

Probabilistic Numerical Methods 2024–2025

Lecture 6: Stochastic Differential Equations

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We fix integers $n, d \geq 1$, a probability space $(\Omega, \mathcal{A}, \mathbb{P})$ provided with a filtration $(\mathcal{F}_t)_{t \geq 0}$ such that \mathcal{F}_0 contains all negligible events, and $(B_t)_{t \geq 0}$ a d -dimensional $(\mathcal{F}_t)_{t \geq 0}$ -Brownian motion.

Given a time-interval $I = [0, T]$ or $I = [0, +\infty)$, we let $b : I \times \mathbb{R}^n \rightarrow \mathbb{R}^n$ and $\sigma : I \times \mathbb{R}^n \rightarrow \mathbb{R}^{n \times d}$ be measurable functions which are bounded on bounded subsets of $I \times \mathbb{R}^n$.

We are interested in the Stochastic Differential Equation

$$dX_t = b(t, X_t)dt + \sigma(t, X_t)dB_t, \quad (\text{SDE})$$

complemented with the initial condition

$$X_0 = \xi, \quad (\text{IC})$$

where ξ is an \mathcal{F}_0 -measurable random variable in \mathbb{R}^n .

The function b is called the *drift* of the SDE, and σ is the *dispersion matrix*. The $n \times n$ matrix

$$a(t, x) := \sigma(t, x)\sigma^\top(t, x)$$

is called the *diffusion* matrix.

1 Solution to (SDE)–(IC)

1.1 Notion of solution and associated differential operator

Definition 1.1 (Solution to (SDE)–(IC)). *A solution to (SDE)–(IC) is an n -dimensional Ito process such that, almost surely¹,*

$$\forall t \in I, \quad \forall i \in \{1, \dots, n\}, \quad X_t^i = \xi^i + \int_{s=0}^t b_i(s, X_s)ds + \sum_{k=1}^d \int_{s=0}^t \sigma_{ik}(s, X_s)dB_s^k.$$

An important object related with (SDE) is the differential operator L_t defined by, for all C^2 functions $\phi : \mathbb{R}^n \rightarrow \mathbb{R}$,

$$L_t \phi(x) = \sum_{i=1}^n b_i(t, x) \frac{\partial \phi}{\partial x_i}(x) + \frac{1}{2} \sum_{i,j=1}^n a_{ij}(t, x) \frac{\partial^2 \phi}{\partial x_i \partial x_j}(x).$$

The reason for the importance of this operator is that if $(X_t)_{t \in I}$ is a solution to (SDE), then when one wants to apply the Ito formula to $\phi(X_t)$, one gets

$$d\phi(X_t) = L_t \phi(X_t)dt + \sigma^\top(t, X_t) \nabla \phi(X_t) \cdot dB_t.$$

¹Throughout the chapter we systematically work with continuous versions of Ito processes.

1.2 Existence and uniqueness for globally Lipschitz continuous coefficients

Theorem 1.2 (Ito). *Assume that there exists $K \geq 0$ such that:*

- (i) *for any $t \in I$, for any $x, y \in \mathbb{R}^n$, $|b(t, x) - b(t, y)| + |\sigma(t, x) - \sigma(t, y)| \leq K|x - y|$;*
- (ii) *for any $t \in I$, for any $x \in \mathbb{R}^n$, $|b(t, x)| + |\sigma(t, x)| \leq K(1 + |x|)$.*

Then (SDE)–(IC) admits a unique solution².

Notice that if b and σ do not depend on t , then the condition (ii) is implied by (i) so it does not need to be checked.

The proof of Theorem 1.2 can be decomposed in 4 steps, see lecture notes for details.

1. If $|\xi| \in \mathbf{L}^2$ and $I = [0, T]$, then (SDE)–(IC) has a unique solution in $\mathbf{L}^2([0, T])$: this follows from a fixed point argument.
2. If $|\xi| \in \mathbf{L}^2$ and $I = [0, T]$, then in fact any solution to (SDE)–(IC) is in $\mathbf{L}^2([0, T])$: this is an a priori estimate, which follows from the Gronwall Lemma, and therefore proves the statement of the Theorem if $|\xi| \in \mathbf{L}^2$ and $I = [0, T]$.
3. If $|\xi|$ is no longer assumed to be in \mathbf{L}^2 , one may still construct a solution as follows: the first two steps provide a collection of processes $\{(X_t^x)_{t \in I}, x \in \mathbb{R}^n\}$ which solve (SDE) with deterministic (and a fortiori \mathbf{L}^2) initial condition $X_0^x = x$. Then setting $X_t(\omega) := X_t^{\xi(\omega)}(\omega)$ yields a solution to (SDE)–(IC), and uniqueness follows from the Lipschitz condition.
4. If $I = [0, +\infty)$, then the extension of the construction is straightforward.

Example: Ornstein–Uhlenbeck process, explicit solution, law at time t , limit when $t \rightarrow +\infty$.

