

Regularisation by fractional noise: density of SDEs and McKean-Vlasov equations

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Overview

Introduction

Regularity of laws of SDEs

Gaussian bounds

McKean-Vlasov equations

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Consider the equation

$$\boxed{dX_t = \varphi(t, X_t) dt + dB_t,} \quad (\text{E})$$

where $\varphi(t, \cdot)$ is a distribution in some Besov space and B is a fractional Brownian motion.

We look for solutions of the form

$$X_t = X_0 + K_t + B_t,$$

where in case φ is regular enough, $K_t = \int_0^t \varphi(r, X_r) dr$.

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Typical examples

- ▶ $\varphi = \alpha \delta_0$: corresponds formally to an SDE involving the local time of the solution, see [Le Gall'84] in the Brownian case.
- ▶ $\varphi = \alpha |\cdot|^{-s}$: Bessel-like processes and Riesz-type kernels in mathematical physics (e.g. Coulomb gases, Keller-Segel model, etc.).

Noise is your friend

Without noise, classical theory requires

- ▶ $\varphi \in L_t^1 C_b^1$ for well-posedness;
- ▶ $\varphi \in L_t^1 C_b^0$ for mere existence (Peano).

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$$dX_t = \text{sign}(X_t) \sqrt{|X_t|} dt \quad , \quad X_0 = 0,$$

whose solutions are given, for any $t^* \in \mathbb{R}_+$, by

$$(X_t^{t^*})_{t \in \mathbb{R}_+} := t \mapsto (t - t^*)_+^2.$$

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Now add **noise** to the equation. Due to the forcing, solution leaves 0 immediately. But away from 0, Lipschitz drift \implies uniqueness. For almost each trajectory of $(B_t)_{t \geq 0}$, we have a unique solution.

How is the noise helping?

Heuristics – In situations where the ODE $\dot{x}_t = \varphi(x_t)$ lacks uniqueness, adding noise might restore uniqueness \leadsto **regularisation by noise**.

Consider $\tilde{X} = X - B$ which now solves the random ODE:

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In $X = \tilde{X} + B$, \tilde{X} gives slow oscillations and B fast oscillations. Freezing \tilde{X} , consider

$$x \mapsto \int_0^t \varphi(x + B_r) \, dr$$

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In fact, for L the local time of B ,

$$\int_0^t \varphi(x + B_r) \, dr = \int_{\mathbb{R}} \varphi(x + y) L_t(y) \, dy = \varphi * \check{L}_t(x).$$

$\implies \varphi * \check{L}_t$ is more regular than φ !

Rougher noise, smoother local time

For a **Hurst** parameter $H \in (0, 1) \setminus \{\frac{1}{2}\}$, fractional Brownian motion (fBm) is given by:

$$B_t = c_H \int_{\mathbb{R}} \left((t-s)_+^{H-1/2} - (-s)_+^{H-1/2} \right) dW_s, \quad t \in \mathbb{R}.$$

Introduced in the 40's by Kolmogorov as a toy model for turbulence. Since then, many applications in hydrology, telecommunications, physics, finance, ...

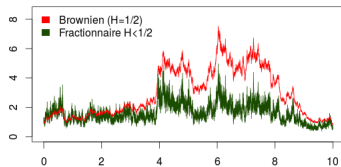
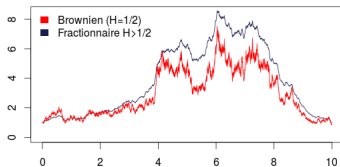
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► Trajectories:



► Gaussian process with memory:

- $H > \frac{1}{2}$: more regular than Bm, long-range dependence.
- Rough regime $H < \frac{1}{2}$: negatively correlated increments, strong oscillations.

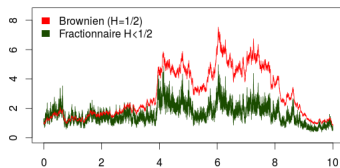
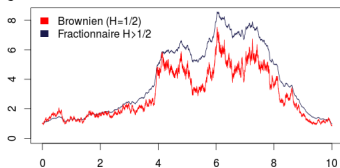
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- $H > \frac{1}{2}$: more regular than Bm, long-range dependence.
- Rough regime $H < \frac{1}{2}$: negatively correlated increments, strong oscillations.

► **Local time**: $x \mapsto L_t(x)$ has regularity $\frac{1}{2H} - \frac{1}{2} - \varepsilon$ a.s.

