

Université Paris Dauphine - PSL

# Stochastic Calculus

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# **Brownian Motion**

- 1.1 Definition
- 1.2 Properties of Brownian Sample Paths
- 1.3 Markov property of Brownian Motion

# Stopping times and Martingales

# 2.1 Stopping times

# 2.2 Martingales

# 2.3 Local Martingales

**Definition 2.1.** An adapted process  $M = (M_t)_{t \geqslant 0}$  with continuous sample paths and such that  $M_0 = 0$  a.s. is called a continuous local martingale if there exists a nondecreasing sequence  $(T_n)_{n\geqslant 0}$  of stopping times such that  $T_n \uparrow \infty$  and, for every n, the stopped process  $M^{T_n}$  is a uniformly integrable martingale.

More generally, when we do not assume that  $M_0 = 0$  a.s., we say that M is a continuous local martingale if the process  $N_t = M_t - M_0$  is a continuous local martingale.

In all cases, we say that the sequence of stopping times  $(T_n)$  reduces M if  $T_n \uparrow \infty$  and, for every n, the stopped process  $M^{T_n}$  is a uniformly integrable martingale.

### Proposition 2.2.

- (i) A nonnegative continuous local martingale M such that  $M_0 \in L^1$  is a supermartingale.
- (ii) A continuous local martingale M such that there exist a random variable  $Z \in L^1$  with  $|M_t| \leq Z$  for every  $t \geq 0$  (in particular a bounded continuous local martingale) is a uniformly integrable martingale.
- (iii) If M is a continuous local martingale and  $M_0 = 0$  (or more generally  $M_0 \in L^1$ ), the sequence of stopping times

$$\tau_n := \inf\{t \geqslant 0 : |M_t| \geqslant n\}$$

reduces M.

# 2.4 Finite variation processes

# 2.5 Quadratic variation

**Definition 2.3.** Let M be a continuous local martingale, then there exist a unique finite variation process denoted  $\langle M \rangle$  such that the process  $M^2 - \langle M \rangle$  is a continuous local martingale. Moreover we

have, for every  $t \ge 0$ ,

$$\sum_{i=1}^{p_n} (M_{t_i^n} - M_{t_{i-1}^n})^2 \xrightarrow[n \to \infty]{\mathbb{P}} \langle M \rangle_t,$$

where  $0 = t_0^n < \cdots < t_{p_n}^n = t$  is a sequence of subdivision of [0, t] whose mesh tend to 0.

**Proposition 2.4.** Let M be a continuous local martingale such that  $M_0 = 0$ . Then we have  $\langle M, M \rangle = 0$  if and only if M = 0.

**Theorem 2.5.** Let M be a continuous local martingale with  $M_0 \in L^2$ .

- (i) The following are equivalent:
  - (a) M is a martingale bounded in  $L^2$ .
  - (b)  $\mathbb{E}\left[\langle M, M \rangle_{\infty}\right] < \infty$ .

Furthermore, if these properties hold, the process  $M^2 - \langle M, M \rangle$  is a uniformly integrable martingale.

- (ii) The following are equivalent:
  - (a) M is a  $L^2$  martingale.
  - (b)  $\mathbb{E}\left[\langle M, M \rangle_t\right] < \infty$  for every  $t \geqslant 0$ .

Furthermore, if these properties hold, the process  $M^2 - \langle M, M \rangle$  is a martingale.

**Definition 2.6.** If M and N are two continuous local martingales, the bracket  $\langle M, N \rangle$  is the finite variation process defined by setting, for every  $t \ge 0$ ,

$$\langle M,N\rangle_t = \frac{1}{2}\left(\langle M+N,M+N\rangle_t - \langle M,M\rangle_t - \langle N,N\rangle_t\right)$$

**Definition 2.7.** Two continuous local martingales M and N are said to be orthogonal if  $\langle M, N \rangle = 0$ , which holds if and only if MN is a continuous local martingale.

