Convergence rate of EM algorithm for SDEs under integrability condition

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- Motivations
- Convergence rate of EM scheme with integrable drift
- Convergence rate of EM scheme with non-integrable drift
- Proofs of main results

This talk is based on a joint paper with Xing Huang & Shaoqin Zhang.

- Strong/weak convergence of numerical schemes for SDEs with regular coefficients has been investigated extensively; Kloeden-Platen (1992);
- EM scheme is valid merely to SDEs with linear growth; Jentzen, et al. (CMS, 2016);
- Backward EM scheme (Higham, et al., 2002), tamed EM scheme (Dareiotis, et al., 2016; Sabanis, 2016), truncated EM scheme (Mao, 2015), ...;

Convergence of numerical algorithms for SDEs with irregular coefficients:

- Gyöngy-Rásonyi (SPA, 2011): SDEs with Hölder continuous diffusions via Yamada-Watanabe approach;
- Yan (AOP, 2002): SDEs with Hölder continuous drifts by Meyer-Tanaka formula & estimates on local times;
- Pamen-Taguchi (SPA, 2017); SDEs with Hölder continuous drift using the regularity of backward Kolmogrov equations (i.e., Zvonkin's transformation);

Numerical approximations of SDEs with discontinuous drifts:

 Leobacher-Szölgyenyi (AAP, 2017) & Leobacher-Szölgyenyi (Numer. Math., 2018) under piecewise Lipschitz assumption.

Motivations

SDEs with integrable coefficients:

- Zvonkin (1974): pioneer work;
- Krylov-Röckner (PTRF, 2005): additive noise;
- Zhang (SPA, 2005; EJP, 2011): multiplicative noise;
- Huang-Wang (SPA, 2019) & Röckner-Zhang (2018): McKean-Vlasov SDEs;
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Approximations of SDEs with integrable coefficients:

• Ngo-Taguchi (2016): integrability condition & one-sided Lipschitz;

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• Ngo-Taguchi (2017): integrability condition & 1-dimension.

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Framework

In this talk, we consider

$$dX_t = b(X_t)dt + \sigma(X_t)dW_t, \quad t \ge 0, \quad X_0 = x,$$
(1)

where

•
$$b: \mathbb{R}^d \to \mathbb{R}^d$$
, $\sigma: \mathbb{R}^d \to \mathbb{R}^d \otimes \mathbb{R}^d$;

• $(W_t)_{t\geq 0}$ is a *d*-dimensional Brownian motion.

EM scheme associated with (1): For any $\delta \in (0, 1)$,

$$dX_t^{(\delta)} = b(X_{t_{\delta}}^{(\delta)})dt + \sigma(X_{t_{\delta}}^{(\delta)})dW_t, \quad t \ge 0, \quad X_0^{(\delta)} = X_0,$$
(2)

where $t_{\delta} := \lfloor t/\delta \rfloor \delta$.

(A1) $||b||_{\infty} := \sup_{x \in \mathbb{R}^d} |b(x)| < \infty$ and there is $p > \frac{d}{2}$ such that $|b|^2 \in L^p$; (A2) There exist a constant $\alpha > 0$ and a locally integrable function $\phi \in C(\mathbb{R}_+; \mathbb{R}_+)$ such that

$$\frac{1}{s^{\frac{d}{2}}} \int_{\mathbb{R}^d} |b(x+y) - b(x+z)|^2 e^{-\frac{1}{s}|x|^2} dx \le \phi(s) |y-z|^{\alpha}, \ y, z \in \mathbb{R}^d, \ s > 0;$$

(A3) There exist constants $\check{\lambda}_0, \hat{\lambda}_0, L_0 > 0$ such that

$$|\check{\lambda}_0|\xi|^2 \le \langle (\sigma\sigma^*)(x)\xi,\xi \rangle \le \widehat{\lambda}_0|\xi|^2, \ x,\xi \in \mathbb{R}^d \qquad (UE)$$
 (3)

and

$$\|\sigma(x) - \sigma(y)\|_{\mathrm{HS}} \le L_0 |x - y|, \quad x, y \in \mathbb{R}^d.$$
(4)

- Under (A1) and (A3), has a unique strong solution;
- $(X_{k\delta}^{(\delta)})_{k\geq 0}$ is a homogeneous Markov chain;
- (A2) is imposed to reveal the convergence rate of EM scheme;
- $b(x) = \mathbb{1}_{[a,b]}(x)$ satisfies (A2).

