

Non-asymptotic Gaussian Estimates for the Recursive Approximation of the Invariant Measure of a Diffusion

PDE & Probability Methods for Interactions (Sophia)



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Steady regime of diffusion

▷ Let $(X_t)_{t \geq 0}$ be the unique strong solution to the stochastic differential equation (SDE)

$$dX_t = b(X_t)dt + \sigma(X_t)dW_t.$$

where $b : \mathbf{R}^d \rightarrow \mathbf{R}^d$ and $\sigma : \mathbf{R}^d \rightarrow \mathcal{M}(d, q)$ are Lipschitz continuous. Let L be its **infinitesimal generator** defined on twice differentiable functions $g : \mathbf{R}^d \rightarrow \mathbf{R}$ by

$$Lg = (b|\nabla g) + \frac{1}{2}\text{Tr}(\sigma^* D^2 g \sigma).$$

▷ *Existence of a stationary regime (Mean-reversion):*

If there exists a \mathcal{C}^2 Lyapunov function $V : \mathbf{R}^d \rightarrow \mathbf{R}_+$ such that

$$\lim_{|x| \rightarrow +\infty} V(x) = +\infty \quad \text{and} \quad \overline{\lim}_{|x| \rightarrow +\infty} LV(x) = -\infty.$$

then there exists a distribution ν such that $(X_t)_{t \geq 0}$ is **\mathbf{P}_ν -stationary** i.e

$$X_0 \stackrel{d}{=} \nu \quad \text{and} \quad \forall \theta > 0, \quad (X_{t_1}, \dots, X_{t_p}) \stackrel{d}{=} (X_{t_1+\theta}, \dots, X_{t_p+\theta}).$$

Connection and examples

- Connection with stationary Fokker-Planck equation:

$$\nu \text{ invariant} \iff \forall g \in \mathcal{C}_K^2(\mathbf{R}^d, \mathbf{R}), \quad \nu(Ag) = 0$$

(made rigorous through Echeverria-Weiss theorem) i.e. if $\nu(dx) = p(x)dx$, if p solution to

$$A^*p = 0.$$

- **Setting 2 (Stat. mechanics):** If V is \mathcal{C}^2 and $\lim_{|x| \rightarrow +\infty} V(x) = +\infty$

$$dX_t = -\nabla V(X_t)dt + \sigma dW_t \quad \sigma \in (0, +\infty) \rightsquigarrow \nu_\sigma(dx) = C_V e^{-\frac{V(x)}{2\sigma^2}} dx.$$

Then

$$\nu_\sigma \xrightarrow{\text{weakly}} \nu = \text{Unif}(\text{argmin} V) \quad \text{as} \quad \sigma \rightarrow 0.$$

Classical starting result of the theory of simulated annealing ...

Other applications

- **Stationary** stochastic volatility models: Heston model
 \rightsquigarrow γ -distribution, multi-asset Heston models for the pricing and hedging of path dependent options.
- Pricing of swing and spark options, real options for gas plants, gas storages.
- **Ergodic** stochastic control problems (long term investments, etc).

Ergodicity, stability

▷ **Ergodicity:** If ν is an **extremal invariant measure**, then X is \mathbf{P}_ν -ergodic:

$$\nu(dx)\text{-a.s.}, \mathbf{P}_x\text{-a.s.} \quad \nu_t(\omega, x, d\xi) = \frac{1}{t} \int_0^t \delta_{X_s^x(\omega)} ds \xrightarrow{\text{(weakly)}} \nu \text{ as } t \rightarrow +\infty$$

▷ **Stability:** Let $\mathcal{I}_{SDE} = \{\nu : \mathbf{P}_\nu \text{ stationary}\}$.

Theorem (Stability result)

If

$$LV \leq \beta - \alpha V^\rho, \quad \alpha > 0, \rho \in (0, 1],$$

then $\mathcal{I}_{SDE} = \{\nu, \text{ invariant distribution}\} \neq \emptyset$, convex, weakly compact and

$$\forall x \in \mathbf{R}^d, \mathbf{P}_x\text{-a.s.} \quad d_{\text{weak}}(\nu_t(\omega, x, d\xi), \mathcal{I}_{SDE}) \xrightarrow{\text{weakly}} 0.$$

If $\mathcal{I}_{SDE} = \{\nu\}$, for every continuous f with $f = o(V^\rho)$ at infinity,

$$\forall x \in \mathbf{R}^d, \mathbf{P}_x\text{-a.s.} \quad \frac{1}{t} \int_0^t f(X_s^x) ds \longrightarrow \nu(f)$$

Uniqueness of ν

- $\mathcal{L}(X_t^x) = p_t(x, y)\mu(dy)$ with $(\forall x > 0, p_t(x, y) > 0) \mu(dy)$ -a.s.
[\Leftarrow hypo-ellipticity and controlability]

or

- (b, σ) confluence:

$$(b(x) - b(y)|x - y) + \frac{1}{2}\text{Tr}((\sigma(x) - \sigma(y))(\sigma(x) - \sigma(y))^*) < 0$$

then

ν is unique.

