

Spatial ergodicity and central limit theorem for stochastic heat equation

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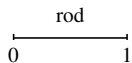
September 3, 2020

Outline

- Introduction
 - ▶ Stochastic Heat Equation (SHE)
 - ▶ Poincaré inequality
- Spatial ergodicity for SHE (\Rightarrow law of large numbers)
- Spatial central limit theorem for SHE

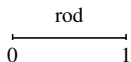
Stochastic heat equation

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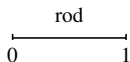
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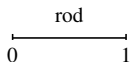
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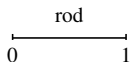
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- Add random external heat density $W(t, x)$ at (t, x) ,

$$\partial_t u(t, x) = \frac{1}{2} \partial_x^2 u(t, x) + W(t, x). \quad (\text{additive})$$

- **Space-time white noise:** $W(t, x)$, $t > 0, x \in [0, 1]$, independent centered Gaussian random variables.

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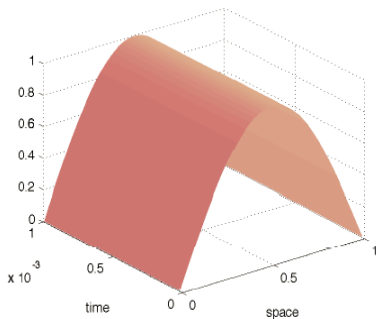
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- Interaction.

$$\partial_t u(t, x) = \frac{1}{2} \partial_x^2 u(t, x) + \sigma(u(t, x)) W(t, x). \quad (\text{multiplicative})$$

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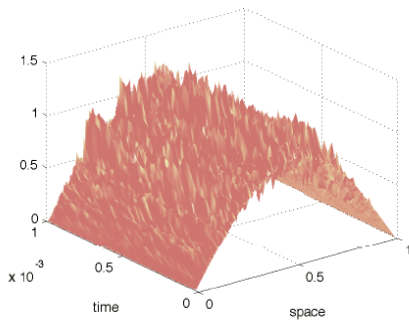
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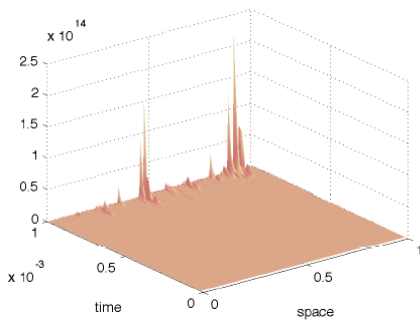
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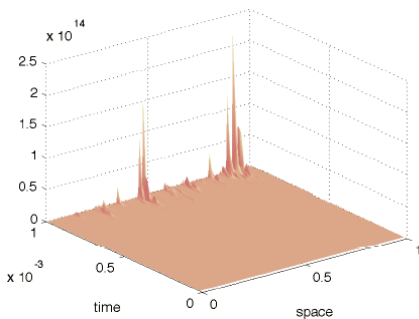
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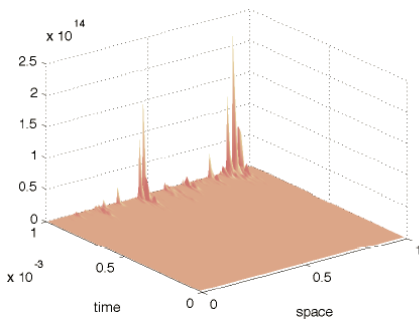


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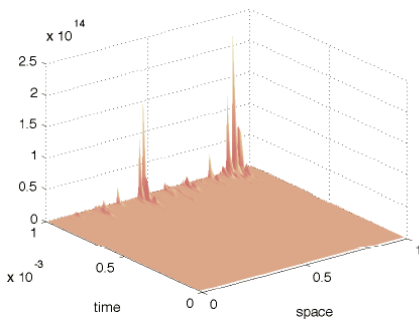


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- High peaks; others are small. What is average? [Law of large numbers](#).
- Normalization of average. [Central limit theorem](#).
- Number of peaks higher than a certain level. [Poisson limit theorem](#).

