

# Stochastic PDEs with heavy-tailed noise

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## Stochastic heat equation:

$$\begin{aligned}\partial_t Y(t, x) &= \Delta Y(t, x) + \sigma(Y(t, x)) \dot{\Lambda}(t, x) \\ Y(0, x) &= Y^0(x)\end{aligned}$$

$(t, x)$  time and space coordinate:  $t \geq 0$ ,  $x \in \mathbb{R}^d$

$\dot{\Lambda}$  Lévy white noise on  $\mathbb{R}_+ \times \mathbb{R}^d$

$\sigma$  a Lipschitz function

$Y^0$  initial condition (in this talk  $Y^0 \equiv 1$ )

## Approach by Walsh (1986)<sup>1</sup>:

$$Y(t, x) = Y_0(t, x) + \int_0^t \int_{\mathbb{R}^d} G(t-s, x-y) \sigma(Y(s, y)) \Lambda(ds, dy)$$

(SHE)

- ①  $G$  is the heat kernel on  $\mathbb{R}_+ \times \mathbb{R}^d$ :

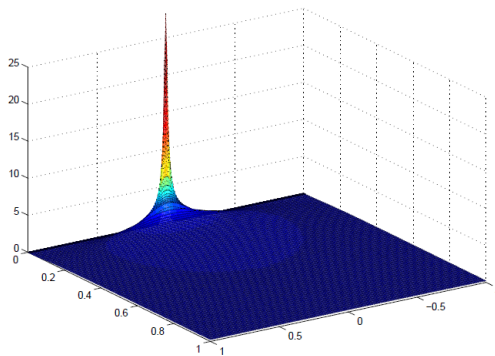
$$G(t, x) := \frac{\exp(-|x|^2/(4t))}{(4\pi t)^{d/2}} \mathbf{1}_{\{t>0\}}$$

- ②  $Y_0(t, x) := \int_{\mathbb{R}^d} G(t, x-y) Y^0(y) dy$  (= 1 in this talk)

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<sup>1</sup> Walsh, J.B. (1986): An introduction to stochastic partial differential equations. In Hennequin, P.L., editor, École d'Été de Probabilités de Saint Flour XIV - 1984, pages 265–439. Springer, Berlin.

# Singularity and integrability



$$\int_0^T \int_{\mathbb{R}^d} G^p(t, x) d(t, x) < \infty$$



$$0 < p < 1 + \frac{2}{d}$$

Figure: The 1D heat kernel for  $(t, x) \in [0, 1] \times [-1, 1]$

## $\wedge$ Gaussian white noise

- Function-valued solutions only for  $d = 1$
- A lot of work on generalizations and properties

→ Dalang, Hairer, Khoshnevisan, Mueller, Walsh ...

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## Our focus:

Noises with jumps

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## Our focus:

Noises with jumps

## In fact:

Noises with **heavy-tailed** jumps

**Noise:**  $\Lambda$  is a **homogeneous Lévy basis without Gaussian part:**

$$\Lambda(dt, dx) = b dt dx + \int_{\mathbb{R}} z \mathbf{1}_{\{|z| \leq 1\}} (p - q)(dt, dx, dz) + \int_{\mathbb{R}} z \mathbf{1}_{\{|z| > 1\}} p(dt, dx, dz)$$

- $p$  Poisson random measure on  $\mathbb{R}_+ \times \mathbb{R}^d \times \mathbb{R}$  with intensity measure  $q(dt, dx, dz) = dt dx \nu(dz)$
- $\nu$  a **Lévy measure**



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**Space of processes:** define for  $p \in (0, 2]$

$$B^p := \left\{ Y \text{ predictable: } \sup_{(t,x) \in [0, T] \times \mathbb{R}^d} \mathbb{E}[|Y(t, x)|^p] < \infty \quad \forall T \in \mathbb{R}_+ \right\}$$

Theorem: Saint Loubert Bié (1998)<sup>1</sup>

If the Lévy measure  $\nu$  satisfies

$$\int_{\mathbb{R}} |z|^p \nu(dz) < \infty$$

for some  $0 < p < 1 + 2/d$ , then (SHE) has a unique solution in  $B^p$ .

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<sup>1</sup> Saint Loubert Bié, E. (1998): Étude d'une EDPS conduite par un bruit poissonnien. *Probab. Theory Relat. Fields*, 111(2):287–321.

