

# Introduction to stochastic PDEs

## Session 2

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January 2011

# Outline

- The stochastic heat equation in  $[0, 1]$ : large deviations, comparison theorem and some generalizations
- One-dimensional stochastic wave equation
- SPDEs in dimension  $d > 1$ : spatially correlated noise
  - ▶ The stochastic heat equation in  $\mathbb{R}^d$ ,  $d \geq 1$
  - ▶ The stochastic wave equation in  $\mathbb{R}^d$ ,  $d \in \{1, 2\}$
  - ▶ The stochastic wave equation in  $\mathbb{R}^3$ : *extension* of the stochastic integral
  - ▶ General result on existence and uniqueness of solution

# Large deviations

Recall that we are dealing with the SPDE:

$$u(t, x) = \int_0^1 u_0(y) G_t(x, y) dy + \int_0^t \int_0^1 G_{t-s}(x, y) \sigma(u(s, y)) W(ds, dy).$$

Let us consider the **stochastic integral** term:

$$V(t, x) := \int_0^t \int_0^1 G_{t-s}(x, y) \sigma(u(s, y)) W(ds, dy).$$

## Theorem

There exist constants  $C_1, C_2 > 0$  such that, for all  $\lambda > 0$ :

$$P \left( \sup_{(t,x) \in [0, T] \times [0, 1]} |V(t, x)| > \lambda \right) \leq C_1 e^{-C_2 \lambda^2}.$$

The proof is based on a generalization of the so-called **Garsia-Rodemich-Rumsey** lemma, thanks to **Arnold-Imkeller** (1996) and also **Funaki-Kikuchi-Potthof** (2006):

## Lemma

Let  $I = [0, T] \times [0, 1]$  and  $g : I \rightarrow \mathbb{R}$  continuous. Assume that there exist two increasing functions  $\psi$  and  $\rho$  with  $\rho(0) = 0$  and  $\lim_{z \rightarrow \infty} \psi(z) = +\infty$ , and such that:

$$B := \int_I \int_I \psi \left( \frac{|g(t, x) - g(s, y)|}{\rho(|t - s|^{p_1} + |x - y|^{p_2})} \right) ds dy dx dt < \infty,$$

where  $p_1, p_2 > 0$ . Then, for all  $(t, x), (s, y) \in I$ ,

$$|g(t, x) - g(s, y)| \leq K \int_0^{|t-s|^{p_1} + |x-y|^{p_2}} \psi^{-1} \left( \frac{B}{r^{\frac{1}{p_1} + \frac{1}{p_2}}} \right) \rho(dr).$$

- This deterministic result is used to prove the **Kolmogorov criterium**.
- Up to some constant,  $r^{\frac{1}{p_1} + \frac{1}{p_2}}$  is the area of the region determined by the points  $(t, x)$  such that  $|t|^{p_1} + |x|^{p_2} \leq r$ .

## Sketch of the proof (large deviations):

The idea is to apply the above lemma with  $g = V$ ,  $\psi(z) = e^{z^2}$ ,  $\rho(z) = z$  and  $p_1 = \frac{1}{4}$ ,  $p_2 = \frac{1}{2}$ . Then:

$$B = \int_I \int_I \exp \left( \frac{|V(t, x) - V(s, y)|^2}{(|t - s|^{\frac{1}{4}} + |x - y|^{\frac{1}{2}})^2} \right) ds dy dx dt.$$

- Using the **martingale property** of the stochastic integral and the fact that the paths of  $u$  are **Hölder-continuous**, we see that  $E(B) < \infty$ .
- Applying the lemma we end up with:

$$\sup_{(t,x) \in [0, T] \times [0, 1]} |V(t, x)| \leq C |\log B|^{\frac{1}{2}}, \quad \mathbb{P}\text{-a.s.}$$

- Eventually, we apply **Chebychev inequality** with the function  $\psi$ :

$$P \left( \sup_{(t,x) \in [0, T] \times [0, 1]} |V(t, x)| > \lambda \right) \leq C_1 e^{-C_2 \lambda^2}.$$

□

# Comparison theorem

Mueller (1991) and Donati-Martin and Pardoux (1993) prove the following:

## Theorem

Let  $u_i$ ,  $i = 1, 2$ , be solutions of the stochastic heat equation

$$u_i(t, x) = \int_0^1 u_{0,i}(y) G_t(x, y) dy + \int_0^t \int_0^1 G_{t-s}(x, y) \sigma(u_i(s, y)) W(ds, dy),$$

with initial condition  $u_{0,i}$ ,  $i = 1, 2$ , respectively. If  $u_{0,1}(x) \leq u_{0,2}(x)$  for all  $x \in [0, 1]$ , then

$$u_1(t, x) \leq u_2(t, x),$$

for all  $t \geq 0$  and  $x \in [0, 1]$ ,  $\mathbb{P}$ -a.s.

- Discretize the equations  $\rightarrow$  system of stochastic differential equations (not partial),
- apply the comparison theorem for (ordinary) SDEs.

# Generalizations

We have studied the equation:

$$\frac{\partial u}{\partial t}(t, x) = \frac{\partial^2 u}{\partial x^2}(t, x) + \sigma(u(t, x))\dot{W}(t, x), \quad t > 0, x \in [0, 1], \quad (1)$$

$$u(0, x) = u_0(x), \quad x \in [0, 1], \quad u(t, 0) = u(t, 1) = 0, \quad t > 0.$$

We have given a sense to it through the **mild formulation**:

$$u(t, x) = \int_0^1 u_0(y)G_t(x, y)dy + \int_0^t \int_0^1 G_{t-s}(x, y)\sigma(u(s, y))W(ds, dy).$$

- We may add a **non-linear term**  $b(u(t, x))$  (**drift**,  $b$  Lipschitz) to eq. (1):

$$\int_0^t \int_0^1 G_{t-s}(x, y)b(u(s, y)) dy ds \quad (\text{pathwise integral}).$$

- The coefficients  $b$  and  $\sigma$  may depend on  $(t, x, u(t, x))$ .
- **Neumann** boundary conditions:  $\frac{\partial u}{\partial x}(t, 0) = \frac{\partial u}{\partial x}(t, 1) = 0$ .

