

# Joint Hölder continuity of the solutions to a class of SPDEs arising from multi-dimensional superprocesses in the random environment

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# Overview

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# Branching particle system

## Particle motion:

Consider a branching particle system in the random environment. The motion of each particle is described by the SDE:

$$\xi^\alpha(t) = \xi_r^\alpha + B^\alpha(t) - B^\alpha(r) + \int_r^t \int_{\mathbb{R}^d} h(y - \xi^\alpha(s)) W(ds, dy).$$

- $\alpha = (\alpha_0, \dots, \alpha_N) \in \mathbb{N} \times \{1, 2\}^N$ : label of the  $N$ -th generation particles.
- $\xi_r^\alpha \in \mathbb{R}^d$ : the birth position at time  $r \geq 0$ .
- $B^\alpha$ : independent  $d$ -dimensional Brownian motions.
- $W$ :  $d$ -dimensional space-time white noise, independent of  $B^\alpha$ .
- $W$  is regarded as the random environment.

## Branching events:

- For any  $n \in \mathbb{N}$ , the branching happens at fixed time  $\frac{k}{n}$ ,  $k = 1, 2, \dots$
- At any branching time, each particle dies, and randomly generates at most 2 offspring.
- New particles continue the moving/branching mechanism.
- In the  $n$ -th approximation, denote by  $\xi^{\alpha, n}$  the path of particle motion.
- Equip the system an empirical measure

$$X_t^n = \frac{1}{n} \sum_{\alpha \sim_n t} \delta_{\xi_t^{\alpha, n}},$$

where the sum is among all alive particles at time  $t$ .

## Some notations:

- $\rho \in C_b^2(\mathbb{R}^d; \mathbb{R}^d \otimes \mathbb{R}^d)$ , given by

$$\rho(x) = \int_{\mathbb{R}^d} h(x-z)h^*(z)dz.$$

- $A : C_b^2(\mathbb{R}^d) \rightarrow C_b(\mathbb{R}^d)$  given by

$$A\phi(x) = \frac{1}{2} \sum_{i,j=1}^d (\rho^{ij}(0)\partial_{ij}\phi(x)) + \frac{1}{2}\Delta\phi(x),$$

is the generator of one particle motion.

## One dimensional model (known results)

Under the Dawson-Watanabe branching mechanism:

Each particle independently splits into 2 or dies in probabilities  $(\frac{1}{2}/\frac{1}{2})$ .

### Theorem (Wang 97, 98)

Suppose  $X_0^n \Rightarrow X_0 \in M_F(\mathbb{R})$ . Then  $X_t^n \rightarrow X_t$ , that uniquely solves the martingale problem (MP): for any  $\phi \in C_b^2(\mathbb{R})$ ,

$$M_t(\phi) := X_t(\phi) - X_0(\phi) - \int_0^t X_s(A\phi) ds$$

is a continuous square integrable martingale with quadratic variation

$$\langle M(\phi) \rangle_t = \int_0^t \nabla \phi(x) \rho(x-y) \nabla \phi(y) X_s(dx) X_s(dy) ds + \int_0^t X_s(\phi^2) ds.$$

## Theorem (Dawson et al. 00)

$X_t$  has a density  $u_t(x)$  almost surely, that solves the SPDE weakly:

$$u_t(x) = \mu(x) + \int_0^t A^* u_s(x) ds - \int_0^t \int_{\mathbb{R}} \nabla_x [h(y-x) u_s(x)] W(ds, dy) + \int_0^t \sqrt{u_s(x)} \frac{V(ds, dx)}{dx}.$$

where  $\mu$  is the initial “density” that can be any  $L^1$  function or distribution on  $\mathbb{R}$ ,  $V$  is a space-time white noise independent of  $W$ .

## Theorem (Li et al. 12, Hu et al. 14)

Suppose  $\mu \in L^2(\mathbb{R})$  bounded, then  $u$  is almost surely jointly Hölder continuous with time exponent  $\frac{1}{4}$ , and spatial exponent  $\frac{1}{2}$ .

## Remark:

- The density SPDE doesn't have a mild representation. The mild representation is formally written as

$$u_t(x) = \int_{\mathbb{R}} p(t, x - y) \mu(y) dy + \int_0^t \int_{\mathbb{R}} \sqrt{u_s(y)} p(t - s, x - y) V(ds dy) \\ + \int_0^t \int_{\mathbb{R}} \int_{\mathbb{R}} h(y - z) u_s(z) \partial_z p(t - s, x - z) dz W(ds dy),$$

then the last term is not integrable at  $t$ .

