

# Small noise limit for a singular SDE

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June 4, 2024

Joint work with Paul Gassiat

## Motivating question

Consider a function  $b$  and a noise  $(W_t)_{t \geq 0}$  such that SDE (1) is well-defined:

$$X_t = X_0 + \int_0^t b(X_r) dr + \epsilon W_t \quad (1)$$

Our problem:

1. What happens if  $\epsilon \rightarrow 0$ ?
2. Obvious if  $b$  Lipschitz, if not?

# Plan of the talk

1. Regularisation by noise and key ideas why does it work
2. Zero noise limit in 1D
3. work in progress, open questions and future work

## Peano example

Consider  $b(x) = \operatorname{sgn}(x)|x|^\gamma$ ,  $\gamma < 1$ ,  $\epsilon = 0$ . Then:

$$X_t = \int_0^t b(X_r) dr \quad X_t = C_E(t - t_0)_+^{\frac{1}{1-\gamma}}$$

with:

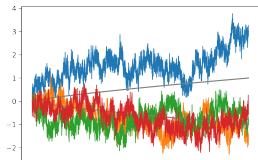
$$C_E = |1 - \gamma|^{\frac{1}{1-\gamma}} \quad t_0 \in \mathbb{R}_+$$

Infinite number of solutions.

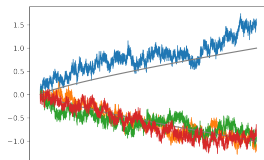
Relevance : fluid dynamics - you tell me !

We will justify these plots...

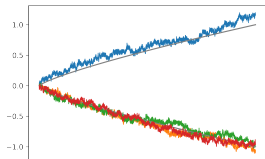
$$dx_t = b(x_t)dt + \epsilon dW_t^H \quad b(x) = \text{sgn}(x)|x|^\gamma, \gamma < 1$$



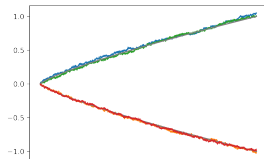
(a)  $\epsilon = 1$



(b)  $\epsilon = 0.3$



(c)  $\epsilon = 0.1$



(d)  $\epsilon = 0.03$

# Fractional Brownian motion

$W^H$  is a centered Gaussian process:

$$\mathbb{E} \left( W_t^H - W_s^H \right)^2 = (t - s)^{2H}$$

- ▶ NOT a martingale
- ▶ NOT a Markov process

$$W_t^H = \int_{-\infty}^t k(t, r) dB_r$$

Lower  $H \Rightarrow$  better regularisation

# Well-posedness with $\epsilon > 0$

## Different contexts

1. Zvonkin 74', Veretennikov 81', Davie 08',  $W$  Brownian motion,  $b \in L^\infty$
2. Catellier-Gubinelli 16',  $W$  fractional Brownian motion.  $b \in \mathcal{C}^\gamma$  (Hölder continuous functions).

$$\gamma > 1 - \frac{1}{2H} \quad W = W^H, \mathbb{E} \left( W_t^H - W_s^H \right)^2 = |t - s|^{2H}$$

3. Krylov-Röckner 05',  $W$  Brownian motion,  $b \in L_q L_p$  for  $\frac{1}{q} + \frac{d}{p} < 1$ , strong solutions (different scheme of proof)

## Well-posedness with $\epsilon > 0$

$$X_t = X_0 + \int_0^t b(X_r) dr + W_t$$

Let  $X_t, Y_t$  be two solutions with  $X_0 = Y_0$  and  $W = 0$

$$|X_t - Y_t| \leq \|\nabla b\|_{L^\infty} \int_0^t |X_r - Y_r| dr$$

so:

$$\|X - Y\|_\infty \leq \|\nabla b\|_{L^\infty} \|X - Y\|_\infty t$$

Can we find a substitute for Lipschitz condition in the presence of the noise  $W_t$ ? If so:

$$\|X - Y\| \leq \|b\|_{C^\gamma} \|X - Y\| t^\alpha$$

$\gamma < 1, \alpha > 0$ , then  $\|b\|_{C^\gamma} t^\alpha < 1/2$

## Well-posedness with $\epsilon > 0$

We change variables:

$$\theta^X = X_t - W_t = \int_0^t b(\theta_r^X + W_r) dr$$

Can we justify ?:

$$\left| \int_0^t b(W_r + \theta_r^X) - b(W_r + \theta_r^Y) dr \right| \leq \left\| \nabla \int_0^\cdot b(W_r + \cdot) dr \right\|_\infty \left\| \theta^X - \theta^Y \right\|$$