Main Result I

Theorem

Assume (A1)-(A3). Then, for $\beta \in (0,2)$ and q > 2, there exist constants $C_1, C_2 > 0$ such that

$$\mathbb{E}\left(\sup_{0 \le t \le T} |X_t - X_t^{(\delta)}|^{\beta}\right) \le C_1 e^{C_2(1+||b|^2||_{L^p}^{\gamma_0})} \delta^{\frac{\beta}{2}(1\wedge\frac{\alpha}{2})},$$
(5)
where $\gamma_0 := \frac{1}{1-1/q-d/2p}.$

Remarks:

• Investigate rate of EM scheme for SDEs with integrability condition;

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- Drop the piecewise Lipschitz continuity on the drifts;
- Extend to high dimensional setting;

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Further Remarks

(A2) can be replaced by (A2') below

(A2') There exist constants $\beta, \theta > 0$ such that

$$\frac{1}{(rs)^{d/2}} \sup_{z \in \mathbb{R}^d} \int_{\mathbb{R}^d \times \mathbb{R}^d} |b(x) - b(y)|^2 \mathrm{e}^{-\frac{|x-z|^2}{s}} \mathrm{e}^{-\frac{|y-x|^2}{r}} \mathrm{d}y \mathrm{d}x \le Cr^{\theta} s^{\beta-1}$$
 for some constant $C > 0$.

- The drift b satisfying (A2') is said to the Gaussian-Besov class with the index (β, θ), denoted by GB²_{β,θ}(ℝ^d).
- For $\theta>0$ and $p\in [2,\infty)\cap (d,\infty),$ if the Gagliardo seminorm

$$[b]_{W^{p,\theta}} := \left(\int_{\mathbb{R}^d \times \mathbb{R}^d} \frac{|b(x) - b(y)|^p}{|x - y|^{d + p\theta}} \mathrm{d}x \mathrm{d}y \right)^{\frac{1}{p}} < \infty,$$

then $b \in GB^2_{1-\frac{d}{p},\theta}(\mathbb{R}^d).$

Under $||b||_{\infty} < \infty$ and (A3), the transition density p of $(X_t)_{t \ge s}$ satisfies

$$p(s, t, x, x') \le e^{\frac{\|b\|_{\infty}^2 T}{2\lambda_0}} \sum_{i=0}^{\infty} \frac{\beta_T^i}{\Gamma(1+\frac{i}{2})} p_0(t-s, x, x'), \quad 0 \le s < t \le T,$$
(6)

where $\Gamma(\cdot)$ is the Gamma function, and

$$p_0(t, x, x') := \frac{\mathrm{e}^{-\frac{|x-x'|^2}{16\lambda_0 t}}}{(2\pi \breve{\lambda}_0 t)^{d/2}}.$$
(7)

- The proof is due to parametrix method (Konakov-Mammen, PTRF,2000);
- β_T is explicit.

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Under $(\mathbf{A1})$ and $(\mathbf{A3})\text{,}$

$$p^{(\delta)}(j\delta,t,x,y) \le \frac{\Lambda_3 \mathrm{e}^{-\frac{|y-x|^2}{\kappa_0(t-j\delta)}}}{(2\pi\check{\lambda}_0(t-j\delta))^{d/2}}, \quad x,y \in \mathbb{R}^d, \ t > j\delta, \ \delta \in (0,1).$$
(8)

Ideas for the proof

- parametrix method (Konakov-Mammen, PTRF,2000)+C-K equation;
- Λ_3 and κ_0 are explicit.

Under (A1)-(A3), for any T > 0, there exists a constant C > 0 such that

$$\int_0^T \mathbb{E}|b(X_t^{(\delta)}) - b(X_{t_{\delta}}^{(\delta)})|^2 \mathrm{d}t \le C\delta^{1\wedge\frac{\alpha}{2}}.$$
(9)

- It is easy to obtain (9) whenever b is Hölder continuity;
- The proof of (9) is based on Heat kernel estimate for numerical scheme;
- (9) is used to reveal the convergence rate of EM scheme.