## Crucial assumptions in previous work:

- 1 The Lévy measure satisfies

$$\int_{\mathbb{R}} |z|^p \nu(dz) < \infty \quad \text{for some } p < 1 + \frac{2}{d},$$

Stable noises are excluded!

or

- 2 Noise has compact support in space (e.g. Balan (2014))

## Our goal:

$$\int_{|z|\leq 1} |z|^p \nu(dz) + \int_{|z|>1} |z|^q \nu(dz) < \infty \quad \text{with} \quad \boxed{q < p}$$

## Problems:

- Unlike SDEs with heavy-tailed noise, the classical stopping technique does **not** work
- $p$ -th order moment estimates are **infinite**
- $q$ -th order moment estimates exist and are finite, but do **not** produce contraction

**Modified stopping strategy:**

$\tau_N :=$  First time a jump of size  $> N(1 + |x|^\eta)$  occurs at some  $x \in \mathbb{R}^d$

**Lemma**

$\tau_N$  strictly positive and increasing to  $+\infty$  if  $\eta > d/q$

$$J(\phi)(t, x) := Y_0(t, x) + \int_0^t \int_{\mathbb{R}^d} G(t-s, x-y) \sigma(\phi(s, y)) \Lambda(ds, dy).$$

**New problem:**  $p$ -th moments blow up in  $x$  in each iteration:

$$\mathbb{E}[|J(\phi)(t, x)|^p] \leq C_T \int_0^t \int_{\mathbb{R}^d} G^p(t-s, x-y) (1 + \mathbb{E}[|\phi(s, y)|^p]) (1 + |y|^n) d(s, y)$$

( **If**  $q = p$ :

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Still manageable, but one loses uniqueness  
(and global moment estimates)

## Theorem<sup>1</sup>: Existence under heavy-tailed noise

If the Lévy measure  $\nu$  satisfies

$$\int_{|z| \leq 1} |z|^p \nu(dz) + \int_{|z| > 1} |z|^q \nu(dz) < \infty$$

with some

$$0 < p < 1 + \frac{2}{d} \quad \text{and} \quad q > \frac{p}{1 + (1 + \frac{2}{d} - p)}$$

then (SHE) has a solution  $Y$  such that for all  $T, R \in \mathbb{R}_+$  and  $N \in \mathbb{N}$

$$\sup_{(t,x) \in [0,T] \times [-R,R]^d} \mathbb{E}[|Y(t,x)|^p \mathbf{1}_{[0,\tau_N]}(t)] < \infty$$

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<sup>1</sup> Chong, C. (2016): Stochastic PDEs with heavy-tailed noise. Preprint at arXiv:1602.00257 [math.PR].



**Noise truncation:** Let  $h(x) := 1 + |x|^\eta$  and define

$$\begin{aligned}\Lambda^N(dt, dx) &:= b d(t, x) + \int_{\mathbb{R}} z \mathbf{1}_{\{|z| \leq 1\}} (\mathfrak{p} - \mathfrak{q})(dt, dx, dz) \\ &\quad + \int_{\mathbb{R}} z \mathbf{1}_{\{1 < |z| \leq Nh(x)\}} \mathfrak{p}(dt, dx, dz), \quad N \in \mathbb{N}.\end{aligned}$$

**Goal:** Solution with  $\Lambda = \Lambda^N$ , then extend solution from  $\tau_N$  to  $\tau_{N+1}$

**Picard iteration:**

$$Y^0(t, x) := 1, \quad Y^n(t, x) = J(Y^{n-1})(t, x).$$

→ Need moment estimates!

## Moment estimates:

$$\begin{aligned} & \mathbb{E}[|Y^n(t, x) - Y^{n-1}(t, x)|^p] \\ & \leq C_T^n \int_0^t \int_{\mathbb{R}^d} \dots \int_0^{t_{n-1}} \int_{\mathbb{R}^d} G^p(t - t_1, x - x_1) \dots G^p(t_{n-1} - t_n, x_{n-1} - x_n) \\ & \quad \times h(x_1)^{p-q} \dots h(x_n)^{p-q} d(t_n, x_n) \dots d(t_1, x_1) \\ & \leq C_T^n \int_0^t \dots \int_0^{t_{n-1}} \int_{\mathbb{R}^d} \dots \int_{\mathbb{R}^d} G^p(t - t_1, x_1) h(x - x_1)^{p-q} \dots \\ & \quad \times G^p(t_{n-1} - t_n, x_n) h(x - x_1 - \dots - x_n)^{p-q} dx_n \dots dx_1 dt_n \dots dt_1. \end{aligned}$$