**Proposition 2.8** (Kunita-Watanabe). Let M and N be two continuous local martingales and let H and K be two measurable processes. Then, a.s.,

$$\int_0^\infty |H_s| |K_s| |d\langle M, N\rangle_s| \leqslant \left( \int_0^\infty H_s^2 d\langle M, M\rangle_s \right)^{1/2} \left( \int_0^\infty K_s^2 d\langle N, N\rangle_s \right)^{1/2}.$$

# 2.6 Semi-Martingales

**Definition 2.9.** A process  $X = (X_t)_{t \ge 0}$  is a continuous semimartingale if it can be written in the form

$$X_t = M_t + A_t,$$

where M is a continuous local martingale and A is a finite variation process.

The decomposition X = M + A is then unique up to indistinguishability. We say that this is the canonical decomposition of X.

**Definition 2.10.** Let X = M + A and X' = M' + A' be the canonical decompositions of two continuous semimartingales X and X'. The bracket  $\langle X, X' \rangle$  is the finite variation process defined by

$$\langle X, X' \rangle_t = \langle M, M' \rangle_t$$
.

In particular, we have  $\langle X,X\rangle_t=\langle M,M\rangle_t.$ 

# Stochastic Integration

# 3.1 Construction of stochastic integrals

## 3.2 Itô's formula

**Theorem 3.1** (Itô's formula). Let  $X = (X^1, ..., X^d)$  be d continuous semi-martingales, and let F be a twice continuously differentiable real function on  $\mathbb{R}^d$ . Then for every  $t \ge 0$ ,

$$F(X_t) = F(X_0) + \sum_{i=1}^d \int_0^t \frac{\partial F}{\partial x^i}(X_s) dX_s^i + \frac{1}{2} \sum_{i,j=1}^d \int_0^t \frac{\partial^2 F}{\partial x^i \partial x^j}(X_s) d\langle X^i, X^j \rangle_s.$$

Corollary 3.2 (Integration by part). Taking d = 2 and F(x,y) = xy, then if X and Y are two continuous semi-martingales,

$$X_t Y_t = X_0 Y_0 + \int_0^t Y_s dX_s + \int_0^t X_s dY_s + \langle X, Y \rangle_t.$$

Corollary 3.3 (Itô's formula for functions depending on time). By taking  $X_t^1 = t$  and  $X_t^2 = B_t$ , we also get for every  $F : \mathbb{R}_+ \times \mathbb{R} \to \mathbb{R} \in \mathscr{C}^2$ 

$$F(t, B_t) = F(0, B_0) + \int_0^t \frac{\partial F}{\partial x}(s, B_s) dB_s + \int_0^t \left(\frac{\partial F}{\partial t} + \frac{1}{2} \frac{\partial^2 F}{\partial x^2}\right)(s, B_s) ds.$$

**Proposition 3.4.** Let M be a continuous local martingale and, for every  $\lambda \in \mathbb{C}$ , let

$$\mathscr{E}(\lambda M)_t = \exp\left(\lambda M - \frac{\lambda^2}{2} \langle M, M \rangle_t\right).$$

The the process  $\mathcal{E}(\lambda M)$  is a complex continuous local martingale, which can be written in the form

$$\mathscr{E}(\lambda M)_t = e^{\lambda M_0} + \lambda \int_0^t \mathscr{E}(\lambda M)_s dM_s.$$

## A Few Consequences of Itô's Formula

**Theorem 3.5** (Lévy's Characterization of Brownian Motion). Let  $X = (X^1, ..., X^d)$  be an adapted process with continuous sample paths. The following are equivalent:

- (i) X is a d-dimensional  $(\mathscr{F}_t)$ -Brownian motion.
- (ii) The processes  $X^1, \ldots, X^d$  are continuous local martingales, and  $\langle X^i, X^j \rangle_t = \delta_{ij}t$  for every  $i, j \in \{1, \ldots, d\}$ .

In particular, a continuous local martingale M is an  $(\mathcal{F}_t)$ -Brownian motion if and only if  $\langle M, M \rangle_t = t$ , for every  $t \ge 0$ , or equivalently if and only if  $M_t^2 - t$  is a continuous local martingale.

**Theorem 3.6** (Dambis-Dubins-Schwarz). Let M be a continuous local martingale such that almost surely  $\langle M, M \rangle_{\infty} = \infty$ . There exists a Brownian motion  $\beta$  such that

a. s. 
$$\forall t \geq 0$$
,  $M_t = \beta_{\langle M, M \rangle_t}$ .