Weak rate of convergence

Still assume $\mathcal{I}_{SDE} = \{\nu\}$.

Theorem (Weak rate of convergence, Bhattacharya's CLT)

If $f - \nu(f) = L\varphi$, $\varphi \in \mathcal{C}_b^2$, then

$$\sqrt{t}(\nu_t(\omega, x)(f) - \nu(f)) \xrightarrow{\text{(weakly)}} \mathcal{N}(0, \sigma(f)^2) \quad \text{as } t \rightarrow +\infty.$$

with

$$\sigma(f)^2 = \int_{\mathbf{R}^d} |\sigma^* \varphi|^2 d\nu.$$

Langevin MC: mimicking the ergodic theorem

- ▷ **Aim:** Computing by Langevin Monte Carlo simulation $\nu(f)$.

Langevin/ergodic MC simulation = 1 path of length n .

The constant step approach (Talay,'96)

▷ Euler scheme with constant step $\gamma > 0$:

$$\bar{X}_{n+1}^\gamma = \bar{X}_n^\gamma + \gamma b(\bar{X}_n^\gamma) + \sigma(\bar{X}_n^\gamma)(W_{n\gamma} - W_{(n-1)\gamma}), \quad n \geq 0, \quad \bar{X}_0^\gamma = x.$$

▷ Markov chain $(\bar{X}_n^\gamma)_{n \geq 0}$ shares the properties of the diffusions for a small enough step γ .

- (Unique) invariant distribution ν^γ .
- stability/positive recurrence of the chain.

▷ "Regular"/uniform empirical measures:

$$\mu_n^\gamma(\omega, dx) = \frac{1}{n\gamma} \sum_{k=1}^n \gamma \delta_{\bar{X}_{k-1}^\gamma(\omega)} \xrightarrow{\mathbf{R}^d} \nu^\gamma \quad \text{as } n \rightarrow +\infty \mathbf{P}\text{-a.s..}$$

and (Talay 96): $\nu^\gamma \xrightarrow{\mathbf{R}^d} \nu$ at rate $O(\gamma)$.

Unbiased estimation

(Lamberton-P.'00, Lemaire '05, Panloup '06, etc.)

▷ Switch to the **Euler scheme with decreasing step**:

$$\bar{X}_{n+1} = \bar{X}_n + \gamma_{n+1}b(\bar{X}_n) + \sqrt{\gamma_n}\sigma(\bar{X}_n)U_{n+1}, \quad n \geq 0, \quad \bar{X}_0 = x$$

where $(U_n)_{n \geq 1}$ is a(n L^2) white noise and

$$\gamma_n > 0, \quad \gamma_n \downarrow 0 \quad \text{and} \quad \Gamma_n := \gamma_1 + \dots + \gamma_n \rightarrow +\infty.$$

▷ For numerics $U_n = \frac{W_{\Gamma_{n+1}} - W_{\Gamma_n}}{\sqrt{\gamma_n}} \sim \mathcal{N}(0; I_q)$ or **Bernoulli** $(1/2)^{\otimes q}$.

▷ Weighted empirical measures by **mimicking the ergodic continuous time empirical measure**:

$$\nu_n(\omega, dx) = \frac{1}{\Gamma_n} \sum_{k=1}^n \gamma_k \delta_{\bar{X}_{k-1}(\omega)} \simeq \frac{1}{\Gamma_n} \int_0^{\Gamma_n} \delta_{\bar{X}_{N(s)}(\omega)} ds.$$

Theorem (Lamberton-P. ('02) & ('03))

Mean-reversion: $\rho = 1$ & $\mathcal{I}_{inv} = \{\nu\}$ & $\gamma_n = \gamma_1 n^{-\theta}$, $0 < \theta \leq 1$.

(a) Convergence: $\forall x \in \mathbf{R}^d$, $\mathbf{P}_x(d\omega)$ -a.s. $\nu_n^\gamma(\omega, dx) \xrightarrow{\text{(weakly)}} \nu$.