## Moment estimates:

$$\begin{aligned} & \mathbb{E}[|Y^n(t, x) - Y^{n-1}(t, x)|^p] \\ & \leq C_T^n \int_0^t \int_{\mathbb{R}^d} \dots \int_0^{t_{n-1}} \int_{\mathbb{R}^d} G^p(t - t_1, x - x_1) \dots G^p(t_{n-1} - t_n, x_{n-1} - x_n) \\ & \quad \times h(x_1)^{p-q} \dots h(x_n)^{p-q} d(t_n, x_n) \dots d(t_1, x_1) \\ & \leq C_T^n \int_0^t \dots \int_0^{t_{n-1}} \int_{\mathbb{R}^d} \dots \int_{\mathbb{R}^d} G^p(t - t_1, x_1) h(x - x_1)^{p-q} \dots \\ & \quad \times G^p(t_{n-1} - t_n, x_n) h(x - x_1 - \dots - x_n)^{p-q} dx_n \dots dx_1 dt_n \dots dt_1. \end{aligned}$$

## The red integrals:

$$\int_{(\mathbb{R}^d)^n} G^p(t-t_1, x_1) \dots G^p(t_{n-1}-t_n, x_n) h(x-x_1-\dots-x_n)^{n(p-q)} d(x_1, \dots, x_n)$$

## Special property of heat kernel:

$$Ct^{\frac{d}{2}(p-1)} G^p(t, \cdot) = \text{density of the } N(0, tI_d)\text{-distribution}$$

**Thus:** With  $X \sim N(0, tI_d)$  we have

The red integrals

$$\begin{aligned} &= C^n \prod_{j=1}^n (t_{j-1} - t_j)^{-\frac{d}{2}(p-1)} \mathbb{E}[h(x - X)^{n(p-q)}] \\ &\leq C^n \prod_{j=1}^n (t_{j-1} - t_j)^{-\frac{d}{2}(p-1)} \Gamma\left(\frac{1 + n\eta(p-q)}{2}\right) \end{aligned}$$

for  $|x| \leq R$

## Putting everything together:

$$\begin{aligned}
 & \mathbb{E}[|Y^n(t, x) - Y^{n-1}(t, x)|^p] \\
 & \leq C_T^n \int_0^t \cdots \int_0^{t_{n-1}} \int_{\mathbb{R}^d} \cdots \int_{\mathbb{R}^d} G^p(t - t_1, x_1) h(x - x_1)^{p-q} \cdots \\
 & \quad \times G^p(t_{n-1} - t_n, x_n) h(x - x_1 - \dots - x_n)^{p-q} dx_n \dots dx_1 dt_n \dots dt_1. \\
 & \leq C_T^n \int_0^t \cdots \int_0^{t_{n-1}} \prod_{j=1}^n (t_{j-1} - t_j)^{-\frac{d}{2}(p-1)} \Gamma\left(\frac{1 + m\eta(p-q)}{2}\right) dt_n \dots dt_1 \\
 & = C_T^n \underbrace{\Gamma\left(\frac{1 + m\eta(p-q)}{2}\right)}_{=\text{blow up due to stopping}} \bigg/ \underbrace{\Gamma\left(1 + \left(1 - \frac{d}{2}(p-1)\right)n\right)}_{=\text{size of iterated integrals}}
 \end{aligned}$$

Summable in  $n$  under the stated hypotheses



# Byproducts: Uniqueness?

No uniqueness in a “nice” space!

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No uniqueness in a “nice” space! But:

## Theorem

The constructed solution  $Y$  to (SHE) in the previous theorem is the unique solution in the space of predictable processes for which there exist a sequence of stopping times  $(T_N)_{N \in \mathbb{N}}$  increasing to  $+\infty$  a.s. and a process  $\phi_0$  such that for arbitrary  $T, R \in \mathbb{R}_+$  and  $K \in \mathbb{N}$  we have

$$\sup_{(t,x) \in [0,T] \times \mathbb{R}^d} \mathbb{E}[|\phi_0(t,x)|^p \mathbf{1}_{[0,T_K]}(t)] < \infty,$$

and

$$\sup_{(t,x) \in [0,T] \times [-R,R]^d} \mathbb{E}[|(Y - J^{(n)}(\phi_0))(t,x)|^p \mathbf{1}_{[0,T_K]}(t)] \rightarrow 0, \quad n \rightarrow \infty.$$

## Theorem<sup>1</sup>

If  $Y^N$  is the solution to (SHE) with noise

$$\Lambda^N(dt, dx) = b dt dx + \int_{\mathbb{R}} z \mathbf{1}_{\{1 < |z| < N\}} p(dt, dx, dz) \\ + \int_{\mathbb{R}} z \mathbf{1}_{\{|z| \leq 1\}} (p - q)(dt, dx, dz),$$

then for all  $T, R \in \mathbb{R}_+$  and  $K \in \mathbb{N}$

$$\sup_{(t,x) \in [0, T] \times [-R, R]^d} \mathbb{E}[|Y(t, x) - Y^N(t, x)|^p \mathbf{1}_{[0, \tau_K]}(t)] \rightarrow 0, \quad N \rightarrow \infty.$$

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<sup>1</sup> Chong, C. (2016): Stochastic PDEs with heavy-tailed noise. Preprint at arXiv:1602.00257 [math.PR].