Rule of thumb: rougher noise, better regularisation!

A few results - Brownian case

- Works of Zvonkin, Veretennikov, [Krylov & Röckner'05]: Strong WP for $\varphi(t, x) \in L^q([0, T]; L^p(\mathbb{R}^d))$

$$\text{if } p \geq 2, q > 2, \frac{2}{q} + \frac{d}{p} < 1.$$

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- Hölder setting: $\varphi \in \mathcal{C}^\gamma$,
 - [Bass & Chen'01] proved strong WP for $\gamma > -\frac{1}{2}$, $d = 1$, counter-examples for $\gamma < -\frac{1}{2}$.
 - Weak WP for $\gamma > -\frac{2}{3}$, $d = 1$ [Delarue & Diel'16];
weak WP for $\gamma > -\frac{1}{2}$, $d \geq 1$ [Flandoli, Issoglio & Russo'17];
Canizzaro-Chouk, Coutin-Duboscq-Réveillac, etc.
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These results rely crucially on the **Markov property** of the BM, and subsequently on **PDE techniques** (martingale problem and/or Zvonkin transform).

A few results - fBm case

But fBm is neither Markov, nor a semimartingale.

- Early work by [Nualart & Ouknine'02]. Then [Catellier & Gubinelli'16] used *nonlinear Young integration* to prove that there is a unique solution if

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- ▶ Recently, thanks to the **Stochastic Sewing Lemma** of [Lê'20],
 - For $\varphi \in \mathcal{B}_p^\gamma(\mathbb{R}^d)$, $p < \infty$, $\gamma - \frac{d}{p} = 1 - \frac{1}{2H}$, strong WP of the fBm-driven SDE [Anzeletti, R. & Tanré'23];
 - Weak well-posedness: weak existence in [Anzeletti, R. & Tanré'23] for $\gamma > \frac{1}{2} - \frac{1}{2H}$, uniqueness in law in [Butkovsky & Mytnik '24].

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Theorem ([Galeati & Gerencsér'24] – *Time-dependent drift*)

Strong WP holds for (E) when $\varphi \in L^q([0, T]; \mathcal{C}^\gamma(\mathbb{R}^d))$ with

$$\gamma > 1 - \frac{1}{H(q' \vee 2)} \text{ and } q \in (1, \infty].$$

McKean-Vlasov equations

As for “linear” SDEs, it is possible to exploit the regularising effect of the noise for McKean-Vlasov SDEs. Consider specifically convolution-type equations

$$\begin{cases} dY_t = \psi_t * \mu_t(Y_t) dt + dB_t \\ \mu_t = \text{Law}(Y_t). \end{cases} \quad (\text{McKV})$$

This eq. arises formally as the limit of interacting particle systems.

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Theorem ([Galeati, Harang & Mayorcas'23], [Galeati & Gerencsér'24])

Strong WP holds for (McK-V) when $\psi \in L^q C^\gamma$ with $\gamma > 1 - \frac{1}{H(q'\sqrt{2})}$ and $q \in (1, \infty]$.

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Remark: a similar statement holds for more general drift $\Psi(t, x, \mu)$.

Objectives:

- Obtain the regularity of the law of a *linear* SDE;
- Exploit this regularity for (McK-V) to go below the $1 - \frac{1}{H(q'\sqrt{2})}$ threshold.

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- For $\varphi \in L^\infty([0, T]; \mathcal{C}^\gamma(\mathbb{R}^d))$ and $H > \frac{1}{2}$, [Olivera & Tudor'19] : X_t has a density with some Besov regularity.
- For $\varphi \in L^\infty([0, T]; \mathcal{C}^\gamma(\mathbb{R}^d))$, [Galeati, Harang & Mayorcas'23] show that $\mathcal{L}(X_\cdot) \in L^{\tilde{q}}([0, T]; \mathcal{B}_1^\alpha)$ for $\alpha < \frac{1}{H}(\frac{1}{\tilde{q}} - \frac{1}{2})$.