- When  $d \geq 2$ , it is proved (Dawson and Hochberg 79) that the super Brownian motion ( $h \equiv 0$ ) involves a singular measure.



# New branching mechanism (Mytnik 96, Sturm 03)

Let  $\{\eta(x) \in \mathbb{R} : x \in \mathbb{R}^d\}$  be the random field on  $\mathbb{R}^d$ :

- $\eta$  is symmetric: for any  $x$ , the distribution of  $\eta(x)$  is symmetric.
- $\exists p > 2$ , s.t.  $\sup_x \mathbb{E}|\eta(x)|^p < \infty$ .
- The correlation in different points:

$$\mathbb{E}(\eta(x)\eta(y)) = \kappa(x, y) \in C_b^2(\mathbb{R}^{2d}).$$

For any  $n \in \mathbb{N}$ , let  $\eta_n = (\sqrt{n} \wedge \eta) \vee -\sqrt{n}$ . The offspring distribution is described by  $\xi_n$ :

In the  $n$ -th approximation:

- Let  $\{\eta_n^i\}_{i \in \mathbb{Z}_+}$  be i.i.d. copies of  $\eta_n$ .
- At the branching time  $\frac{i+1}{n}$ , the offspring distribution of the  $i$ -th generation particle, conditioned on  $\eta_n^i$  and its position  $x$ , is given by

$$P(N^{\alpha,n} = 2 | \eta_n, x) = \frac{1}{\sqrt{n}} \eta_n^+(x),$$

$$P(N^{\alpha,n} = 0 | \eta_n, x) = \frac{1}{\sqrt{n}} \eta_n^-(x),$$

$$P(N^{\alpha,n} = 1 | \eta_n, x) = 1 - \frac{1}{\sqrt{n}} |\eta_n|(x).$$

# Lebesgue density and SPDEs

## Theorem (Hu, nualart, X.)

Suppose  $X_0^n \rightarrow X_0 \in M_F(\mathbb{R}^d)$  that has a bounded Lebesgue density  $\mu$ .  
Then,

- $X^n$  converges weakly to  $X$  in  $D([0, T], M_F(\mathbb{R}^d))$ .
- $X_t$  has a Lebesgue density  $u_t$  for all  $t \in [0, T]$ .
- $u_t$  is the unique weak solution to the following SPDE:

$$u_t = \mu + \int_0^t A^* u_s ds + \int_0^t \int_{\mathbb{R}^d} \nabla^* \cdot [u_s(y) h(y-x)] W(ds, dy) + \int_0^t u_s(x) \frac{V(ds, dx)}{dx}. \quad (1)$$

Here  $A^*$  is the adjoint of  $A$ ,  $V$  is an independent noise, that is white in time, and colored in space with correlation  $\kappa$ .

# Martingale Problems

- By the typical tightness argument, we show that  $\{X^n\}_{n \in \mathbb{N}}$  is tight in  $D([0, T]; M_F(\mathbb{R}^d))$ :

- By Itô's formula,  $\phi \in C_b^2(\mathbb{R}^d)$ , one can decompose

$$X_t^n(\phi) := \int_{\mathbb{R}^d} \phi(x) X_t^n(dx) = X_0^n(\phi) + Z_t^n(\phi) + B_t^n(\phi) + U^n(\phi) + M^{r,n}(\phi),$$

where  $Z_t^n(\phi)$  denotes the drift term,  $B_t^n(\phi)$  denotes the martingale from  $B^\alpha$ ,  $U^n(\phi)$  denotes the random environment martingale, and  $M^{r,n}(\phi)$  denotes the branching martingale.

- $X^n(\phi)$ ,  $Z^n(\phi)$ ,  $M^{b,n}(\phi)$ ,  $U^n(\phi)$  are C-Tight in  $D([0, T], \mathbb{R})$ .

$B_t^n(\phi) \rightarrow 0$  in  $L^2(\Omega)$  for all  $t \in [0, T]$ .

- (Dawson)  $X^n$  is C-tight in  $D([0, T], M_F(\mathbb{R}^d))$ .