# The stochastic heat equation in $\mathbb{R}$

Using the same techniques, we can deal with the SPDE:

$$\frac{\partial u}{\partial t}(t, x) = \frac{\partial^2 u}{\partial x^2}(t, x) + b(u(t, x)) + \sigma(u(t, x))\dot{W}(t, x), \quad t > 0, x \in \mathbb{R},$$

$$u(0, x) = u_0(x), \quad x \in \mathbb{R}.$$

The associated stochastic integral reads:

$$\begin{aligned} u(t, x) &= \int_{\mathbb{R}} u_0(y) G_t(x - y) dy + \int_0^t \int_{\mathbb{R}} G_{t-s}(x - y) \sigma(u(s, y)) W(ds, dy) \\ &\quad + \int_0^t \int_{\mathbb{R}} G_{t-s}(x - y) b(u(s, y)) dy ds. \end{aligned}$$

where here  $G_t(x)$  is the **fundamental solution** of the heat equation in  $\mathbb{R}$  (**heat kernel**):

$$G_t(x) = \frac{1}{\sqrt{4\pi t}} e^{-\frac{x^2}{2t}}.$$

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# The stochastic wave equation

Almost all the results obtained for the stochastic heat equation are also valid for the **one-dimensional stochastic wave equation**:

$$\frac{\partial^2 u}{\partial t^2}(t, x) = \frac{\partial^2 u}{\partial x^2}(t, x) + b(u(t, x)) + \sigma(u(t, x))\dot{W}(t, x), \quad t > 0, x \in [0, 1],$$

$$u(0, x) = \frac{\partial u}{\partial t}(0, x) = 0, \quad x \in [0, 1],$$

with Dirichlet (or Neumann) boundary conditions:

$$u(t, 0) = u(t, 1) = 0, \quad t \geq 0.$$

As before,  $\dot{W}$  denotes the **space-time white noise**. The **mild solution** is given by:

$$\begin{aligned} u(t, x) = & \int_0^t \int_0^1 \Gamma_{t-s}(x, y) \sigma(u(s, y)) W(ds, dy) \\ & + \int_0^t \int_0^1 \Gamma_{t-s}(x, y) b(u(s, y)) dy ds, \end{aligned}$$

where  $\Gamma_t(x, y)$  is the **Green function** of the wave equation in  $[0, 1]$  with Dirichlet boundary conditions (its expression is a bit nasty).

Or one may also consider the **stochastic wave equation in  $\mathbb{R}$** :

$$u(t, x) = \int_0^t \int_{\mathbb{R}} \Gamma_{t-s}(x - y) \sigma(u(s, y)) W(ds, dy) \\ + \int_0^t \int_{\mathbb{R}} \Gamma_{t-s}(x - y) b(u(s, y)) dy ds,$$

where  $\Gamma_t(x)$  is the **fundamental solution** of the wave equation in  $\mathbb{R}$ :

$$\Gamma_t(x) = \frac{1}{2} \mathbf{1}_{\{|x| < t\}}, \quad t \geq 0 \quad x \in \mathbb{R}.$$

Thus,  $(s, y) \mapsto \Gamma_{t-s}(x - y)$  is the indicator function of the (inverted) **light cone** with apex  $(t, x)$

→ **R. Carmona and D. Nualart**. Random nonlinear wave equations: smoothness of the solutions. Probab. Theory Related Fields 79 (1988), no. 4, 469–508.

# What if $x \in \mathbb{R}^2$ ?

Consider the following **stochastic heat equation**:

$$\frac{\partial u}{\partial t}(t, x) = \left[ \frac{\partial^2 u}{\partial x_1^2}(t, x) + \frac{\partial^2 u}{\partial x_2^2}(t, x) \right] + \dot{W}(t, x), \quad t > 0, \quad x = (x_1, x_2)^* \in \mathbb{R}^2,$$

$$u(0, x) = 0, \quad x \in \mathbb{R}^2.$$

- $\dot{W}$  stands for a white noise in  $\mathbb{R}_+ \times \mathbb{R}^2$ .
- In general, on  $\mathbb{R}^d$ , the operator  $\Delta := \sum_{i=1}^d \frac{\partial^2}{\partial x_i^2}$  is called the **Laplacian**.

The **mild solution** would simply be

$$u(t, x) = \int_0^t \int_{\mathbb{R}^2} G_{t-s}(x - y) W(ds, dy),$$

where  $G_t(x)$  is the heat kernel in  $\mathbb{R}^2$ :

$$G_t(x) = \frac{1}{4\pi t} e^{-\frac{|x|^2}{2t}}, \quad x \in \mathbb{R}^2,$$

and  $|\cdot|$  denotes the euclidean norm in  $\mathbb{R}^2$ .

In order that  $\{u(t, x), (t, x) \in \mathbb{R}_+ \times \mathbb{R}^2\}$  defines a mild solution, we only need to check that the stochastic integral

$$u(t, x) = \int_0^t \int_{\mathbb{R}^2} G_{t-s}(x - y) W(ds, dy)$$

is well-defined.

- for all  $(t, x)$ ,  $u(t, x)$  is the **Wiener integral** of  $G_{t-\cdot}(x - \star)$ :

$$u(t, x) = \dot{W}(G_{t-\cdot}(x - \star)).$$

- Hence,  $u(t, x)$  is a mean zero **Gaussian** random variable.

Its variance is given by:

$$\begin{aligned} E(|u(t, x)|^2) &= \int_0^t \int_{\mathbb{R}^2} |G_{t-s}(x - y)|^2 dy ds = \int_0^t \int_{\mathbb{R}^2} |G_s(y)|^2 dy ds \\ &= \int_0^t \int_{\mathbb{R}^2} \frac{1}{16\pi^2 s^2} e^{-\frac{|y|^2}{2s}} dy ds = C \int_0^t s^{-1} ds = +\infty!!! \end{aligned}$$

$\implies$  **there is not** a mild solution of the stochastic heat equation in  $\mathbb{R}^2$ .

We have exactly the same problem for the **stochastic wave equation** in  $\mathbb{R}^2$ :

$$\frac{\partial^2 u}{\partial t^2}(t, x) = \Delta u(t, x) + \dot{W}(t, x), \quad t > 0, x \in \mathbb{R}^2,$$

$$u(0, x) = \frac{\partial u}{\partial t}(0, x) = 0, \quad x \in \mathbb{R}^2.$$

Mild solution:

$$u(t, x) = \int_0^t \int_{\mathbb{R}^2} \Gamma_{t-s}(x-y) W(ds, dy) = \dot{W}(\Gamma_{t-\cdot}(x-\star)),$$

where here  $\Gamma_t(x)$  is the fundamental solution of the wave equation in  $\mathbb{R}^2$ :

$$\Gamma_t(x) = \frac{1}{2\pi} \frac{1}{\sqrt{t^2 - |x|^2}} \mathbf{1}_{\{|x| < t\}}$$

Then

$$E|u(t, x)|^2 = \frac{1}{4\pi^2} \int_0^t \int_{\mathbb{R}^2} \frac{1}{(t-s)^2 - |x-y|^2} \mathbf{1}_{\{|x-y| < t-s\}} dy ds = +\infty.$$

# Spatially correlated noise (Dalang and Frangos 1998, Dalang 1999)

In order to obtain well-defined **random field** solutions, we consider a slightly **more regular** noise  $\dot{W}(t, x)$ :

- Mean zero Gaussian noise  $\dot{W}(t, x)$  with covariance (formal):

$$E(\dot{W}(t, x)\dot{W}(s, y)) = \delta_0(t - s)f(x - y), \quad (t, x) \in \mathbb{R}_+ \times \mathbb{R}^d \quad (d \geq 1)$$

- **White** correlation in time (Brownian) and **colored** in space.