Besov regularity

$$\mathcal{B}_1^\alpha(\mathbb{R}^d) = \{f \in \mathcal{S}'(\mathbb{R}^d) : \|f\|_{\mathcal{B}_1^\alpha} < \infty\},$$

where $\|\cdot\|_{\mathcal{B}_1^\alpha}$ has the equivalent thermic representation:

$$\|\mathcal{F}^{-1}(\phi \mathcal{F}f)\|_{L^1(\mathbb{R}^d)} + \sup_{s \in (0,1]} s^{\frac{n-\alpha}{2}} \|\partial_s^n g_s * f\|_{L^1(\mathbb{R}^d)},$$

for any $n \geq \alpha$, $n \in \mathbb{N}$.

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for any $n \geq \alpha$, $n \in \mathbb{N}$.

For instance, one gets for the fBm B of Hurst parameter $H \in (0, 1)$ that

$$\|\mathcal{L}(B_t)\|_{\mathcal{B}_1^\alpha} = \|g_{t^{2H}}\|_{\mathcal{B}_1^\alpha} \lesssim \frac{1}{1 \wedge t^{\alpha H}}, \quad \forall t > 0.$$

In particular, $\mathcal{L}(B_\cdot) \in L^{\tilde{q}}([0, T]; \mathcal{B}_1^\alpha)$ when $\alpha < \frac{1}{H\tilde{q}}$.

$$X_t = X_0 + \int_0^t \varphi(s, X_s) \, ds + B_t, \quad t \in [0, T]. \quad (\text{E})$$

Definition

► *Solution:*

- $(\varphi^n)_{n \in \mathbb{N}}$ in $L^q([0, T]; \mathcal{C}_b^\infty)$, $\varphi^n \rightarrow \varphi$ in $L^q([0, T]; \mathcal{C}^{\gamma-})$.
- $\forall n \in \mathbb{N}$, denote X^n the solution of (E) with drift φ^n .
- If $(X^n)_{n \in \mathbb{N}}$ converges in $L^2(\Omega; \mathcal{C}_{[0, T]})$, call the limit $(X_t)_{t \in [0, T]}$ a *solution* to (E).

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Assumption: $\varphi \in L^q([0, T]; \mathcal{C}^\gamma(\mathbb{R}^d))$ with

$$H \in (0, +\infty) \setminus \mathbb{N}, \quad \gamma > 1 - \frac{1}{H(q' \vee 2)} \quad \text{and} \quad q \in (1, +\infty]. \quad (\text{A})$$

Time-space regularity of the density

Theorem (Anzeletti, Galeati, R. & Tanré '25)

Under (A), let X be the solution to (E). Let $\tilde{q} \in [1, \infty)$ and

$$0 \leq \alpha < \min \left\{ \frac{1}{H\tilde{q}}, \gamma - 1 + \frac{1}{H} \right\}.$$

Then for any $0 \leq s < t \leq T$, ($\gamma \leq 0$.)

$$\|\mathcal{L}(X.)\|_{L_{[s,t]}^{\tilde{q}} \mathcal{B}_1^\alpha} \lesssim (t-s)^{\frac{1}{\tilde{q}} - \alpha H} + (\|\varphi\|_{L_{[s,t]}^q} c_\gamma + \|\varphi\|_{L_{[s,t]}^q}^{1+\eta} c_\gamma) (t-s)^\varepsilon,$$

where

$$\varepsilon = \frac{1}{q'} + \frac{1}{\tilde{q}} - H(\alpha + 1) + \min \left(-\frac{\eta}{q}, \gamma H \right) > 0$$

and

$$\eta = \frac{-\gamma H}{1 + H\gamma - H} \in (0, 1).$$

Theorem (More general version)

Under (A), let X be the solution to (E) starting from an \mathcal{F}_0 -measurable random variable X_0 .

(a) For

$$0 < \alpha < \gamma - 1 + \frac{1}{Hq'},$$

then for any $0 \leq u < t \leq T$, the conditional law $\mathcal{L}(X_t \mid \mathcal{F}_u)$ has a density which satisfies

$$\left\| \|\mathcal{L}(X_t \mid \mathcal{F}_u)\|_{\mathcal{B}_1^\alpha} \right\|_{L^\infty_\Omega} \leq C(1 + (t - u)^{-\alpha H}).$$

(b) Let (\tilde{q}, α) satisfying

$$\tilde{q} \in (1, +\infty), \quad 0 < \alpha < \min \left\{ \frac{1}{H\tilde{q}}, \gamma - 1 + \frac{1}{H} \right\},$$

then for any $u \in [0, T)$, $t \mapsto \mathcal{L}(X_t \mid \mathcal{F}_u)$ belongs a.s. to $L^{\tilde{q}}([u, T]; \mathcal{B}_1^\alpha)$ and satisfies

$$\left\| \|\mathcal{L}(X. \mid \mathcal{F}_u)\|_{L^{\tilde{q}}([u, T]; \mathcal{B}_1^\alpha)} \right\|_{L^\infty_\Omega} \leq C(T - u)^{\frac{1}{\tilde{q}} - \alpha H}.$$