(Mitoma)  $Z^n$ ,  $M^n$ ,  $U^n$ , and  $B^n$  are C-tight in  $D([0, T], \mathcal{S}'(\mathbb{R}^d))$ .

- Any limit  $X$  of a convergent subsequence  $X^{n_k}$  is a solution to the MP: for any  $\phi \in C_b^2(\mathbb{R}^d)$ ,

$$M_t(\phi) = X_t(\phi) - X_0(\phi) - \int_0^t X_s(A\phi) ds \quad (2)$$

is a continuous square integrable martingale with quadratic variation

$$\begin{aligned} \langle M(\phi) \rangle_t &= \int_0^t \int_{\mathbb{R}^d \times \mathbb{R}^d} \nabla \phi(x)^* \rho(x-y) \nabla \phi(y) X_s(dx) X_s(dy) ds \\ &+ \int_0^t \int_{\mathbb{R}^d \times \mathbb{R}^d} \kappa(x,y) \phi(x) \phi(y) X_s(dx) X_s(dy) ds. \end{aligned} \quad (3)$$

# Absolute continuity w.r.t. the Lebesgue measure

## Lemma

Let  $X_t$  be any solution to the MP (2) - (3).  $X_0 \in M_F(\mathbb{R}^d)$  has a bounded Lebesgue density  $\mu$ . Then  $X_t$  also has Lebesgue density a.s.

*Sketch of the proof:* We prove the theorem by showing

$$\sup_{h \in (0,1)} \int_0^T \int_{\mathbb{R}^d} \mathbb{E} |X_t(p_h^x)|^2 dx dt < \infty,$$

$$\lim_{h_1, h_2 \downarrow 0} \int_0^T \int_{\mathbb{R}^d} \mathbb{E} |X_t(p_{h_1}^x) - X_t(p_{h_2}^x)|^2 dx dt = 0,$$

where  $p_h^x$  is the heat kernel:

$$p_h^x(\cdot) = (4\pi h)^{-\frac{d}{2}} \exp\left(-\frac{|x - \cdot|^2}{4h}\right).$$

- $\{X_t(p_h^x)\}_{h>0}$  is Cauchy in  $L^2(\Omega \times [0, T] \times \mathbb{R}^d)$ .
- The limit  $u_t(x)$  is the Lebesgue density of  $X_t$  a.s.

By using the moment duality, one can obtain the moment formula:

### Lemma (Moment formula)

For any  $f \in C_b^2(\mathbb{R}^{nd})$ ,

$$\mathbb{E}X_t^{\otimes n}(f) = X_0^{\otimes n}(v(t, \cdot)),$$

where  $v$  is the solution to the PDE

$$\partial_t v(t, x) - \left[ A^{(n)} v(t, x) + \frac{1}{2} \sum_{\substack{1 \leq i, j \leq n \\ i \neq j}} \kappa(x_i, x_j) v(t, x) \right] = 0, \quad (4)$$

with  $v(0, x) = f(x)$ ,  $A^{(n)}$  is the generator of the  $n$ -particle motion:

$$A^{(n)} f = \frac{1}{2} \Delta f + \frac{1}{2} \sum_{i_1, i_2=1}^n \sum_{j_1, j_2=1}^d \rho^{j_1 j_2}(x_{i_1} - x_{i_2}) \frac{\partial^2 f}{\partial x_{i_1}^{j_1} \partial x_{i_2}^{j_2}}(x_1, \dots, x_n).$$

# Density Equation and Conditional Mild Formulation

- Any solution to the MP has a Lebesgue density.
- Every limit process of the particle approximation is a solution to the MP.

Thus every limit process has a density. It solves the SPDE (1) weakly.

- **Uniqueness:** Let  $d_t(x) = u_t^1(x) - u_t^2(x)$ , then  $d$  is a solution to the SPDE with initial condition  $\mu \equiv 0$ , and thus a solution to the MP. By the moment duality,

$$\mathbb{E}\langle d_t, \phi \rangle = \mathbb{E}\langle \mu, \phi_t \rangle = 0,$$

for all  $\phi \in C_b^2(\mathbb{R}^d)$ , which means  $d \equiv 0$ , a.s..

The uniqueness of the solution to the SPDE implies the convergence of empirical measures.



