

UNIVERSITÉ PARIS DAUPHINE - PSL

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# Stochastic Calculus

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# Contents

1	Brownian Motion	3
2	Stopping times and Martingales	6
3	Stochastic Integration	11
4	Stochastic differential equations	16

# Chapter 1

## Brownian Motion

### Exercise 1

Justify that  $B_t - B_s$  is independent of  $B_1$  for all  $1 \leq s \leq t$  and determine its law.

### Exercise 2

Compute the conditional expectations  $\mathbb{E}[B_t] \mathcal{F}_s$  and  $\mathbb{E}[B_t^2] \mathcal{F}_s$  for all  $t \geq 1$ .

### Exercise 3

Let  $\xi = \int_0^1 B_s ds$ . What is the law of  $\xi$  ?

### Exercise 4

Let  $\eta = \int_0^2 B_s ds$ . Compute the conditional expectation  $\mathbb{E}[B_1] \eta$ .

### Exercise 5

Let  $T := \inf \{t \geq 0 \mid B_t = 1\}$ . Show that  $\mathbb{P}(T < \infty) \geq \frac{1}{2}$ .

### Exercise 6

Let  $T := \inf \{t \geq 0 \mid |B_t| = 1\}$ .

1. Prove that  $T < \infty$  a.s.
2. Show that  $T$  and  $\mathbb{1}_{B_T=1}$  are independent.

### Exercise 7

Justify that almost surely,  $(B_t)_{t \geq 0}$  is not monotone on any interval.

### Exercise 8

1. Show that for all  $a > 0$ ,  $(\frac{1}{\sqrt{a}} B_{at})_{t \geq 0}$  is a Brownian motion.
2. Show that for all  $a > 0$ ,  $(B_a - B_{a-t})_{0 \leq t \leq a}$  is a Brownian motion.

3. Define  $X_0 = 0$  and  $X_t = tB_{1/t}$  for all  $t > 0$ . Show that  $(X_t)_{t \geq 0}$  is a Brownian motion.
4. Deduce that  $\lim_{t \rightarrow \infty} \frac{B_t}{t} = 0$  a.s.

**Exercise 9**

Prove that  $\int_0^\infty |B_s| ds = \infty$  a.s.

**Exercise 10**

Prove that for all  $a > 0$ ,  $(B_{t+a} - B_a)_{t \geq 0}$  is a Brownian motion independent of  $\mathcal{F}_a$ .

**Exercise 11**

1. Show that for all  $t \geq 0$ ,  $|B_t|$  has the same distribution as  $\sqrt{t}|B_1|$ .
2. Do the processes  $(|B_t|)_{t \geq 0}$  and  $(\sqrt{t}|B_1|)_{t \geq 0}$  have the same distribution ?

**Exercise 12**

1. Show that  $\int_0^1 \frac{B_s}{s} ds$  is well-defined a.s.
2. Let  $\beta_t := B_t - \int_0^1 \frac{B_s}{s} ds$ . Show that  $(\beta_t)_{t \geq 0}$  is a Brownian motion.

**Exercise 13 (Brownian bridge)**

We define the *Brownian bridge* as the process  $Z_t := B_t - tB_1$  for all  $0 \leq t \leq 1$ .

1. Show that  $Z$  is a Gaussian process independent of  $B_1$ .
2. Prove that  $Z$  has the same law as the process  $Y$  defined by

$$Y_t = \begin{cases} (1-t)B_{\frac{t}{1-t}} & \text{if } 0 \leq t < 1, \\ 0 & \text{if } t = 1. \end{cases}$$

**Exercise 14**

Let  $T$  be a random variable distributed according to the exponential distribution of mean 1. What is the law of  $B_T$  ?

**Exercise 15**

In this exercise,  $\mathbf{B}$  is a  $d$ -dimensional ( $d \in \mathbb{N}$ ) standard Brownian motion, that is  $\mathbf{B}_t = (B_t^1, \dots, B_t^d)$ , where  $B^i$ 's are independent standard Brownian motions. Let  $U \in \mathbb{R}^{d \times d}$  be an orthogonal matrix. Prove that the process  $(\mathbf{W}_t)_{t \geq 0} = (U\mathbf{B}_t)_{t \geq 0}$  is a  $d$ -dimensional standard Brownian motion.

**Exercise 16**

Let  $\tau$  be a stopping time that is almost surely finite. Show that the process  $(B_{t+\tau} - B_\tau)_{t \geq 0}$  is a Brownian motion independent of  $\mathcal{F}_\tau$ .

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**Exercise 17**

Prove that almost surely,  $\sup_{0 \leq t \leq s} B_t > 0$  for all  $s > 0$  and  $\sup_{t \geq 0} B_t = \infty$ .

**Exercise 18**

1. Show that for all  $a \geq 0$ ,  $\tau_a = a^2 \tau_1$  in distribution.
2. Let  $0 \leq a \leq b < \infty$ . Justify that  $\tau_b - \tau_a$  has the same distribution as  $\tau_{b-a}$  and is independent of  $\mathcal{F}_{\tau_a}$ .

## Chapter 2

# Stopping times and Martingales

For each  $x \in \mathbb{R}$ , we define the stopping time  $\tau_x := \inf \{t \geq 0 : B_t = x\}$ .

### Exercise 1

1. Show that  $(B_t^2 - t)_{t \geq 0}$  is a martingale.
2. Construct a martingale from  $(B_t^3)_{t \geq 0}$ . Same question with  $B_t^4$ .
3. Prove that  $(e^{\lambda B_t - \frac{\lambda^2}{2}t})_{t \geq 0}$  is a martingale for all  $\lambda \in \mathbb{R}$ .

### Solution

1. To show that  $(B_t^2 - t)$  is a martingale, we compute the conditional expectation:

$$\mathbb{E}[B_t^2 \mid \mathcal{F}_s] = \mathbb{E}[(B_t - B_s + B_s)^2 \mid \mathcal{F}_s] = B_s^2 + \mathbb{E}[(B_t - B_s)^2] = B_s^2 + (t - s),$$

since  $B_t - B_s$  is independent of  $\mathcal{F}_s$  and normally distributed with mean zero and variance  $t - s$ . Therefore, we have

$$\mathbb{E}[B_t^2 - t \mid \mathcal{F}_s] = B_s^2 - s,$$

confirming that  $(B_t^2 - t)_{t \geq 0}$  is indeed a martingale.

