

Quantitative approximation of kinetic SDEs: from discrete to continuum

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New trends of stochastic nonlinear systems: well-posedness, dynamics and numerics

CIRM-Marseille

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Kinetic models

- ▶ Newton's Second Law of Motion:

$$\dot{V}_t = \ddot{X}_t = F(X_t, V_t)$$

- ▶ The distribution $\mu_t = \mu_t(dx, dv) := \mathbb{P} \circ (X_t, V_t)^{-1}$ solves the following kinetic equation:

$$\partial_t \mu_t + v \cdot \nabla_x \mu_t + \operatorname{div}_v (F \mu_t) = 0$$

Kinetic models

- ▶ Newton's Second Law of Motion:

$$\dot{V}_t = \ddot{X}_t = F(X_t, V_t) + \sqrt{2}\dot{B}_t.$$

- ▶ The distribution $\mu_t = \mu_t(dx, dv) := \mathbb{P} \circ (X_t, V_t)^{-1}$ solves the following kinetic equation:

$$\partial_t \mu_t + v \cdot \nabla_x \mu_t + \operatorname{div}_v (F \mu_t) = \Delta_v \mu_t.$$

- ▶ **Mesoscopic** (Boltzmann equation) and **macroscopic** (Navier-Stokes equations), ..., Villani, Lions, Golse, Bouchut, Imbert, Mouhot, Silvestre, Guo, Mourrat, ...
- ▶ **Microscopic** (Kinetic SDEs) and **mesoscopic** (Boltzmann equation): Tanaka 1978, PTRF., Mischler-Mouhot 2013, Invent.,...
- ▶ **Difficulty**: degenerate parabolic PDE.

Kinetic SDEs and Euler-Maruyama scheme

- ▶ Kinetic SDEs:

$$\begin{cases} dX_t = V_t dt, \\ dV_t = b(X_t, V_t) dt + dB_t. \end{cases}$$

- ▶ Let $n \in \mathbb{N}$ and $k_n(t) := [nt]/n$. Euler-Maruyama scheme:

$$\begin{cases} dX_t^n = V_t^n dt, \\ dV_t^n = b(X_{k_n(t)}^n, V_{k_n(t)}^n) dt + dB_t. \end{cases}$$

Euler-Maruyama scheme

- ▶ Let $n \in \mathbb{N}$ and $k_n(t) := [nt]/n$. Euler-Maruyama scheme:

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- ▶ Let $h := 1/n$. For $s \in [kh, (k+1)h]$,

$$\begin{aligned} X_{(k+1)h}^n &= X_{kh}^n + \int_{kh}^{(k+1)h} V_s^n ds, \\ V_s^n &= V_{kh}^n + (s - kh)b(X_{kh}^n, V_{kh}^n) + W_s - W_{kh}. \end{aligned}$$

- ▶ Recursive relation:

$$\begin{pmatrix} X_{(k+1)h}^n \\ V_{(k+1)h}^n \end{pmatrix} = F_h^n(X_{kh}^n, V_{kh}^n) + \xi_k^n, \quad (1)$$

where $k = 0, 1, 2, \dots$,

$$F_h^n(x, v) := \begin{pmatrix} x + hv + \frac{h^2}{2} b(r, x, v) \\ v + hb(x, v) \end{pmatrix}, \quad \xi_k^n = \begin{pmatrix} \int_{kh}^{(k+1)h} (W_s - W_{kh}) ds \\ W_{(k+1)h} - W_{kh} \end{pmatrix}.$$

- ▶ ξ_k^n are i.i.d. Gaussian random variables with covariance matrix

$$\begin{pmatrix} \frac{h^3}{3} \mathbb{I}_{d \times d} & \frac{h^2}{2} \mathbb{I}_{d \times d} \\ \frac{h^2}{2} \mathbb{I}_{d \times d} & h \mathbb{I}_{d \times d} \end{pmatrix}.$$

Selected results

- ▶ (Lemaire-Menozzi, 2010. EJP) b is bounded:
Weak convergence: $\mathbb{P} \circ (X^n, V^n)^{-1} \rightarrow \mathbb{P} \circ (X, V)^{-1}$, as $n \rightarrow \infty$.
(Heat kernel estimates)

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- ▶ (Leobacher-Szölgyenyi, 2018. Numer. Math.)
 b is piecewise continuous and bounded:

$$\mathbb{E} \left(\sup_{t \in [0,1]} |(X_t, V_t) - (X_t^n, V_t^n)| \right) \lesssim n^{-\frac{1}{4}+}.$$