- For $q = \tilde{q} = 2$, the condition on γ is $\gamma > 1 - \frac{1}{2H}$ and the density estimate becomes

$$\|\mathcal{L}(X.)\|_{L^2_{[s,t]}\mathcal{B}_1^\alpha} \lesssim (t-s)^\varepsilon,$$

for any $\alpha < \frac{1}{2H}$.

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- Similarly for $\tilde{q} = 1+$, $\mathcal{L}(X.) \in L^{1+}_{[s,t]}\mathcal{B}_1^\alpha$ for $\alpha < \frac{1}{H}$.

Scheme of proof - 1

Fix $s < t$. By a duality argument,

$$\begin{aligned} \|\mathcal{L}(X.)\|_{L^{\tilde{q}}_{[s,t]} \mathcal{B}_1^\alpha} &\lesssim \sup_{\substack{f \in L^{\tilde{q}'}_{[s,t]} \mathcal{C}^{-\alpha}, \|f\| \leq 1 \\ f \text{ smooth}}} \left| \int_s^t \langle f_r, \mathcal{L}(X_r) \rangle dr \right| \\ &\lesssim \sup_{\substack{f \in L^{\tilde{q}'}_{[s,t]} \mathcal{C}^{-\alpha}, \|f\| \leq 1 \\ f \text{ smooth}}} \left| \mathbb{E} \int_s^t f_r(X_r) dr \right|. \end{aligned}$$

The above expectation of $\int_s^t f_r(X_r) dr$ can now be studied *via sewing techniques*.

Scheme of proof - 2

Lemma

Assume (A), $\gamma < 0$. Let $\tilde{q} \in [1, \infty)$ and

$$0 \leq \alpha < \min \left\{ \frac{1}{H\tilde{q}}, \gamma - 1 + \frac{1}{H} \right\}.$$

For any $f \in L^{\tilde{q}'}([0, T]; \mathcal{C}_b^\infty(\mathbb{R}^d))$ and any $0 \leq s < t \leq T$,

$$\begin{aligned} \left| \mathbb{E} \int_s^t f_r(X_r) dr \right| &\lesssim \|f\|_{L_{[s,t]}^{\tilde{q}'}} \mathcal{C}^{-\alpha} \\ &\times \left((t-s)^{\frac{1}{\tilde{q}} - \alpha H} + (\|\varphi\|_{L_{[s,t]}^q} \mathcal{C}^\gamma + \|\varphi\|_{L_{[s,t]}^{q,1+\eta}} \mathcal{C}^\gamma)(t-s)^\varepsilon \right). \end{aligned}$$

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Gaussian bounds for the density of X

Recent results giving Gaussian bounds on the density of SDEs:

- ▶ [Besalú et al.'16], [Baudoin et al.'16] : Rough differential equations driven by fBm, smooth vector fields, using **Malliavin calculus**;
- ▶ [Li, Panloup & Sieber'23]: Differential equations with additive fractional noise, irregular drift **function** in the Catellier-Gubinelli regime, i.e. imposes restrictions when $H < 1/2$;
- ▶ [Perkowski & van Zuijlen'23]: upper and lower bound on the density of SDEs, distributional drift with reg. $> -\frac{1}{2}$.

Gaussian bounds

$$dX_t = \varphi(t, X_t)dt + dB_t. \quad (\text{E})$$

For $H \geq 1/2$, $\gamma > 1 - 1/(2H)$ and $\varphi \in L^\infty([0, T]; \mathcal{C}^\gamma(\mathbb{R}^d))$,
[Li, Panloup & Sieber'23] proved upper and lower Gaussian bounds.

Theorem

Let $H \leq 1/2$, $\gamma > 1 - 1/(2H)$ and $\varphi \in L^\infty([0, T]; \mathcal{C}^\gamma(\mathbb{R}^d))$.