Assume (A1) and (A3). Then, for $f \in L^p_q(T)$ with $(p,q) \in \mathscr{K}_1$,

$$\mathbb{E}\left(\int_{s}^{t} |f_{r}(\boldsymbol{X}_{r}^{(\boldsymbol{\delta})})| \mathrm{d}r \Big| \mathscr{F}_{s}\right) \leq \alpha_{0} \|f\|_{L^{p}_{q}(T)}(t-s)^{\frac{1}{\gamma_{0}}}, \quad 0 \leq s \leq t \leq T,$$
(10)

where $\gamma_0 := \frac{1}{1 - 1/q - d/2p}$.

• α_0 is explicit;

• The proof of (10) is due to heat kernel estimate of numerical scheme.

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Khasminskill's Estimate for Numerical Scheme

Lemma

Assume (A1) and (A3). Then, for $f \in L^p_q(T)$ with $(p,q) \in \mathscr{K}_1$, the following Khasminskill type estimate

$$\mathbb{E}\exp\left(\lambda\int_{0}^{T}|f_{t}(\boldsymbol{X}_{t}^{(\boldsymbol{\delta})})|\mathrm{d}t\right) \leq 2^{1+T(2\lambda\alpha_{0}\|f\|_{L^{p}_{q}(T)})^{\gamma_{0}}}, \quad \lambda > 0$$
(11)

holds, where $\gamma_0 := rac{1}{1-1/q-d/2p}$ and

$$\alpha_0 := \frac{(1-1/p)^{\frac{d}{2}(1-1/p)}}{(\check{\lambda}_0(2\pi)^{\frac{1}{p}})^{\frac{d}{2}}} \Big\{ \widehat{\lambda}_0^{\frac{d}{2}(1-1/p)} + \Lambda_3(\gamma_0(1-1/q))^{\frac{q-1}{q}}(\kappa_0/2)^{\frac{d}{2}(1-1/p)} \Big\}.$$

Remark: Follow the line of Xie & Zhang (2017).

- The Krylov estimate associated with $(X_{t_{\delta}}^{(\delta)})_{t\geq 0}$ no longer holds true.
- Take $s,t\in [k\delta,(k+1)\delta)$ for some integer $k\geq 1.$ Then,

$$\mathbb{E}\left(\int_{s}^{t} |f_{r_{\delta}}(X_{r_{\delta}}^{(\delta)})| \mathrm{d}r \Big| \mathscr{F}_{s}\right) = |f_{k\delta}(X_{k\delta}^{(\delta)})|(t-s)$$
(12)

which is a random variable. Hence, it is impossible to control the quantity on the left hand side of (12) by $||f||_{L^p_a(T)}$ up to a constant.

Regularity of Kolmogrov Backward Equation

For any $\lambda > 0$, consider the following PDE for $u^{\lambda} : [0,T] \times \mathbb{R}^d \to \mathbb{R}^d$:

$$\partial_t u^{\lambda} + \frac{1}{2} \sum_{i,j=1}^d \langle \sigma \sigma^* e_i, e_j \rangle \nabla_{e_i} \nabla_{e_j} u^{\lambda} + b + \nabla_b u^{\lambda} = \lambda u^{\lambda}.$$
(13)

• (13) has a unique solution $u^{\lambda} \in \mathscr{H}_{2p}^{2,2q}(0,T)$ satisfying $(1 \lor \lambda)^{1-\frac{d}{4p}-\frac{1}{2q}} \|\nabla u^{\lambda}\|_{T,\infty} + \|\nabla^2 u^{\lambda}\|_{L^{2p}_{2q}(T)} \le C_1 \||b|^2\|_{L^p}$ (14)

for some constant $C_1 > 0$; see e.g. Xie-Zhang (2017).

• There is a constant $\lambda_0 \ge 1$ such that

$$\|\nabla u^{\lambda}\|_{T,\infty} \le \frac{1}{2}, \quad \lambda \ge \lambda_0.$$
(15)

• We can apply Itô's formula to u by a standard approximation argument.

