

### Final exam (2h30)

Let  $T > 0$ ,  $x \in \mathbb{R}$ ,  $b : \mathbb{R} \mapsto \mathbb{R}$  a  $C_b^2$  function (can be derived twice and its derivatives are all bounded) and, on a probability space  $(\Omega, \mathcal{F}, \mathbb{P})$ ,  $(W_t)_{t \in [0, T]}$  a standard Brownian motion in  $\mathbb{R}$ . We are interested in the following stochastic differential equation

$$\begin{cases} X_0 &= x, \\ dX_t &= dW_t + b(X_t)dt. \end{cases}$$

We fix  $N \in \mathbb{N}^*$  and for  $0 \leq k \leq N$ , we set  $t_k = k\Delta t$  where  $\Delta t = T/N$ . The Euler scheme with  $N$  steps is defined recursively by

$$\begin{cases} \bar{X}_0 = x \\ \forall 0 \leq k \leq N-1, \forall t \in [t_k, t_{k+1}], \bar{X}_t = \bar{X}_{t_k} + (W_t - W_{t_k}) + b(\bar{X}_{t_k})(t - t_k) \end{cases}$$

- (1) For  $s \in [0, T]$ , we set  $\underline{s} = [s/\Delta t] \times \Delta t$  ( $[\dots] =$  integer part), this is the last discretisation time before  $s$ . Show that

$$\forall t \in [0, T], |X_t - \bar{X}_t| \leq \sup_{x \in \mathbb{R}} |b'(x)| \times \int_0^t |X_{\underline{s}} - \bar{X}_{\underline{s}}| ds + \left| \int_0^t b(X_s) - b(X_{\underline{s}}) ds \right|.$$

(Hint:  $b(X_s) - b(\bar{X}_{\underline{s}}) = b(X_s) - b(X_{\underline{s}}) + b(X_{\underline{s}}) - b(\bar{X}_{\underline{s}})$ .)

(2)

- (a) Show that, for all  $1 \leq k \leq N$ ,

$$\int_{t_{k-1}}^{t_k} \int_{t_{k-1}}^s b'(X_u) dW_u ds = \int_{t_{k-1}}^{t_k} (t_k - u) b'(X_u) dW_u.$$

- (b) Show that for  $1 \leq k \leq N$ ,

$$\begin{aligned} \int_{t_{k-1}}^{t_k} b(X_s) - b(X_{t_{k-1}}) ds &= \int_{t_{k-1}}^{t_k} (t_k - r) (b(X_r) b'(X_r) + \frac{1}{2} b''(X_r) dr \\ &\quad + \int_{t_{k-1}}^{t_k} (t_k - r) b'(X_r) dW_r). \end{aligned}$$

(Hint: start by computing  $b(X_s) - b(X_{t_{k-1}})$  using Itô's formula.)

- (3) Check that, for all  $0 \leq t \leq T$ ,

$$\left| \int_0^t b(X_s) - b(X_{\underline{s}}) ds - \int_0^{\underline{t}} b(X_s) - b(X_{\underline{s}}) ds \right| \leq 2 \sup_{x \in R} |b(x)| \times \Delta t.$$

(4)

- (a) Show that, for all  $0 \leq k \leq N$ ,

$$\begin{aligned} \int_0^{t_k} b(X_s) - b(X_{\underline{s}}) ds &= \int_0^{t_k} (\underline{r} + \Delta t - r) \left( b(X_r) b'(X_r) + \frac{1}{2} b''(X_r) \right) dr \\ &\quad + \int_0^{t_k} (\underline{r} + \Delta t - r) b'(X_r) dW_r, \end{aligned}$$

- (b) And that, for all  $0 \leq t \leq T$ ,

$$\begin{aligned} \left| \int_0^t b(X_s) - b(X_{\underline{s}}) ds - \int_0^{\underline{t}} (\underline{s} + \Delta t - s) \left( b(X_s) b'(X_s) + \frac{1}{2} b''(X_s) \right) ds \right. \\ \left. - \int_0^{\underline{t}} (\underline{s} + \Delta t - s) b'(X_s) dW_s \right| \leq 2 \sup_{x \in R} |b(x)| \times \Delta t. \end{aligned}$$

(5) Show that

$$\begin{aligned} \mathbb{E} \left( \sup_{t \in [0, T]} \left| \int_0^t b(X_s) - b(\underline{X}_s) ds \right| \right) \\ \leq \Delta t \left( 2 \sup_{x \in R} |b(x)| + \int_0^T \mathbb{E} \left( \left| b(X_s) b'(X_s) + \frac{1}{2} b''(X_s) \right| \right) + \sqrt{\int_0^T \mathbb{E}((b'(X_s))^2) ds} \right). \end{aligned}$$

(6) We set  $z(t) = \mathbb{E} \left( \sup_{u \in [0, t]} |X_u - \bar{X}_u| \right)$ . Show that

$$\forall t \in [0, T], z(t) \leq C \left( \Delta t + \int_0^t z(s) ds \right),$$

where the constant  $C$  does not depend on  $N$ .

(7) Conclude that

$$\mathbb{E} \left( \sup_{t \in [0, T]} |X_t - \bar{X}_t| \right) \leq \frac{C}{N},$$

for some constant  $C$  that does not depend on  $N$ .

- (8) Suppose that  $f$  is a function defined in a `python` code. Write a `python` function that returns a simulation  $f(\bar{X}_T)$  (anything vaguely looking like `python` is enough). You will define the constants you need.
- (9) Write a `python` code that returns a Monte-Carlo computation of  $\mathbb{E}(f(\bar{X}_T))$  (anything vaguely looking like `python` is enough).