**Theorem 3.7** (Burkholder-Davis-Gundy inequalities). For every real p > 0, there exist two constants  $c_p, C_p > 0$  depending only on p such that, for every continuous local martingale M with  $M_0 = 0$ , and every stopping time T,

$$c_p \mathbb{E}\left[\langle M, M \rangle_T^{p/2}\right] \leqslant \mathbb{E}\left[(M_T^*)^p\right] \leqslant C_p \mathbb{E}\left[\langle M, M \rangle_T^{p/2}\right].$$

Corollary 3.8. Let M be a continuous local martingale such that  $M_0 = 0$ . The condition

$$\mathbb{E}\left[\langle M,M\rangle_{\infty}^{1/2}\right]<\infty$$

implies that M is a uniformly integrable martingale.

## 3.3 Girsanov's theorem

**Proposition 3.9.** Assume that  $\mathbb{Q}$  is a probability measure on  $(\Omega, \mathscr{F})$ , which is absolutely continuous with respect to  $\mathbb{P}$  on the  $\sigma$ -field  $\mathscr{F}_{\infty}$ . For every  $t \in [0, \infty]$ , let

$$D_t = \left. \frac{d\mathbb{Q}}{d\mathbb{P}} \right|_{\mathscr{F}_t}$$

be the Radon-Nikodym derivative of  $\mathbb{Q}$  with respect to  $\mathbb{P}$  on the  $\sigma$ -field  $\mathscr{F}_t$ . The process  $(D_t)_{t\geqslant 0}$  is a uniformly integrable martingale. Furthermore, for every stopping time T, we have

$$D_T = \left. \frac{d\mathbb{Q}}{d\mathbb{P}} \right|_{\mathscr{F}_T}.$$

Finally, if we assume that  $\mathbb{P}$  and  $\mathbb{Q}$  are mutually absolutely continuous on  $\mathscr{F}_{\infty}$ , we have

$$\inf_{t>0} D_t > 0, \quad \mathbb{P} - \text{a. s.}$$

**Proposition 3.10.** Let D be a continuous local martingale taking positive values. There exists a unique continuous local martingale L such that

$$D_t = \exp\left(L_t - \frac{1}{2}\langle L, L \rangle_t\right) = \mathscr{E}(L)_t.$$

Moreover, L is given by the formula

$$L_t = \log D_0 + \int_0^t \frac{1}{D_s} dD_s.$$

**Theorem 3.11** (Girsanov). Assume that the probability measures  $\mathbb{P}$  and  $\mathbb{Q}$  are mutually absolutely continuous on  $\mathscr{F}_{\infty}$ . Let  $(D_t)_{t\geqslant 0}$  be the martingale with càdlàg sample paths such that, for every  $t\geqslant 0$ ,

$$D_t = \left. \frac{d\mathbb{Q}}{d\mathbb{P}} \right|_{\mathscr{F}_t}$$

Assume that D has continuous sample paths, and let L be the unique continuous local martingale such that  $D = \mathcal{E}(L)$ . Then, if M is a continuous local martingale under  $\mathbb{P}$ , the process

$$\tilde{M} := M - \langle M, L \rangle$$

is a continuous local martingale under  $\mathbb{Q}$ .

**Theorem 3.12.** Let L be a continuous local martingale such that  $L_0 = 0$ . Consider the following properties:

- (i)  $\mathbb{E}\left[\exp\frac{1}{2}\langle L,L\rangle_{\infty}\right]<\infty$  (Novikov's criterion);
- (ii) L is a uniformly integrable martingale, and  $\mathbb{E}\left[\exp\frac{1}{2}L_{\infty}\right]<\infty$  (Kazamaki's criterion);
- (iii)  $\mathcal{E}(L)$  is a uniformly integrable martingale.

Then, 
$$(i) \implies (ii) \implies (iii)$$
.