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- ▶ (Question)
 - ▷ What about the singular case like

$$b(x, v) = x/|x|^{-1-\gamma}, \quad \gamma > 0?$$

- ▷ Can the convergence rate be improved?
- ▶ Let $p_x, p_v \in [1, \infty]$. Define the mix L^p space with the norm:

$$\|b\|_{L_x^{p_x} L_v^{p_v}} := \left(\int_{\mathbb{R}^d} \left| \int_{\mathbb{R}^d} |b(x, v)|^{p_x} dx \right|^{\frac{p_v}{p_x}} dv \right)^{\frac{1}{p_v}}.$$

Kinetic semi-group and transport

- ▶ Let $b = 0$.

$$(X_t, V_t) = \left(\int_0^t B_s ds, B_t \right).$$

- ▶ Heat kernel estimate: (Kolmogorov, 1934): $p_t(x, v) \sim (X_t, V_t)$,

$$p_t(x, v) = (4\pi t^4/3)^{-d/2} \exp\left(- (3|x|^2 + |3x - 2tv|^2)/(4t^3)\right).$$

- ▶ (Chaudru-Menozzi-Pesce-Zhang, 2023. Bull. Sci. Math.),
(Ren-Zhang, 2025. Bernoulli) ...

- ▶ Kinetic semi-group:

$$P_t f(x, v) = \mathbb{E}f(x + tv + X_t, v + V_t) = \Gamma_t(p_t * f)(x, v) = (\Gamma_t p_t) * (\Gamma_t f)(x, v),$$

where the transport operator Γ_t is defined by

$$\Gamma_t f(x, v) := f(x + tv, v).$$

- ▶ $\partial_t P_t f = (\frac{1}{2}\Delta_v + v \cdot \nabla_x) P_t f$.

Kinetic scaling

- ▶ Recall kinetic SDEs:

$$\begin{cases} dX_t = V_t dt, \\ dV_t = b(X_t, V_t) dt + dB_t. \end{cases}$$

- ▶ Scaling ($b = 0$):

$$\begin{aligned} (X_{\varepsilon^2 t}, V_{\varepsilon^2 t}) &= \left(\int_0^{\varepsilon^2 t} B_s ds, B_{\varepsilon^2 t} \right) = \left(\int_0^t \varepsilon^2 B_{\varepsilon^2 s'} ds', B_{\varepsilon^2 t} \right) \\ &\stackrel{(d)}{=} (\varepsilon^3 \int_0^t B_{s'} ds', \varepsilon B_t) = (\varepsilon^3 X_t, \varepsilon V_t). \end{aligned}$$

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- ▶ Anisotropic scaling: $X : V = 3:1$.



$$p_t(x, v) = t^{-2d} p_1(t^{-3/2} x, t^{-1/2} v).$$

Well-posedness

- ▶ Kinetic SDEs: $V = \dot{X}$,

$$d\dot{X}_t = b(X_t, \dot{X}_t)dt + dB_t.$$

- ▶ Weak well-posedness [[Weak existence + Uniqueness in law](#)]

- ▶ ([Chaudru de Raynal-Menozi, 2021. TAMS](#)) $b \in L_{x,v}^p$, $\frac{4d}{p} < 1$.

- ▶ ([Ren-Zhang, 2025. Bernoulli](#)) Kato class, $b \in L_x^{p_x} L_v^{p_v}$, $\frac{3d}{p_x} + \frac{d}{p_v} < 1$.

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- ▶ Strong well-posedness [[Strong existence + Pathwise uniqueness](#)]

- ▶ ([Chaudru de Raynal, 2017. AIHP](#)) $b \in C_x^{\frac{2}{3}+} \cap C_v^{0+}$.

- ▶ ([Wang-Zhang, 2016. SIAM](#)) Dini continuous.

- ▶ ([Fedrizzi-Flandoli-Priola-Vovelle, 2017. EJP](#))

$$(1 - \Delta_x)^{\frac{1}{3}+} b \in L_{x,v}^p, p > 6d.$$

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- ▶ (Zhang, 2018. Sci. China) $(1 - \Delta_x)^{\frac{1}{3}} b \in L_{x,v}^p, p > 6d$.

- ▶ Distribution drift: (H.-Zhang-Zhu-Zhu, 2024. AOP), (Issoglio-Pagliarani-Russo-Trevisani, 2024.), (Chen-H.-Zhang, 2025) ...

