

# Chapter 6

## Diffusion processes

### 6.1 SDEs with random coefficients

We recall in this section the basic results for stochastic differential equations

$$dX_t = b_t(X_t)dt + \sigma_t(X_t)dW_t, \quad t \in [0, T]. \quad (6.1.1)$$

Here,  $b$  and  $\sigma$  are  $\mathbb{F} \otimes \mathcal{B}(\mathbb{R}^n)$ -progressively measurable functions from  $[0, T] \times \Omega \times \mathbb{R}^n$  to  $\mathbb{R}^n$  and  $\mathbb{R}^{n \times d}$  respectively. In particular, for every fixed  $x \in \mathbb{R}^n$ , the processes  $\{b_t(x), \sigma_t(x), t \in [0, T]\}$  are  $\mathbb{F}$ -progressively measurable.

**Definition 6.1.1** *A strong solution of (6.1.1) is an  $\mathbb{F}$ -progressively measurable process  $X$  such that  $\int_0^T (|b_t(X_t)| + |\sigma_t(X_t)|^2)dt < \infty$  a.s. and*

$$X_t = X_0 + \int_0^t b_s(X_s)ds + \int_0^t \sigma_s(X_s)dW_s, \quad t \in [0, T].$$

The main existence and uniqueness result is the following.

**Theorem 6.1.1** *Let  $X_0$  be a square integrable r.v. independent of  $W$ . Assume that the processes  $b.(0)$  and  $\sigma.(0)$  are in  $\mathbb{H}^2$ , and there exists a constant  $L > 0$  such that*

$$|b_t(x) - b_t(y)| + |\sigma_t(x) - \sigma_t(y)| \leq L|x - y| \quad (6.1.2)$$

*for all  $t \in [0, T]$  and  $x, y \in \mathbb{R}^n$ .*

Then there exists a unique strong solution of (6.1.1) in  $\mathbb{H}^2$ . Moreover, we have

$$\mathbb{E} \left[ \sup_{t \in [0, T]} |X_t|^2 \right] \leq C(1 + \mathbb{E}[|X_0|^2]) \quad (6.1.3)$$

for some constant  $C = C(T, L)$  depending only on  $T$  and  $L$ .

**Proof.** We proceed in two steps. We first prove the existence and uniqueness of a solution to the SDE and then establish (6.1.3).

**Step 1.** Let  $c$  be a positive constant to be fixed later. We define the norm  $\|\cdot\|_{\mathbb{H}_c^2}$  by

$$\|\phi\|_{\mathbb{H}_c^2} = \mathbb{E} \left[ \int_0^T e^{-ct} |\phi_s|^2 ds \right]^{1/2}$$

for  $\phi \in \mathbb{H}^2$ . We notice that the norms  $\|\cdot\|_{\mathbb{H}_c^2}$  and  $\|\cdot\|_{\mathbb{H}^2}$  are equivalent on  $\mathbb{H}^2$ . Consider now the map  $U : \mathbb{H}^2 \rightarrow \mathbb{H}^2$  defined by

$$U(Y)_t = X_0 + \int_0^t b_s(Y_s) ds + \int_0^t \sigma_s(Y_s) dW_s$$

From the Lipschitz property of  $b$  and  $\sigma$  and since  $b.(0), \sigma.(0) \in \mathbb{H}^2$ , the map  $U$  is well defined on  $\mathbb{H}^2$ . We next notice that  $X$  is solution to (6.1.1) if and only if  $X$  is fixed point of  $U$ . Therefore, it is sufficient to prove that  $U$  is a contraction for some convenient  $c$  to get the result. For  $Y, Z \in \mathbb{H}^2$ , we have

$$\begin{aligned} \mathbb{E} [|U(Y)_t - U(Z)_t|^2] &= \\ \mathbb{E} \left[ \left| \int_0^t (b_s(Y_s) - b_s(Z_s)) ds + \int_0^t (\sigma_s(Y_s) - \sigma_s(Z_s)) dW_s \right|^2 \right]. \end{aligned}$$

From Young's inequality we get

$$\begin{aligned} \mathbb{E} [|U(Y)_t - U(Z)_t|^2] &\leq 2\mathbb{E} \left[ \left| \int_0^t (b_s(Y_s) - b_s(Z_s)) ds \right|^2 \right] \\ &\quad + 2\mathbb{E} \left[ \left| \int_0^t (\sigma_s(Y_s) - \sigma_s(Z_s)) dW_s \right|^2 \right]. \end{aligned}$$