# Stochastic differential equations

Example 4.1 (Langevin equation).

$$dX_t = -bX_t dt + \sigma dB_t$$

# Exercises

## 5.1 Brownian motion

## Exercise 1.1

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

#### Exercise 1.2

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

## Exercise 1.3

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

## Exercise 1.4

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

## Exercise 1.5

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

## Exercise 1.6

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

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

#### Exercise 1.7

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

### Exercise 1.8

- **1.** Show that for all a > 0,  $(\frac{1}{\sqrt{a}}B_{at})_{t \ge 0}$  is a Brownian motion.
- **2.** Show that for all a > 0,  $(B_a B_{a-t})_{0 \le t \le a}$  is a Brownian motion.
- **3.** Define  $X_0 = 0$  and  $X_t = tB_{1/t}$  for all t > 0. Show that  $(X_t)_{t \ge 0}$  is a Brownian motion.
- **4.** Deduce that  $\lim_{t\to\infty} \frac{B_t}{t} = 0$  a.s.

### Exercise 1.9

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

## Exercise 1.10

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

#### Exercise 1.11

- 1. Show that for all  $t \ge 0$ ,  $|B_t|$  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 1.12

- **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  $(\beta_t)_{t \ge 0}$  is a Brownian motion.

# Exercise 1.13 (Brownian bridge)

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

- 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

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

#### Exercise 1.14

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

#### Exercise 1.15

In this exercise, **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 1.16

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  $\mathscr{F}_{\tau}$ .

### Exercise 1.17

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

#### Exercise 1.18

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

# 5.2 Stopping times and martingales

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

#### Exercise 2.1

- 1. Show that  $(B_t^2 t)_{t \ge 0}$  is a martingale.
- **2.** Construct a martingale from  $(B_t^3)_{t\geqslant 0}$ . Same question with  $B_t^4$ .
- **3.** Prove that  $(e^{\lambda B_t \frac{\lambda^2}{2}t})_{t \ge 0}$  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:

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

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

$$\mathbb{E}\left[B_t^2 - t \,\middle|\, \mathscr{F}_s\right] = B_s^2 - s,$$

confirming that  $(B_t^2 - t)_{t \ge 0}$  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  $\mathscr{F}_s$ , we find:

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

Therefore, the process  $(B_t^3 - 3B_s t)_{t \ge 0}$  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:

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

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

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

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

since  $B_t - B_s$  is independent of  $\mathscr{F}_s$  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  $(e^{\lambda B_t - \frac{\lambda^2}{2}t})_{t\geqslant 0}$  is a martingale.

#### Exercise 2.2

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

- 1. Using the martingale  $(B_t^2 t)_{t \ge 0}$  compute the expectation of  $T_a^{\star}$ .
- 2. Using a well-chosen martingale, compute the variance of  $T_a^{\star}$  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  $T_a^{\star}$  is almost surely finite, the stopped martingale  $(B_{t \wedge T_a^{\star}}^2 - t \wedge T_a^{\star})$  converges almost surely to  $B_{T_a^{\star}}^2 - T_a^{\star} = a^2 - T_a^{\star}$ . Moreover,  $B_{t \wedge T_a^{\star}}^2$  is bounded by  $a^2$ , ensuring  $L^1$  convergence by the dominated convergence theorem. In the other hand,  $t \wedge T_a^{\star}$  converge in  $L^1$  to  $T_a^{\star}$  by monotone convergence. Therefore,

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

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

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

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

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

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

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

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

Therefore, the variance is:

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

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,  $\mathbb{E}\left[M_{T_a^*}\right]=2$ . Evaluating  $M_{T_a^*}$ , we have:

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

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

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,

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

#### Exercise 2.3

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

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

2. Deduce the law of

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

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

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

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

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

#### Exercise 2.4

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

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

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

(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  $\lambda > 0$ ,

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

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

#### Exercise 2.5

Let  $M = (M_t)_{0 \le t \le 1}$  be a sub-martingale. Let  $(\mathcal{G}_s)_{s \ge 0}$  be a sub-filtration of  $(\mathcal{F}_s)_{s \ge 0}$ . 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  $0 \le s \le t$ :

$$\begin{split} \mathbb{E}\left[N_t \,|\, \mathcal{G}_s\right] &= \mathbb{E}\left[\mathbb{E}\left[M_t \,|\, \mathcal{G}_t\right] \,|\, \mathcal{G}_s\right] \\ &= \mathbb{E}\left[M_t \,|\, \mathcal{G}_s\right] \\ &= \mathbb{E}\left[\mathbb{E}\left[M_t \,|\, \mathcal{F}_s\right] \,|\, \mathcal{G}_s\right] \\ &\geqslant \mathbb{E}\left[M_s \,|\, \mathcal{G}_s\right] = N_s. \end{split}$$

