge 0$  and any  $y \in \mathbb{R}^d$ , and the identity holds in particular for s = t - r and for y = B(r). Therefore

$$u(t-s,B(s))V(s) = u(t,x) + \int_0^s \frac{\partial u}{\partial x}(t-r,B(r))V(r)\,dB(r)\;,$$

and in particular for s = t it holds

$$u(t,x) + \int_0^t \frac{\partial u}{\partial x}(t-r,B(r)) V(r) \, dB(r) = u(0,B(t)) V(t) = g(B(t)) \, \exp\{-\int_0^t c(B(r)) \, dr\} \, .$$

(ii) Show that

$$u(t,x) = \mathbb{E}_{0,x}[g(B(t)) \exp\{-\int_0^t c(B(r)) \, dr\}] \; .$$

#### (iii) Check a posteriori that the derivative w.r.t. the space variable is bounded.

[Hint: write  $B(t) = x + B_0(t)$  with another standard *d*-dimensional Brownian motion starting from  $B_0(0) = 0$ .]

**Exercise 5** [Wong–Zakai approximation] Let B be a one–dimensional standard Brownian motion, and let

$$B_n(t) = \int_0^t \dot{B}_n(s) \, ds \; ,$$

be an absolutely continuous approximation, such that  $B_n(t) \to B(t)$  almost surely as  $n \uparrow \infty$ . Let f be a twice differentiable function, and let u = f'.

#### (i) Write the usual chain rule (change of variable formula) for $f(B_n(t))$ .

\_\_\_\_\_ Solution \_\_\_\_

The usual chain rule yields

$$\frac{d}{dt}f(B_n(t)) = f'(B_n(t))\dot{B}_n(t) ,$$

or in integrated form

$$f(B_n(t)) = f(B_n(0)) + \int_0^t f'(B_n(s)) \dot{B}_n(s) \, ds \; .$$

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#### (ii) Write the Itô formula for f(B(t)).

 $\_$  Solution  $\_$ 

The Itô formula yields

$$f(B(t)) = f(B(0)) + \int_0^t f'(B(s)) \, dB(s) + \frac{1}{2} \int_0^t f''(B(s)) \, ds \; .$$

(iii) What is the limit as  $n \uparrow \infty$  of

$$\int_0^t u(B_n(s)) \, dB_n(s) = \int_0^t u(B_n(s)) \, \dot{B}_n(s) \, ds \ ?$$

\_\_\_\_\_ Solution \_\_\_\_\_

Clearly,  $f(B_n(t)) - f(B_n(0)) \rightarrow f(B(t)) - f(B(0))$ , hence

$$\int_0^t u(B_n(s)) \dot{B}_n(s) \, ds \to \int_0^t u(B(s)) \, dB(s) + \frac{1}{2} \, \int_0^t u'(B(s)) \, ds \; ,$$

as  $n \uparrow \infty$ .

As an illustration, one can consider the polygonal approximation

$$B_n(s) = \frac{B(t_{i-1}^n) \left(t_i^n - s\right) + B(t_i^n) \left(s - t_{i-1}^n\right)}{t_i^n - t_{i-1}^n} \qquad \text{for any } t_{i-1}^n \le s \le t_i^n$$

associated with a convergent partition  $0 = t_0^n < t_1^n < \cdots < t_n^n = t$  of [0, t].

Clearly

$$\int_{0}^{t} u(B_{n}(s)) \dot{B}_{n}(s) ds = \sum_{i=1}^{n} \int_{t_{i-1}^{n}}^{t_{i}^{n}} u(B_{n}(s)) \frac{B(t_{i}^{n}) - B(t_{i-1}^{n})}{t_{i}^{n} - t_{i-1}^{n}} ds$$
$$= \sum_{i=1}^{n} \left[\frac{1}{t_{i}^{n} - t_{i-1}^{n}} \int_{t_{i-1}^{n}}^{t_{i}^{n}} u(B_{n}(s)) ds\right] \left(B(t_{i}^{n}) - B(t_{i-1}^{n})\right)$$

fails to converge to the stochastic integral, simply because

$$\frac{1}{t_i^n - t_{i-1}^n} \int_{t_{i-1}^n}^{t_i^n} u(B_n(s)) \, ds = \frac{1}{t_i^n - t_{i-1}^n} \int_{t_{i-1}^n}^{t_i^n} u(\frac{B(t_{i-1}^n) \left(t_i^n - s\right) + B(t_i^n) \left(s - t_{i-1}^n\right)}{t_i^n - t_{i-1}^n}) \, ds$$

is a (complicated) function of the two random variables  $B(t_{i-1}^n)$  and  $B(t_i^n)$ , and cannot be measurable w.r.t.  $\mathcal{F}(t_{i-1}^n)$ , for any  $i = 1, \dots, n$ .

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