Then the solution to (E) has a density for any $t \in (0, T]$ and $\exists C > 0$ s.t.
 $\forall t \in (0, T], \forall x \in \mathbb{R}^d$,

$$\frac{C^{-1}}{t^{dH}} \exp\left(-C \frac{|x - x_0|^2}{t^{2H}}\right) \leq \frac{d\mathcal{L}(X_t)}{dx}(x) \leq \frac{C}{t^{dH}} \exp\left(-C^{-1} \frac{|x - x_0|^2}{t^{2H}}\right).$$

Girsanov's formula for fBm gives for the density of X_1 :

$$\frac{d\mathcal{L}(X_1)}{dy}(y) = (2\pi)^{-dH} e^{-\frac{|y|^2}{2}} \Psi(y),$$

where

$$\Psi(y) = \mathbb{E} \left[\exp \left(\int_0^1 (K_H^{-1} Z)_s \cdot dW_s - \frac{1}{2} \int_0^1 |(K_H^{-1} Z)_s|^2 ds \right) \mid B_1 = y \right],$$

K_H is a nonlocal operator from the definition of fBm, and $Z_\cdot = \int_0^\cdot \varphi(s, B_s) ds$.

Girsanov's formula for fBm gives for the density of X_1 :

$$\frac{d\mathcal{L}(X_1)}{dy}(y) = (2\pi)^{-dH} e^{-\frac{|y|^2}{2}} \Psi(y),$$

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- ▶ Study above exponential functionals using regularising effects (SSL+LND) of those Gaussian bridges (recall that φ can still be distributional!).

Overview

Introduction

Regularity of laws of SDEs

Gaussian bounds

McKean-Vlasov equations

Consider convolution-type McKean-Vlasov SDEs:

$$\begin{cases} dY_t = \psi_t * \mu_t(Y_t) dt + dB_t \\ \mu_t = \text{Law}(Y_t), \quad t \geq 0. \end{cases} \quad (\text{McK-V})$$

Arises formally as the limit of interacting particle systems, as $N \rightarrow +\infty$:

$$\begin{cases} dY_t^{i,N} = \frac{1}{N} \sum_{j=1}^N \psi_t(Y_t^{i,N} - Y_t^{j,N}) dt + dB_t^i, & i \in \{1, \dots, N\} \\ B^1, \dots, B^N \text{ independent fBm.} \end{cases}$$

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Existence

$$\begin{cases} dY_t = \psi_t * \mu_t(Y_t) dt + dB_t \\ \mu_t = \text{Law}(Y_t), \quad t \geq 0. \end{cases} \quad (\text{McK-V})$$

Theorem

Let $\psi \in L^\infty([0, T]; \mathcal{C}^\theta)$ with

$$\theta > 1 - \frac{1}{H}.$$

There exists Y and a family $(\mu_t)_{t \in [0, T]}$ solution of (McK-V), i.e.:

- ▶ for $\rho \in [1, \infty)$ and $\alpha < \frac{1}{\rho H}$, $\mu \in L^\rho([0, T]; \mathcal{B}_1^\alpha)$;
- ▶ Y is the unique strong solution of the (linear) SDE with drift $\psi * \mu \in L^\rho([0, T]; \mathcal{C}_b^1)$;
- ▶ For any $t \geq 0$, μ_t is the law of Y_t .

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 \leadsto In particular in $d = 2$, $s = 1$ corresponds to Coulombian interaction. In case $H = \frac{1}{2}$ and the kernel is attractive \equiv Keller-Segel model, which is known to have blow-ups in certain regimes.
- ▶ A heuristic scaling argument permits to retrieve the condition $\theta > 1 - \frac{1}{H}$.

Uniqueness

Theorem

Let $H \in (0, +\infty) \setminus \mathbb{N}$ and $\psi \in L^\infty([0, T]; \mathcal{B}_p^\theta)$ for some $\theta \in (-\infty, 1)$, $p \in [1, \infty]$ satisfying

$$\theta > 1 - \frac{1}{2H}, \quad \theta - \frac{d}{p} > 1 - \frac{1}{H}.$$

Further assume that $\mathcal{L}(Y_0) \in L^\infty(\mathbb{R}^d)$. Then *pathwise uniqueness and uniqueness in law hold* for (McK-V), in the class of solutions such that $\psi * \mu \in L^1([0, T]; \mathcal{C}_b^1)$.

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- The condition $\theta - \frac{d}{p} > 1 - \frac{1}{H}$ still permits to reach a subcritical regime, up to working in Besov spaces with $p < \infty$.