## Theorem<sup>1</sup>

If  $Y^N$  is the solution to (SHE) with noise

$$\Lambda^N(dt, dx) = b dt dx + \int_{\mathbb{R}} z \mathbf{1}_{\{|z|>1\}} \mathbf{1}_{[-N, N]^d}(x) \mathfrak{p}(dt, dx, dz) \\ + \int_{\mathbb{R}} z \mathbf{1}_{\{|z|\leq 1\}} (\mathfrak{p} - \mathfrak{q})(dt, dx, dz),$$

then for all  $T, R \in \mathbb{R}_+$  and  $K \in \mathbb{N}$

$$\sup_{(t,x) \in [0, T] \times [-R, R]^d} \mathbb{E}[|Y(t, x) - Y^N(t, x)|^p \mathbf{1}_{[0, \tau_K]}(t)] \rightarrow 0, \quad N \rightarrow \infty.$$

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## Theorem<sup>1</sup>

If we additionally have that

$$|\sigma(x)| \leq C(1 + |x|^\gamma), \quad x \in \mathbb{R},$$

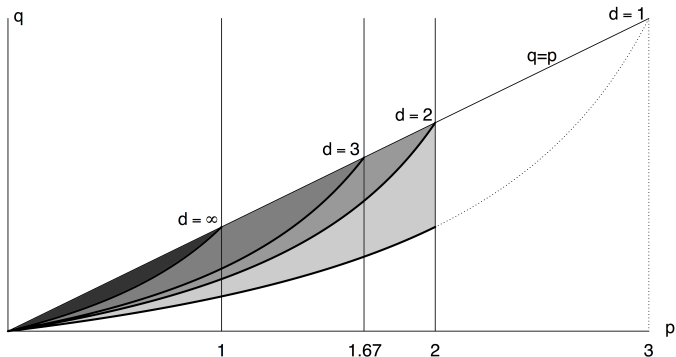
for some  $C \in \mathbb{R}_+$  and  $\gamma \in [0, q/p]$ , then the constructed solution  $Y$  to (SHE) satisfies

$$\sup_{(t,x) \in [0,T] \times \mathbb{R}^d} \mathbb{E}[|Y(t,x)|^q] < \infty, \quad T \geq 0.$$

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## What did we gain?



**Figure:** Constraints on  $p$  and  $q$  dependent on the dimension  $d$ ; new area in grey

## General parabolic SPDEs:

$$\partial_t Y(t, x) = \sum_{|\alpha| \leq 2m} c_\alpha(t, x) \partial^\alpha Y(t, x) + \sigma(Y(t, x)) \dot{\Lambda}(t, x)$$

- 1  $\sigma$  and  $\dot{\Lambda}$  as before
- 2  $m \in \mathbb{N}$ ,  $c_\alpha$  suitable continuous bounded functions

## Result from PDE theory:

$$|G(t, x; s, y)| \leq C_T g(t - s, x - y),$$

$$g(t, x) = \frac{1}{t^{d/(2m)}} \exp\left(C \frac{|x|^{(2m)/(2m-1)}}{t^{1/(2m-1)}}\right)$$

## Theorem<sup>1</sup>

The previous theorems remain valid if the kernel  $G$  satisfies

$$|G(t, x)| \leq C_T t^{-\tau d/\rho} e^{-C|x|^\rho/t^\tau} \mathbf{1}_{(0, \infty)}(t), \quad (t, x) \in [0, T] \times \mathbb{R}^d$$

and

$$0 < \rho < 1 + \frac{\rho}{\tau d} \quad \text{and} \quad \frac{\rho}{1 + \tau(1 + \frac{\rho}{\tau d} - \rho)} < q \leq \rho.$$

**In the parabolic SPDE case:**

- 1  $\rho = (2m)/(2m - 1)$
- 2  $\tau = 1/(2m - 1)$

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**Thank you very much!**