More rigorously: let  $\{\dot{W}(\varphi), \varphi \in \mathcal{D}(\mathbb{R}^{d+1})\}$  be a family of **mean zero Gaussian** r.v. with the following **covariance functional**:

$$\begin{aligned} E(\dot{W}(\varphi)\dot{W}(\psi)) &= \int_{\mathbb{R}_+} dt \int_{\mathbb{R}^d} dx \int_{\mathbb{R}^d} dy \varphi(t, x)f(x - y)\psi(t, y) \\ &= \int_{\mathbb{R}_+} dt \int_{\mathbb{R}_+} ds \int_{\mathbb{R}^d} dx \int_{\mathbb{R}^d} dy \varphi(t, x)\delta_0(t - s)f(x - y)\psi(t, y), \end{aligned}$$

- $f : \mathbb{R}^d \rightarrow \mathbb{R}_+$  is continuous in  $\mathbb{R}^d \setminus \{0\}$ .
- Covariance functional  $\implies f$  **symmetric**.

(usually we write  $W(\varphi)$ )

## Remarks

$$E(\dot{W}(\varphi)\dot{W}(\psi)) = \int_{\mathbb{R}_+} dt \int_{\mathbb{R}^d} dx \int_{\mathbb{R}^d} dy \varphi(t, x) f(x - y) \psi(t, y)$$

- We say that  $\dot{W}$  is a **spatially homogeneous noise**.
- The case  $f$  bounded has been treated by Mueller 1997.
- Applications to physics: spatial correlations of larger order than those of time  $\Rightarrow$  better models than the space-time white noise.
- Indeed, one may consider even a more general correlation: non-negative tempered measure.

### The spectral measure associated to the noise:

There exists a **non-negative tempered measure**  $\mu$  such that  $\mathcal{F}\mu = f$ . Namely:

$$\int_{\mathbb{R}^d} \frac{1}{(1 + |\xi|^2)^k} \mu(d\xi) < \infty, \quad \text{for some } k \geq 1,$$

$$\mathcal{F}\mu = f \iff \int_{\mathbb{R}^d} f(x)\phi(x)dx = \int_{\mathbb{R}^d} \mathcal{F}\phi(\xi)\mu(d\xi), \quad \phi \in \mathcal{S}(\mathbb{R}^d).$$

By means of a change of variables and applying that  $\mathcal{F}\mu = f$ , we obtain:

$$\begin{aligned}
 E(\dot{W}(\varphi)\dot{W}(\psi)) &= \int_{\mathbb{R}_+} dt \int_{\mathbb{R}^d} dx \int_{\mathbb{R}^d} dy \varphi(t, x) f(x - y) \psi(t, y) \\
 &= \int_{\mathbb{R}_+} dt \int_{\mathbb{R}^d} dy f(y) \int_{\mathbb{R}^d} dx \varphi(t, x) \psi(t, x - y) \\
 &= \int_{\mathbb{R}_+} dt \int_{\mathbb{R}^d} dy f(y) (\varphi(t) * \tilde{\psi}(t))(y) \quad (\tilde{\psi}(t, y) := \psi(t, -y)) \\
 &= \int_{\mathbb{R}_+} dt \int_{\mathbb{R}^d} \mu(d\xi) \mathcal{F}\varphi(t)(\xi) \overline{\mathcal{F}\psi(t)(\xi)}.
 \end{aligned}$$

In particular:

$$E(|\dot{W}(\varphi)|^2) = \int_{\mathbb{R}_+} dt \int_{\mathbb{R}^d} \mu(d\xi) |\mathcal{F}\varphi(t)(\xi)|^2, \quad \varphi \in \mathcal{D}(\mathbb{R}^{d+1}).$$

# Examples

- 1 The **space-time white noise**:  $f(x) = \delta_0(x)$

$$E(\dot{W}(\varphi)\dot{W}(\psi)) = \int_{\mathbb{R}_+} dt \int_{\mathbb{R}^d} dx \int_{\mathbb{R}^d} dy \varphi(t, x)\delta_0(x - y)\psi(t, y) = \langle \varphi, \psi \rangle_{L^2(\mathbb{R}_+ \times \mathbb{R}^d)}^2$$

The spectral measure  $\mu$  is the **Lebesgue measure** on  $\mathbb{R}^d$  and

$$E(|\dot{W}(\varphi)|^2) = \|\varphi\|_{L^2(\mathbb{R}_+ \times \mathbb{R}^d)}^2 = \int_{\mathbb{R}_+} \|\mathcal{F}\varphi(t)\|_{L^2(\mathbb{R}^d)}^2 dt \quad (\text{Plancherel}).$$

- 2 The **Riesz kernels**: for  $x \in \mathbb{R}^d$  and  $0 < \alpha < 2 \wedge d$ , let  $f_\alpha(x) = |x|^{-\alpha}$ . It holds that  $f_\alpha = \mathcal{F}f_{d-\alpha}$ , thus:

$$\begin{aligned} E(|\dot{W}(\varphi)|^2) &= \int_{\mathbb{R}_+} dt \int_{\mathbb{R}^d} dx \int_{\mathbb{R}^d} dy \varphi(t, x)|x - y|^{-\alpha}\varphi(t, y) \\ &= \int_{\mathbb{R}_+} dt \int_{\mathbb{R}^d} d\xi |\xi|^{d-\alpha} |\mathcal{F}\varphi(t)(\xi)|^2. \end{aligned}$$

# The isonormal process

- 1 Let  $\mathcal{H}$  be the Hilbert space which is the closure of  $\mathcal{S}(\mathbb{R}^d)$  with respect to the inner product

$$\langle \phi_1, \phi_2 \rangle_{\mathcal{H}} = \int_{\mathbb{R}^d} dx \int_{\mathbb{R}^d} dy \phi_1(x) f(x-y) \phi_2(y), \quad \phi_1, \phi_2 \in \mathcal{S}(\mathbb{R}^d).$$

The space  $\mathcal{H}$  is sometimes called the **reproducing kernel Hilbert space**. Set  $\mathcal{H}_T := L^2([0, T]; \mathcal{H})$ .