2. For all  $t > s$ , we expand  $B_t^3$  using  $B_t = B_s + (B_t - B_s)$ :

$$B_t^3 = (B_s + (B_t - B_s))^3 = B_s^3 + 3B_s^2(B_t - B_s) + 3B_s(B_t - B_s)^2 + (B_t - B_s)^3.$$

Taking the conditional expectation given  $\mathcal{F}_s$ , we find:

$$\mathbb{E}[B_t^3 \mid \mathcal{F}_s] = B_s^3 + 3B_s \mathbb{E}[(B_t - B_s)^2] = B_s^3 + 3B_s(t - s).$$

Therefore, the process  $(B_t^3 - 3B_s t)_{t \geq 0}$  is a martingale.

Similarly, for  $B_t^4$ , we expand:

$$B_t^4 = (B_s + (B_t - B_s))^4 = B_s^4 + 4B_s^3(B_t - B_s) + 6B_s^2(B_t - B_s)^2 + 4B_s(B_t - B_s)^3 + (B_t - B_s)^4.$$

Taking the conditional expectation, we obtain:

$$\mathbb{E}[B_t^4 \mid \mathcal{F}_s] = B_s^4 + 6B_s^2(t - s) + 3(t - s)^2.$$

Thus,  $(B_t^4 - 6B_s^2 t + 3t^2)_{t \geq 0}$  is a martingale.

3. Let  $\lambda \in \mathbb{R}$ . We compute the conditional expectation:

$$\begin{aligned}\mathbb{E} \left[ e^{\lambda B_t - \frac{\lambda^2}{2} t} \middle| \mathcal{F}_s \right] &= e^{\lambda B_s - \frac{\lambda^2}{2} s} \mathbb{E} \left[ e^{\lambda(B_t - B_s) - \frac{\lambda^2}{2}(t-s)} \middle| \mathcal{F}_s \right] \\ &= e^{\lambda B_s - \frac{\lambda^2}{2} s} \mathbb{E} \left[ e^{\lambda(B_t - B_s) - \frac{\lambda^2}{2}(t-s)} \right] \\ &= e^{\lambda B_s - \frac{\lambda^2}{2} s},\end{aligned}$$

since  $B_t - B_s$  is independent of  $\mathcal{F}_s$  and  $\mathbb{E} \left[ e^{\lambda(B_t - B_s) - \frac{\lambda^2}{2}(t-s)} \right] = 1$  due to the properties of the normal distribution. This confirms that  $(e^{\lambda B_t - \frac{\lambda^2}{2} t})_{t \geq 0}$  is a martingale.

## Exercise 2

Let  $a > 0$  and define  $T_a^* := \inf \{t \geq 0 : |B_t| = a\}$ .

1. Using the martingale  $(B_t^2 - t)_{t \geq 0}$  compute the expectation of  $T_a^*$ .
2. Using a well-chosen martingale, compute the variance of  $T_a^*$  and its Laplace transform.
3. Compute the Laplace transform of  $\tau_a$  and find that it has the law as  $(a/B_1)^2$ . What is the value of  $\mathbb{E}[\tau_a]$  ?

## Solution

1. Since  $T_a^*$  is almost surely finite, the stopped martingale  $(B_{t \wedge T_a^*}^2 - t \wedge T_a^*)$  converges almost surely to  $B_{T_a^*}^2 - T_a^* = a^2 - T_a^*$ . Moreover,  $B_{t \wedge T_a^*}^2$  is bounded by  $a^2$ , ensuring  $L^1$  convergence by the dominated convergence theorem. In the other hand,  $t \wedge T_a^*$  converge in  $L^1$  to  $T_a^*$  by monotone convergence. Therefore,

$$0 = \mathbb{E} \left[ B_{t \wedge T_a^*}^2 - t \wedge T_a^* \right] \xrightarrow[t \rightarrow \infty]{} a^2 - \mathbb{E} [T_a^*].$$

Hence,  $\mathbb{E} [T_a^*] = a^2$ .

2. To compute the variance, we consider the martingale  $(B_t^4 - 6B_t^2 t + 3t^2)_{t \geq 0}$ . Applying similar reasoning as before, we have:

$$\mathbb{E} \left[ B_{t \wedge T_a^*}^4 - 6B_{t \wedge T_a^*}^2 (t \wedge T_a^*) + 3(t \wedge T_a^*)^2 \right] = 0.$$

Taking the limit as  $t \rightarrow \infty$  and using  $B_{T_a^*}^2 = a^2$ , we get:

$$a^4 - 6a^2 \mathbb{E} [T_a^*] + 3\mathbb{E} [(T_a^*)^2] = 0.$$

Substituting  $\mathbb{E} [T_a^*] = a^2$ , we solve for  $\mathbb{E} [(T_a^*)^2]$ :

$$a^4 - 6a^4 + 3\mathbb{E} [(T_a^*)^2] = 0 \quad \implies \quad \mathbb{E} [(T_a^*)^2] = \frac{5}{3}a^4.$$

Therefore, the variance is:

$$\text{Var} (T_a^*) = \mathbb{E} [(T_a^*)^2] - (\mathbb{E} [T_a^*])^2 = \frac{5}{3}a^4 - a^4 = \frac{2}{3}a^4.$$



For the Laplace transform, consider the martingale for  $\lambda > 0$  (check it):

$$M_t = \exp\left(\sqrt{2\lambda}B_t - \lambda t\right) + \exp\left(-\sqrt{2\lambda}B_t - \lambda t\right).$$

The stopped martingale  $M^{T_a^*}$  is dominated in  $L^1$  by  $2e^{\sqrt{2\lambda}a}$ , and thus uniformly integrable. Therefore,  $\mathbb{E}[M_{T_a^*}] = 2$ . Evaluating  $M_{T_a^*}$ , we have:

$$\begin{aligned} 2 = \mathbb{E}[M_{T_a^*}] &= \mathbb{E}\left[e^{-\lambda T_a^*}(e^{\sqrt{2\lambda}a} + e^{-\sqrt{2\lambda}a})\mathbb{1}_{\{B_{T_a^*}=a\}}\right] + \mathbb{E}\left[e^{-\lambda T_a^*}(e^{-\sqrt{2\lambda}a} + e^{\sqrt{2\lambda}a})\mathbb{1}_{\{B_{T_a^*}=-a\}}\right] \\ &= 2 \cosh(\sqrt{2\lambda}a)\mathbb{E}[X] \end{aligned}$$

$$\text{Hence } \mathbb{E}[e^{-\lambda T_a^*}] = \frac{1}{\cosh(\sqrt{2\lambda}a)}.$$