Change point of view:

$$\underbrace{x \mapsto b(x)}_{\text{irregular}} \quad \underbrace{x \mapsto \int_0^t b(W_r + x) dr}_{\text{smooth}}$$

# How?

Main tools:

1. stochastic sewing (Le 18, Delarue-Diel 16):

$$\int_0^\cdot (\nabla b)(W_r^H + x) dr =_{L^p} \lim_{|t_{i+1} - t_i| \rightarrow 0} A_{st}(x)$$

where  $A_{st}$  a local approximation

2. nonlinear Young equations (Catellier-Gubinelli 16, Galeati 22):

$$\theta_t = \int_0^t A(dr, \theta_r) \quad A \in \mathcal{C}^\beta \mathcal{C}^1, \beta > 1/2$$

3. convolutions with Gaussian density:

$$g \in L^p \quad \|P_{t^{2H}}(\nabla g)\|_{L^\infty} \lesssim t^{1+H(\frac{1}{p}-1)} \|g\|_{L^p}$$

# Power counting heuristic

$$X_t = x_0 + \int_0^t b(X_r) dr + W_t^H$$

1. Let  $b \in \mathcal{C}^\gamma \cap \mathcal{C}^1(\mathbb{R} \setminus \{0\})$ ,  $\gamma < 1$  - one singularity at zero, ex.:

$$b(x) = \operatorname{sgn}(x)|x|^\gamma$$

2. Local (!) uniqueness almost everywhere except  $x_0 = 0$ .
3. If  $W^H \in \mathcal{C}^{H-}$  then  $X \in \mathcal{C}^{H-}$ .
4.  $b(X) \in \mathcal{C}^{\gamma H}$  and then  $\int_0^\cdot b(X_r) dr \in \mathcal{C}^{1+\gamma H}$
5. As  $t^H > t^{1+\gamma H}$  we have:

$$W_t^H \gg \int_0^t b(X_r) dr$$

6. we escape the bad point

# Did we just cheat?

Not a lot !

Proposition (Galeati-Harang-Mayorcas 22')

For  $X_t$  solution of:

$$X_t = X_0 + \int_0^t b(X_r) dr + W_t^H$$

there holds:

$$\sup_{t < T} t^{-(1+H\gamma)} \left| \int_0^t b(X_r) dr \right| < K \quad \mathbb{E} \exp(K^2) < \infty$$

Proof.

Girsanov + similar estimate for  $\int_0^t b(W_r^H) dr$  via stochastic sewing.



## Back to Peano example - limit of $\epsilon \rightarrow 0$

$$x_t = x_0 + \int_0^t \operatorname{sgn}(x_r) |x_r|^\gamma dr + \epsilon W_t$$

$$y_t^\pm = \pm C_E t^{\frac{1}{1-\gamma}} \quad C_E = \left( \frac{1}{1-\gamma} \right)^{\frac{1}{1-\gamma}}$$

Existing results:

1. Bafico-Baldi 82' -  $X_t^\epsilon \rightarrow y_t^\pm$  with PDE methods, more general  $b$ :

$$b(x) \operatorname{sgn}(x) \geq 0$$

2. Gradinaru-Herrmann-Roynette 01' - Ito calculus  $\Rightarrow$  exponential rate of convergence
3. Delarue-Flandoli 14' - dynamical proof, without rate of convergence
4. Pilipenko-Proske 18' - convergence for  $H \neq 1/2, \gamma > 0$  (not quantified)

# Transition point

$$X_t^\epsilon = \int_0^t \operatorname{sgn}(X_r) |X_r|^\gamma dr + \epsilon W_t^H \quad (2)$$

Heuristic:

1. Initially  $X_t \simeq \epsilon W_t^H$ , for small  $t$

2.

$$\epsilon W_t^H \simeq \epsilon t^H \quad \int_0^t \operatorname{sgn}(X_r) |X_r|^\gamma dr \simeq \epsilon^\gamma t^{1+H\gamma}$$

3. Noise dominates  $t \simeq \epsilon^{\frac{1-\gamma}{1+H(\gamma-1)}} =: t_\epsilon \dots$

4. ...and drags the path away from zero, above  $x_\epsilon = t_\epsilon^{\frac{1}{1-\gamma}}$

Gaussian scaling -  $X_\cdot^\epsilon \stackrel{d}{=} x_\epsilon X_{\cdot/t_\epsilon}^1$

## Theorem (M.-Gassiat 24+)

Let  $\gamma > 1 - \frac{1}{2H}$ . There exists the random variable  $\rho$  such that for some  $\kappa \in (0, 1)$ :

$$\forall s \in [0, 1] \quad |X_{t_\epsilon \rho + s}^\epsilon| \rightarrow C_E s^{\frac{1}{1-\gamma}} = y_s^+ \quad \ln \mathbb{P}(\rho > t) \lesssim -t^\kappa$$

Moreover:

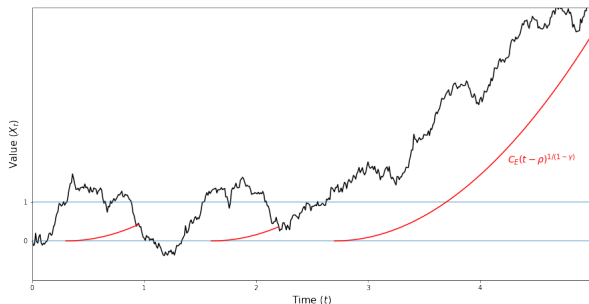
$$|X_{t_\epsilon \rho + s}^\epsilon - \pm y_{t_\epsilon + s}^+| < \epsilon^\alpha \quad \alpha \in (0, 1)$$

and sign is chosen at random.

$\kappa$  can be computed - in the Brownian case ( $H = 1/2$ ) and  $\gamma > 0$  there holds  $\kappa = 1$ .

arXiv - "Zero noise limit for singular ODE regularized by fractional noise", Gassiat, P., Małdry, Ł.

# Ergodic approach



1.  $W$  Brownian motion - strong Markov property  $\Rightarrow$  we can restart the dynamics
2.  $W^H$  fractional Brownian motion - no SMP, but:

$$\forall |t - s| > 1 \quad |\mathbb{E} [W_{s,s+1}^H W_{t,t+1}^H]| \lesssim |t - s|^{2H-2}$$

# Ingredients of the solution

1. Knowledge of the exact solution -

$$\epsilon = 0 \Rightarrow X_t = C_E t^{\frac{1}{1-\gamma}}$$

2. one-dimensionality of the problem
3. Mandelbrot representation with  $(B_t)_{t \in \mathbb{R}}$ :

$$W_t^H = \int_{-\infty}^t (t-r)_+^{H-1/2} - (-r)_+^{H-1/2} dB_r \quad (\cdot)_+ = \max(\cdot, 0)$$

point 3 is critical - quantification of the influence of the past !

# Markovianization (Stage 1)

**Problem** - let  $t, s > 0$ .  $\int_{-\infty}^s (t-r)^{H-1/2} - (s-r)^{H-1/2} dB_r$   
depends on the whole past !

**Solution:**

Find a sequence of times  $(\tau_k)_{k \geq 1}$

$$W_t^H - W_{\tau_k}^H = \int_{\tau_k}^{\tau_k+t} (\tau_k+t-r)^{H-1/2} dB_r + \underbrace{u_t - u_{\tau_k}}_{\text{small}} \quad \|u_{\tau_k+\cdot}\|_{\mathcal{C}^H} < 1$$

tails:

$$\mathbb{E} [\exp(b\tau_k^\alpha)] < (1+b)^k \quad b > 0, \alpha \in (0, 1)$$

Ideas from the study of ergodic properties of fractional SDEs -  
Hairer 05', Panloup-Richard 20'