- 2 Fix  $T > 0$ . Then the map  $\varphi \mapsto \dot{W}(\varphi)$  defines an **isometry**:

$$\left( \mathcal{S}(\mathbb{R}^d), \|\cdot\|_{\mathcal{H}_T} \right) \longrightarrow \left( L^2(\Omega), \|\cdot\|_{L^2(\Omega)} \right)$$

where

$$\|h\|_{\mathcal{H}_T}^2 = \int_0^T \|h(s)\|_{\mathcal{H}}^2 ds.$$

- 3 This let us extend  $\dot{W}(\varphi)$  to a mean-zero Gaussian family  $\{\dot{W}(h), h \in \mathcal{H}_T\}$  such that:

$$E(\dot{W}(h)\dot{W}(g)) = \langle h, g \rangle_{\mathcal{H}_T} = \int_{\mathbb{R}_+} \langle g(s), h(s) \rangle_{\mathcal{H}} ds, \quad g, h \in \mathcal{H}_T.$$

# The martingale measure

Using an approximation argument by means of *smooth* functions, we prove that  $\mathcal{H}$  contains elements of the form  $h = \mathbf{1}_A$  with  $A \in \mathcal{B}_0(\mathbb{R}^d)$ . Then

$$W_t(A) := \dot{W}(\mathbf{1}_{[0,t]}(\cdot)\mathbf{1}_A(\star)), \quad t \geq 0, A \in \mathcal{B}_0(\mathbb{R}^d)$$

defines a **worthy martingale measure** with respect to the *natural* filtration.

The **covariance measure** coincides with the **dominant measure** and they are given by:

$$Q([0, t] \times A \times B) = \langle W(A), W(B) \rangle_t = t \int_{\mathbb{R}^d} dx \int_{\mathbb{R}^d} dy \mathbf{1}_A(x) f(x-y) \mathbf{1}_B(y).$$

Hence, we can define the **stochastic integral**

$$(Y \cdot W)_t(A) = \int_0^t \int_A Y(s, y) W(ds, dy), \quad t \in [0, T], A \in \mathcal{B}_0(\mathbb{R}^d),$$

provided that  $\{Y(s, y), (s, y) \in [0, T] \times \mathbb{R}^d\}$  is a predictable process such that:

$$E \left( \int_0^T ds \int_{\mathbb{R}^d} dx \int_{\mathbb{R}^d} dy |Y(s, x)| f(x-y) |Y(s, y)| \right) < +\infty.$$

# Remarks

- The isonormal process  $\{\dot{W}(h), h \in \mathcal{H}_T\}$  would let us apply the techniques of the **Malliavin calculus**.
- As mentioned before, in the case of the **space-time white noise**, we have  $\mathcal{H} = L^2(\mathbb{R}^d)$ . However, in the present situation,  $\mathcal{H}$  is a **larger space** and indeed may contain distributions:

$$\begin{aligned}\langle \phi_1, \phi_2 \rangle_{\mathcal{H}} &= \int_{\mathbb{R}^d} dx \int_{\mathbb{R}^d} dy \phi_1(x) f(x-y) \phi_2(y) \\ &= \int_{\mathbb{R}^d} \mu(d\xi) \mathcal{F}\phi_1(\xi) \overline{\mathcal{F}\phi_2(\xi)}, \quad \phi_1, \phi_2 \in \mathcal{S}(\mathbb{R}^d).\end{aligned}$$

- The space  $\mathcal{H}$  can be identified with the elements  $\Phi$  such that

$$\|\Phi\|_{\mathcal{H}}^2 = \int_{\mathbb{R}^d} \mu(d\xi) |\mathcal{F}\Phi(\xi)|^2 < +\infty.$$

Later on, we will give examples of  $\Phi$  that are not functions.

# The stochastic heat equation in $\mathbb{R}^d$

$$\frac{\partial u}{\partial t}(t, x) = \Delta u(t, x) + \dot{W}(t, x), \quad t > 0, x \in \mathbb{R}^d,$$

$$u(0, x) = 0, \quad x \in \mathbb{R}^d.$$

Here,  $\dot{W}$  stands for the **spatially homogeneous noise** on  $\mathbb{R}^d$ . The mild solution, if exists, is given by the Wiener integral

$$u(t, x) = \int_0^t \int_{\mathbb{R}^d} G_{t-s}(x-y) W(ds, dy) = \dot{W}(G_{t-\cdot}(x-\star)), \quad (t, x) \in [0, T] \times \mathbb{R}^d,$$

where  $G_t(x) = (4\pi t)^{-d/2} \exp\left(-\frac{|x-y|^2}{2t}\right)$ . This is well-defined provided that

$$\int_0^t ds \int_{\mathbb{R}^d} dy \int_{\mathbb{R}^d} dz G_{t-s}(x-y) f(y-z) G_{t-s}(x-z) < +\infty,$$

because, for any  $t > 0$ ,  $G_t(\star) \in \mathcal{S}(\mathbb{R}^d)$ . This exhibits the relation between the differential operator and the noise.

Let us express this condition in terms of the **spectral measure**  $\mu$  (recall that  $\mathcal{F}f = \mu$ ): again because  $G_t(\star) \in \mathcal{S}(\mathbb{R}^d)$ , we have

$$\begin{aligned} & \int_0^t ds \int_{\mathbb{R}^d} dy \int_{\mathbb{R}^d} dz G_{t-s}(x-y) f(y-z) G_{t-s}(x-z) \\ &= \int_0^t ds \int_{\mathbb{R}^d} \mu(d\xi) |\mathcal{F}G_s(x-\cdot)(\xi)|^2 = \int_0^t ds \int_{\mathbb{R}^d} \mu(d\xi) |\mathcal{F}G_s(\xi)|^2 \\ &= \int_0^t ds \int_{\mathbb{R}^d} \mu(d\xi) e^{-8\pi^2 s |\xi|^2} = \int_{\mathbb{R}^d} \mu(d\xi) \frac{1}{8\pi^2 |\xi|^2} \left(1 - e^{-8\pi^2 t |\xi|^2}\right). \end{aligned}$$

The latter expression is finite if and only if

$$\int_{\mathbb{R}^d} \frac{1}{1 + |\xi|^2} \mu(d\xi) < +\infty \quad (\text{for all } d \geq 1)$$

Recall that  $|\cdot|$  denotes the Euclidean norm in  $\mathbb{R}^d$ .

The condition

$$\int_{\mathbb{R}^d} \frac{1}{1 + |\xi|^2} \mu(d\xi) < +\infty \quad (2)$$

can be expressed in terms of  $f$ , but we do not have such a unified expression:

- If  $d = 1$ , (2) is always true.
- If  $d = 2$ , (2) is equivalent to

$$\int_{|x| \leq 1} f(x) \log \frac{1}{|x|} dx < +\infty.$$

- If  $d \geq 3$ , (2) is equivalent to

$$\int_{|x| \leq 1} f(x) \frac{1}{|x|^{d-2}} dx < +\infty.$$

This has been proved by Dalang 1999.