1.3 The case of locally Lipschitz continuous coefficients

Theorem 1.3 (Local existence and uniqueness). *Let D be an open subset of \mathbb{R}^n , and assume that there exists $K_D \geq 0$ such that:*

- (i) *for any $t \in I$, for any $x, y \in D$, $|b(t, x) - b(t, y)| + |\sigma(t, x) - \sigma(t, y)| \leq K_D|x - y|$;*
- (ii) *for any $t \in I$, for any $x \in D$, $|b(t, x)| + |\sigma(t, x)| \leq K_D(1 + |x|)$.*

Then there exists an Ito process $(X_t)_{t \in I}$ such that, letting $\tau_D := \inf\{t \in I : X_t \notin D\}$ ³, we have, almost surely,

$$\forall t < \tau_D, \quad X_t = \xi + \int_{s=0}^t b(s, X_s) ds + \int_{s=0}^t \sigma(s, X_s) dB_s.$$

Moreover, if there exists another Ito process $(X'_t)_{t \in I}$ satisfying the same properties (with exit time from D denoted by τ'_D), then almost surely,

$$\tau_D = \tau'_D \quad \text{and} \quad \forall t < \tau_D, \quad X_t = X'_t.$$

This theorem follows from the combination of Theorem 1.2 and Theorem 2.1, p. 102 in Friedman (SDEs vol. 1).

Assume for simplicity that $I = [0, +\infty)$ and that b and σ satisfy the assumptions of Theorem 1.3 on every open ball with radius M , namely $D = B(0, M)$. An important such case is when b and σ do not depend on t and are C^1 in x . Denoting by τ_M the corresponding exit time, we therefore have the existence and uniqueness of a solution up to the *explosion* time $\tau_* = \sup_M \tau_M$. There are situations in which τ_* is finite, so that X_t indeed explodes when t reaches τ_* (see the SDE $dX_t = \frac{1}{2}e^{2X_t} dt + e^{X_t} dB_t$ in the exercise sheet). On the other hand, it is useful to have criteria ensuring that $\tau_* = \infty$, almost surely, so that existence and uniqueness of a (global-in-time) solution still holds even if the coefficients of the SDE are not globally Lipschitz continuous. An example of such a criterion is provided by the next statement.

²Uniqueness is understood here as: the continuous versions of any two solutions are indistinguishable.

³If $I = [0, T]$ and $X_t \in D$ for all $t \in [0, T]$, we set $\tau_D = T$.

Proposition 1.4 (Global existence by Lyapunov function). *Assume that $I = [0, +\infty)$ and that the assumptions of Theorem 1.3 hold on every open ball $D = B(0, M)$, with corresponding exit time denoted by τ_M . Assume moreover that there exists a C^2 function $\Phi : \mathbb{R}^n \rightarrow \mathbb{R}$ such that:*

- (i) $\Phi \geq 0$ and $\lim_{|x| \rightarrow +\infty} \Phi(x) = +\infty$;
- (ii) $\mathbb{E}[\Phi(\xi)] < +\infty$;
- (iii) *there exists $c \geq 0$ such that for all $t \geq 0$, $L_t \Phi(x) \leq c\Phi(x)$.*

Then $\tau_ = \infty$, almost surely (so (SDE)–(IC) has a unique global-in-time solution), and moreover we have the estimate*

$$\forall t \geq 0, \quad \mathbb{E}[\Phi(X_t)] \leq e^{ct} \mathbb{E}[\Phi(\xi)].$$

Proof. Applying Ito's formula to $\Phi(X_t)e^{-ct}$ for $t < \tau_M$, we get

$$\Phi(X_{t \wedge \tau_M}) e^{-ct \wedge \tau_M} = \Phi(\xi) + \int_{s=0}^{t \wedge \tau_M} e^{-cs} (L_s \Phi(X_s) - c\Phi(X_s)) ds + \int_{s=0}^{t \wedge \tau_M} e^{-cs} \sigma^\top(s, X_s) \nabla \Phi(X_s) \cdot dB_s.$$

Since $(s, x) \mapsto e^{-cs} \sigma^\top(s, x) \nabla \Phi(x)$ is bounded on the bounded set $[0, t] \times B(0, M)$, we deduce that the stochastic integral is integrable and has expectation 0. On the other hand, by (iii), the time integral is almost surely nonpositive. Therefore

$$\mathbb{E} [\Phi (X_{t \wedge \tau_M}) e^{-ct \wedge \tau_M}] \leq \mathbb{E}[\Phi(\xi)],$$

and since $t \wedge \tau_M \leq t$ we deduce that

$$\mathbb{E} [\Phi (X_{t \wedge \tau_M})] \leq e^{ct} \mathbb{E}[\Phi(\xi)].$$

We now show that $\tau_M \rightarrow +\infty$, almost surely. Writing

$$\mathbb{E} [\Phi (X_{t \wedge \tau_M})] \geq \mathbb{E} [\Phi (X_{\tau_M}) \mathbf{1}_{\{\tau_M \leq t\}}] \geq \inf_{|x|=M} \Phi(x) \mathbb{P}(\tau_M \leq t),$$

we deduce that

$$\mathbb{P}(\tau_M \leq t) \leq \frac{e^{ct} \mathbb{E}[\Phi(\xi)]}{\inf_{|x|=M} \Phi(x)}.$$

Using (i) and (ii) we get that the right-hand side goes to 0 when $M \rightarrow +\infty$. This shows that $\tau_M \rightarrow +\infty$ and therefore that $\tau_* = \sup_M \tau_M = +\infty$, almost surely. The final estimate now follows from Fatou's Lemma. \square

2 Discretisation

See lecture notes for details: Euler–Maruyama scheme, strong error, weak error, computation for the Ornstein–Uhlenbeck process.