Using the Lipschitz properties of  $b$  and  $\sigma$ , Jensen's inequality and the Ito Isometry, we get

$$\mathbb{E}\left[|U(Y)_t - U(Z)_t|^2\right] \leq 2(T+1)L\mathbb{E}\left[\int_0^t |Y_s - Z_s|^2 ds\right].$$

Therefore we get

$$\|U(Y) - U(Z)\|_{\mathbb{H}_c^2} \leq \frac{2(T+1)L}{c} \|Y - Z\|_{\mathbb{H}_c^2}$$

and  $U$  is a  $\|\cdot\|_{\mathbb{H}_c^2}$ -contraction for  $c$  large enough.

**Step 2.** We now prove (6.1.3). We have from Young and Jensen inequalities

$$\begin{aligned} \mathbb{E}\left[\sup_{s \in [0,t]} |X_s|^2\right] &\leq \mathbb{E}\left[\sup_{s \in [0,t]} |X_0 + \int_0^s b_u(X_u)du + \int_0^s \sigma_u(X_u)dW_u|^2\right] \\ &\leq 3\left(\mathbb{E}[|X_0|^2] + t \int_0^t \mathbb{E}[|b_u(X_u)|^2]du + \mathbb{E}\left[\sup_{s \in [0,t]} \left|\int_0^s \sigma_u(X_u)dW_u\right|^2\right]\right). \end{aligned}$$

Using Doob's maximal inequality, we get

$$\mathbb{E}\left[\sup_{s \in [0,t]} |X_s|^2\right] \leq 3\left(\mathbb{E}[|X_0|^2] + t \int_0^t \mathbb{E}[|b_u(X_u)|^2]du + 4\mathbb{E}\left[\int_0^t |\sigma_u(X_u)|^2 du\right]\right).$$

From Lipschitz properties of  $b$  and  $\sigma$ , we get a constant  $C(T, L)$  such that

$$\mathbb{E}\left[\sup_{s \in [0,t]} |X_s|^2\right] \leq C(T, K)\left(1 + \mathbb{E}[|X_0|^2] + \int_0^t \mathbb{E}\left[\sup_{s \in [0,u]} |X_s|^2\right]du\right),$$

and we get the result from Gronwall's lemma.  $\square$

The previous result can be easily extended to any initial time  $t \in [0, T]$  instead of 0. In the sequel, we shall denote by  $X^{t,x} = \{X_s^{t,x}, t \leq s \leq T\}$  the process solution to

$$X_s = x + \int_t^s b_u(X_u)du + \int_t^s \sigma_u(X_u)dW_u, \quad u \in [t, T],$$

for  $t \in [0, T]$  and  $x \in \mathbb{R}^n$ .

In addition to the estimate (6.1.3) of Theorem 6.1.1, we have the following flow continuity results of the solution of the SDE.

**Theorem 6.1.2** Suppose that assumptions of Theorem 6.1.1 hold.

(i) There exists a constant  $C > 0$  such that

$$\mathbb{E} \left[ \sup_{t \leq s \leq t'} |X_s^{t,x} - X_s^{t,x'}|^2 \right] \leq C e^{Ct'} |x - x'|^2$$

for  $t, t' \in [0, T]$  such that  $t \leq t'$  and  $x, x' \in \mathbb{R}^n$ .

(ii) Moreover, if we have

$$B := \sup_{0 \leq s < s' \leq T} (s' - s)^{-1} \mathbb{E} \int_s^{s'} (|b_r(0)|^2 + |\sigma_r(0)|^2) dr < +\infty$$

then, there exists a constant  $C > 0$  such that

$$\mathbb{E} \left[ \sup_{t' \leq s \leq T} |X_s^{t,x} - X_s^{t',x'}|^2 \right] \leq C e^{CT} (B + |x|^2)(t' - t)$$

for  $t, t' \in [0, T]$  such that  $t \leq t'$  and  $x \in \mathbb{R}^n$ .