#### Exercise 2.6

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

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

## Exercise 2.7

Let  $M = (M_t)_{0 \le t \le 1}$  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  $\mathbb{E}[M_t - M_s \mid \mathscr{F}_s]$  is non negative and of expectation 0, therefore  $\mathbb{E}[M_t - M_s \mid \mathscr{F}_s] = 0$  and M is a martingale.

### Exercise 2.8

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

#### Solution

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

$$\mathbb{E}\left[M_{t+1} \,\middle|\, \mathscr{F}_{t+\frac{1}{n}}\right] = M_{t+\frac{1}{n}}.$$

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

### Exercise 2.9

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

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

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

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

#### Solution

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

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

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

2. We have

$$\mathbb{P}\left(\sup_{t\geqslant 0}|M_t|>a\right) = \mathbb{P}\left(\sup_{t\geqslant 0}|M_t|>a, \sigma_a<\infty\right) + \mathbb{P}\left(\sup_{t\geqslant 0}|M_t|>a, \sigma_a=\infty\right)$$
 
$$\leqslant \mathbb{P}(\sigma_a<\infty) + \mathbb{P}\left(\sup_{t\in [0,\sigma_a]}|M_t|>a\right).$$

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

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

- **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 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  $\mathbb{R}_+$ .

# 5.3 Stochastic integration

#### Exercise 3.1

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  $M_{\tau} = \mathbb{E}[M_{\infty} | \mathscr{F}_{\tau}]$  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  $A \in \mathscr{F}_{s \wedge \tau_n} \subseteq \mathscr{F}_s$ , we know that

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

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

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

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

#### Exercise 3.2

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

#### Solution

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

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

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

#### Exercise 3.3

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

#### Exercise 3.4

Let  $\alpha : \mathbb{R}_+ \to \mathbb{R}$  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  $dX_t = \alpha(t)X_tdt + dB_t$ .

#### Solution

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

$$X_{t} = \int_{0}^{t} \exp\left(\int_{0}^{t} \alpha(u)du - \int_{0}^{s} \alpha(u)du\right) dB_{s}$$

$$= \exp\left(\int_{0}^{t} \alpha(s)ds\right) \int_{0}^{t} \exp\left(-\int_{0}^{s} \alpha(u)du\right) dB_{s}$$

$$= e^{Y_{t}} \int_{0}^{t} e^{-Y_{s}} dB_{s}$$

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

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

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

#### Exercise 3.5

Let  $f: \mathbb{R} \to \mathbb{R}$  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 3.6

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

#### Solution

Since,  $\varphi \in L^2_{loc}$ ,  $\int_0^t \varphi(s) dB_s$  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  $t \ge 0$ . Let  $0 \le s \le t$  we have

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

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  $\mathscr{F}_s$  and is a centered gaussian variable with variance  $\int_s^t \varphi(s)^2 ds$ , we thus obtain

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

$$= e^{\frac{1}{2}\int_s^t \varphi(s)^2 ds} \cdot e^{\int_0^s \varphi(u)dB_u - \frac{1}{2}\int_0^t \varphi(u)^2 du}$$

$$= e^{\int_0^s \varphi(u)dB_u - \frac{1}{2}\int_0^s \varphi(u)^2 du}.$$

Hence the result.

