

Martingale methods in stochastic differential equations

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The concentration of diffusing particles in \mathbb{R}^d is described by the **forward Kolmogorov equation**

$$D_t v(t, x) = \frac{1}{2} \sum_{i,j=1}^d D_i D_j (a_{ij}(x) v(t, x)) - \sum_{i=1}^d D_i (b_i(x) v(t, x)).$$

Here

$$a : \mathbb{R}^d \rightarrow \mathbb{S}^d \quad (d \times d \text{ real symmetric matrices})$$

$$b : \mathbb{R}^d \rightarrow \mathbb{R}^d$$

are the **diffusion** and **drift** parameters.

To solve this equation, a_{ij} and b_j must be sufficiently smooth.

I: PDE approach

Consider the adjoint **backward Kolmogorov equation**

$$D_t u(t, x) = \frac{1}{2} \sum_{i,j=1}^d a_{ij}(x) D_i D_j u(t, x) + \sum_{i=1}^d b_i(x) D_i u(t, x).$$

Theorem. Suppose a and b are **bounded and Hölder continuous**, and a satisfies the **nondegeneracy condition**

$$\langle a(x)y, y \rangle \geq \lambda |y|^2 \quad (y \in \mathbb{R}^d).$$

Then the above problem admits a fundamental solution p such that for any terminal value $u(t, x) = f(x)$ with $f \in C_b(\mathbb{R}^d)$, the solution is given by

$$u(s, x) = \int_{\mathbb{R}^d} p_{t-s}(x, y) f(y) dy \quad (0 \leq s < t).$$

Define the operators $P_t : B_b(\mathbb{R}^d) \rightarrow B_b(\mathbb{R}^d)$ by

$$P_t f(x) = \int_{\mathbb{R}^d} p_t(x, y) f(y) dy.$$

Then P_t is a **Feller semigroup** on $C_0(\mathbb{R}^d)$, i.e.,

- $P_0 = I, P_t \circ P_s = P_{t+s}$
- $0 \leq f \leq 1 \implies 0 \leq P_t f \leq 1$
- $f \in C_0(\mathbb{R}^d) \implies P_t f \in C_0(\mathbb{R}^d)$
- $\lim_{t \downarrow 0} P_t f = f$ for all $f \in C_0(\mathbb{R}^d)$

P_t is **conservative**, i.e., $P_t 1 = 1$.

Its generator extends the operator L , i.e., $C_c^2(\mathbb{R}^d) \subseteq \mathcal{D}(L)$ and

$$P_t f = f + \int_0^t P_r Lf dr \quad (f \in C_c^2(\mathbb{R}^d)).$$

Let P_t be a conservative Feller semigroup with **local** generator L :

$$f \equiv 0 \text{ on } U \implies Lf \equiv 0 \text{ on } U.$$

For each $x \in \mathbb{R}^d$ there exists a unique probability measure \mathbb{P}^x on

$$\Omega := C([0, \infty); \mathbb{R}^d)$$

such that the coordinate process

$$X_t(\omega) := \omega_t \quad (\omega \in \Omega)$$

is a **Markov process** starting at x with transition semigroup P_t , i.e., \mathbb{P}^x -a.s. one has $X_0 = x$ and

$$\mathbb{E}^x(f(X_{t+s}) | \mathcal{F}_s) = P_t f(X_s).$$

For all $x \in X$ and $f \in C_c^2(\mathbb{R}^d)$,

$$M_t^x := f(X_t) - \int_0^t Lf(X_r) dr$$

is a **martingale** with respect to \mathbb{P}^x , i.e.,

$$\mathbb{E}^x(M_t^x | \mathcal{F}_s) = M_s^x \quad (t \geq s).$$

Indeed,

$$\begin{aligned} & \mathbb{E}^x(f(X_t) - f(X_s) | \mathcal{F}_s) - \int_s^t \mathbb{E}^x(Lf(X_r) | \mathcal{F}_s) dr \\ &= P_{t-s}f(X_s) - f(X_s) - \int_s^t P_{r-s}Lf(X_s) dr \\ &= P_{t-s}f(X_s) - f(X_s) - \int_0^{t-s} P_rLf(X_s) dr \\ &= 0. \end{aligned}$$

II: SDE approach

Let B_t be a Brownian motion in \mathbb{R}^d .