3. To compute the Laplace transform of  $\tau_a$ , we utilize the martingale  $M_t = e^{\sqrt{2\lambda}B_t - \lambda t}$ . At the stopping time  $\tau_a$ , we have  $B_{\tau_a} = a$ , so:

$$M_{\tau_a} = e^{\sqrt{2\lambda}a - \lambda\tau_a}.$$

Since  $M_t$  is a martingale,  $\mathbb{E}[M_{\tau_a}] = M_0 = 1$ . Therefore,

$$\mathbb{E}[e^{-\lambda\tau_a}] = e^{-\sqrt{2\lambda}a}.$$

### Exercise 3

1. Let  $M$  be a continuous martingale such that  $M_0 = x \geq 0$ . Suppose that  $M_t \geq 0$  for all  $t \geq 0$  and that  $M_t \rightarrow 0$  as  $t \rightarrow \infty$ , a.s. Show that, for all  $y > x$ ,

$$\mathbb{P}\left(\sup_{t \geq 0} M_t \geq y\right) = \frac{x}{y}.$$

2. Deduce the law of

$$\sup_{t \leq T_0} B_t,$$

when  $B$  is a Brownian motion started at  $x > 0$  and  $T_0 = \inf\{t \geq 0 : B_t = 0\}$ .

3. Suppose now that is started at 0, and let  $\mu > 0$ . Using a well-chosen exponential martingale, prove that

$$\sup_{t \geq 0} (B_t - \mu t)$$

follows the exponential distribution of parameter  $2\mu$ .

### Exercise 4

Let  $a < 0 < b$  and set  $T = \tau_a \wedge \tau_b$ .

1. Prove that, for all  $\lambda > 0$ ,

$$\mathbb{E}[\exp(-\lambda T)] = \frac{\cosh\left(\frac{b+a}{2}\sqrt{2\lambda}\right)}{\cosh\left(\frac{b-a}{2}\sqrt{2\lambda}\right)}.$$

(Hint: introduce the martingale

$$M_t = \exp\left(\sqrt{2\lambda}(B_t - \alpha) - \lambda t\right) + \exp\left(-\sqrt{2\lambda}(B_t - \alpha) - \lambda t\right),$$

with a well-chosen  $\alpha$ .)

2. Similarly, prove that, for all  $\lambda > 0$ ,

$$\mathbb{E} [\exp(-\lambda T) \mathbf{1}_{\{T=\tau_a\}}] = \frac{\sinh(b\sqrt{2\lambda})}{\sinh((b-a)\sqrt{2\lambda})}.$$

3. With the help of question 2, give an expression of  $\mathbb{P}(\tau_a < \tau_b)$ .

### Exercise 5

Let  $M = (M_t)_{0 \leq t \leq 1}$  be a sub-martingale. Let  $(\mathcal{G}_s)_{s \geq 0}$  be a sub-filtration of  $(\mathcal{F}_s)_{s \geq 0}$ . Prove that  $N_t = \mathbb{E}[M_t | \mathcal{G}_t]$  is a  $(\mathcal{G}_s)$ -sub-martingale.

### Solution

First note that  $N$  is clearly adapted and integrable. We prove the martingale property, let  $0 \leq s \leq t$ :

$$\begin{aligned} \mathbb{E}[N_t | \mathcal{G}_s] &= \mathbb{E}[\mathbb{E}[M_t | \mathcal{G}_t] | \mathcal{G}_s] \\ &= \mathbb{E}[M_t | \mathcal{G}_s] \\ &= \mathbb{E}[\mathbb{E}[M_t | \mathcal{F}_s] | \mathcal{G}_s] \\ &\geq \mathbb{E}[M_s | \mathcal{G}_s] = N_s. \end{aligned}$$

### Exercise 6

Let  $\sigma \leq \tau$  be two bounded stopping times. Show that

$$\mathbb{E}[(B_\tau - B_\sigma)^2] = \mathbb{E}[B_\tau^2] - \mathbb{E}[B_\sigma^2] = \mathbb{E}[\tau - \sigma].$$

### Exercise 7

Let  $M = (M_t)_{0 \leq t \leq 1}$  be a sub-martingale such that  $\mathbb{E}[M_0] = \mathbb{E}[M_1]$ . Prove that  $M$  is a martingale.

### Solution

Since  $M$  is a sub-martingale we have for all  $t \in [0, 1]$ ,  $\mathbb{E}[M_0] \leq \mathbb{E}[M_t] \leq \mathbb{E}[M_1]$ , hence  $\mathbb{E}[M_0] = \mathbb{E}[M_t]$  for all  $t$ . Let  $0 \leq s \leq t \leq 1$ ,  $M$  is a sub-martingale so that  $\mathbb{E}[M_t - M_s | \mathcal{F}_s]$  is non negative and of expectation 0, therefore  $\mathbb{E}[M_t - M_s | \mathcal{F}_s] = 0$  and  $M$  is a martingale.

### Exercise 8

Let  $M$  be a càd martingale. Let  $t \geq 0$ . Prove that  $M_{t+\varepsilon} \xrightarrow[\varepsilon \rightarrow 0]{L^1} M_t$ .

### Solution

The càd property give us  $M_{t+\varepsilon} \xrightarrow[\varepsilon \rightarrow 0]{\text{a.s.}} M_t$ . For every  $n \geq 1$  we have

$$\mathbb{E}\left[M_{t+1/n} \mid \mathcal{F}_{t+\frac{1}{n}}\right] = M_{t+\frac{1}{n}}.$$

Hence the sequence  $(M_{t+1/n})_{n \geq 1}$  is uniformly integrable and  $M_{t+1/n} \xrightarrow[n \rightarrow \infty]{L^1} M_t$ .

**Exercise 9**

Let  $M$  be a local continuous martingale such that  $M_0 = 0$  a.s.