# Markovianization

Let  $G(t, s, r) = (t - r)^{H-1/2} - (s - r)^{H-1/2}$  and

$$W_t^H - W_s^H = \int_s^t (t - r)^{H-1/2} dB_r + \int_{-\infty}^s G(t, s, r) dB_r$$

Remote and recent past:

$$\int_{-\infty}^{s-\Delta} G(t, s, r) dB_r + \int_{s-\Delta}^s G(t, s, r) dB_r = I + J$$

For  $J$ :

1. fractional integration\*:

$$J \lesssim \|B\|_{[s-\Delta, s]}$$

- 2.

$$c > 0 \quad \mathbb{P}^{\mathcal{F}_{s-\Delta}}(J < c) > \lambda > 0$$

\*Picard 11, "Representation formulae for fBm"

# Markovianization - forgetting property

Let  $k \in \mathbb{N}$ ,  $z^k < w^k$ , for  $[z^k, w^k]$  failed attempt and use integration by parts for fractional integrals\*:

$$\sup_{u>0} u^{-1} \left| \int_{z^k}^{w^k} G(u+v, v, r) dB_r \right| \lesssim |v - w^k|^{H-1} \underbrace{\left( \sup_{s \in [z^k, w^k]} \frac{|B_{w^k} - B_s|}{(1 + w^k - s)^{1/2+\delta}} \right)}_{L(z^k, w^k)}$$

Takeaway - for any interval  $[z^k, w^k]$  we can wait until rhs is small (integrable):

$$v - w^k > k^{\mu_1} + L(z^k, w^k)^{\mu_2} \quad \mu_1, \mu_2 \geq 1$$

\*Picard 11, "Representation formulae for fBm" or Stage 1 in MG24 "Zero noise..."

## Leaving critical strip (Stage 2)

After Markovianization:

$$X_t = \int_0^t b(X_r) dr + \int_0^t (t-r)^{H-1/2} dB_r + u_t$$

Recall (Galeati-Harang-Mayorcas 22', Prop. 3.8):

$$\sup_{t < 1} t^{-(1+H\gamma)} \left| \int_0^t b(X_r) dr \right| < K \quad \mathbb{E}^{\mathcal{F}_0} \exp(K^2) < \infty$$

For any \*fixed\*  $t_e$  there exists  $\lambda > 0$ :

$$\mathbb{P}^{\mathcal{F}_0} \left( \int_0^{t_e} (t_e - r)^{H-1/2} dB_r > 3t_e^H \right) = \mathbb{P}(\mathcal{N}(0, 1) > 3) > \lambda > 0$$

## Leaving critical strip

Fix  $t_e$  small and let  $A$  be the event such that:

1.  $\int_0^{t_e} (t_e - r)^{H-1/2} dB_r > 3t_e^H$
2.  $|\int_0^{t_e} b(X_r) dr| < K t_e^{1+H\gamma} < t_e^H \Rightarrow K < t_e^{-(1+H(\gamma-1))}$

$$\mathbb{P}^{\mathcal{F}_0}(A) > \lambda > 0$$

think about power counting ! Result:

$$X_{t_e} > \underbrace{\int_0^{t_e} (\cdot - r)^{H-1/2} dB_r}_{3t_e^H} - \underbrace{\int_0^{t_e} b(X_r) dr}_{t_e^H} - \underbrace{t_e^H}_u > t_e^H$$

More technicalities -  $\mathbb{P}^{\mathcal{F}_0}(X_1 > 1) > \lambda > 0$

## Drift wins with the noise (Stage 3)

Let  $A > 0$ ,  $X_0 > 1$  and  $b(x) = \text{sgn}(x)|x|^\gamma$ .

$$X_t = X_0 + A \int_0^t b(X_r) dr + w_t \quad \varphi(x_0, t) = x_0 + \int_0^t b(\varphi(x_0, r)) dr$$