#### Exercise 3.7

Find a progressive process  $X = (X_t)_{t \ge 0}$  such that the process  $Z = (Z_t)_{t \ge 0}$  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

$$Z_{t} = 1 + \int_{0}^{t} Z_{s} dX_{s} - 2 \int_{0}^{t} Z_{s} B_{s} dB_{s} + \frac{1}{2} \int_{0}^{t} Z_{s} d\langle X, X \rangle_{s} - 2 \int_{0}^{t} Z_{s} B_{s} d\langle X, B \rangle_{s} + \int_{0}^{t} Z_{s} (2B_{s}^{2} - 1) ds$$

$$= 1 + \int_{0}^{t} Z_{s} (\sigma_{s} - 2B_{s}) dB_{s} + \int_{0}^{t} Z_{s} \left( \frac{1}{2} \sigma_{s}^{2} - 2B_{s} \sigma_{s} + 2B_{s}^{2} - 1 + b_{s} \right) ds.$$

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 3.8

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

$$\beta_t = \int_0^t \cos(H_s) dX_s - \int_0^t \sin(H_s) dY_s,$$
$$\gamma_t = \int_0^t \sin(H_s) dX_s + \int_0^t \cos(H_s) dY_s.$$

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

### Exercise 3.9

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

## Exercise 3.10

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

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

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

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

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

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

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

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

**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  $\{s\geqslant 0: |B_s|=t\}=\int_0^t B_s^2 ds+\int_0^t \tilde{B}_s^2 ds$  in law.

# Solution

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

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

### Exercise 3.11

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  $\mathbb{E}[\tau]$ .

# 5.4 Stochastic Differential Equations

#### Exercise 4.1

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

- 1. Justify that M is a martingale and express M as a stochastic integral.
- **2.** Show that for all  $b \ge 0$ , the exponential local martingale  $\mathscr{E}(-bM)$  is a martingale. For all T > 0, justify that  $\mathbb{Q} := \mathscr{E}(-bM)_T \cdot \mathbb{P}$  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  $a, b \ge 0$ ,

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

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

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

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,  $\mathbb{E}\left[\langle M \rangle_t\right] = \mathbb{E}\left[\int_0^t B_s^2 ds\right] = \int_0^t \mathbb{E}\left[B_s^2\right] ds = \frac{t^2}{2} < \infty$ , hence M is a martingale.
- **2.** For every  $T \geqslant 0$ ,  $t \in [0,T]$ , we have  $\mathscr{E}(-bM)_t = e^{-\frac{b}{2}B_t^2 + \frac{bt}{2} \frac{b^2}{2}\langle M \rangle_t} \leqslant e^{\frac{bT}{2}}$ . Hence,  $\mathscr{E}(-bM)$  is a martingale,  $\mathbb{E}\left[\mathscr{E}(-bM)_T\right] = 1$  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

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

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  $a, b \ge 0$  we have

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

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

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

Therefore,

$$\mathbb{E}_{\mathbb{Q}}\left[e^{(\frac{b}{2}-a)(\frac{1-e^{-2bt}}{2b})Z-\frac{b}{2}t}\right] = \frac{e^{-\frac{bt}{2}}}{\sqrt{1-(\frac{b}{2}-a)(\frac{1-e^{-2bt}}{b})}}$$

$$= \frac{1}{\sqrt{e^{bt}(1-(\frac{b}{2}-a)(\frac{1-e^{-2bt}}{b}))}}$$

$$= \left(\frac{b}{be^{bt}-(\frac{b}{2}-a)(e^{bt}-e^{-bt})}\right)^{1/2}$$

$$= \left(\frac{b}{b\cosh(bt)+2a\sinh(bt)}\right)^{1/2}.$$

#### Exercise 4.2

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

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

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

 $X_t^0 + x(1-t)$  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 continous 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

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

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

#### Exercise 4.3

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

- **1.** Find a probability measure  $\mathbb{Q}$  on  $\mathscr{F}_{\infty}$  such that  $(X_t)_{t\geqslant 0}$  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 4.4

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

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

Show that  $(f(X_t)e^{-\lambda t})_{t\geqslant 0}$  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  $\lambda \geqslant 0$ . 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  $T_x = \inf\{t \ge 0 : X_t = x\}$ . Show that for every  $\lambda \ge 0$ , 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 4.5

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  $\beta \ge 0$  are constants. The process X is called Ornstein-Uhlenbeck process.

## Solution

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

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

Therefore, using again Itô formula, we have

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

We deduce that

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

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

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