

Wasserstein-1 distance between SDEs driven by Brownian motion and stable process

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Based on a joint work with R. Schilling (Dresden) and L. Xu (Macau):
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1. Background and motivation

2. Main result

3. Sketch of proof

Background

Consider

$$dX_t = b(X_t) dt + dL_t, \quad X_0 = x,$$

$$dY_t = b(Y_t) dt + dB_t, \quad Y_0 = x,$$

where L_t is an α -stable process and B_t is a B.M. in \mathbb{R}^d .

- $\mathbb{E} e^{i\epsilon L_t} = e^{-t|\epsilon|^\alpha} \xrightarrow{\alpha \uparrow 2} e^{-t|\epsilon|^2} = \mathbb{E} e^{i\epsilon B_t}$
- $L_t \xrightarrow{\alpha \uparrow 2} B_t$
- Natural question: $X_t \xrightarrow{\alpha \uparrow 2} Y_t?$
Liu (JMAA 2022, SPA 2022, JDE 2022)
- Further (informal) question: $X_\infty \xrightarrow{\alpha \uparrow 2} Y_\infty?$

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- Ergodicity: $\mathcal{L}_{X_t} \rightarrow \mu_\alpha$ and $\mathcal{L}_{Y_t} \rightarrow \mu_2$, as $t \rightarrow \infty$.
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Main result

$$dX_t = b(X_t) dt + dL_t, \quad X_0 = x,$$

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Assumption: $\|\nabla b\|_\infty < \infty$, $\|\nabla^2 b\|_\infty < \infty$, $\|\nabla^3 b\|_\infty < \infty$, and

$$\limsup_{|x-y| \rightarrow \infty} \frac{\langle x - y, b(x) - b(y) \rangle}{|x - y|^2} < 0.$$

Typical Exam.: $b(x) = -x +$ 'small perturbation'.

Theorem (D.-Schilling-Xu, 2023+)

For any $x, y \in \mathbb{R}^d$ and $t > 0$,

$$W_1(\mathcal{L}_{X_t^x}, \mathcal{L}_{Y_t^y}) \leq C_1 e^{-C_2 t} |x - y| + C(2 - \alpha).$$

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Corollary (D.-Schilling-Xu, 2023+)

$$W_1(\mu_\alpha, \mu_2) \leq C(2 - \alpha).$$

Proof:

$$W_1(\mu_\alpha, \mu_2) \leq W_1(\mu_\alpha, \mathcal{L}_{X_t^x}) + W_1(\mathcal{L}_{X_t^x}, \mathcal{L}_{Y_t^y}) + W_1(\mathcal{L}_{Y_t^y}, \mu_2).$$

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Optimal rate: $2 - \alpha$? O-U case

$$dX_t = -X_t dt + dL_t, \quad \mu_\alpha = \mathcal{L}_{\alpha^{-1/\alpha} L_1},$$

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$$\begin{aligned} W_1(\mu_\alpha, \mu_2) &= \inf_{\Pi \in \mathcal{C}(\mu_\alpha, \mu_2)} \iint |x - y| \Pi(dx, dy) \\ &\geq \inf_{\Pi \in \mathcal{C}(\mu_\alpha, \mu_2)} \left| \iint |x| \Pi(dx, dy) - \iint |y| \Pi(dx, dy) \right| \\ &= \left| \int |x| \mu_\alpha(dx) - \int |y| \mu_2(dy) \right| \\ &= \left| \mathbb{E}|\alpha^{-1/\alpha} L_1| - \mathbb{E}|2^{-1/2} B_1| \right| \\ &\asymp (2 - \alpha). \end{aligned}$$

Remark: The rate $2 - \alpha$ is sharp for the O-U case.

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Sketch of proof

$$\begin{aligned}dX_t &= b(X_t) dt + dL_t, & P_t, \mathcal{A}^P, \\dY_t &= b(Y_t) dt + dB_t, & Q_t, \mathcal{A}^Q.\end{aligned}$$

Proof: It suffices to bound $|\mathcal{A}^P - \mathcal{A}^Q|$ since (Duhamel formula)

$$\begin{aligned}W_1(\mathcal{L}_{X_t^x}, \mathcal{L}_{Y_t^y}) &= \sup_{h \in \text{Lip}(1)} |P_t h(x) - Q_t h(x)| \\&= \sup_{h \in \text{Lip}(1)} \left| \int_0^t \frac{d}{ds} Q_{t-s} P_s h(x) ds \right| \\&= \sup_{h \in \text{Lip}(1)} \left| \int_0^t Q_{t-s} (\mathcal{A}^P - \mathcal{A}^Q) P_s h(x) ds \right|.\end{aligned}$$

$$\mathcal{A}^P f = \langle \nabla f, b \rangle + \int_{\mathbb{R}^d \setminus \{0\}} [f(\cdot + z) - f(\cdot) - \langle \nabla f(\cdot), z \rangle \mathbf{1}_{\{|z| \leq 1\}}] \frac{C_{d,\alpha}}{|z|^{d+\alpha}} dz,$$

$$\mathcal{A}^Q f = \langle \nabla f, b \rangle + \frac{1}{2} \Delta f.$$

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In order to bound $|(\mathcal{A}^P - \mathcal{A}^Q)P_s h|$, we need the following gradient estimate.

Lemma

For any $h \in \text{Lip}(1)$ and $t \in (0, 1]$,

$$\|\nabla P_t h\|_\infty \leq C,$$

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Remark: For $h \in \mathcal{B}_b(\mathbb{R}^d)$ and $t \in (0, 1]$,

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See e.g. Zhang, SPA 2013.

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$$\limsup_{|x-y| \rightarrow \infty} \frac{\langle x-y, b(x) - b(y) \rangle}{|x-y|^2} < 0.$$

Our result: $W_1(\mu_\alpha, \mu_2) \leq C(2 - \alpha).$

Question: 1) Other distance $d(\mu_\alpha, \mu_2)$?

2) More general coefficients?

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$$dX_t = b(X_t) dt + dL_t, \quad \mu_\alpha = \mathcal{L}_{X_\infty},$$

$$dY_t = b(Y_t) dt + dB_t, \quad \mu_2 = \mathcal{L}_{Y_\infty}.$$

Assumption: $\|\nabla b\|_\infty < \infty$, $\|\nabla^2 b\|_\infty < \infty$, $\|\nabla^3 b\|_\infty < \infty$, and

$$\limsup_{|x-y| \rightarrow \infty} \frac{\langle x - y, b(x) - b(y) \rangle}{|x - y|^2} < 0.$$

Our result: $W_1(\mu_\alpha, \mu_2) \leq C(2 - \alpha)$.

Question: 1) Other distance $d(\mu_\alpha, \mu_2)$?

2) More general coefficients?

3)

Thanks for Your Attention!