Key Points for the Proof of Main Result I

• For $\theta_t^\lambda(x) := x + u_t^\lambda(x), x \in \mathbb{R}^d$, and $Z_t^{(\delta)} := X_t - X_t^{(\delta)}$, by Itô's formula,

$$d\theta_t^{\lambda}(X_t) = \lambda u^{\lambda}(X_t) dt + \nabla \theta_t^{\lambda}(X_t) \sigma(X_t) dW_t$$

$$d\theta_t^{\lambda}(X_t^{(\delta)}) = \left\{ \lambda u^{\lambda}(X_t^{(\delta)}) + \nabla \theta_t^{\lambda}(X_t^{(\delta)}) (b(X_{t\delta}^{(\delta)}) - b(X_t^{(\delta)})) + \frac{1}{2} \sum_{i,j=1}^d \langle ((\sigma\sigma^*)(X_{t\delta}^{(\delta)}) - (\sigma\sigma^*)(X_t^{(\delta)})) e_i, e_j \rangle \nabla_{e_i} \nabla_{e_j} u_t^{\lambda}(X_t^{\delta})) \right\} dt$$

$$+ \nabla \theta_t^{\lambda}(X_t^{(\delta)}) \sigma(X_{t\delta}^{(\delta)}) dW_t.$$

Key Point: Choose the same θ_t^{λ} .

Key Points for the Proof of Main Result I (Cont.)

• Again, by Itô's formula,

$$\begin{split} |Z_t^{(\delta)}|^2 &\leq 8\lambda \int_0^t \langle \theta_s^{\lambda}(X_s) - \theta_s^{\lambda}(X_s^{(\delta)}), u^{\lambda}(X_s) - u^{\lambda}(X_s^{(\delta)}) \rangle \mathrm{d}s \\ &+ 8 \int_0^t \langle \theta_s^{\lambda}(X_s) - \theta_s^{\lambda}(X_s^{(\delta)}), \nabla \theta_s^{\lambda}(X_s^{(\delta)})(b(X_s^{(\delta)}) - b(X_{s_{\delta}}^{(\delta)})) \rangle \mathrm{d}s \\ &+ \sum_{i,j=1}^d \int_0^t \langle ((\sigma\sigma^*)(X_{s_{\delta}}^{(\delta)}) - (\sigma\sigma^*)(X_s^{(\delta)}))e_i, e_j \rangle \\ &\times \langle \theta_s^{\lambda}(X_s) - \theta_s^{\lambda}(X_s^{(\delta)}), \nabla_{e_i} \nabla_{e_j} u_s^{\lambda}(X_s^{\delta})) \rangle \mathrm{d}s \\ &+ 4 \int_0^t \| \nabla \theta_s^{\lambda}(X_s) \sigma(X_s) - \nabla \theta_s^{\lambda}(X_s^{(\delta)}) \sigma(X_{s_{\delta}}^{(\delta)}) \|_{\mathrm{HS}}^2 \mathrm{d}s + M_t \\ =: I_{1,\delta}(t) + I_{2,\delta}(t) + I_{3,\delta}(t) + I_{4,\delta}(t) + M_t. \end{split}$$

Key Points for the Proof of Main Result I (Cont.)

- Handle $I_{1,\delta}(t)$ via (15);
- Deal with $I_{2,\delta}(t)$ by (15) & Lemma 4;
- Cope with $I_{3,\delta}(t)$ by (A3) & (15).
- Estimate $I_{4,\delta}(t)$ by the Hardy-Littlewood maximum theorem;
- Obtain the estimate

$$\left(\mathbb{E}\Big(\sup_{0\leq s\leq t}|Z_t^{(\delta)}|^{2\kappa'}\Big)\right)^{1/\kappa'}, \quad \kappa'\in(0,1)$$

by stochastic Gronwall inequality (Xie-Zhang, 2017) and Khasminskill's estimate.

Main Result II

- In Theorem 8, the integrable condition (i.e., |b|² ∈ L^p) seems to be a little bit restrictive;
- It rules out some typical examples, e.g., $b(x) = \mathbf{1}_{[0,\infty)}(x)$;
- By implementing a truncation argument, the integrable condition can indeed be dropped.