1. Let  $a > 0$  and let  $\sigma_a := \inf\{t \geq 0 : \langle M, M \rangle_t \geq a^2\}$ . Show that

$$\mathbb{P} \left( \sup_{s \in [0, \sigma_a]} |M_s| > a \right) \leq \frac{1}{a^2} \mathbb{E} [a^2 \wedge \langle M, M \rangle_\infty].$$

2. Show that  $\mathbb{P}(\sup_{t \geq 0} |M_t| > a) \leq \mathbb{P}(\langle M, M \rangle_\infty \geq a^2) + a^{-2} \mathbb{E} [a^2 \wedge \langle M, M \rangle_\infty]$ .
3. Show that  $\mathbb{E} [\sup_{t \geq 0} |M_t|] \leq 3 \mathbb{E} [\sqrt{\langle M, M \rangle_\infty}]$ .
4. Show that if  $\mathbb{E} [\sqrt{\langle M, M \rangle_\infty}] < \infty$ , then  $M$  is a uniformly integrable martingale.
5. Show that if  $\mathbb{E} [\sqrt{\langle M, M \rangle_t}] < \infty$  for every  $t \geq 0$ , then  $M$  is an integrable martingale.

**Solution**

1. We observe that  $\mathbb{P} \left( \sup_{s \in [0, \sigma_a]} |M_s| > a \right) = \mathbb{P}(\sup_{t \geq 0} |M_{t \wedge \sigma_a}| > a)$  and since  $\mathbb{E} [\langle M \rangle_{t \wedge \sigma_a}] \leq a^2$ ,  $M^{\sigma_a}$  and  $(M^{\sigma_a})^2 - \langle M^{\sigma_a} \rangle$  are uniformly integrable martingales. Therefore we can apply Doob maximal inequality:

$$a^2 \mathbb{P} \left( \sup_{t \geq 0} |M_{t \wedge \sigma_a}| > a \right) \leq \sup_{t \geq 0} \mathbb{E} [M_{t \wedge \sigma_a}^2] \leq \mathbb{E} [\langle M \rangle_{\sigma_a}] \leq \mathbb{E} [a^2 \wedge \langle M \rangle_\infty].$$

Indeed,  $\langle M \rangle_{\sigma_a} = a^2 \mathbf{1}_{\sigma_a < \infty} + \langle M \rangle_\infty (1 - \mathbf{1}_{\sigma_a < \infty}) \leq a^2$  and  $a^2 \mathbf{1}_{\sigma_a < \infty} + \langle M \rangle_\infty (1 - \mathbf{1}_{\sigma_a < \infty}) \leq \langle M \rangle_\infty$ .

2. We have

$$\begin{aligned} \mathbb{P} \left( \sup_{t \geq 0} |M_t| > a \right) &= \mathbb{P} \left( \sup_{t \geq 0} |M_t| > a, \sigma_a < \infty \right) + \mathbb{P} \left( \sup_{t \geq 0} |M_t| > a, \sigma_a = \infty \right) \\ &\leq \mathbb{P}(\sigma_a < \infty) + \mathbb{P} \left( \sup_{t \in [0, \sigma_a]} |M_t| > a \right). \end{aligned}$$

3. By integrating with respect to  $a$  we have

$$\begin{aligned} \mathbb{E} \left[ \sup_{t \geq 0} |M_t| \right] &\leq \int_0^\infty \mathbb{P}(\langle M, M \rangle_\infty \geq a^2) + a^{-2} \mathbb{E} [a^2 \wedge \langle M, M \rangle_\infty] da \\ &\leq \int_0^\infty \mathbb{P}(\sqrt{\langle M, M \rangle_\infty} \geq a) da + \mathbb{E} \left[ \int_0^\infty 1 \wedge \frac{\langle M, M \rangle_\infty}{a^2} da \right] \\ &\leq \mathbb{E} \left[ \sqrt{\langle M, M \rangle_\infty} \right] + \mathbb{E} \left[ \int_0^{\sqrt{\langle M, M \rangle_\infty}} da \right] + \mathbb{E} \left[ \langle M, M \rangle_\infty \int_{\sqrt{\langle M, M \rangle_\infty}}^\infty a^{-2} da \right] \\ &\leq 3 \mathbb{E} \left[ \sqrt{\langle M, M \rangle_\infty} \right]. \end{aligned}$$

4. Using the previous question we have that  $\sup_{t \geq 0} |M_t| = C < \infty$  which implies that the local martingale  $M$  is dominated in  $L^1$  and thus a uniformly integrable martingale.
5. By the same reasoning we have that for every  $T > 0$ , the process  $(M_t)_{t \in [0, T]}$  is dominated in  $L^1$ , that is a martingale on  $[0, T]$  for every  $T > 0$  and thus a martingale on  $\mathbb{R}_+$ .

## Chapter 3

# Stochastic Integration

### Exercise 1

Let  $M$  be a local martingale. Show that the family  $\{M_\tau, \tau < \infty\}$  is uniformly integrable if, and only if,  $M$  is a uniformly integrable martingale.

### Solution

Assume first that  $M$  is a uniformly integrable martingale, therefore for all stopping time  $\tau$  we have that  $M_\tau = \mathbb{E}[M_\infty | \mathcal{F}_\tau]$  hence  $\{M_\tau, \tau < \infty\}$  is uniformly integrable.

Assume now that the family  $\{M_\tau, \tau < \infty\}$  is uniformly integrable, in particular the family  $\{M_t\}_{t \geq 0}$  is uniformly integrable, we have to prove the martingale property. Let  $\tau_n$  be a localizing sequence, for every  $0 \leq s \leq t$  and every  $A \in \mathcal{F}_{s \wedge \tau_n} \subseteq \mathcal{F}_s$ , we know that

$$\mathbb{E}[M_{s \wedge \tau_n} \mathbf{1}_A] = \mathbb{E}[M_{t \wedge \tau_n} \mathbf{1}_A].$$

By assumption, we have that  $(M_{x \wedge \tau_n})_{x \geq 0}$  is uniformly integrable, so we have  $L^1$  converge in both terms, we have

$$\mathbb{E}[M_s \mathbf{1}_A] = \mathbb{E}[M_t \mathbf{1}_A]$$

for all  $A \in \mathcal{F}_s$  and  $M$  is a uniformly integrable martingale.