Our aim

- Let $p_x, p_v \in [2, \infty]$ with

$$\frac{3d}{p_x} + \frac{d}{p_v} < 1.$$

	$b \in L_x^{p_x} L_v^{p_v}$	$(1 - \Delta_x)^{\frac{1}{3}} b \in L_x^{p_x} L_v^{p_v}$
Well-known:	Weak well-posedness	Strong well-posedness if $p_x = p_v > 6d$
Our aim:	Weak convergence	Strong well-posedness & Strong convergence

Tamed Euler-Maruyama scheme with time transport

- ▶ Recall $\Gamma_t f(x, v) := f(x + tv, v)$ and let $b_n := b * \phi_n$, where for a smooth probability density function ϕ and $\theta > 0$,

$$\phi_n(x, v) := n^{4d\theta} \phi(n^{3\theta} x, n^\theta v).$$

- ▶ Tamed Euler-Maruyama scheme:

$$\begin{cases} dX_t^n = V_t^n dt, \\ dV_t^n = \Gamma_{t-k_n(t)} b_n(X_{k_n(t)}^n, V_{k_n(t)}^n) dt + dB_t. \end{cases}$$

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- ▶ Benefits of adding $\Gamma_{t-k_n(t)}$:

Let $X_t := \int_0^t B_s ds$ and $f(x, v) = f(x)$ be a Lipschitz function.

$$\triangleright \text{(Without } \Gamma_{t-k_n(t)}) : \mathbb{E}|f(X_t) - f(X_{k_n(t)})| \leq \|f\|_{Lip} \int_{k_n(t)}^t s^{1/2} ds \lesssim n^{-1}.$$

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- ▶ (With $\Gamma_{t-k_n(t)}$):

$$\begin{aligned} \mathbb{E}|f(X_t) - \Gamma_{t-k_n(t)} f(X_{k_n(t)})| &\leq \|f\|_{Lip} \int_{k_n(t)}^t \mathbb{E}|W_s - W_{k_n(s)}| ds \\ &\lesssim \int_{k_n(t)}^t (s - k_n(s))^{1/2} ds \lesssim n^{-3/2}. \end{aligned}$$

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- ▶ $P_t f - f$ is replaced with $P_t f - \Gamma_t f$ (Imbert-Silvestre, 2021. Anal. PDE), (H.-Zhang-Zhu-Zhu, 2024. AOP), (Zhang-Zhang, 2024. AAP), ...

Weak convergence

Main results

- ▶ Assume that $b \in L_x^{p_x} L_v^{p_v}$ with

$$\alpha := \frac{3d}{p_x} + \frac{d}{p_v} < 1.$$

Theorem 1

Let $\theta \in (0, \frac{1}{2\alpha}]$. For any $T > 0$,

$$\sup_{t \in [0, T]} t^{(\theta \vee \frac{1}{2})\alpha} \|\mathbb{P} \circ (X_t, V_t)^{-1} - \mathbb{P} \circ (X_t^n, V_t^n)^{-1}\|_{\text{var}} \lesssim n^{-\frac{1}{2}} + n^{-\theta}.$$

- ▶ When b is bounded, we can replace b_n with b and have

$$\sup_{t \in [0, T]} \|\mathbb{P} \circ (X_t, V_t)^{-1} - \mathbb{P} \circ (X_t^n, V_t^n)^{-1}\|_{\text{var}} \lesssim n^{-\frac{1}{2}}.$$

Remarks:

- (i) We don't need any continuous assumption on b .
- (ii) When $b = b(v)$, the kinetic SDE reduces to

$$dV_t = b(V_t)dt + dB_t. \quad (2)$$

We can replace the taming $b_n := b * \phi_n$ with

$$b_n(v) := \frac{|b(v)| \wedge (Cn^{\alpha\theta})}{|b(v)|} b(v)$$

for some constant $C > 0$.

In (Jourdain-Menzio, 2024. AAP), where $b \in L^p$ with $p > d$ and $\theta = 1/(2\alpha)$, they obtained the convergence of the density, which in particular implies

$$\|\mathbb{P} \circ (V_t)^{-1} - \mathbb{P} \circ (V_t^n)^{-1}\|_{\text{var}} \lesssim n^{-\frac{1-d/p}{2}}.$$

In contrast, our weak convergence rate is $1/2$, which is **independent** of p .

Strong convergence

Main results

- ▶ Assume that $b \in \mathbf{B}_{(p_x, p_v); a}^\beta$ (anisotropic Besov space) with $\beta \in (0, 1)$,

$$\gamma := -\beta + \frac{3d}{p_x} + \frac{d}{p_v} < 1,$$

and assume

$$(1 - \Delta_x)^{\frac{1}{3}} b \in \cup_{\varepsilon > 0} \mathbf{B}_{(p_x, p_v); a}^\varepsilon. \quad (3)$$

- ▶ $\mathbf{B}_{(p_x, p_v); a}^\beta \approx \left[(1 - \Delta_x)^{-\frac{\beta}{6}} L_x^{p_x} L_v^{p_v} \right] \cap \left[(1 - \Delta_v)^{-\frac{\beta}{2}} L_x^{p_x} L_v^{p_v} \right]$.