### Conditional mild representation: (Li et al. 12)

The SPDE (1) has a unique weak solution that is the unique strong solution to the equation:

$$u_t(x) = \int_{\mathbb{R}^d} \mu(z) p^W(r, z; t, x) dz + \int_0^t \int_{\mathbb{R}^d} p^W(r, z; t, x) u_s(z) V(ds, dz),$$

where  $p^W$  is the transition density density of one-particle motion conditional on  $W$ .

The well-posedness of the equation can be proved by using the moment estimates of  $p^W$  below.

## Moment estimate of one-particle motion

Recall the one-particle motion  $\xi_t = \xi_t^{r,x}$ , described by the SDE:

$$\xi_t = x + B_t - B_r + \int_0^t \int_{\mathbb{R}^d} h(y - \xi_u) W(du, dy),$$

where  $B$  is a Brownian motion,  $W$  is a space-time white noise, independent of  $B$ , and  $h \in H_2^3(\mathbb{R}^d; \mathbb{R}^d \otimes \mathbb{R}^d)$ .

### Lemma

Let  $p^W(r, x; t, y)$  be the transition density of  $\xi_t$  conditioned on  $W$ . Then

$$\|p^W(r, x; t, y)\|_{2p} \leq C \exp\left(-\frac{|x-y|^2}{t-r}\right) (t-r)^{-\frac{d}{2}},$$

$$\int_{\mathbb{R}^d} \left\| p^W(r, z; t, y_2) - p^W(r, z; t, y_1) \right\|_{2p} dz \leq C (t-r)^{-\frac{1}{2}\beta} |y_2 - y_1|^\beta,$$

$$\int_{\mathbb{R}^d} \left\| p^W(r, z; t, y) - p^W(r, z; s, y) \right\|_{2p} dz \leq C (s-r)^{-\frac{1}{2}\beta} (t-s)^{\frac{1}{2}\beta}.$$

## Preliminaries and notations

- Let  $B = \{B_t : 0 \leq t \leq T\}$  be the standard  $d$ -dimensional Brownian motion.

$H = L^2([0, T]; \mathbb{R}^d)$  is then the basic Hilbert space with  $B$ .

- Denote by  $D$  the Malliavin derivative operator and  $\delta$  the divergence operator.

For any  $n$ -dimensional random variable  $F$ , denoted by  $\gamma_F$ ,  $\sigma_F$  the  $n \times n$  Malliavin matrix and its inverse respectively, if exists.

- The integration by parts formula:

$$\mathbb{E}(\partial_i \phi(F) G) = \sum_{j=1}^n \mathbb{E}[\phi(F) \delta(G \sigma_F^{ij} D F_j)],$$

for all  $F \in \mathbb{D}^{1,2}(\mathbb{R}^n)$ , and  $G \in \mathbb{D}^{1,2}(\mathbb{R})$ .

## Lemma (Malliavin and Thalmaier 05, Bally and Caramellino 11)

Let  $F \in \mathbb{D}^{2,p}(\mathbb{R}^n)$ , and  $\sigma_F^{ij} \in L^p(\Omega)$  for all  $p \geq 1$ . Then  $F$  has the Lebesgue density:

$$\rho(x) = -\mathbb{E} \sum_{i,j=1}^n \mathbf{1}_{B(x,\rho)} \partial_i Q_n(F-x) \delta[\phi_\rho^x(F) \sigma_F^{ij} DF_j],$$

where  $\partial_i Q_n(x) = C_n \frac{x_i}{|x|^{n-1}}$ , and  $\phi_\rho^x \in C^1(\mathbb{R}^d)$ :  $\mathbf{1}_{B(x,\rho)} \leq \phi_\rho^x \leq \mathbf{1}_{B(x,2\rho)}$  and  $|\nabla \phi_\rho^x| \leq 1/\rho$ .

For simplification, we denote by

$$H_{(i)}(F, G) := \sum_{j=1}^n \delta(G \sigma_F^{ij} DF_j)$$

## Non-degeneracy of the Malliavin matrix

- By differentiating the equation of particle motion:

$$D_{\theta}^{(k)} \xi_t = I - \int_{\theta}^t \int_{\mathbb{R}^d} D_{\theta}^{(k)} \xi_s^* dM_s,$$

where  $M$  is a matrix-valued martingale, with entries

$$M_t^{ij} := \sum_{k=1}^d \int_{\theta}^t \int_{\mathbb{R}^d} \partial_i h^{jk}(y - \xi_t) W^k(ds, dy).$$