### Exercise 2

Let  $M$  be a bounded local martingale. Show that  $\langle M \rangle_\infty < \infty$  a.s.

### Solution

Clearly,  $M$  is a bounded martingale. Hence  $\lim_{t \rightarrow \infty} M_t^2 = M_\infty^2$  in  $L^1$  and  $\lim_{t \rightarrow \infty} \mathbb{E}[\langle M \rangle_t] = \mathbb{E}[\langle M \rangle_\infty]$  by monotone convergence. We obtain

$$0 = \mathbb{E}[M_t^2 - \langle M \rangle_t] \xrightarrow[t \rightarrow \infty]{} \mathbb{E}[M_\infty^2 - \langle M \rangle_\infty].$$

This implies that  $\mathbb{E}[\langle M \rangle_\infty] = \mathbb{E}[M_\infty^2] < \infty$ .

**Exercise 3**

Let  $B$  be a standard Brownian motion with  $B_0 = x > 0$ . Set  $T = \inf\{t \geq 0 : B_t = 0\}$ . Let  $f : \mathbb{R}_+ \rightarrow \mathbb{R}$  compactly supported. Compute  $\mathbb{E} \left[ \int_0^T f(B_s) ds \right]$ .

**Exercise 4**

Let  $\alpha : \mathbb{R}_+ \rightarrow \mathbb{R}$  be a continuous function. With the help of the stochastic integration by parts formula, check that the process

$$X_t := \int_0^t \exp \left( \int_s^t \alpha(u) du \right) dB_s,$$

satisfies the stochastic differential equation  $dX_t = \alpha(t)X_t dt + dB_t$ .

**Solution**

First, we have for every  $t \geq 0$

$$\begin{aligned} X_t &= \int_0^t \exp \left( \int_0^t \alpha(u) du - \int_0^s \alpha(u) du \right) dB_s \\ &= \exp \left( \int_0^t \alpha(s) ds \right) \int_0^t \exp \left( - \int_0^s \alpha(u) du \right) dB_s \\ &= e^{Y_t} \int_0^t e^{-Y_s} dB_s \end{aligned}$$

Where  $Y_t = \int_0^t \alpha(s) ds$ . Now using integration by part we obtain

$$\begin{aligned} dX_t &= d \left( e^{Y_t} \right) \int_0^t e^{-Y_s} dB_s + e^{Y_t} d \left( \int_0^t e^{-Y_s} dB_s \right) \\ &= \alpha(t) e^{Y_t} dt \int_0^t e^{-Y_s} dB_s + e^{Y_t} e^{-Y_t} dB_t. \end{aligned}$$

Hence,  $dX_t = \alpha(t)X_t dt + dB_t$ .

**Exercise 5**

Let  $f : \mathbb{R} \rightarrow \mathbb{R}$  be continuously twice-differentiable. Show that the process

$$X_t := f(B_t) - \frac{1}{2} \int_0^t f''(B_s) ds,$$

is a continuous local martingale. Give a sufficient condition for  $X$  to be a martingale.

**Exercise 6**

Let  $\varphi : \mathbb{R}_+ \rightarrow \mathbb{R}$  be a (deterministic) function in  $L_{\text{loc}}^2(\mathbb{R}_+)$  (i.e.  $\int_0^t \varphi^2(s) ds < \infty$  for every  $t \geq 0$ ) and  $Z^\varphi = (Z_t^\varphi)_{t \geq 0}$  the associated Doléans-Dade exponential process. Check that  $Z^\varphi$  is a martingale.

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## Solution

Since,  $\varphi \in L^2_{\text{loc}}$ ,  $\int_0^t \varphi(s)dB_s$  is well-defined and is a centered gaussian random variable with variance  $\int_0^t \varphi(s)^2 ds < \infty$ , this implies that  $Z_t^\varphi \in L^1$  for every  $t \geq 0$ . Let  $0 \leq s \leq t$  we have

$$\mathbb{E} \left[ e^{\int_0^t \varphi(u)dB_u - \frac{1}{2} \int_0^t \varphi(u)^2 du} \middle| \mathcal{F}_s \right] = \mathbb{E} \left[ e^{\int_0^t \varphi(u)dB_u - \int_0^s \varphi(u)dB_u} \middle| \mathcal{F}_s \right] e^{\int_0^s \varphi(u)dB_u - \frac{1}{2} \int_0^t \varphi(u)^2 du}.$$

Recall that,  $\int_0^t \varphi(u)dB_u - \int_0^s \varphi(u)dB_u = \int_s^t \varphi(u)dB_u$  is independent of  $\mathcal{F}_s$  and is a centered gaussian variable with variance  $\int_s^t \varphi(s)^2 ds$ , we thus obtain

$$\begin{aligned} \mathbb{E} \left[ e^{\int_0^t \varphi(u)dB_u - \frac{1}{2} \int_0^t \varphi(u)^2 du} \middle| \mathcal{F}_s \right] &= \mathbb{E} \left[ e^{\int_s^t \varphi(u)dB_u} \right] e^{\int_0^s \varphi(u)dB_u - \frac{1}{2} \int_0^t \varphi(u)^2 du} \\ &= e^{\frac{1}{2} \int_s^t \varphi(s)^2 ds} \cdot e^{\int_0^s \varphi(u)dB_u - \frac{1}{2} \int_0^t \varphi(u)^2 du} \\ &= e^{\int_0^s \varphi(u)dB_u - \frac{1}{2} \int_0^s \varphi(u)^2 du}. \end{aligned}$$

Hence the result.

## Exercise 7

Find a progressive process  $X = (X_t)_{t \geq 0}$  such that the process  $Z = (Z_t)_{t \geq 0}$  defined by  $Z_t = \exp(X_t - B_t^2)$  is a martingale.