Sketch of proof (existence)

- Consider the smooth approximations

$$\begin{cases} Y_t^n = Y_0 + \int_0^t \psi_s^n * \mu_s^n(Y_s^n) ds + B_t \\ \mu_t^n = \mathcal{L}(Y_t^n), \quad t \geq 0, \end{cases}$$

which have a pathwise unique, strong solution for any $n \in \mathbb{N}$.

- Apply the density Theorem with $q = \tilde{q} = 2$,
 $\gamma = \alpha + \theta \approx 1 - 1/(2H)$ which gives us the condition $\alpha < 1/(2H)$:

$$\begin{aligned} \|\mu^n\|_{L^2_{[s,t]} \mathcal{B}_1^\alpha} &\lesssim (t-s)^\varepsilon + (t-s)^\varepsilon \|\psi^n * \mu^n\|_{L^2_{[s,t]}}^{1+\eta} \mathcal{C}^{\theta+\alpha} \\ &\lesssim (t-s)^\varepsilon \left(1 + \|\psi_n\|_{L^\infty_{[0,t]}}^{1+\eta} \mathcal{C}^\theta \|\mu^n\|_{L^2_{[s,t]} \mathcal{B}_1^\alpha}^{1+\eta} \right) \end{aligned}$$

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- $\eta \leq 1$, with an argument borrowed from rough paths, then for $(t-s)$ small enough, $\|\mu^n\|_{L^2_{[s,t]} \mathcal{B}_1^\alpha} \leq C(t-s)^\varepsilon$.
- Proceed with Kolmogorov's tightness criterion for $(Y_n)_{n \in \mathbb{N}}$.
- Identify the limit points as solutions of the McKean-Vlasov equation.

Thank you!

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Time-dependent drift – What scaling γ ?

$$dX_t = \varphi(t, X_t)dt + dB_t. \quad (\text{E})$$

Following the scaling argument of [Galeati & Gerencsér'24], consider

$B_t^{(\lambda)} = \lambda^{-H} B_{\lambda t}$ and $\varphi^{(\lambda)}(t, x) = \lambda^{1-H} \varphi(\lambda t, \lambda^H x)$. Then

$X_t^{(\lambda)} = \lambda^{-H} X_{\lambda t}$ solves

$$dX_t^{(\lambda)} = \varphi^{(\lambda)}(t, X_t^{(\lambda)}) dt + dB_t^{(\lambda)}.$$

Now observe that

$$\|\varphi^{(\lambda)}\|_{L^{q'} C^\gamma} = \lambda^{1-H-\frac{1}{q'}+\gamma H} \|\varphi\|_{L^{q'} C^\gamma}.$$

As $\lambda \rightarrow 0$, we want to keep $\lambda^{1-H-\frac{1}{q'}+\gamma H}$ bounded, so heuristically,

$$\gamma > 1 - \frac{1}{Hq'}.$$

Theorem ([Galeati & Gerencsér'24])

Strong WP holds for (E) when $\varphi \in L^{q'} C^\gamma$ with $\gamma > 1 - \frac{1}{Hq'}$ and $q' \geq 2$.

Elements of proof 1/3

Lemma

Assume (A), $\gamma < 0$. Let $\tilde{q} \in [1, \infty)$ and

$$0 \leq \alpha < \min \left\{ \frac{1}{H\tilde{q}}, \gamma - 1 + \frac{1}{H} \right\}.$$

For any $f \in L^{\tilde{q}'}([0, T]; \mathcal{C}_b^\infty(\mathbb{R}^d))$ and any $0 \leq s < t \leq T$,

$$\begin{aligned} \left| \mathbb{E} \int_s^t f_r(X_r) dr \right| &\lesssim \|f\|_{L_{[s,t]}^{\tilde{q}'}} \mathcal{C}^{-\alpha} \\ &\times \left((t-s)^{\frac{1}{\tilde{q}} - \alpha H} + (\|\varphi\|_{L_{[s,t]}^q} \mathcal{C}^\gamma + \|\varphi\|_{L_{[s,t]}^{q,1+\eta}} \mathcal{C}^\gamma)(t-s)^\varepsilon \right). \end{aligned}$$

Elements of proof 2/3

Sketch of proof of the Lemma:

We introduce a Sewing Lemma with shifting (deterministic version of [Gerencsér'23]) and control functions.