Theorem

Assume (A1)-(A3) without $|b|^2 \in L^p$. Then, for $\beta \in (0,2)$, and p,q > 2 with $\frac{d}{p} + \frac{1}{q} < 1$, there exist constants $C_1, C_2 > 0$ such that

$$\mathbb{E}\Big(\sup_{0\leq t\leq T}|X_t - X_t^{(\delta)}|^\beta\Big) \leq C_1\Big\{\mathrm{e}^{C_2(-\frac{\beta}{2}(1\wedge\frac{\alpha}{2})\log\delta)^{\frac{d\gamma_0}{2p}}} + 1\Big\}\delta^{\frac{\beta}{2}(1\wedge\frac{\alpha}{2})}.$$
 (16)

Key Points for the Proof of Main Result II

- Let $\psi : \mathbb{R}_+ \to [0,1]$ be a smooth function such that $\psi(r) = 1, r \in [0,1]$, and $\psi(r) \equiv 0, r \geq 2$.
- For each integer $k \ge 1$, let $b_k(x) = b(x)\psi(|x|/k)$, $x \in \mathbb{R}^d$, which satisfies

$$\|b_k\|_{\infty} \le \|b\|_{\infty}$$
 and $\||b_k|^2\|_{L^p} \le \left(\frac{2^d \pi^{\frac{d}{2}}}{\Gamma(\frac{d}{2}+1)}\right)^{1/p} k^{\frac{d}{p}} \|b\|_{\infty}^2$. (17)

• Consider the following truncated SDE corresponding to (1)

$$dX_{t}^{k} = b_{k}(X_{t}^{k})dt + \sigma(X_{t}^{k})dW_{t}, \quad t \ge 0, \quad X_{0}^{k} = X_{0}.$$
(18)

Key Points for the Proof of Main Result II (Cont.)

• For $q \in (0,2)$, observe that

$$\begin{split} \mathbb{E} \| X - X^{(\delta)} \|_{T,\infty}^q &\leq 3^{0 \vee (q-1)} \{ \mathbb{E} \| X - X^k \|_{T,\infty}^q + \mathbb{E} \| X^{(\delta)} - X^{k,(\delta)} \|_{T,\infty}^q \\ &+ \mathbb{E} \| X_t^k - X^{k,(\delta)} \|_{T,\infty}^q \} =: 3^{0 \vee (q-1)} \{ I_1 + I_2 + I_3 \} \end{split}$$

• According to martingale inequality (Shigekawa, 2004),

$$I_1, I_2 \le C_2 \exp\left(\frac{(|x| + \|b\|_{\infty}T)^2}{8d^2\hat{\lambda}_0 T}\right) e^{-\frac{k^2}{16d^2\hat{\lambda}_0 T}}$$
(19)

• By virtue of Main Result I,

$$I_{3} \le C_{4} \mathrm{e}^{C_{6} \|b\|_{\infty}^{2\gamma_{0}} k^{\frac{d\gamma_{0}}{p}}} \delta^{\frac{q}{2}(1 \wedge \frac{\alpha}{2})}$$
(20)

• The desired assertion is available by taking

$$\mathbf{k} = \left(-8qd^2\widehat{\lambda}_0 T\left(1 \wedge \frac{\alpha}{2}\right)\log\delta\right)^{\frac{1}{2}}.$$

Summaries

Theorem

Assume (A1)-(A3). Then, for $\beta \in (0,2)$ and q > 2, there exist constants $C_1, C_2 > 0$ such that, for $\gamma_0 := \frac{1}{1-1/q-d/2p}$,

$$\mathbb{E}\Big(\sup_{0 \le t \le T} |X_t - X_t^{(\delta)}|^{\beta}\Big) \le C_1 \mathrm{e}^{C_2(1+\||b|^2\|_{L^p}^{\gamma_0})} \delta^{\frac{\beta}{2}(1\wedge\frac{\alpha}{2})}.$$
 (21)

Theorem

Assume (A1)-(A3) without $|b|^2 \in L^p$. Then, for $\beta \in (0,2)$, and p,q > 2with $\frac{d}{p} + \frac{1}{q} < 1$, there exist constants $C_1, C_2 > 0$ such that

$$\mathbb{E}\Big(\sup_{0\leq t\leq T}|X_t - X_t^{(\delta)}|^\beta\Big) \leq C_1\Big\{\mathrm{e}^{C_2(-\frac{\beta}{2}(1\wedge\frac{\alpha}{2})\log\delta)^{\frac{d\gamma_0}{2p}}} + 1\Big\}\delta^{\frac{\beta}{2}(1\wedge\frac{\alpha}{2})}.$$
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Thanks A Lot !

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