- Denote by

$$g_k^{ij}(t, y) = \partial_i h^{jk}(y - \xi_t).$$

- Let  $\lambda_{\theta}^{ij}(t) = \sum_{k=1}^d D_{\theta}^{(k)} \xi_t^i D_{\theta}^{(k)} \xi_t^j$ , then

$$\begin{aligned} \lambda_{\theta}(t) = & I - \int_{\theta}^t \lambda_{\theta}(s) dM_s - \int_{\theta}^t dM_s^* \cdot \lambda_{\theta}(s) \\ & + \sum_{k=1}^d \int_{\theta}^t \int_{\mathbb{R}^d} g_k^*(s, y) \lambda_{\theta}(s) g_k(s, y) dy ds. \end{aligned}$$

- It turns out that fix  $\theta \in [r, t)$ ,  $\lambda$  is invertible, with  $\lambda_\theta(t)^{-1} = \beta_\theta(t)$ , satisfies the SDE:

$$\begin{aligned} \beta_\theta(t) = & I + \int_\theta^t \beta_\theta(s) dM_s^* + \int_\theta^t dM_s \cdot \beta_\theta(s) \\ & + \sum_{k=1}^d \int_\theta^t \int_{\mathbb{R}^n} [g_k(s, y)^2 \beta_\theta(s) + g_k(s, y) \beta_\theta(s) g_k^*(s, y) \\ & + \beta_\theta(s) g_k^*(s, y)^2] dy ds. \end{aligned}$$

- By Jensen's inequality (Stroock 83.), almost surely we have

$$\|\sigma_t\|_{H.S.} = \left\| \left( \int_r^t \lambda_\theta(t) d\theta \right)^{-1} \right\|_{H.S.} \leq \frac{1}{(t-r)^2} \left\| \int_r^t \lambda_\theta(t)^{-1} d\theta \right\|_{H.S.}$$

Sketch of the proof: Let  $\rho = \sqrt{t-r}$ . By the density lemma:

$$p^W(r, x; t, y) = - \sum_{i=1}^d \mathbb{E}^W \left( \mathbf{1}_{B(y, 2\rho)}(\xi_t) \partial_i Q_d(\xi_t - y) H_{(i)}(\xi_t, \phi_\rho^y(\xi_t)) \right).$$

- $\xi_t$  is a Gaussian process:

$$\mathbb{P}(\xi_t \in B(y, 2\rho)) \leq C \exp\left(-\frac{k|x-y|^2}{t-r}\right).$$

- (Nualart and Nualart) By integration by parts formula:

$$\sup_{y \in \mathbb{R}^d} \|\partial_i Q_d(\xi_t - y)\|_p \leq C \max_{1 \leq i \leq d} \|H_{(i)}(\xi_t; 1)\|_{\frac{p}{p-1}}^{d-1}.$$

- For  $H_{(i)}(\xi_t, \phi_\rho^y(\xi_t))$ , we have the identity by integration by parts formula:

$$H_{(i)}(\xi_t, \phi(\xi_t)) = \phi_\rho^y(\xi_t) H_{(i)}(\xi_t, 1) + \partial_i \phi_\rho^y(\xi_t).$$

- By Meyer's inequality and the estimates of Sobolev norms of  $\xi_t$ , and  $\sigma_t$ :

$$\|H_{(i)}(\xi_t, 1)\|_p \leq C(t-r)^{-\frac{1}{2}}.$$

The moment estimation of  $p^W(r, x; t, y)$  follows from Hölder's inequality.



## Final result






Based on the conditional mild representation and moment estimates, we easily show:

### Theorem (Hu, Nualart, X.)

For any  $\beta_1, \beta_2 \in (0, 1)$ ,  $p > 1$ ,  $0 < s < t \leq T$ , there exists  $C = C_{p, T, \beta_1, \beta_2}$ , s.t.

$$\|u(t, x) - u(s, y)\|_{2p} \leq Cs^{-\frac{1}{2}} \left( |x - y|^{\beta_1} + (t - s)^{\frac{1}{2}\beta_2} \right).$$

Hence by Kolmogorov's continuity criteria,  $u(t, x)$  is almost surely jointly Hölder continuous, with exponent  $\beta_1 \in (0, 1)$  in space and  $\beta_2 \in (0, \frac{1}{2})$  in time.

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# Thanks!