Quantitative approximation of kinetic SDEs: from discrete to continuum

Zimo Hao

Joint work with Khoa Lê and Chengcheng Ling

Bielefeld University

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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:

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

Maxwell 1867; Boltzmann 1872; Vlasov 1938; Landau 1936;

Chapman-Enskog method 1970:

Mesoscopic (Boltzmann equation) and macroscopic (Navier-Stokes equations),, Villani, Lions, Golse, Bouchut, Imbert, Mouhot, Silvestre, Guo, Mourrat, ...

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- Microscopic (Kinetic SDEs) and mesoscopic (Boltzmann equation): Tanaka 1978, PTRF., Mischler-Mouhot 2013, Invent.,...
- (Difficulty): degenerate parabolic PDE; the transfer of regularity between x and v.

Kinetic SDEs and scaling

► Kinetic SDEs:

$$\begin{cases} \mathrm{d}X_t = V_t \mathrm{d}t, \\ \mathrm{d}V_t = b(X_t, V_t) \mathrm{d}t + \mathrm{d}B_t. \end{cases}$$

• Scaling (b = 0):

$$(X_{\varepsilon^{2}t}, V_{\varepsilon^{2}t}) = \left(\int_{0}^{\varepsilon^{2}t} B_{s} \mathrm{d}s, B_{\varepsilon^{2}t}\right) = \left(\int_{0}^{t} \varepsilon^{2} B_{\varepsilon^{2}s'} \mathrm{d}s', B_{\varepsilon^{2}t}\right)$$
$$\stackrel{(d)}{=} \left(\varepsilon^{3} \int_{0}^{t} B_{s'} \mathrm{d}s', \varepsilon B_{t}\right) = \left(\varepsilon^{3} X_{t}, \varepsilon V_{t}\right).$$

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- Anisotropic scaling: X : V = 3:1.
- ▶ Heat kernel estimate: (Kolmogorov, 1934): $p_t(x, v) \sim (X_t, V_t)$ with b = 0,

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

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

(Chaudru-Menozzi-Zhang, 2023. Bull. Sci. Math.), ...

• Kinetic SDEs: $V = \dot{X}$,

$$\mathrm{d}\dot{X}_t = b(X_t, \dot{X}_t)\mathrm{d}t + \mathrm{d}B_t.$$

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

 $\triangleright \quad \text{(Chaudru de Raynal-Menozzi, 2021. TAMS)} \quad b \in L^q_t L^p_{x,v}, \left| \frac{2}{q} + \frac{4d}{p} < 1 \right|.$

▷ (Ren-Zhang, 2024. Bernoulli) Kato class, $b \in L^q_t L^{p_x}_x L^{p_y}_v$, $\frac{2}{q} + \frac{3d}{p_x} + \frac{d}{p_y} < 1$.

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

▷ (Zhang, 2018. Sci. China) $(1 - \Delta_x)^{\frac{1}{3}} b \in L^p_{t,x,v}, \quad \frac{2}{p} + \frac{3d}{p} + \frac{d}{p} < 1$.

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▷ (Our results) There is some $\beta > 0$,

$$(1+(1-\Delta_x)^{\frac{1}{3}}-\Delta_v)^{\frac{\beta}{2}}(1-\Delta_x)^{\frac{1}{3}}b\in L^{p_x}_x L^{p_y}_v, \quad \frac{3d}{p_x}+\frac{d}{p_v}<1.$$
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- Distribution drift: (H.-Zhang-Zhu-Zhu, 2024. AOP), (Issoglio-Pagliarani-Russo-Trevisani, 2024.) ...
- Lévy processes cases: (Chen-Zhang, 2018. JMPA), (H.-Wu-Zhang, 2020. JMPA), (Marino-Menozzi, 2023. SPA)...

Euler-Maruyama scheme

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

$$\begin{cases} \mathrm{d}X_t^n = V_t^n \mathrm{d}t, \\ \mathrm{d}V_t^n = b(X_{k_n(t)}^n, V_{k_n(t)}^n) \mathrm{d}t + \mathrm{d}B_t. \end{cases}$$

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

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

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

Tamed Euler-Maruyama scheme with time transport

► Let $\Gamma_t f(x, v) := f(x + tv, v)$ and $b_n := b * \phi_n$, 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:

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

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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. ▷ (Without $\Gamma_{t-k_n(t)}$) : $\mathbb{E}[f(X_t) - f(X_{k_n(t)})] \le ||f||_{Lip} \int_{k_n(t)}^t s^{1/2} ds \lesssim n^{-1}$. ▷ (With $\Gamma_{t-k_n(t)}$) : $\mathbb{E}[f(X_t) - \Gamma_{t-k_n(t)}] \le ||f||_{Lip} \int_{t-k_n(t)}^t \mathbb{E}[W_t - W_{k_n(t)}] ds$

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

Main results

• Assume that $b \in L^{p_x}_x L^{p_y}_v$ with

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

Let $\theta \in (0, \frac{1}{2\alpha})$.

Theorem 1

Weak convergence:

$$\int_0^1 \|\mathbb{P} \circ (X_t, V_t)^{-1} - \mathbb{P} \circ (X_t^n, V_t^n)^{-1}\|_{\text{var}} dt \lesssim n^{-\frac{1}{2}} + n^{-\theta}.$$

Strong convergence: Under the condition (H_{β}) ,

$$\mathbb{E}\left[\sup_{t\in[0,1]}|(X_t,V_t)-(X^n_t,V^n_t)|\right]\lesssim n^{-\frac{1+\beta/3}{2}+\varepsilon}+n^{-(\beta+1-\alpha)\theta+\varepsilon},\quad\forall\varepsilon>0.$$

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- When b is bounded, we can replace b_n with b and let $\theta = +\infty$.
- ▶ The key point in the proof: Littlewood-Paley's type estimate for the heat kernel.
- Difficulty: $p_t * p_s \neq p_{t+s}$; $p_t * (\Gamma_s p_s) = \Gamma_s p_{t+s}$.

Remarks:

- (i) We don't need any continuous assumption on b.
- (ii) In contrast to (Jourdain-Menozzi, 2024. AAP) for $b \in L^p$ and

$$\mathrm{d}X_t = b(X_t)\mathrm{d}t + \mathrm{d}W_t,\tag{1}$$

$$\|\mathbb{P}\circ (X_t)^{-1}-\mathbb{P}\circ (X_t^n)^{-1}\|_{\mathrm{var}}\lesssim n^{-rac{1-d/p}{2}},$$

our weak convergence rate is 1/2 when $\theta \in [1/2, 1/(2\alpha)]$, which is independent of (p_x, p_y) .

(iii) The strong convergence rate is 1/2 when $\alpha < 1/2$ and $\theta \in (1/(2(1-\alpha)), 1/(2\alpha))$, which coincides the results in (Lê-Ling, 2021) for SDE (1).

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Open questions:

- (i) Is the convergence rate 1/2 optimal (especially for the weak case)? (Ellinger-Müller-Gronbach-Yaroslavtseva 2024)
- (ii) When b(x, v) = b(x), would it be possible to improve the convergence rate from $\frac{1}{2}$ to $\frac{3/2}{2}(=3/4)$?

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Thank you!