(b) Rate of convergence: Assume $f - \nu(f) = -L\varphi$, $\varphi \in \mathcal{C}_b^3$, $\mathbf{E}U_1^{\otimes 3} = 0$:

- If $\gamma_n = \gamma_1 n^{-\theta}$, $\theta \in (\frac{1}{3}, 1)$ (fast decreasing step): *Bhattacharya's CLT for diffusion holds*:

$$n^{\frac{1-\theta}{2}} (\nu_n(f) - \nu(f)) \xrightarrow{\text{(weakly)}} c_{\gamma_1, \theta} \cdot \mathcal{N}(0, \sigma_f^2).$$

- If $\gamma_n = \gamma_1 n^{-\theta}$, $\theta = \frac{1}{3}$ (optimal rate): *a bias m_f appears*:

$$n^{\frac{1}{3}} (\nu_n(f) - \nu(f)) \xrightarrow{\text{(weakly)}} c_{\gamma_1, 1/3} \cdot \mathcal{N}(\tilde{\gamma} m_f; \sigma_f^2).$$

- If $\gamma_n = \gamma_1 n^{-\theta}$, $\theta \in (0, \frac{1}{3})$ (slowly decreasing step): *the discretization effect slows down the convergence*.

$$c_{\gamma_1, \theta} n^\theta (\nu_n(f) - \nu(f)) \xrightarrow{\mathbf{P}_x/\text{a.s.}} c_{\gamma_1, \theta} \cdot m_f.$$

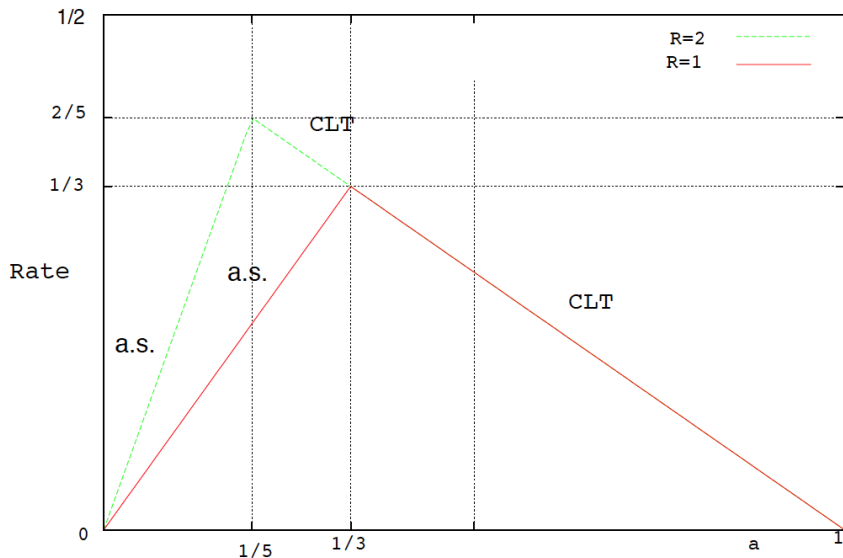


Figure: Rate of convergence depending on a when $\gamma_n \sim \gamma_1 n^{-a}$.

Looking for non-asymptotic confidence intervals

- How to devise non-asymptotic confidence interval in presence of a CLT ?

Let $\gamma_n \asymp n^{-\theta}$, $\theta \in [\frac{1}{3}, 1]$.

$$\mathbf{P}\left(n^{\frac{1-\theta}{2}} (\nu_n(f) - \nu(f)) \geq a\right) \leq Ce^{-ca^2}?$$

- ▶ Malrieu-Talay (*MCQMC² Proc.*, 2006), for the constant step Euler scheme by a direct approach the Euler semi-group (log-Sob, Poincaré) implemented.
- ▶ Frikha-Menozzi (*ECP*, 2012) by a martingale approach based on the Gaussian deviation inequalities.
- ▶ For every Lipschitz continuous function $f : \mathbf{R}^r \rightarrow \mathbf{R}$ and every $\lambda > 0$.

$$\mathbf{E}\left[\exp(\lambda f(U_1))\right] \leq \exp\left(\lambda \mathbf{E}[f(U_1)] + \frac{\lambda^2 [f]_{\text{Lip}}^2}{2}\right).$$

Remark: $\|\sigma\|^2 = \text{Tr}(\sigma\sigma^*)$ denotes the Fröbenius norm.

Theorem (Fast Decreasing Step: $\gamma_k = \gamma_1 k^{-\theta}$, $\frac{1}{3} < \theta \leq 1$)

Assume $U_1 \in L^4$. Let $f : \mathbf{R}^d \rightarrow \mathbf{R}$ be a *coboundary* such that

$$f - \nu(f) = -L\varphi, \quad \varphi = O(\sqrt{V}), \quad \varphi \in \mathcal{C}_{b,Lip}^2(\mathbf{R}^d, \mathbf{R})$$

For every $a > 0$,

$$\mathbf{P}\left[\sqrt{\Gamma_n}|\nu_n(f) - \nu(f)| \geq a\right] \leq 2C_n \exp\left(-c_n \frac{a^2}{2\|\sigma\|_\infty^2 \|\nabla\varphi\|_\infty^2}\right)$$

where the explicit sequences $C_n \downarrow 1$ and $c_n \uparrow 1$.

