

### Exercise 6.1

Let  $\alpha, \sigma > 0$  and  $x_0 \in \mathbb{R}$ . Use the ansatz  $X_t = a(t)(X_0 + \int_0^t b(s) dB_s)$  for differentiable functions  $a, b : [0, 1] \rightarrow \mathbb{R}$  to find the solution to the following stochastic differential equations:

(i)  $dX_t = -\alpha X_t dt + \sigma dB_t, \quad t \in [0, 1], \quad X_0 = x_0,$

(ii)  $dX_t = -\frac{X_t}{1-t} dt + dB_t, \quad t \in [0, 1), \quad X_0 = 0.$

[Remark: You can apply Proposition 5.2 but then you lose one point per part.]

### Exercise 6.2

Let  $X = (X_t)_{t \in [0, T]}$  be an Itô process with  $X_0 = 0$  and absolutely continuous quadratic variation  $\langle X \rangle$ . Consider the stochastic differential equation (SDE)

$$Z_t = 1 + \int_0^t Z_s dX_s, \quad t \in [0, T]. \quad (1)$$

Show that the unique solution  $(Z_t)_{t \in [0, T]}$  of the SDE (1) is given by

$$Z_t = \mathcal{E}(X)_t := \exp\left(X_t - \frac{1}{2}\langle X \rangle_t\right), \quad t \in [0, T].$$

We call  $\mathcal{E}(X)$  the *stochastic exponential* of  $X$ .

### Exercise 6.3

For  $\mu, \sigma \in C^1(\mathbb{R})$  let  $X$  be a solution to

$$X_t = x_0 + \int_0^t \mu(X_s) ds + \int_0^t \sigma(X_s) dB_s, \quad t \in [0, T].$$

Let  $u \in C^{2,1}(\mathbb{R} \times [0, T])$  be a bounded solution to Kolmogorov's backward equation

$$\begin{cases} \partial_t u(x, t) = \mathcal{L}u(x, t), & x \in \mathbb{R}, t \in (0, T] \\ u(x, 0) = f(x), & x \in \mathbb{R}, \end{cases}$$

where  $f \in C(\mathbb{R})$  is compactly supported and  $\mathcal{L}g := \frac{1}{2}\sigma^2 g_{xx} + \mu g_x$  for  $g \in C^{2,1}(\mathbb{R} \times [0, T])$ . Prove that

$$u(x_0, t) = \mathbb{E}^{x_0}[f(X_t)], \quad \text{for } t \in [0, T] \text{ and any initial value } x_0 \in \mathbb{R}.$$

[Hint: Fix  $t_0 \in (0, T]$  and prove that  $M = (M_t)_{t \in [0, t_0]}$ , with  $M_t = u(X_t, t_0 - t)$ ,  $t \in [0, t_0]$ , is a martingale using Itô's formula.]

## Programming exercise 6

Let  $X = (X_t)_{t \in [0, T]}$  be a solution of the SDE

$$dX_t = \mu(t, X_t) dt + \sigma(t, X_t) dB_t, \quad t \in [0, T], \quad X_0 = x_0$$

where  $(B_t)_{t \in [0, T]}$  is a one-dimensional Brownian motion. For the equidistant grid given by  $t_i = \frac{i}{n}T$ ,  $i = 0, \dots, n$ , the *Euler-Maruyama approximation*  $Y = (Y_i, i \in \{0, \dots, n\})$  of  $X$  is given by  $Y_0 = x_0$  and

$$Y_{i+1} = Y_i + \mu(t_i, Y_i) \frac{T}{n} + \sigma(t_i, Y_i) (B_{t_{i+1}} - B_{t_i}), \quad i = 0, \dots, n-1,$$

Implement this approximation scheme for

(i)  $dX_t = X_t dt + X_t dB_t$ ,  $t \in [0, T]$  with  $X_0 = 1$  and

(ii)  $dX_t = -X_t dt + dB_t$ ,  $t \in [0, T]$  with  $X_0 = 1$ .

Choose an appropriate time horizon  $T$  and step size  $n$ . For (i), also visualize the exact solution that you know from the lecture in the plot and compare it to the *Euler-Maruyama approximation*.