## Solution

Assume that the process  $X$  has the form

$$X_t = \int_0^t b_s ds + \int_0^t \sigma_s dB_s,$$

and  $X_0 = 0$ , then by Itô's formula we have

$$\begin{aligned} Z_t &= 1 + \int_0^t Z_s dX_s - 2 \int_0^t Z_s B_s dB_s + \frac{1}{2} \int_0^t Z_s d\langle X, X \rangle_s - 2 \int_0^t Z_s B_s d\langle X, B \rangle_s + \int_0^t Z_s (2B_s^2 - 1) ds \\ &= 1 + \int_0^t Z_s (\sigma_s - 2B_s) dB_s + \int_0^t Z_s \left( \frac{1}{2} \sigma_s^2 - 2B_s \sigma_s + 2B_s^2 - 1 + b_s \right) ds. \end{aligned}$$

Then, by taking  $X_t = \int_0^t (1 - B_s^2/2) ds + \int_0^t B_s dB_s$ ,  $Z_t = 1 - \int_0^t Z_s B_s dB_s$  is a local martingale. Using the fact that  $B_t^2 - t = 2 \int_0^t B_s dB_s$ , we have

$$Z_t = e^{-\int_0^t B_s dB_s - \frac{1}{2} \int_0^t B_s^2 ds}.$$

## Exercise 8

Let  $X$  and  $Y$  be two  $(\mathcal{F}_t)$  independent Brownian motions and let  $H$  be a progressive process. We set

$$\begin{aligned} \beta_t &= \int_0^t \cos(H_s) dX_s - \int_0^t \sin(H_s) dY_s, \\ \gamma_t &= \int_0^t \sin(H_s) dX_s + \int_0^t \cos(H_s) dY_s. \end{aligned}$$

Show that  $\beta$  and  $\gamma$  are independent  $(\mathcal{F}_t)$  Brownian motions.

**Exercise 9**

Let  $B$  be a Brownian motion. Show that  $\int_0^t \mathbf{1}_{\{B_s=0\}} dB_s = 0$ .

**Exercise 10**

1. Let  $g : \mathbb{R}_+ \rightarrow \mathbb{R}_+$  be a continuous function. Let  $f : \mathbb{R}_+ \rightarrow (0, \infty)$  be a  $\mathcal{C}^2$  function such that  $f'' = 2gf$  on  $\mathbb{R}_+$  and  $f(0) = 1$ ,  $f'(1) = 0$ . We set

$$u(t) := \frac{f'(t)}{2f(t)}, \quad t \geq 0.$$

Show that  $u' + 2u^2 = g$  on  $\mathbb{R}_+$ .

2. Let  $\beta$  a  $(\mathcal{F}_t)$  standard Brownian motion. Let  $x_0 \geq 0$ ,  $a \geq 0$  and  $X$  an adapted, continuous and non negative process such that

$$X_t = x_0 + 2 \int_0^t \sqrt{X_s} d\beta_s + at.$$

Show that  $u(t)X_t - \int_0^t g(s)X_s ds = u(0)X_0 + \int_0^t u(s)dX_s - 2 \int_0^t u(s)^2 X_s ds$ ,  $t \geq 0$ .

3. Set  $M_t := u(0)x_0 + 2 \int_0^t u(s)\sqrt{X_s}d\beta_s$ ,  $t \geq 0$ . Show that

$$f(t)^{-a/2} \exp \left( u(t)X_t - \int_0^t g(s)X_s ds \right) = \mathcal{E}(M)_t.$$

4. Show that  $f$  is non increasing on  $[0, 1]$  and show that

$$\mathbb{E} \left[ \exp \left( - \int_0^1 g(s)X_s ds \right) \right] = f(1)^{a/2} e^{x_0 f'(0)/2}.$$

5. Show that

$$\mathbb{E} \left[ \exp \left( - \frac{\theta^2}{2} \int_0^1 X_s ds \right) \right] = \frac{1}{\cosh(\theta)^{a/2}} \exp \left( - \frac{x_0}{2} \theta \tanh(\theta) \right), \quad \forall \theta \in \mathbb{R}.$$

6. Let  $B$  be a standard Brownian motion. For every  $x \in \mathbb{R}$ , show that,

$$\mathbb{E} \left[ \exp \left( - \frac{\theta^2}{2} \int_0^1 (B_s + x)^2 ds \right) \right] = \frac{1}{\cosh(\theta)^{1/2}} \exp \left( - \frac{x^2}{2} \theta \tanh(\theta) \right), \quad \forall \theta \in \mathbb{R}.$$

7. Let  $B$  and  $\tilde{B}$  be two independent standard Brownian motions. For every  $t > 0$  show that  $\inf \{s \geq 0 : |B_s| = t\} = \int_0^t B_s^2 ds + \int_0^t \tilde{B}_s^2 ds$  in law.

**Solution**

4. Use definition of  $f$ , that is, we have domination in  $L^1$  for every  $t \in [0, 1]$  ( $X \geq 0, u \leq 0$ ).

5. Take  $f(t) = \frac{\cosh(\theta(t-1))}{\cosh(\theta)}$ .

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**Exercise 11**

Let  $B$  be a Brownian motion and let  $S_t := \sup_{s \in [0, t]} B_s$ . We set  $X_t := S_t - B_t$ .

1. Show that  $\int_0^t \mathbb{1}_{\{X_u \neq 0\}} dS_u = 0$ .
2. Show that  $Y_t := X_t^2 - t$  is a martingale.
3. Let  $\tau := \inf\{t \geq 0 : X_t = 1\}$ . Compute  $\mathbb{E}[\tau]$ .



# Chapter 4

## Stochastic differential equations

### Exercise 1

Let  $M_t := \frac{1}{2}(B_t^2 - t)$ ,  $t \geq 0$ .

1. Justify that  $M$  is a martingale and express  $M$  as a stochastic integral.
2. Show that for all  $b \geq 0$ , the exponential local martingale  $\mathcal{E}(-bM)$  is a martingale. For all  $T > 0$ , justify that  $\mathbb{Q} := \mathcal{E}(-bM)_T \cdot \mathbb{P}$  defines a probability measure.
3. Determine the SDE satisfied by  $(B_t)_{t \in [0, T]}$  on  $\mathbb{Q}$ . Deduce the distribution of  $B_t$ ,  $t \in [0, T]$  on  $\mathbb{Q}$ .
4. Deduce that for all  $a, b \geq 0$ ,

$$\mathbb{E} \left[ \exp \left\{ -aB_t^2 - \frac{b^2}{2} \int_0^t B_s^2 ds \right\} \right] = \left( \frac{b}{b \cosh(bt) + 2a \sinh(bt)} \right)^{1/2}.$$