**Proof.** (i) To simplify notation we set  $\delta x = x - x'$ ,  $\delta X = X^{t,x} - X^{t,x'}$ ,  $\delta b = b(X^{t,x}) - b(X^{t,x'})$  and  $\delta \sigma = \sigma(X^{t,x}) - \sigma(X^{t,x'})$ . We next have from Young and Jensen inequalities

$$|\delta X_s|^2 \leq 3 \left( |\delta x|^2 + (s - t) \int_t^s |\delta b_s|^2 ds + \left| \int_t^s \delta \sigma_s dW_s \right|^2 \right)$$

for  $s \in [t, T]$ . Using Doob's maximal inequality and the Lipschitz properties of  $b$  and  $\sigma$  we get

$$\begin{aligned} \mathbb{E} \left[ \sup_{t \leq s \leq t'} |X_s^{t,x} - X_s^{t,x'}|^2 \right] &\leq \\ 3 \left( |\delta x|^2 + L^2(t' + 4) \int_t^{t'} \mathbb{E} \left[ \sup_{t \leq s \leq u} |X_s^{t,x} - X_s^{t,x'}|^2 \right] du \right) \end{aligned}$$

for  $t' \in [t, T]$ . The result follows from Gronwall's Lemma.

(ii) We set  $X = X^{t,x}$ ,  $X' = X^{t',x}$ ,  $\delta t = t' - t$ ,  $\delta X = X' - X$ ,  $\delta b = b(X') - b(X)$  and  $\delta \sigma = \sigma(X') - \sigma(X)$ . Following the same arguments as in (i) we get

$$\mathbb{E} \left[ \sup_{t' \leq s \leq u} |\delta X_s|^2 \right] \leq 3 \left( \mathbb{E} |\delta X_{t'}|^2 + L^2(T + 4) \int_{t'}^u \mathbb{E} \left[ \sup_{t \leq s \leq r} |\delta X_s|^2 \right] dr \right) \quad (6.1.4)$$

for  $u \in [t', T]$ . We now concentrate on the first term of the left hand side. We have by the same arguments

$$\begin{aligned}\mathbb{E}|\delta X_{t'}|^2 &= \mathbb{E}|X_{t'}^{t,x} - x|^2 \\ &\leq 2\left(T \int_t^{t'} \mathbb{E}|b(X_u^{t,x})|^2 du + \int_t^{t'} \mathbb{E}|\sigma(X_u^{t,x})|^2 du\right) \\ &\leq 6(T+1)\left(\int_t^{t'} (L^2 \mathbb{E}|X_u^{t,x} - x|^2 + L^2|x|^2 + \mathbb{E}[|b_u(0)|^2 + |\sigma_u(0)|^2]) du\right) \\ &\leq 6(T+1)\left((t' - t)(L^2|x|^2 + B) + L^2 \int_t^{t'} \mathbb{E}|X_u^{t,x} - x|^2 du\right).\end{aligned}$$

By Gronwall's lemma we get

$$\mathbb{E}|\delta X_{t'}|^2 \leq C(t' - t)(|x|^2 + B)e^{C(t' - t)}.$$

Putting this last inequality in (6.1.4), we get the result using Gronwall's lemma another time.  $\square$

## 6.2 Markov SDEs

In this section, we restrict the coefficients  $b$  and  $\sigma$  to be deterministic functions of  $(t, x)$ . In this context, we suppose that  $b$  and  $\sigma$  are continuous functions Lipschitz in  $x$  uniformly in  $t$ , that is (6.1.2) holds true. We recall that  $X^{t,x}$  denotes the solution of the stochastic differential equation

$$X_s = x + \int_t^s b_u(X_u)du + \int_t^s \sigma_u(X_u)dW_u, \quad u \in [t, T],$$

for  $t \in [0, T]$  and  $x \in \mathbb{R}^n$ . We then have the two following properties.

- From the pathwise uniqueness, we have the flow property:

$$X_s^{t,x} = X_s^{u, X_u^{t,x}}$$

for  $t \leq u \leq s$ . We notice that this flow property also holds for  $u$  being a stopping time valued in  $[t, s]$ .

- Since  $X^{t,x}$  is adapted to the filtration generated by the process  $\{W_s - W_t, s \geq t\}$  we get from the previous flow continuity and Doob's representation Theorem the existence of a measurable function  $F$  such that

$$X_s^{t,x} = F(t, x, s, W_u - W_t, u \in [t, s])$$

for  $s \leq T$ .

**Proposition 6.2.1 (Markov property)** *We have*

$$\mathbb{E}[\Phi(X_r^{t,x}, u \leq r \leq s) | \mathcal{F}_u] = \mathbb{E}[\Phi(X_r^{t,x}, u \leq r \leq s) | X_u^{t,x}]$$

for any  $u \in [t, s]$  and any measurable bounded (or nonnegative) function  $\Phi : C([u, s]) \rightarrow \mathbb{R}$ .