Let $\tilde{X} = X - B$ and for $u < v \leq T$ with $u - (v - u) \geq 0$,

$$A_{u,v} := \mathbb{E} \int_u^v f_r(B_r + \mathbb{E}^{u-(v-u)} \tilde{X}_r) \, dr$$

Idea: $\mathcal{A}_t = \mathbb{E} \int_s^t f_r(X_r) \, dr \approx \sum A_{u_k, u_{k+1}}$.

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(a) $|A_{u,v}| \lesssim \|f\|_{L_{[u,v]}^{\tilde{q}'}} c^{-\alpha} (v-u)^{\frac{1}{\tilde{q}} - \alpha H}.$

(b) for $\xi = \frac{u+v}{2}$,

$$\begin{aligned} & |A_{u,v} - A_{u,\xi} - A_{\xi,v}| \\ & \lesssim (\|\varphi\|_{L_{[u,v]}^q} c^\gamma + \|\varphi\|_{L_{[u,v]}^q}^{1+\eta} c^\gamma) \|f\|_{L_{[u,v]}^{\tilde{q}'}} c^{-\alpha} (v-u)^{H(\gamma-1 + \frac{1}{Hq'} + \frac{1}{H\tilde{q}} - \alpha)}. \end{aligned}$$

(c) For any $t \in [0, T]$, the convergence in probab. of $\sum_{t_i^n \in \Pi^n} A_{t_i^n, t_{i+1}^n}$ to $\mathbb{E} \int_0^t f_r(X_r) \, dr$, \forall partitions of $[0, t]$ s.t. $|\Pi^n| \rightarrow 0$.

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$$\Rightarrow |\mathbb{E} \int_s^t f_r(X_r) dr| \lesssim \|f\|_{L_{[s,t]}^{\tilde{q}'}} c^{-\alpha} ((t-s)^{\frac{1}{\tilde{q}} - \alpha H} + C_\varphi (t-s)^{H(\gamma-1 + \frac{1}{Hq'} + \frac{1}{H\tilde{q}} - \alpha)}).$$

Elements of proof 3/3

Proving (b) boils down to control

$$\left| \mathbb{E} \int_u^\xi \underbrace{f_r(B_r + \mathbb{E}^{u-(v-u)} \tilde{X}_r) - f_r(B_r + \mathbb{E}^{u-(\xi-u)} \tilde{X}_r)}_{=: \tilde{f}_r(B_r)} \, dr \right|.$$

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Proving (b) boils down to control

$$\left| \mathbb{E} \int_u^\xi \underbrace{f_r(B_r + \mathbb{E}^{u-(v-u)} \tilde{X}_r) - f_r(B_r + \mathbb{E}^{u-(\xi-u)} \tilde{X}_r)}_{=: \tilde{f}_r(B_r)} \, dr \right|.$$

► Use that $|\mathbb{E}^{u-(\xi-u)} \tilde{f}_r(B_r)| = |g_{\sigma^2} * \tilde{f}_r(\mathbb{E}^{u-(\xi-u)}[B_r])|$ with $\sigma^2 = \text{Var}(B_r \mid \mathcal{F}_{u-(\xi-u)}) \gtrsim (r - u + \xi - u)^{2H}$ \leftarrow use LND!

Elements of proof 3/3

Proving (b) boils down to control

$$\left| \mathbb{E} \int_u^\xi \underbrace{f_r(B_r + \mathbb{E}^{u-(v-u)} \tilde{X}_r) - f_r(B_r + \mathbb{E}^{u-(\xi-u)} \tilde{X}_r)}_{=: \tilde{f}_r(B_r)} dr \right|.$$

► Use that $|\mathbb{E}^{u-(\xi-u)} \tilde{f}_r(B_r)| = |g_{\sigma^2} * \tilde{f}_r(\mathbb{E}^{u-(\xi-u)}[B_r])|$ with $\sigma^2 = \text{Var}(B_r \mid \mathcal{F}_{u-(\xi-u)}) \gtrsim (r - u + \xi - u)^{2H}$ \leftarrow use LND!

► Thus $|\mathbb{E}^{u-(\xi-u)} \tilde{f}_r(B_r)| \lesssim \|\tilde{f}_r\|_{\mathcal{C}^{-\alpha-1}} (r - u + \xi - u)^{-(\alpha+1)H}$ \leftarrow use smoothing of Gaussian kernel.

