

# The Stochastic heat and wave equation

with irregular noise coefficients: well-posedness, long time behavior and small mass limit

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- ▶ "*On quasi-linear stochastic partial differential equations*", Gyöngy and Pardoux, 1993.

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- ▶ Strong uniqueness: imposing Sobolev regularity conditions on  $\sigma$  and use PDE theory.
- ▶ None of them works in infinite dimensions.

# Notion of solutions to stochastic PDEs

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- ▶ Evolution equations on Hilbert spaces: Da Prato-Zabczyk.  
Consider orthogonal basis  $(e_n(x))_{n \in \mathbb{N}}$ ,

$$W(dxdt) = \sum_{n=1}^{\infty} e_n(x) dB_t^{(n)} dx.$$

$$N(t, x) = \sum_{n=1}^{\infty} \int_0^t \int_{\mathbb{R}^d} S(t-s, x-y) g(s, y) e_n(y) dy dB_s^{(n)}.$$

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- ▶ Rough path theory, regularity structure, paracontrolled calculus.

$$\partial_t \Phi = \Delta \Phi + C\Phi - \Phi^3 + \xi$$

Only for additive noise or sufficiently regular noise coefficient.



# Known results for the stochastic Heat equation

- ▶ Additive noise case  $\partial_t u = \Delta u + f(u) + \frac{\partial^2 W}{\partial_t \partial x}$ .
  - ▶ Hilbert space valued SPDEs: strong solution for Hölder continuous  $f(u)$ . Strong solution for a.e. initial value for bounded measurable  $f$ . "*Strong uniqueness for SDEs in Hilbert spaces with nonregular drift.*"

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  - ▶ Random field SPDEs: strong solutions for distributional  $f(u)$ . "*Well-posedness of stochastic heat equation with distributional drift and skew stochastic heat equation.*"
- ▶ Multiplicative, Hölder continuous noise coefficient:  $\sigma$  being  $\frac{3}{4} + \epsilon$ -Hölder continuous in  $X(t, x)$ , the random field case:

$$\frac{\partial}{\partial t} X(t, x) = \frac{1}{2} \Delta X(t, x) dt + \sigma(t, x, X(t, x)) dW(t, x)$$

*"Pathwise Uniqueness for Stochastic Heat Equations with Hölder Continuous Coefficients: the White Noise Case", 77 p".*

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- ▶  $B$  has linear growth.

# Well-Posedness

## Theorem (Well-posedness of Stochastic Heat equation)

*Under the assumptions in the previous slide, there exists a unique (probabilistic weak) mild solution to*

$$dX_t = AX_t dt + B(X_t)dt + \sigma(X_t)dW_t, \quad X_0 \in H. \quad (4)$$

*Given  $\frac{1}{2} + \epsilon$ -Hölder  $F : H \rightarrow H$ , unique weak-mild solution to*

$$dX_t = AX_t dt + (-A)^{1/2}F(X_t)dt + B(X_t)dt + \sigma(X_t)dW_t, \quad (5)$$

- ▶ Examples: Burgers type equations,  $\xi \in (0, 2\pi)$

$$du(t, \xi) = \frac{\partial^2}{\partial \xi^2} u(t, \xi) dt + \frac{\partial}{\partial \xi} h(u(t, \xi)) dt + \sigma(u(t, \xi)) dW_t(\xi).$$

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- ▶ Cahn-Hilliard equations in dimensions 1,2,3:

$$du(t, \xi) = -\Delta_\xi^2 u(t, \xi)dt + \Delta_\xi h(u(t, \xi))dt + \sigma(u(t, \xi))dW_t(\xi)$$

# Long-time behaviour

## Theorem (Exponential ergodicity)

*Assume the drift  $B : H \rightarrow H$  is Hölder continuous, and the Lyapunov condition hold: for some  $V : H \rightarrow \mathbb{R}_+$  and some  $\lambda \in (0, 1)$  infinity at infinity,*

$$\mathbb{E}[V(X_t)] \leq \lambda V(X_0) + M \quad (6)$$

*for some given  $t > 0$  and  $M > 0$ . Then there exists a unique invariant measure, and the solution converges to the invariant measure exponentially fast with respect to (some specific) Wasserstein distance on  $\mathcal{P}(H)$ .*

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- ▶ No applicable Itô formula for cylindrical noise
- ▶ Lyapunov assumption satisfied when  $B, F, \sigma$  are bounded, and  $A$  is a negative operator.

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- ▶ In order to compare  $X_t$  to a process

$$dX_t^n = AX_t^n dt + B^n(X_t^n)dt + \sigma^n(X_t^n)dW_t,$$

consider auxiliary process with  $\lambda > 0$  and stopping time  $\tau$

$$d\tilde{X}_t^n = A\tilde{X}_t^n dt + B^n(\tilde{X}_t^n)dt + \lambda 1_{t \leq \tau} dt + \sigma^n(\tilde{X}_t^n)dW_t,$$



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- ▶ Use pathwise estimate to compare  $X|_{[0, T]}$  with  $\tilde{X}^n|_{[0, T]}$ .

Some technical challenges in infinite dimensions:

- ▶ Derive maximal inequality for the process, when  $\lambda$  is large:

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- ▶ Lipschitz approximation in infinite dimensions: compactness of heat semigroup.

# Stochastic wave equation: well-posedness

Our method works not only for the parabolic systems, but also for hyperbolic systems.

Consider the (abstract) damped stochastic wave equation

$$\mu \frac{\partial^2 u_\mu(t)}{\partial t^2} = Au_\mu(t) - \frac{\partial u_\mu(t)}{\partial t} + B(t, u_\mu(t)) + G(t, u_\mu(t))dW_t,$$

and the stochastic wave equation without damping term

$$\mu \frac{\partial^2 u_\mu(t)}{\partial t^2} = Au_\mu(t) + B(t, u_\mu(t)) + G(t, u_\mu(t))dW_t,$$

## Theorem (Well-posedness of stochastic wave equation)

*Under the same assumption on  $A$ ,  $B$  and  $G$  as in the case of the stochastic heat equation, there exists a unique weak-mild solution to both these equations.*

# Stochastic wave equation: small mass limit

## Theorem

Assume moreover that  $B$  is Hölder continuous in  $u_\mu$ . Then as  $\mu$  tends to 0, the solution to the damped stochastic wave equation

$$\mu \frac{\partial^2 u_\mu(t)}{\partial t^2} = Au_\mu(t) - \frac{\partial u_\mu(t)}{\partial t} + B(t, u_\mu(t)) + G(t, u_\mu(t))dW_t$$

converges in distribution on path space to the solution of the stochastic heat equation

$$\frac{\partial u_\mu(t)}{\partial t} = Au_\mu(t) + B(t, u_\mu(t)) + G(t, u_\mu(t))dW_t.$$

*"On the Smoluchowski-Kramers approximation for a system with an infinite number of degrees of freedom", Freidlin and Cerrai, 2006.*



# Thanks