5. Using that for all  $\alpha, \beta > 0$  and  $s \geq 0$ ,

$$\frac{1}{\Gamma(\alpha)} \int_0^\infty \beta^\alpha x^{\alpha-1} e^{-(\beta+s)x} dx = \left( \frac{\beta}{s+\beta} \right)^\alpha,$$

compute

$$\mathbb{E} \left[ \exp \left\{ -\frac{b^2}{2} \int_0^t B_s^2 ds \right\} \middle| B_t = y \right], \quad b > 0, y \in \mathbb{R}.$$

### Solution

1. By Itô's formula we have that  $B_t^2 = 2 \int_0^t B_s dB_s + t$ , hence  $M_t = \int_0^t B_s dB_s$  is a local martingale. Moreover,  $\mathbb{E} [\langle M \rangle_t] = \mathbb{E} \left[ \int_0^t B_s^2 ds \right] = \int_0^t \mathbb{E} [B_s^2] ds = \frac{t^2}{2} < \infty$ , hence  $M$  is a martingale.
2. For every  $T \geq 0$ ,  $t \in [0, T]$ , we have  $\mathcal{E}(-bM)_t = e^{-\frac{b}{2}B_t^2 + \frac{bt}{2} - \frac{b^2}{2}\langle M \rangle_t} \leq e^{\frac{bT}{2}}$ . Hence,  $\mathcal{E}(-bM)$  is a martingale,  $\mathbb{E} [\mathcal{E}(-bM)_T] = 1$  and thus  $\mathbb{Q}$  define a probability measure.
3. By Girsanov theorem we have that  $\beta_t := B_t + b \langle B, M \rangle_t$  is a Brownian motion under  $\mathbb{Q}$ , i.e.  $B$  solve

$$dB_t = d\beta_t - bB_t dt.$$

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We recognize the SDE satisfied by the Ornstein-Uhlenbeck process and thus,

$$B_t = \int_0^t e^{-b(t-s)} d\beta_s.$$

Therefore,  $B$  is a centered gaussian process with variance  $\frac{1-e^{-2bt}}{2b}$ .

4. For every  $a, b \geq 0$  we have

$$\begin{aligned} \mathbb{E} \left[ \exp \left\{ -aB_t^2 - \frac{b^2}{2} \int_0^t B_s^2 ds \right\} \right] &= \mathbb{E} \left[ e^{-aB_t^2 + b(B_t^2 - t)} \mathcal{E}(-bM)_t \right] \\ &= \mathbb{E}_{\mathbb{Q}} \left[ e^{(\frac{b}{2}-a)B_t^2 - \frac{b}{2}t} \right] \\ &= \mathbb{E}_{\mathbb{Q}} \left[ e^{(\frac{b}{2}-a)(\frac{1-e^{-2bt}}{2b})Z - \frac{b}{2}t} \right] \end{aligned}$$

Where  $Z \sim \mathcal{N}(1)$ , and the expectation is well defined if  $(\frac{b}{2} - a)(\frac{1-e^{-2bt}}{2b}) < \frac{1}{2}$ , indeed

$$\left( \frac{b}{2} - a \right) \left( \frac{1 - e^{-2bt}}{2b} \right) \leq \frac{b - 2a}{4b} < \frac{1}{2}.$$

Therefore,

$$\begin{aligned} \mathbb{E}_{\mathbb{Q}} \left[ e^{(\frac{b}{2}-a)(\frac{1-e^{-2bt}}{2b})Z - \frac{b}{2}t} \right] &= \frac{e^{-\frac{bt}{2}}}{\sqrt{1 - (\frac{b}{2} - a)(\frac{1-e^{-2bt}}{b})}} \\ &= \frac{1}{\sqrt{e^{bt} \left( 1 - (\frac{b}{2} - a)(\frac{1-e^{-2bt}}{b}) \right)}} \\ &= \left( \frac{b}{be^{bt} - (\frac{b}{2} - a)(e^{bt} - e^{-bt})} \right)^{1/2} \\ &= \left( \frac{b}{b \cosh(bt) + 2a \sinh(bt)} \right)^{1/2}. \end{aligned}$$

## Exercise 2

1. Justify that for all  $T < 1$  and  $x \in \mathbb{R}$ , there exists almost surely a solution to the SDE:

$$X_t^x = x + B_t - \int_0^t \frac{X_s^x}{1-s} ds, \quad t \in [0, T].$$

2. By applying the Itô formula to  $(\frac{X_t^0}{1-t})_{t \in [0, T]}$ , find an explicit formula for  $(X_t^0)_{t \in [0, T]}$ .

3. Show that  $X_t^x = X_t^0 + x(1-t)$ ,  $t \in [0, T]$  and determine its distribution.

4. Show that  $X^x$  can be extended to a continuous process on  $[0, 1]$ .

5. What does  $(X_t^x)_{t \in [0, 1]}$  represent ?

**Solution**

1. The coefficients  $\sigma(t, x) = 1$  and  $b(t, x) = -\frac{x}{1-t}$  are Lipschitz in space and bounded in time, thus there exists a unique solution.

2. By Itô formula we have

$$\frac{X_t^0}{1-t} = \int_0^t \frac{1}{1-s} dX_s^0 + \int_0^t \frac{X_s^0}{(1-s)^2} ds = \int_0^t \frac{1}{1-s} dB_s.$$

Hence,  $X_t^0 = \int_0^t \frac{1-t}{1-s} dB_s$  and  $X_t^0 \sim \mathcal{N}\left(0, \int_0^t \left(\frac{1-t}{1-s}\right)^2 ds\right) = \mathcal{N}(0, t(1-t))$  for every  $t \in [0, T]$ .

3. Let  $t \in [0, T]$ ,

$$x + B_t - \int_0^t \frac{X_s^0 + x(1-t)}{1-s} ds = B_t - \int_0^t \frac{X_s^0}{1-s} ds + x - \int_0^t x ds = X_t^0 + x(1-t).$$

$X_t^0 + x(1-t)$  solve the SDE and by uniqueness we have that  $X_t^x = X_t^0 + x(1-t)$ . Moreover  $X_t^x \sim \mathcal{N}(x(1-t), t(1-t))$ .