### 6.3 Connection with PDEs

Let  $X^{t,x}$  be the unique solution to the SDE

$$X_s = x + \int_t^s b_u(X_u)du + \int_t^s \sigma_u(X_u)dW_u, \quad s \in [t, T], \quad (6.3.5)$$

for  $t \in [0, T]$  and  $x \in \mathbb{R}^n$ . We next define the operator  $\mathcal{L}$  by

$$\mathcal{L}\varphi(t, x) = \lim_{h \rightarrow 0} \frac{\mathbb{E}[\varphi(X_{t+h}^{t,x})] - \varphi(x)}{h}$$

for any function  $\varphi : \mathbb{R}^n \rightarrow \mathbb{R}$  such that  $\mathcal{L}\varphi$  is well defined.  $\mathcal{L}$  is called the generator of the diffusion. From Ito's formula,  $\mathcal{L}\varphi$  is well defined for any  $\varphi$  bounded  $C^2$  with bounded derivatives and we have

$$\mathcal{L}\varphi(t, x) = b_t(x) \cdot \nabla \varphi(x) + \frac{1}{2} \text{Tr}(\sigma_t(x) \sigma_t(x)^\top \nabla^2 \varphi(x))$$

for  $t \in [0, T]$  and  $x \in \mathbb{R}^n$ . The generator provides a connection between diffusion processes and linear partial differential equations.

**Proposition 6.3.2** Suppose that the function  $v$  defined by

$$v(t, x) = \mathbb{E}[g(X_T^{t,x})], \quad (t, x) \in [0, T] \times \mathbb{R}^n,$$

is in  $C^{1,2}([0, T] \times \mathbb{R}^n)$ . Then,  $v$  solves the partial differential equation:

$$\partial_t v + \mathcal{L}v = 0 \text{ on } [0, T) \times \mathbb{R}^n$$

with terminal condition

$$v(T, .) = g \text{ on } \mathbb{R}^n.$$

**Proof.** Fix  $(t, x) \in [0, T] \times \mathbb{R}^n$  and let  $\tau := T \wedge \inf\{s > t : |X_s^{t,x} - x| \geq 1\}$ . By the tower property and the Markov property of  $X^{t,x}$  we have

$$v(t, x) = \mathbb{E}[v(s \wedge \tau, X_{s \wedge \tau}^{t,x})]$$

for any  $s \in [t, T]$ . Since  $v \in C^{1,2}([0, T] \times \mathbb{R}^n)$  we have from Ito's formula

$$\begin{aligned} 0 &= \mathbb{E}\left[\int_t^{s \wedge \tau} (\partial_t v + \mathcal{L}v)(u, X_u^{t,x}) du\right] + \mathbb{E}\left[\int_t^{s \wedge \tau} \nabla v \cdot \sigma(u, X_u^{t,x}) dW_u\right] \\ &= \mathbb{E}\left[\int_t^{s \wedge \tau} (\partial_t v + \mathcal{L}v)(u, X_u^{t,x}) du\right] \end{aligned}$$

where the last inequality comes from the boundedness of  $X^{t,x}$  on  $[t, \tau]$ . We now take  $s = t + h$  with  $h > 0$  and we get

$$0 = \mathbb{E}\left[\frac{1}{h} \int_t^{(t+h) \wedge \tau} (\partial_t v + \mathcal{L}v)(u, X_u^{t,x}) du\right].$$

From the mean value theorem and the dominated convergence theorem, we get the result by sending  $h$  to  $0+$ .  $\square$

**Feynman-Kac representation of Cauchy problem** We consider the following linear partial differential equation called Cauchy problem

$$\begin{cases} \partial_t v + \mathcal{L}v - kv + f = 0, & \text{on } [0, T) \times \mathbb{R}^n, \\ v(T, .) = g, & \text{on } \mathbb{R}^n. \end{cases} \quad (6.3.6)$$

with  $k$  and  $f$  two functions from  $[0, T] \times \mathbb{R}^n$  to  $\mathbb{R}$ . The next results provides a representation of this purely deterministic problem by means of stochastic differential equations.