Stochastic heat equation

- Stochastic heat equation

$$\begin{cases} \partial_t u(t, x) = \frac{1}{2} \Delta u(t, x) + \sigma(u(t, x)) W(t, x), & t > 0, x \in \mathbb{R}^d, \\ u(0) \equiv 1 \end{cases} \quad (\text{SHE})$$

- $\sigma : \mathbb{R} \mapsto \mathbb{R}$ Lipschitz continuous.

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- $\sigma : \mathbb{R} \mapsto \mathbb{R}$ Lipschitz continuous.
- White noise $\{W(t, x) : (t, x) \in \mathbb{R}_+ \times \mathbb{R}^d\}$: Gaussian process with covariance

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

where f is a nonnegative-definite measure on \mathbb{R}^d .

- Space-time white noise: $f = \delta_0$.

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- Space-time white noise: $f = \delta_0$.
- Mild solution.

$$u(t, x) = 1 + \int_0^t \int_{\mathbb{R}^d} \mathbf{p}_{t-s}(x - y) \sigma(u(s, y)) W(s, y) ds dy,$$

where $\mathbf{p}_t(x) = \frac{1}{(2\pi t)^{d/2}} e^{-\frac{\|x\|^2}{2t}}$, $t > 0, x \in \mathbb{R}^d$.

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- $W(ds dy)$ denotes Walsh / Itô integral (Walsh 86).

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- Existence and uniqueness:

$$\int_{\mathbb{R}^d} \frac{\hat{f}(dx)}{1 + \|x\|^2} < \infty, \quad (\text{Dalang's condition 99})$$

where \hat{f} denotes the Fourier transform of f .

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 - ▶ Cole-Hopf solution to KPZ equation: $H(t, x) := \log u(t, x)$, then formally

$$\partial_t H(t, x) = \frac{1}{2} \partial_x^2 H(t, x) + \frac{1}{2} (\partial_x H(t, x))^2 + W(t, x). \quad (\text{KPZ})$$

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How does f determine the spatial ergodicity and central limit theorem?

Poincaré inequality

- $d\gamma(x) = (2\pi)^{-n/2} e^{-\frac{\|x\|^2}{2}}$ Gaussian probability measure on \mathbb{R}^n . $F : \mathbb{R}^n \mapsto \mathbb{R}$ smooth enough,

$$\text{Var}_\gamma(F) := \int_{\mathbb{R}^n} F^2 d\gamma - \left(\int_{\mathbb{R}^n} F d\gamma \right)^2 \leq \int_{\mathbb{R}^n} |\nabla F|^2 d\gamma.$$

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- $\Omega = C(\mathbb{R}_+)$, Wiener measure \mathbb{W} , Brownian motion: $B_t(\omega) = \omega_t, \omega \in \Omega$.

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Fix $t > 0$ and g Lipschitz continuous, and suppose $f(\mathbb{R}^d) < \infty$. Then,

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- ▶ Malliavin-Stein method (Nourdin, Peccati):

If $F = \delta(v)$ and $\mathbb{E}[F^2] = 1$,

$$d_{\text{TV}}(F, \mathbf{N}(0, 1)) \leq 2\sqrt{\text{Var}(\langle DF, v \rangle_{\mathcal{H}})}.$$

Total variation distance: $d_{\text{TV}}(F, G) := \sup_{A \in \mathcal{B}(\mathbb{R})} |\mathbb{P}(F \in A) - \mathbb{P}(G \in A)|$

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Choose $a, b \in \mathbb{R}$,

$$\left| \mathbb{E} \left[e^{i(a(X_N(r_2) - X_N(r_1)) + b(X_N(r_4) - X_N(r_3)))} \right] - \mathbb{E} \left[e^{ia(X_N(r_2) - X_N(r_1))} \right] \mathbb{E} \left[e^{ib(X_N(r_4) - X_N(r_3))} \right] \right| \rightarrow 0,$$

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- [3] Chen, L., Khoshnevisan, D., Nualart, D. and Pu, F: Central limit theorems for spatial averages of the stochastic heat equation via Malliavin-Stein's method. arXiv:2008.02408 (2020)