### Lemma

Let  $(w_t)_{t \geq 0}$  be a continuous path such that for some  $\beta > 0$ :

$$\sup_{s < t} \frac{|w_t - w_s|}{(1 + t - s)^\beta} \leq A^\beta X_0$$

Then:

$$\forall t > 0 \quad X_t > \varphi(X_0, t) - c(1 + t)^\beta \quad c \in (0, 1)$$

### Proof.

$b$  is Lipschitz away from zero, so we can use comparison principles and exact flow expressions

## Drift wins with the noise - local approximation

$$X_t = \theta_t + w_t = \int_0^t b(X_r) dr + w_t = \int_0^t b(\theta_r + w_r) dr + w_t$$

Let  $\bar{w}_T = \sup_{r < T} |w_r|$  and for  $b \in \mathcal{C}^1, \partial_x b < 0$

$$\theta_t > \theta_0 + \int_0^t b(\theta_r + \bar{w}_T) dr$$

Therefore:

$$\theta_t > \varphi(\theta_0 + \bar{w}, t) - \bar{w}_T$$

And:

$$X_t > \underbrace{\varphi(X_0 + \bar{w}, t)}_{\simeq T^{\frac{1}{1-\gamma}}} - \underbrace{2\bar{w}_T}_{\simeq (1+T)^{H+\delta}}$$

## Drift wins with the noise - condition check

We need:

$$\sup_{s < t} \frac{|w_t - w_s|}{(1 + t - s)^\beta} \leq A^\beta X_0$$

In our case:

$$w_t - w_s = \underbrace{\int_s^t (t - r)^{H-1/2} dB_r + \int_0^s G(t, s, r) dB_r}_{I(t, s)} + \underbrace{\int_{-\infty}^0 G(t, s, r) dB_r}_{J(t, s)}$$

1.  $\forall s < t \ |J(t, s)| \leq (1 + t - s)^{H+\delta}$  a.s.

2.

$$\mathbb{P}^{\mathcal{F}_0} \left( \forall s < t \ |I(t, s)| \leq (1 + t - s)^{H+\delta} \right) > \lambda > 0$$

Result - we can apply Lemma from previous slide.

# Construction of $\rho$ from the main theorem

Recall that  $\rho$  is a random time such that:

$$\forall t > 0 \ X_{\rho t_\epsilon + t} > \varphi(x_\epsilon, t) - t^{H+\delta}$$

Therefore:

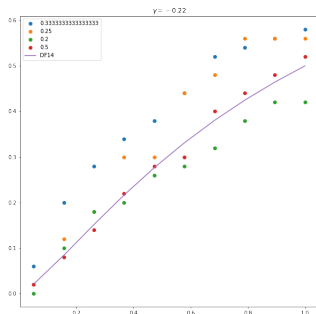
$$\rho_k = \sum_{j=0}^k \Delta_j^1 + \Delta_j^2 + \Delta_j^3$$

1.  $\Delta_j^1$  - "neutralize" the past noise
2.  $\Delta_j^2$  - escape from  $[-1, 1]$
3.  $\Delta_j^3$  - stay above the solution

$$\rho = \rho_{k^*} \quad k^* = \inf\{k \in \mathbb{N} : \Delta_{k+1}^3 = \infty\}$$

# Open questions

1. Optimal rate ? "True" rate probably  $\kappa = 1$  for all  $H \in (0, 1)$
2. Probability of choice of each trajectory ? Far from clear



**Figure:**  $A^+/A^-$ ,  $\gamma = -.2$ ,  $\epsilon = 1$ ,  $T = 10000$

## Future work

1. Multidimensional case - work in progress (some partial progress, many more challenges)
2. Distribution-dependent SDEs ?
3. Vortex systems - burst and collapse:

$$dx_t^i = \sum_{j \neq i}^N b(x_t^i - x_t^j) dt + \epsilon dw_t^i \quad b(x) = x^\perp / |x|^2$$

Numerical results - Grotto, Romito, Viviani 23' for  $N = 3$ .  
Rigorous proof missing.

4. Ergodicity -  $b(x) = g(x) - x^\alpha$ ,  $\alpha \geq 1$ ,  $g$  repulsive and singular.  
Large time behaviour, uniqueness and existence of invariant measure

Thank you for your attention !