Theorem 2

- ▶ Under the assumption (3) with $\alpha < 1$, the strong well-posedness holds.
- ▶ If $\gamma \geq 0$, then for any $\theta \in (0, \frac{1}{2\gamma}]$, $\varepsilon > 0$, $T > 0$ and $m \geq 1$,

$$\left(\mathbb{E} \left[\sup_{t \in [0, T]} |(X_t, V_t) - (X_t^n, V_t^n)|^m \right] \right)^{\frac{1}{m}} \lesssim n^{-\frac{1+\beta}{2} + \varepsilon} + n^{-(1-\gamma)\theta + \varepsilon}.$$

- ▶ If $\gamma < 0$, then $b \in L_{x,v}^\infty$, we can replace the taming b_n with b and have

$$\left(\mathbb{E} \left[\sup_{t \in [0, T]} |(X_t, V_t) - (X_t^n, V_t^n)|^m \right] \right)^{\frac{1}{m}} \lesssim n^{-\frac{1+\beta}{2} + \varepsilon}.$$

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Remarks:

- (i) Strong well-posedness holds without assuming $p_x = p_v$, covering both cases $b = b(x)$ and $b = b(v)$.
- (i) The strong convergence rate exceeds $1/2$ when

$$\gamma < 1/2 \quad \text{and} \quad \theta \in (1/(2(1-\gamma)), 1/(2\gamma)).$$
- (ii) When $b = b(v)$, our results coincide with those in (Dareiotis-Gerencsér-Lê, 2023. AAP, Lê-Ling, 2025. AOP) for non-degenerate SDE.
- (iv) When $b = b(x)$, the condition becomes to $b \in \mathbf{B}_{p_x}^{\frac{\beta}{3}}$. In other words, we achieve a convergence rate of $\frac{1+\beta}{2}$ assuming that b has regularity $\frac{\beta}{3}$.

Main results

Theorem 3

Assume $b = b(x) \in \mathbf{C}^{\frac{\beta}{3}}$ with some $\beta \in (2, 3)$. Then we can replace b_n with b and have

$$\left(\mathbb{E} \left[\sup_{t \in [0, T]} |(X_t, V_t) - (X_t^n, V_t^n)|^m \right] \right)^{\frac{1}{m}} \lesssim n^{-\frac{1+\beta}{2} + \epsilon}.$$

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- ▶ **Strong convergence:** The strong convergence exceeds $3/2$ and approaches 2 as $\gamma \rightarrow 1$.
- ▶ **Reason:** This is due to the second-order construction and our scheme:

$$\begin{aligned} dV_t^n &= \Gamma_{t-k_n(t)} b(X_{k_n(t)}^n, V_{k_n(t)}^n) dt + dB_t \\ &= b(X_{k_n(t)}^n - (t - k_n(t)) \dot{X}_{k_n(t)}^n) dt + dB_t. \end{aligned}$$

- ▶ **Relation to the Milstein scheme:**

$$dV_t^n = \left[b(X_{k_n(t)}^n) - (t - k_n(t)) V_{k_n(t)}^n \cdot \nabla b(X_{k_n(t)}^n) \right] dt + dB_t;$$

- ▶ **Related work:** A similar scheme was studied in [Vanden-Eijnden and Ciccotti \(2006, Chemical Physics Letters\)](#) for very smooth b :

$$dV_t^n = \frac{1}{2} \left[b(X_{k_n(t)}^n) + b(X_{k_n(t) + \frac{1}{n}}^n) \right] dt + dB_t.$$

Purely discrete scheme

- ▶ Assume b is bounded. Consider the purely discrete scheme:

$$\begin{cases} X_t^n = \xi + \int_0^t V_s^n ds, \\ V_t^n = \eta + \int_0^t \Gamma_{k_n, 2}(s) - k_n(s) b(X_{k_n(s)}^n, V_{k_n(s)}^n) ds + B_t. \end{cases}$$

- ▶ Weak convergence:

$$\sup_{t \in [0, T]} \left\| \mathbb{P} \circ (X_t^n, V_t^n)^{-1} - \mathbb{P} \circ (X_t, V_t)^{-1} \right\|_{\text{var}} \lesssim n^{-\frac{1}{2}}.$$

- ▶ Strong convergence: assume that $b \in \mathbf{C}_x^{\frac{\beta}{3}} \cap \mathbf{C}_v^\beta$ with $\beta \in (0, 1)$ and $b \in \mathbf{C}_x^{\frac{2}{3}+}$, then for any $\varepsilon > 0$,

$$\left\| \sup_{t \in [0, T]} |(X_t, V_t) - (X_t^n, V_t^n)| \right\|_{L^m(\Omega)} \lesssim n^{-\frac{1+\beta}{2} + \varepsilon}.$$

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Thank you for your time and attention!

Merci beaucoup!