**Theorem 6.3.3** Assume that the coefficients  $b$  and  $\sigma$  satisfy the assumptions of Theorem 6.1.1. Assume further that the function  $k$  is uniformly lower bounded, and  $f$  has quadratic growth in  $x$  uniformly in  $t$ . Let  $v \in C^{1,2}([0, T] \times \mathbb{R}^n) \cap C^0([0, T] \times \mathbb{R}^n)$  be a solution of (6.3.6) with quadratic growth in  $x$  uniformly in  $t$ . Then

$$v(t, x) = \mathbb{E} \left[ \int_t^T \beta_s^{t,x} f(s, X_s^{t,x}) ds + \beta_T^{t,x} g(X_T^{t,x}) \right]$$

where  $X^{t,x}$  is the unique solution to (6.3.5) and

$$\beta_s^{t,x} = e^{-\int_t^s k(u, X_u^{t,x}) du}, \quad s \in [t, T],$$

for  $(t, x) \in [0, T] \times \mathbb{R}^d$ .

**Proof.** We introduce the sequence of stopping times  $(\tau_n)_n$  defined by

$$\tau_n = \left( T - \frac{1}{n} \right) \wedge \inf\{s > t : |X_s^{t,x} - x| \geq n\}, \quad n \geq 1.$$

We notice that  $\tau_n \uparrow T-$   $\mathbb{P}$ -a.s. as  $n \uparrow +\infty$ . We then have from Ito's formula

$$\begin{aligned} d(\beta_s^{t,x} v(s, X_s^{t,x})) &= \beta_s^{t,x} (\partial_t v + \mathcal{L}v - kv)(s, X_s^{t,x}) ds \\ &\quad + \beta_s^{t,x} \nabla v \cdot \sigma(s, X_s^{t,x}) dW_s, \end{aligned}$$

for  $s \in [t, T)$ . Since  $v$  is a solution of (6.3.6), we get

$$d(\beta_s^{t,x} v(s, X_s^{t,x})) = -\beta_s^{t,x} f(s, X_s^{t,x}) ds + \beta_s^{t,x} \nabla v \cdot \sigma(s, X_s^{t,x}) dW_s.$$

Therefore, we get

$$\begin{aligned} v(t, x) - \mathbb{E}[\beta_{\tau_n}^{t,x} v(\tau_n, X_{\tau_n}^{t,x})] &= \mathbb{E} \left[ \int_t^{\tau_n} \beta_s^{t,x} f(s, X_s^{t,x}) ds \right] \\ &\quad - \mathbb{E} \left[ \int_t^{\tau_n} \beta_s^{t,x} \nabla v \cdot \sigma(s, X_s^{t,x}) dW_s \right]. \end{aligned}$$

We then notice that the expectation of the stochastic integral term vanishes as the stochastic integrand is bounded since the function  $v$  is continuous,

the function  $\sigma$  and satisfies (6.1.2) and the function  $k$  is lower bounded. Therefore, we get

$$v(t, x) = \mathbb{E} \left[ \beta_{\tau_n}^{t,x} v(\tau_n, X_{\tau_n}^{t,x}) + \int_t^{\tau_n} \beta_s^{t,x} f(s, X_s^{t,x}) ds \right]. \quad (6.3.7)$$

Using the continuity of  $v$ , the terminal condition of (6.3.6) and the continuity of  $X^{t,x}$  we have

$$\beta_{\tau_n}^{t,x} v(\tau_n, X_{\tau_n}^{t,x}) + \int_t^{\tau_n} \beta_s^{t,x} f(s, X_s^{t,x}) ds \xrightarrow[n \rightarrow +\infty]{\mathbb{P}-a.s.} \beta_T^{t,x} g(X_T^{t,x}) + \int_t^T \beta_s^{t,x} f(s, X_s^{t,x}) ds.$$

Moreover, since  $k$  is lower bounded and the functions  $f$  and  $v$  have quadratic growth in  $x$  uniformly in  $t$ , there exists a constant  $C$  such that

$$\left| \beta_{\tau_n}^{t,x} v(\tau_n, X_{\tau_n}^{t,x}) + \int_t^{\tau_n} \beta_s^{t,x} f(s, X_s^{t,x}) ds \right| \leq C \left( 1 + \sup_{s \in [t, T]} |X_s^{t,x}|^2 \right),$$

for any  $n \geq 1$ . We can then apply the dominated convergence theorem and we get the result by sending  $n$  to  $\infty$  in (6.3.7).  $\square$