Can this result be improved?

- We have $\|\sigma\|_\infty^2 \|\nabla\varphi\|_\infty^2$ instead of $\int |\sigma^*\varphi|^2 d\nu$.
- $\|\nabla\varphi\|_\infty^2$ looks somewhat *intrinsic*.
- What about $\|\sigma\|_\infty^2$? In particular, what happens if we also assume that

$\|\sigma\|^2 - \nu(\|\sigma\|^2)$ is a coboundary...

If $\|\sigma\|^2 - \nu(\|\sigma\|^2)$ is a coboundary. . .

Theorem

Still with $\gamma_n = \gamma_1 n^{-\theta}$, $\theta \in (1/3, 1)$. If furthermore

$$\|\sigma\|^2 - \nu(\|\sigma\|^2) = -L\varsigma, \varsigma \in \mathcal{C}_{\text{Lip}}^3(\mathbf{R}^d, \mathbf{R}) [\dots]$$

Then, for every $a \in \left(0, \varepsilon_0 \frac{\sqrt{\Gamma_n}}{\Gamma_n^{(2)}}\right)$ [note that $\frac{\sqrt{\Gamma_n}}{\Gamma_n^{(2)}} \rightarrow +\infty$].

$$\mathbf{P} \left[\left| \sqrt{\Gamma_n} (\nu_n(f) - \nu(f)) \right| \geq a \right] \leq 2\tilde{C}_n \exp \left(-\tilde{c}_n \frac{a^2}{2\nu(\|\sigma\|^2) \|\nabla\varphi\|_\infty^2} \right)$$

where $\tilde{c}_n \uparrow 1$ and $\tilde{C}_n \downarrow 1$ for large enough n respectively.

Critical case (with bias!), $\gamma_n = \frac{\gamma_1}{n^\theta}$, $\theta = \frac{1}{3}$

Theorem (Critical Decreasing step (case $\theta = \frac{1}{3}$))

- Let f be s.t. $f - \nu(f) = -L\varphi$, $\varphi \in \mathcal{C}^4(\mathbf{R}^d, \mathbf{R})$. Then $\forall n \geq 1, \forall a > 0$:

$$\mathbf{P} \left[\left| \sqrt{\Gamma_n}(\nu_n(f) - \nu(f)) - \underbrace{\tilde{\gamma} m}_{\text{bias}} + E_n \right| \geq a \right] \leq 2C_n \exp \left(-c_n \frac{a^2}{2\|\sigma\|_\infty^2 \|\nabla\varphi\|_\infty^2} \right)$$

where $E_n \xrightarrow{n \rightarrow +\infty} 0$ and $c_n \uparrow 1$ and $C_n \downarrow 1$, for large enough n (free of a).

- Moreover, if $\|D^2\varphi(x)\| \leq 1/(1+|x|)$, then $(E_n)_{n \geq 1}$ is **square exponentially tight**:

$$\exists \lambda_0 > 0, \forall \lambda \leq \lambda_0, \sup_{n \geq 1} \mathbf{E} \left[\exp(\lambda |E_n|^2) \right] < +\infty$$

If $\gamma_n = \gamma_1 n^{-1/3}$, $\theta = \frac{1}{3}$ and $\|\sigma\|^2 - \nu(\|\sigma\|^2)$ is a coboundary...

Theorem

If $\gamma_n = \gamma_1 n^{-1/3}$ and furthermore

$$\|\sigma\|^2 - \nu(\|\sigma\|^2) = -L\varsigma, \quad \varsigma \in \mathcal{C}_{\text{Lip}}^3(\mathbf{R}^d, \mathbf{R}) [\dots]$$

Then, for every $a > 0$,

$$\mathbf{P}\left[|\sqrt{\Gamma_n}(\nu_n(f) - \nu(f)) - \tilde{\gamma}m + E_n| \geq a\right] \leq 2\tilde{C}_n \exp\left(-\tilde{c}_n \frac{\Phi_n(a)}{2\nu(\|\sigma\|^2)\|\nabla\varphi\|_\infty^2}\right)$$

where $\tilde{c}_n \uparrow 1$ and $\tilde{C}_n \downarrow 1$ for large enough n respectively and

$$\Phi_n(a) = \alpha_n a^2 - \beta_n a^4, \quad \alpha_n \rightarrow 1, \beta_n \geq 0, \beta_n \rightarrow 0 \quad \text{as } n \rightarrow \infty.$$