Elements of proof 3/3

Proving (b) boils down to control

$$\left| \mathbb{E} \int_u^\xi \underbrace{f_r(B_r + \mathbb{E}^{u-(v-u)} \tilde{X}_r) - f_r(B_r + \mathbb{E}^{u-(\xi-u)} \tilde{X}_r)}_{=: \tilde{f}_r(B_r)} dr \right|.$$

- Use that $|\mathbb{E}^{u-(\xi-u)} \tilde{f}_r(B_r)| = |g_{\sigma^2} * \tilde{f}_r(\mathbb{E}^{u-(\xi-u)}[B_r])|$ with $\sigma^2 = \text{Var}(B_r \mid \mathcal{F}_{u-(\xi-u)}) \gtrsim (r - u + \xi - u)^{2H}$ \leftarrow use LND!
- Thus $|\mathbb{E}^{u-(\xi-u)} \tilde{f}_r(B_r)| \lesssim \|\tilde{f}_r\|_{C^{-\alpha-1}} (r - u + \xi - u)^{-(\alpha+1)H}$ \leftarrow use smoothing of Gaussian kernel.
- Now $\|\tilde{f}_r\|_{C^{-\alpha-1}} \leq \|f_r\|_{C^{-\alpha}} |\mathbb{E}^{u-(v-u)} \tilde{X}_r - \mathbb{E}^{u-(\xi-u)} \tilde{X}_r|.$

Elements of proof 3/3

Proving (b) boils down to control

$$\left| \mathbb{E} \int_u^\xi \underbrace{f_r(B_r + \mathbb{E}^{u-(v-u)} \tilde{X}_r) - f_r(B_r + \mathbb{E}^{u-(\xi-u)} \tilde{X}_r)}_{=: \tilde{f}_r(B_r)} dr \right|.$$

► Use that $|\mathbb{E}^{u-(\xi-u)} \tilde{f}_r(B_r)| = |g_{\sigma^2} * \tilde{f}_r(\mathbb{E}^{u-(\xi-u)}[B_r])|$ with $\sigma^2 = \text{Var}(B_r \mid \mathcal{F}_{u-(\xi-u)}) \gtrsim (r-u+\xi-u)^{2H} \leftarrow$ use **LND!**

► Thus $|\mathbb{E}^{u-(\xi-u)} \tilde{f}_r(B_r)| \lesssim \|\tilde{f}_r\|_{C^{-\alpha-1}} (r-u+\xi-u)^{-(\alpha+1)H} \leftarrow$ use **smoothing of Gaussian kernel**.

► Now $\|\tilde{f}_r\|_{C^{-\alpha-1}} \leq \|f_r\|_{C^{-\alpha}} |\mathbb{E}^{u-(v-u)} \tilde{X}_r - \mathbb{E}^{u-(\xi-u)} \tilde{X}_r|$.

► It remains to control $|\mathbb{E}^{u-(v-u)} \tilde{X}_r - \mathbb{E}^{u-(\xi-u)} \tilde{X}_r|$: using *Stochastic sewing with controls*,

$$\|\tilde{X}_r - \mathbb{E}^{u-(v-u)} \tilde{X}_r\|_{L^\infty_\Omega} \leq C(\|\varphi\|_{L^q_{[u-,r]}} c^\gamma + \|\varphi\|_{L^q_{[u-,r]}}^{1+\eta} c^\gamma)(r-u+v-u)^{\frac{1}{q'}+H\gamma}.$$

Elements of proof - Conclusion

Denote $\mathcal{S}_{u,v}$ the set of functions $f \in L^{\tilde{q}'}([u, v]; \mathcal{C}_b^\infty)$ s.t.

$$\|f\|_{L^{\tilde{q}'}_{[u,v]} \mathcal{C}^{-\alpha}} \leq 1.$$

By a density argument, it is sufficient to take the supremum over $f \in \mathcal{S}_{u,v}$, to get

$$\begin{aligned} \|\mathcal{L}(X.)\|_{L^{\tilde{q}}_{[u,v]} \mathcal{B}_{1,1}^\alpha} &\leq C \sup_{f \in \mathcal{S}_{u,v}} \left| \int_u^v \langle f_s, \mathcal{L}(X_s) \rangle ds \right| \\ &\leq C \sup_{f \in \mathcal{S}_{u,v}} \left| \mathbb{E} \int_u^v f_s(X_s) ds \right|. \end{aligned}$$

It remains to use the lemma.