## Mild solution

$$\frac{\partial u}{\partial t}(t, x) = \Delta u(t, x) + b(u(t, x)) + \sigma(u(t, x))\dot{W}(t, x), \quad t > 0, x \in \mathbb{R}^d,$$
$$u(0, x) = u_0, \quad x \in \mathbb{R}^d$$

- $\dot{W}$  is the **spatially homogeneous noise** on  $\mathbb{R}^d$ .
- $b, \sigma : \mathbb{R} \rightarrow \mathbb{R}$  are **Lipschitz** functions.
- The initial condition  $u_0 : \mathbb{R}^d \rightarrow \mathbb{R}$  is **measurable and bounded**.

By definition, the **mild solution** is an adapted process  $\{u(t, x), (t, x) \in [0, T] \times \mathbb{R}^d\}$  satisfying:

$$u(t, x) = (G_t * u_0)(x) + \int_0^t \int_{\mathbb{R}^d} G_{t-s}(x-y) \sigma(u(s, y)) W(ds, dy) \\ + \int_0^t \int_{\mathbb{R}^d} G_{t-s}(x-y) b(u(s, y)) dy ds. \quad (3)$$

# Existence and uniqueness of solution

## Theorem

Assume that  $b, \sigma$  are Lipschitz,  $u_0$  is measurable and bounded and that

$$\int_{\mathbb{R}^d} \frac{\mu(d\xi)}{1 + |\xi|^2} < \infty \quad \left( \Leftrightarrow \int_0^T dt \int_{\mathbb{R}^d} \mu(d\xi) |\mathcal{F}G_t(\xi)|^2 < \infty \right)$$

Then, there exists a unique mild solution to equation (3). Moreover, for all  $p \geq 1$ :

$$\sup_{(t,x) \in [0,T] \times \mathbb{R}^d} E(|u(t,x)|^p) < +\infty.$$

→ Dalang 1999, Dalang and QS 2010.

The proof is based on a Picard iteration scheme and the following **version of the Gronwall lemma**:

### Lemma (Dalang 1999)

Let  $g : [0, T] \rightarrow \mathbb{R}$  be a non-negative function such that

$$\int_0^T g(s) ds < +\infty.$$

Let  $\{f_n, n \geq 1\}$  be a sequence of non-negative functions on  $[0, T]$  and  $k_1, k_2 > 0$  such that, for all  $t \in [0, T]$ :

$$f_n(t) \leq k_1 + \int_0^t (k_2 + f_{n-1}(s))g(t-s)ds.$$

Then

$$\sup_{n \geq 0} \sup_{t \in [0, T]} f_n(t) < \infty$$

and if  $k_1 = k_2 = 0$ , then  $\sum_{n \geq 0} f_n(t)$  converges uniformly  $[0, T]$ .

# The stochastic wave equation with $d \in \{1, 2\}$

$$\frac{\partial^2 u}{\partial t^2}(t, x) = \Delta u(t, x) + b(u(t, x)) + \sigma(u(t, x))\dot{W}(t, x), \quad t > 0, x \in \mathbb{R}^d,$$

$$u(0, x) = \frac{\partial u}{\partial t}(0, x) = 0, \quad x \in \mathbb{R}^d \quad (\text{for simplicity}).$$

The **mild form** is given by:

$$\begin{aligned} u(t, x) &= \int_0^t \int_{\mathbb{R}^d} \Gamma_{t-s}^d(x-y) \sigma(u(s, y)) W(ds, dy) \\ &\quad + \int_0^t \int_{\mathbb{R}^d} \Gamma_{t-s}^d(x-y) b(u(s, y)) dy ds, \end{aligned}$$

where the **fundamental solution**  $\Gamma_t^d(x)$ ,  $d \in \{1, 2\}$  is given by:

$$\Gamma_t^1(x) = \frac{1}{2} \mathbf{1}_{\{|x| < t\}}, \quad \Gamma_t^2(x) = \frac{1}{2\pi} \frac{1}{\sqrt{t^2 - |x|^2}} \mathbf{1}_{\{|x| < t\}}.$$

**Dalang 1999:** under the condition

$$\int_0^T dt \int_{\mathbb{R}^d} \mu(d\xi) |\mathcal{F}\Gamma_t^d(\xi)|^2 < +\infty \quad (d = 1, 2) \quad (4)$$

there **exists a unique mild solution** with uniformly finite moments of any order. Indeed, condition (4) is also equivalent to

$$\int_{\mathbb{R}^d} \frac{1}{1 + |\xi|^2} \mu(d\xi) < +\infty.$$

This is because there is a unified expression of  $\mathcal{F}\Gamma_t^d$  for any dimension  $d \geq 1$ :

$$\mathcal{F}\Gamma_t^d(\xi) = \frac{\sin(2\pi t|\xi|)}{2\pi|\xi|},$$

so that one proves:

$$\frac{C_1}{1 + |\xi|^2} \leq \int_0^T \frac{\sin^2(2\pi t|\xi|)}{4\pi^2|\xi|^2} dt \leq \frac{C_2}{1 + |\xi|^2}.$$

# Remarks

- Stochastic heat equation with  $d \geq 1$ : the **path properties** have been studied in Sanz-Solé and Sarrà 2002.
- Stochastic wave equation with  $d \in \{1, 2\}$ :
  - ▶  $d = 1$  with the **space-time white noise**: Carmona and Nualart 1988.
  - ▶  $d = 2$  with the spatially homogeneous noise: Mueller 1997, Dalang and Frangos 1998 and Dalang 1999.
  - ▶ Path properties: Dalang and Sanz-Solé 2005.
- Extension to other types of SPDEs: **the damped wave equation** (Dalang 1999):

$$\frac{\partial^2 u}{\partial t^2}(t, x) + c \frac{\partial u}{\partial t}(t, x) - \Delta u(t, x), \quad x \in \mathbb{R}^d, \quad d \geq 1,$$

where  $c \neq 0$ . The condition  $\int_{\mathbb{R}^d} (1 + |\xi|^2)^{-1} \mu(d\xi) < \infty$  is also sufficient to have existence and uniqueness of mild solution.

# Recall

- We have considered the **stochastic heat and wave equations** with space dimension  $d \geq 1$  and  $d \in \{1, 2\}$ , respectively.
- The noise  $\dot{W}$  is white in time and with a **spatially homogeneous correlation**  $f$ , whose spectral measure is  $\mu$ :

$$E(\dot{W}(t, x)\dot{W}(s, y)) = \delta_0(t - s)f(x - y)$$

- We have obtained existence and uniqueness of **mild solution** provided that

$$\int_{\mathbb{R}^d} \frac{1}{1 + |\xi|^2} \mu(d\xi) < +\infty.$$

- The stochastic integral in the mild form is with respect to the **martingale measure** associated to the noise.