Suppose that

$$a(x) = \sigma(x)\sigma^*(x)$$

and consider the stochastic differential equation

$$\begin{cases} dU_t = b(U_t) dt + \sigma(U_t) dB_t \\ U_0 = x \end{cases}$$

A **solution** is a continuous adapted process U_t^x in \mathbb{R}^d such that

$$U_t^x = x + \int_0^t b(U_r^x) dr + \int_0^t \sigma(U_r^x) dB_r$$

Theorem. (Itô) *Assume*

- $b : \mathbb{R}^d \rightarrow \mathbb{R}^d$ is bounded and Lipschitz continuous
- $\sigma : \mathbb{R}^d \rightarrow \mathcal{L}(\mathbb{R}^d)$ is bounded and Lipschitz continuous
- $x \in \mathbb{R}^d$.

Then the stochastic differential equation

$$\begin{cases} dU_t = b(U_t) dt + \sigma(U_t) dB_t \\ U_0 = x \end{cases}$$

admits a unique continuous solution U_t^x .

Proof: Picard iteration.

By Itô's formula, for $f \in C_c^2(\mathbb{R}^d)$ one has

$$\begin{aligned} df(U_t^x) &= Df(U_t^x) dU_t^x + \frac{1}{2} \sum_{i,j=1}^d D^2 f(U_t^x) d[U^x]_t \\ &= Df(U_t^x) \sigma(U_t^x) dB_t + Lf(U_t^x) dt \end{aligned}$$

since $[U^x]_t = \int_0^t \sigma(U_r^x) \sigma^*(U_r^x) dr = \int_0^t a(U_r^x) dr$.

It follows that

$$f(U_t^x) - \int_0^t Lf(U_r^x) dr = f(x) + \int_0^t Df(U_r^x) \sigma(U_r^x) dB_r$$

is a martingale.

III: Martingale approach

Let

$$Lf = \frac{1}{2} \sum_{i,j=1}^d a_{ij} D_i D_j f + \sum_{j=1}^d b_j D_j f$$

for $f \in C_c^2(\mathbb{R}^d)$, with $a : \mathbb{R}^d \rightarrow \mathbb{S}^d$ and $b : \mathbb{R}^d \rightarrow \mathbb{R}^d$ as before.

A probability measure \mathbb{P} on Ω **solves the martingale problem** for L starting at x if

- $X_0 = x$ \mathbb{P} -a.s.
- For all $f \in C_c^2(\mathbb{R}^d)$

$$f(X_t) - \int_0^t Lf(X_r) dr$$

is a martingale with respect to \mathbb{P} .

The martingale problem is **well-posed** if for every $x \in \mathbb{R}^d$ there is a unique solution \mathbb{P}^x .

Example. Take $a_{ij} = \delta_{ij}$ and $b_j = 0$, so

$$L = \frac{1}{2}\Delta.$$

If B_t^x is a Brownian motion starting at x , then by Itô's formula, for $f \in C_c^2(\mathbb{R}^d)$

$$f(B_t^x) - \int_0^t \frac{1}{2} \Delta f(B_r^x) dr = f(x) + \int_0^t Df(B_r^x) dB_r$$

is a martingale.

Hence the law of B_t^x solves the martingale problem.

Conversely, suppose that \mathbb{P}^x solves the martingale problem for $L = \frac{1}{2}\Delta$.

By a stopping time argument, for all $f \in C^2(\mathbb{R}^d)$

$$f(X_t) - \int_0^t \frac{1}{2} \Delta f(X_r) dr$$

is a continuous local martingale.

Taking $f_j(x) = x_j$ and $g_j(x) = x_j^2$,

$$X_t^j \quad \text{and} \quad (X_t^j)^2 - t$$

are continuous local martingales with respect to \mathbb{P}^x .