4. It suffices to show that  $X^0$  extend to a continuous process on  $[0, 1]$ . We know that  $\frac{X_t^0}{t-1}$  is a centered gaussian process with covariance  $\min\left(\frac{t}{1-t}, \frac{s}{1-s}\right)$ , thus  $X_t^0 = (1-t)B_{\frac{t}{1-t}}$  as processes on  $[0, T]$ . And by time inversion we have that

$$\lim_{t \rightarrow 1} (1-t)B_{\frac{t}{1-t}} = \lim_{t \rightarrow \infty} \frac{B_t}{t} = 0.$$

5.  $X^x$  is a Brownian bridge between  $x$  and 0.

**Exercise 3**

Let  $b \in \mathbb{R}$ ,  $a > 0$  and set  $X_t = B_t - bt$ . Let  $T = \inf\{t \geq 0 : B_t = a\}$ .

1. Find a probability measure  $\mathbb{Q}$  on  $\mathcal{F}_\infty$  such that  $(X_t)_{t \geq 0}$  is a Brownian motion.
2. Deduce the value of  $\mathbb{P}(T \leq t)$  and the distribution of  $Z = \sup_{t \geq 0} X_t$  under  $\mathbb{P}$ .

**Exercise 4**

1. Let  $X = (X_t)_{t \geq 0}$  solution to  $E(\sigma, b)$  with value in an open set  $D \subset \mathbb{R}^d$ . Let  $\lambda \in \mathbb{R}$ . Let  $f : D \rightarrow \mathbb{R}$  be twice continuously differentiable and such that  $\mathcal{L}f = \lambda f$ , where

$$(\mathcal{L}f)(x) := b(x)^\top \nabla f(x) + \frac{1}{2} \text{tr} \left( (\sigma \sigma^\top)(x) D^2 f(x) \right).$$

Show that  $(f(X_t)e^{-\lambda t})_{t \geq 0}$  is a continuous local martingale.

2. Let  $B = (B^1, B^2, B^3)$  be a Brownian motion with value in  $\mathbb{R}^3$ , and  $B_0 := a \in \mathbb{R}^3 \setminus \{0\}$ . Let  $X = \|B\|^2$ . Show that  $X$  solves some SDE  $E(\sigma, b)$  and give the coefficients  $\sigma$  and  $b$ .
3. We now assume that  $\lambda \geq 0$ . Show that  $2tf''(t) + 3f'(t) = \lambda f(t)$ ,  $t > 0$ , for  $f(t) = \frac{\sinh(\sqrt{2\lambda t})}{\sqrt{2\lambda t}}$ .
4. Let  $x > \|a\|^2$ , and let  $T_x = \inf\{t \geq 0 : X_t = x\}$ . Show that for every  $\lambda \geq 0$ , we have

$$\mathbb{E} \left[ e^{-\lambda T_x} \right] = \frac{\sinh(\sqrt{2\lambda \|a\|^2})}{\sqrt{2\lambda \|a\|^2}} \frac{\sqrt{2\lambda x}}{\sinh(\sqrt{2\lambda x})}.$$

## Exercise 5

1. Let  $H, Z$  and  $X$  be continuous semimartingales, such that  $X_t = H_t + \int_0^t X_s dZ_s$ . Express  $X$  as a function of  $H$  and  $Z$ .

**Hint:** start with  $H \equiv 1$  and use a variation of the constant method as in classical ODE.

2. Solve  $X_t = x + B_t - \beta \int_0^t X_s ds$ , where  $x \in \mathbb{R}$  and  $\beta \geq 0$  are constants. The process  $X$  is called Ornstein-Uhlenbeck process.

## Solution

1. By setting  $H \equiv 1$ , we have that  $X = \mathcal{E}(Z)$ , so we need to find a process  $Y$  such that  $X_t = Y_t \mathcal{E}(Z)_t$ . First by Itô formula we have

$$\begin{aligned} \frac{1}{\mathcal{E}(Z)_t} &= 1 - \int_0^t \frac{1}{\mathcal{E}(Z)_s^2} d\mathcal{E}(Z)_s + \int_0^t \frac{1}{\mathcal{E}(Z)_s^3} d\langle \mathcal{E}(Z) \rangle_s \\ &= 1 - \int_0^t \frac{1}{\mathcal{E}(Z)_s} dZ_s + \int_0^t \frac{1}{\mathcal{E}(Z)_s} d\langle Z \rangle_s. \end{aligned}$$

Therefore, using again Itô formula, we have

$$\begin{aligned} dY_t &= d\left(\frac{X_t}{\mathcal{E}(Z)_t}\right) = X_t d\left(\frac{1}{\mathcal{E}(Z)_t}\right) + \frac{dX_t}{\mathcal{E}(Z)_t} + d\left\langle X, \frac{1}{\mathcal{E}(Z)} \right\rangle_t \\ &= X_t \left(-\frac{1}{\mathcal{E}(Z)_t} dZ_t + \frac{1}{\mathcal{E}(Z)_t} d\langle Z \rangle_t\right) + \frac{1}{\mathcal{E}(Z)_t} (dH_t + X_t dZ_t) + d\left\langle X, \frac{1}{\mathcal{E}(Z)} \right\rangle_t \\ &= \frac{1}{\mathcal{E}(Z)_t} (dH_t + X_t d\langle Z \rangle_t) + d\left\langle X, \frac{1}{\mathcal{E}(Z)} \right\rangle_t \\ &= \frac{1}{\mathcal{E}(Z)_t} (dH_t + X_t d\langle Z \rangle_t) - \frac{1}{\mathcal{E}(Z)_t} d\langle H, Z \rangle_t - \frac{X_t}{\mathcal{E}(Z)_t} d\langle Z \rangle_t \\ &= \frac{dH_t - d\langle H, Z \rangle_t}{\mathcal{E}(Z)_t}. \end{aligned}$$

We deduce that

$$X_t = \mathcal{E}(Z)_t \left( H_0 + \int_0^t \frac{1}{\mathcal{E}(Z)_s} dH_s - \int_0^t \frac{1}{\mathcal{E}(Z)_t} d\langle H, Z \rangle_s \right).$$

2. Taking  $H_t = x + B_t$  and  $Z_t = -\beta t$  in the previous question, we obtain

$$X_t = e^{-\beta t} \left( x + \int_0^t e^{\beta s} dB_s \right).$$