# Problem!

Consider the stochastic wave equation in  $\mathbb{R}^3$  (we take  $b = 0$ ):

$$\frac{\partial^2 u}{\partial t^2}(t, x) = \Delta u(t, x) + \sigma(u(t, x))\dot{W}(t, x), \quad t > 0, x \in \mathbb{R}^3,$$

$$u(0, x) = \frac{\partial u}{\partial t}(0, x) = 0, \quad x \in \mathbb{R}^3.$$

The mild solution should be given by the solution of

$$u(t, x) = \int_0^t \int_{\mathbb{R}^d} \Gamma_{t-s}^3(x-y) \sigma(u(s, y)) W(ds, dy)$$

where  $\Gamma_t^3(x)$  is the fundamental solution of the wave equation in  $\mathbb{R}^3$ :

$$\Gamma_t^3 = \frac{1}{4\pi t} \sigma_t,$$

where  $\sigma_t$  is the uniform surface measure on the 3-dimensional sphere of radius  $t$ . Thus  $\Gamma_t^3$  is a **measure!**

With respect to the martingale measure associated to  $\dot{W}$ , we are only able to integrate functions!

In order to be able to study mild solutions to the stochastic wave equation in  $\mathbb{R}^3$ , we need to **extend the stochastic integral**:

$$\int_0^t \int_{\mathbb{R}^3} Y(s, y) W(ds, dy),$$

- the integral takes values in  $\mathbb{R}$ , but
- the process  $Y$  satisfies that, for any  $t > 0$ ,  $Y(t)$  does not need to be a function

### Extensions of the stochastic integral:

- Dalang 1999: extension of Walsh stochastic integral with respect to worthy martingale measures.
- Nualart and QS 2007 (also Dalang and QS 2010): stochastic integral with respect to the **cylindrical Wiener process** associated to  $\dot{W}$ , covering Walsh and Dalang's integral.

# Cylindrical Wiener process associated to $\dot{W}$

- Hilbert space  $\mathcal{H}$ : the completion of  $\mathcal{S}(\mathbb{R}^d)$  with respect to

$$\langle \phi_1, \phi_2 \rangle_{\mathcal{H}} = \int_{\mathbb{R}^d} dx \int_{\mathbb{R}^d} dy \phi_1(x) f(x-y) \phi_2(y), \quad \phi_1, \phi_2 \in \mathcal{S}(\mathbb{R}^d).$$

We set  $\mathcal{H}_T = L^2([0, T]; \mathcal{H})$ .

- We had  $\{\dot{W}(h), h \in \mathcal{H}_T\}$  the **isonormal Gaussian process** associated to  $\dot{W}$ : mean zero Gaussian family with

$$E(\dot{W}(h_1)\dot{W}(h_2)) = \langle h_1, h_2 \rangle_{\mathcal{H}_T}, \quad h_1, h_2 \in \mathcal{H}_T.$$

## Definition

We define:

$$W_t(g) = \dot{W}(\mathbf{1}_{[0,t]}(\cdot)g(\cdot)), \quad t \in [0, T], g \in \mathcal{H}.$$

Then,  $\{W_t(g), t \in [0, T], g \in \mathcal{H}\}$  is called a **cylindrical Wiener process** on the Hilbert space  $\mathcal{H}$ : for any  $g \in \mathcal{H}$ ,  $\{W_t(g), t \in [0, T]\}$  is a Brownian motion with variance  $t\|g\|_{\mathcal{H}}^2$  and

$$E(W_s(g_1)W_t(g_2)) = (s \wedge t)\langle g_1, g_2 \rangle_{\mathcal{H}}.$$

# Stochastic integration

- There is a *natural and simple* procedure to define stochastic integrals with respect to  $W_t(g)$ .
- The space of square-integrable processes taking values in the Hilbert space  $\mathcal{H}$  is denoted by  $L^2(\Omega \times [0, T]; \mathcal{H})$ .
- Let  $(e_n)_{n \geq 1}$  be a complete orthonormal system of  $\mathcal{H}$ .

## Definition

Let  $Y = \{Y(s, y), (s, y) \in [0, T] \times \mathbb{R}^d\}$  be a predictable process in  $L^2(\Omega \times [0, T]; \mathcal{H})$ . The **stochastic integral** of  $Y$  with respect to the cylindrical Wiener process  $W_t(g)$  is defined by

$$(Y \cdot W)_t := \sum_{n=1}^{\infty} \underbrace{\int_0^t \langle Y(s), e_n \rangle_{\mathcal{H}} dW_s(e_n)}_{\text{Itô integral}}$$

One proves that the series converges in  $L^2(\Omega)$  and the limit does not depend on the chosen CONS.

# Remarks

- We also use the notation:

$$\int_0^t \int_{\mathbb{R}^d} Y(s, y) W(ds, dy) := (Y \cdot W)_t, \quad t \in [0, T],$$

- $\{(Y \cdot W)_t, \mathcal{F}_t, t \in [0, T]\}$  is a square-integrable **martingale**.
- **Isometry property**: let  $\mathcal{M}$  be the space square-integrable and continuous martingales  $X = \{X_t, t \in [0, T]\}$  endowed with the norm  $\|X\| := E(|X_T|^2)$ . Then, the integral defines an isometry

$$L^2(\Omega \times [0, T]; \mathcal{H}) \longrightarrow \mathcal{M}$$

That is:

$$E \left( \int_0^T \|Y(s)\|_{\mathcal{H}}^2 ds \right) = E \left( \left| \int_0^T \int_{\mathbb{R}^d} Y(s, y) W(ds, dy) \right|^2 \right).$$

$$(Y \cdot W)_t := \sum_{n=1}^{\infty} \int_0^t \langle Y(s), e_n \rangle_{\mathcal{H}} dW_s(e_n)$$

- This integral could have been defined following the *usual* procedure: simple processes...
- Let  $\{W_t(A), \mathcal{F}_t, t \in [0, T], A \in \mathcal{B}_0(\mathbb{R}^d)\}$  be the **martingale measure** associated to  $\dot{W}$ . Using Walsh integral, recall that we can integrate any **function-valued** process  $Y(s, y)$  such that

$$E \left( \int_0^T ds \int_{\mathbb{R}^d} dx \int_{\mathbb{R}^d} dy |Y(s, x)| f(x - y) |Y(s, y)| \right) < +\infty. \quad (5)$$

### Proposition (Nualart and QS 2007, Dalang and QS 2010)

If  $Y$  is a predictable process satisfying (5), then  $Y$  belongs to  $L^2(\Omega \times [0, T]; \mathcal{H})$  and the stochastic integrals coincide.