Theorem (Lévy) *With respect to \mathbb{P}^x , X_t is a Brownian motion starting at x .*

Theorem. (Stroock-Varadhan) *Consider*

$$Lf = \frac{1}{2} \sum_{i,j=1}^d a_{ij} D_i D_j f + \sum_{j=1}^d b_j D_j f.$$

Assume

- $a : \mathbb{R}^d \rightarrow \mathbb{S}^d$ is bounded and continuous and satisfies

$$\langle a(x)y, y \rangle \geq \lambda |y|^2$$

- $b : \mathbb{R}^d \rightarrow \mathbb{R}^d$ is bounded and measurable.

Then:

- The martingale problem for L is well-posed.
- The process X_t has the strong Markov property with respect to the measures \mathbb{P}^x .
- If b is continuous, there exists a unique Feller semigroup P_t on $C_0(\mathbb{R}^d)$ whose generator extends L .

Sketch of the proof

Existence:

- Discretization
- Weak compactness of families of probability measures on Ω .

Uniqueness:

- By a Cameron-Martin-Girsanov transformation:

WLOG $b \equiv 0$.

- By localisation arguments:

WLOG $|a_{ij}(x) - \delta_{ij}| < \eta$.

Sketch of the proof

To deal with this case, note that

$$\mathbb{P}^x = \tilde{\mathbb{P}}^x \iff R^x(\lambda)f = \tilde{R}^x(\lambda)f \quad (\lambda > 0, f \in C_c^2(\mathbb{R}^d))$$

where

$$R^x(\lambda)f = \mathbb{E}^x \int_0^\infty e^{-\lambda t} f(X_t) dt.$$

By the martingale property and integration by parts,

$$R^x(\lambda)(\lambda - L)f = f(x).$$

Formally,

$$\underbrace{(\lambda - L)^{-1}}_{R(\lambda)} = \underbrace{(\lambda - \frac{1}{2}\Delta)^{-1}}_{R_{BM}(\lambda)} \left(I - \underbrace{(L - \frac{1}{2}\Delta)}_{\frac{1}{2}(a-\delta)} \underbrace{(\lambda - \frac{1}{2}\Delta)^{-1}}_{R_{BM}(\lambda)} \right)^{-1}$$

where, for Brownian motion,

$$R_{BM}^x(\lambda)f = \mathbb{E}_{BM}^x \int_0^\infty e^{-\lambda t} f(X_t) dt.$$

Rewriting, this gives

$$R_{BM}^x(\lambda)f = R^x(\lambda)f - R^x(\lambda)\left(\frac{1}{2}\sum_{i,j=1}^d (a_{ij} - \delta_{ij})D_iD_j\right)U_\lambda f.$$

Subtracting these identities for \mathbb{P}^x and $\tilde{\mathbb{P}}^x$ gives

$$\|R^x(\lambda) - \tilde{R}^x(\lambda)\| \leq \frac{\eta}{2} \left(\max_{1 \leq i, j \leq d} \|D_iD_jU_\lambda\| \right) \|R^x(\lambda) - \tilde{R}^x(\lambda)\|.$$

For $p > d/2$, $R^x(\lambda)$ and $\tilde{R}^x(\lambda)$ are bounded on L^p .

L^p -Boundedness of Riesz transforms \rightarrow L^p -boundedness of $D_iD_jU_\lambda$

For $\eta > 0$ small this gives

$$R^x(\lambda) = \tilde{R}^x(\lambda).$$

Invariance principles

Let $X_n : A \rightarrow \mathbb{R}^d$ be i.i.d. standard normal random variables.

Given $x \in \mathbb{R}^d$ and $h > 0$, define $\Phi^{x,h} : A \rightarrow \Omega$ by

$$\Phi^{x,h} = x + \sqrt{h} \left(\sum_{n=1}^{\lfloor t/h \rfloor} X_n + (t - h\lfloor t/h \rfloor) X_{\lfloor t/h \rfloor + 1} \right).$$

Let $P^{x,h}$ be its law.

Theorem. (Donsker) $\lim_{h \downarrow 0, y \rightarrow x} \mathbb{P}^{y,h} = \mathbb{P}_{BM}^x$ weakly.

Stroock-Varadhan theory implies, more generally, **convergence of Markov chains to diffusions**.

References

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