We have an analogous result for Dalang's extension of Walsh stochastic integral.

- From now on, all the stochastic integrals will be considered with respect to the **cylindrical Wiener process**  $\{W_t(g), t \in [0, T], g \in \mathcal{H}\}$  associated to the noise  $\dot{W}$ .
- Recall that we aim to give sense to integrals of the form

$$\int_0^t \int_{\mathbb{R}^d} S(s, y) Z(s, y) W(ds, dy),$$

where  $S(s)$  may not be a function and  $Z$  is some real-valued process.

- For example, in the case of the stochastic wave equation in  $\mathbb{R}^3$ :

$$S(s) = \frac{1}{4\pi(t-s)} \sigma_{t-s} \quad Z(s, y) = \sigma(u(s, y))$$

and  $\sigma_t$  is a **measure**.

# Examples of integrands

- (H) Let  $t \mapsto S(t)$  be a deterministic function such that, for all  $t$ ,  $S(t)$  is a **non-negative distribution with rapid decrease** such that

$$\int_0^T dt \int_{\mathbb{R}^d} \mu(d\xi) |\mathcal{F}S(t)(\xi)|^2 < +\infty \quad \sup_{t \in [0, T]} S(t, \mathbb{R}^d) < +\infty.$$

- $S(t)$  non-negative distribution with rapid decrease: non-negative measure on  $\mathbb{R}^d$  such that

$$\int_{\mathbb{R}^d} (1 + |x|^2)^k S(t, dx) < +\infty \quad \text{for all } k \in \mathbb{N}.$$

- The fundamental solutions of the heat and wave equations in  $d \geq 1$  and  $d \in \{1, 2, 3\}$ , respectively, satisfy these assumptions and

$$\int_0^T dt \int_{\mathbb{R}^d} \mu(d\xi) |\mathcal{F}S(t)(\xi)|^2 < +\infty \iff \int_{\mathbb{R}^d} \frac{1}{1 + |\xi|^2} \mu(d\xi) < +\infty.$$

## Proposition (Nualart and QS 2007)

- Assume that  $S$  satisfies condition **(H)**.
- Let  $Z = \{Z(t, x), (t, x) \in [0, T] \times \mathbb{R}^d\}$  be a predictable process such that

$$\sup_{(t,x) \in [0,T] \times \mathbb{R}^d} E(|Z(t, x)|^2) < \infty.$$

Then, the random element  $G = G(t, dx) = Z(t, x)S(t, dx)$  defines a predictable process in  $L^2(\Omega \times [0, T]; \mathcal{H})$ . Moreover, we have that

$$\begin{aligned} E \left( \int_0^T \|G(s)\|_{\mathcal{H}}^2 ds \right) &= E \left( \int_0^T ds \int_{\mathbb{R}^d} \mu(d\xi) |\mathcal{F}(G(s)Z(s))(\xi)|^2 \right) \\ &= E(|G \cdot W|^2) \leq \int_0^T ds \sup_{y \in \mathbb{R}^d} E(|Z(s, y)|^2) \int_{\mathbb{R}^d} \mu(d\xi) |\mathcal{F}S(s)(\xi)|^2. \end{aligned}$$

We have used the notation  $G \cdot W := (G \cdot W)_T$ .

# Remarks

- As already pointed out, this proposition let us define the **stochastic integral** of  $G = G(t, dx) = Z(t, x)S(t, dx)$  with respect to the **cylindrical Wiener process** associated to the noise  $W$ :

$$G \cdot W = \int_0^T \int_{\mathbb{R}^d} S(s, y)Z(s, y)W(ds, dy) \quad (\text{formal notation})$$

- If we have that, for some  $p \geq 2$ :

$$\sup_{(t,x) \in [0,T] \times \mathbb{R}^d} E(|Z(t, x)|^p) < +\infty,$$

then

$$E(|G \cdot W|^p) \leq \int_0^T ds \sup_{y \in \mathbb{R}^d} E(|Z(s, y)|^p) \int_{\mathbb{R}^d} \mu(d\xi) |\mathcal{F}S(s)(\xi)|^2.$$

# Proof of the proposition

**Step 1.** Prove that  $S$  belongs to  $L^2([0, T]; \mathcal{H})$ :

- Let  $\psi \in C^\infty(\mathbb{R}^d)$  be non-negative, with compact support contained in  $B_1(0)$ , and such that  $\int_{\mathbb{R}^d} \psi(x) dx = 1$ . For  $n \geq 1$ , we set  $\psi_n(x) := n^d \psi(nx)$ . Then  $(\psi_n)_n$  defines an **approximation of the identity**.

- We smooth  $S$ :

$$S_n(t) := S(t) * \psi_n \in \mathcal{S}(\mathbb{R}^d)$$
$$|\mathcal{F}S_n(t)(\xi)| \leq |\mathcal{F}\psi_n(\xi)| |\mathcal{F}S(t)(\xi)| \leq |\mathcal{F}S(t)(\xi)|.$$

- We have that  $\{S_n, n \geq 1\} \subset L^2([0, T]; \mathcal{H})$  and we check  $S_n \rightarrow S$  in  $L^2([0, T]; \mathcal{H})$ .

**Step 2.** Prove that  $G = G(t, dx) = Z(t, x)S(t, dx)$  belongs  $L^2([0, T] \times \Omega; \mathcal{H})$ :

- We can assume that  $Z$  is non-negative.
- We prove that,  $\mathbb{P}$ -a.s.,  $G = Z(t, x)S(t, dx)$  defines a non-negative distribution with rapid decrease:

$$\int_{\mathbb{R}^d} (1 + |x|^2)^{\frac{k}{2}} Z(t, y) S(t, dy) < +\infty, \quad \text{for all } k \in \mathbb{N}, \mathbb{P}\text{-a.s.}$$

- We regularize  $G$  also by means of an approximation of the identity.



# Back to SPDEs

We are in position to state an **existence and uniqueness theorem** for a general class of SPDEs, including heat and wave equations (the latter with  $d \in \{1, 2, 3\}$ ):

$$Lu(t, x) = b(u(t, x)) + \sigma(u(t, x))\dot{W}(t, x), \quad t > 0, \quad x \in \mathbb{R}^d$$

where  $L$  is a **partial differential operator**.

**Initial conditions:**

- **Parabolic case:**

$$u(0, x) = u_0(x), \quad x \in \mathbb{R}^d.$$

- **Hyperbolic case:**

$$u(0, x) = u_0(x), \quad \frac{\partial u}{\partial t}(0, x) = v_0(x), \quad x \in \mathbb{R}^d.$$

The functions  $b, \sigma$  are Lipschitz and  $\dot{W}$  is the spatially homogeneous noise.

## Mild solution

By definition, the **mild solution** is an adapted process  $\{u(t, x), (t, x) \in [0, T] \times \mathbb{R}^d\}$  satisfying

$$u(t, x) = l_0(t, x) + \int_0^t \int_{\mathbb{R}^d} S_{t-s}(x - y) \sigma(u(s, y)) W(ds, dy) \\ + \int_0^t \int_{\mathbb{R}^d} S_{t-s}(x - y) b(u(s, y)) dy ds, \quad \mathbb{P}\text{-a.s.}$$

The stochastic integral is with respect to the **cylindrical Wiener process** obtained from  $\dot{W}$ ,  $S_t(x)$  is the **fundamental solution** associated to the operator  $L$  and the term  $l_0(t, x)$  is the contribution of the **initial conditions**:

- **Parabolic case:**

$$l_0(t, x) = \int_{\mathbb{R}^d} u_0(x - y) S(t, dy),$$

- **Hyperbolic case:**

$$l_0(t, x) = \int_{\mathbb{R}^d} v_0(x - y) S(t, dy) + \frac{\partial}{\partial t} \left( \int_{\mathbb{R}^d} u_0(x - y) S(t, dy) \right).$$

## Heat equation:

$$l_0(t, x) = \int_{\mathbb{R}^d} u_0(x - y) (4\pi t)^{-d/2} e^{-\frac{|y|^2}{4t}} dy, \quad x \in \mathbb{R}^d.$$

## Wave equation:

- For  $d = 1$ ,  $l_0(t, x)$  is given by the so-called d'Alembert's formula

$$l_0(t, x) = \frac{1}{2} [u_0(x + t) + u_0(x - t)] + \frac{1}{2} \int_{x-t}^{x+t} v_0(y) dy, \quad x \in \mathbb{R}.$$

- For  $d = 2$ :

$$l_0(t, x) = \frac{1}{2\pi t} \int_{|x-y|<t} \frac{u_0(y + tv_0) + \nabla u_0(y) \cdot (x - y)}{(t^2 - |x - y|^2)^{1/2}} dy, \quad x \in \mathbb{R}^2.$$

- For  $d = 3$ ,  $x \in \mathbb{R}^3$ :

$$l_0(t, x) = \frac{1}{4\pi t^2} \int_{\mathbb{R}^3} (tv_0(x - y) + u_0(x - y) + \nabla u_0(x - y) \cdot y) \sigma_t(dy).$$

## Lemma (Dalang and QS 2010)

Consider the following two sets of hypotheses:

- **Heat equation:**  $u_0 : \mathbb{R}^d \rightarrow \mathbb{R}$  is measurable and bounded.
- **Wave equation:** When  $d = 1$ ,  $u_0$  is bounded and continuous, and  $v_0$  is bounded and measurable. When  $d = 2$ ,  $u_0 \in C^1(\mathbb{R}^2)$  and there is  $q \in ]2, \infty]$  such that  $u_0, \nabla u_0, v_0$  all belong to  $L^q(\mathbb{R}^2)$ . When  $d = 3$ ,  $u_0 \in C^1(\mathbb{R}^3)$ ,  $u_0$  and  $\nabla u_0$  are bounded, and  $v_0$  is bounded and continuous.

Then  $(t, x) \mapsto I_0(t, x)$  is continuous and

$$\sup_{(t,x) \in [0,T] \times \mathbb{R}^d} |I_0(t, x)| < +\infty.$$

# Main result

$$u(t, x) = l_0(t, x) + \int_0^t \int_{\mathbb{R}^d} S_{t-s}(x - y) \sigma(u(s, y)) W(ds, dy) \\ + \int_0^t \int_{\mathbb{R}^d} S_{t-s}(x - y) b(u(s, y)) dy ds, \quad \mathbb{P}\text{-a.s.} \quad (6)$$

## Theorem (Dalang 1999, Dalang and QS 2010)

Assume that, for all  $t$ ,  $S(t)$  is a *non-negative distribution with rapid decrease* such that

$$\int_0^T dt \int_{\mathbb{R}^d} \mu(d\xi) |\mathcal{F}S(t)(\xi)|^2 < +\infty \quad \sup_{t \in [0, T]} S(t, \mathbb{R}^d) < +\infty,$$

$l_0$  is continuous and uniformly bounded, and  $b$  and  $\sigma$  are Lipschitz functions. Then there exists a unique solution  $\{u(t, x), (t, x) \in [0, T] \times \mathbb{R}^d\}$  of equation (6). Moreover, the process  $u$  is  $L^2$ -continuous and for all  $p \geq 1$ ,

$$\sup_{(t, x) \in [0, T] \times \mathbb{R}^d} E(|u(t, x)|^p) < +\infty.$$

# Remarks

- The proof is based on a Picard iteration scheme.
- The fundamental solution of the heat and wave equations in any space dimension and with  $d \in \{1, 2, 3\}$ , respectively, satisfy the assumptions of the theorem provided that

$$\int_0^T dt \int_{\mathbb{R}^d} \mu(d\xi) |\mathcal{FS}(t)(\xi)|^2 < +\infty \Leftrightarrow \int_{\mathbb{R}^d} \frac{1}{1 + |\xi|^2} \mu(d\xi) < +\infty.$$

- This result is also valid for the **damped wave equation** and a class of **parabolic equations with time-dependent coefficients**.
- We are not able to go beyond  $d = 3$  in the stochastic wave equation, because the fundamental solution becomes more irregular and it is no more non-negative definite:

$$\Gamma_t^d = C \left( \frac{1}{t} \frac{\partial}{\partial t} \right)^{(d-3)/2} \frac{\sigma_t}{t}, \quad d \geq 3 \quad \text{odd},$$

$$\Gamma_t^d(x) = C \left( \frac{1}{t} \frac{\partial}{\partial t} \right)^{(d-3)/2} (t^2 - |x|^2)_+^{-1/2}, \quad d \geq 2 \quad \text{even}.$$

## Even more extensions

- **Conus and Dalang** 2008 have extended the stochastic integral with respect to the martingale measure associated to the spatially homogeneous noise, in order to be able to integrate random elements of the form  $S(s, y)Z(s, y)$ , where
- $S$  is a function in  $[0, T]$  with values in some space of distributions, **without the restriction of being non-negative definite**.
- With this integral, they prove existence and uniqueness of solution for the stochastic wave equation in any space dimension.
- But they are not able to prove that the solution has moments of order  $p > 2$ .

→ SPDEs with spatially homogeneous noise in an **infinite dimensional setting**: works by Da Prato and Zabczyk, Karczewska and Zabczyk, Peszat and Zabczyk.

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