

Sticky-reflected stochastic heat equation driven by colored noise

Vitalii Konarovskiy

Bielefeld University &
Institute of Mathematics of NAS of Ukraine

Malliavin Calculus and its Applications – 2021



Sticky-reflected stochastic heat equation

Sticky-reflected stochastic heat equation on $[0, 1]$

$$\frac{\partial X_t}{\partial t} = \frac{1}{2} \frac{\partial^2 X_t}{\partial u^2} + \lambda \mathbb{I}_{\{X_t=0\}} + \mathbb{I}_{\{X_t>0\}} \dot{W}_t$$

$$X_0 = g \geq 0, \quad X_t(0) = X_t(1) = 0$$

where $\lambda > 0$

It is similar to the SDE for sticky-reflected Brownian motion:

$$\begin{aligned} dx(t) &= \lambda \mathbb{I}_{\{x(t)=0\}} dt + \mathbb{I}_{\{x(t)>0\}} dw(t), \\ x(0) &= x_0 \geq 0 \end{aligned}$$

– only weak existence and uniqueness in law! (Engelbert and Peskir, 2014)

Sticky-reflected stochastic heat equation

Sticky-reflected stochastic heat equation on $[0, 1]$

$$\frac{\partial X_t}{\partial t} = \frac{1}{2} \frac{\partial^2 X_t}{\partial u^2} + \lambda \mathbb{I}_{\{X_t=0\}} + \mathbb{I}_{\{X_t>0\}} \dot{W}_t$$

$$X_0 = g \geq 0, \quad X_t(0) = X_t(1) = 0$$

where $\lambda > 0$

It is similar to the SDE for sticky-reflected Brownian motion:

$$\begin{aligned} dx(t) &= \lambda \mathbb{I}_{\{x(t)=0\}} dt + \mathbb{I}_{\{x(t)>0\}} dw(t), \\ x(0) &= x_0 \geq 0 \end{aligned}$$

– only weak existence and uniqueness in law! (Engelbert and Peskir, 2014)

Reason of investigation

- Sticky-reflected SHE vs. Reflected SHE (Nulart and Pardoux, 1992)

$$\frac{\partial X_t}{\partial t} = \frac{1}{2} \frac{\partial^2 X_t}{\partial u^2} + \delta_0(X_t) + \dot{W}_t$$

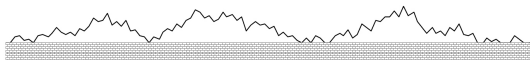
- Possible connection with wetting dynamics (Deuschel, Giacomin, Zambotti, 2004)
- A new method of solving SDEs with discontinuous coefficients.

Reason of investigation

- Sticky-reflected SHE vs. Reflected SHE (Nulart and Pardoux, 1992)

$$\frac{\partial X_t}{\partial t} = \frac{1}{2} \frac{\partial^2 X_t}{\partial u^2} + \delta_0(X_t) + \dot{W}_t$$

- Possible connection with wetting dynamics (Deuschel, Giacomin, Zambotti, 2004)



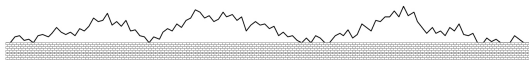
- A new method of solving SDEs with discontinuous coefficients.

Reason of investigation

- Sticky-reflected SHE vs. Reflected SHE (Nulart and Pardoux, 1992)

$$\frac{\partial X_t}{\partial t} = \frac{1}{2} \frac{\partial^2 X_t}{\partial u^2} + \delta_0(X_t) + \dot{W}_t$$

- Possible connection with wetting dynamics (Deuschel, Giacomin, Zambotti, 2004)



- A new method of solving SDEs with discontinuous coefficients.

Formulation of the main result

Sticky-reflected SHE:

$$\begin{aligned}\frac{\partial X_t}{\partial t} &= \frac{1}{2} \frac{\partial^2 X_t}{\partial u^2} + \lambda \mathbb{I}_{\{X_t=0\}} + \mathbb{I}_{\{X_t>0\}} \dot{W}_t \\ X_0 &= g \geq 0, \quad X_t(0) = X_t(1) = 0,\end{aligned}$$

where Q is non-negative definite self-adjoint Hilbert-Schmidt operator in $L_2[0, 1]$

Solution to sticky-reflected SHE

A continuous process $X : [0, \infty) \times [0, 1] \rightarrow \mathbb{R}$ is called a **weak solution** to the sticky-reflected SHE if for any $\varphi \in C^2[0, 1]$ with $\varphi(0) = \varphi(1) = 0$

$$M_t^\varphi := \langle X_t, \varphi \rangle - \langle X_0, \varphi \rangle - \frac{1}{2} \int_0^t \langle X_s, \varphi'' \rangle ds - \int_0^t \langle \lambda \mathbb{I}_{\{X_s=0\}}, \varphi \rangle ds$$

is a martingale with quadratic variation

$$[M^\varphi]_t = \int_0^t \|Q(\mathbb{I}_{\{X_s>0\}}\varphi)\|_{L_2}^2 ds.$$

Formulation of the main result

Sticky-reflected SHE:

$$\begin{aligned}\frac{\partial X_t}{\partial t} &= \frac{1}{2} \frac{\partial^2 X_t}{\partial u^2} + \lambda \mathbb{I}_{\{X_t=0\}} + \mathbb{I}_{\{X_t>0\}} Q \dot{W}_t \\ X_0 &= g \geq 0, \quad X_t(0) = X_t(1) = 0,\end{aligned}$$

where Q is non-negative definite self-adjoint Hilbert-Schmidt operator in $L_2[0, 1]$

Solution to sticky-reflected SHE

A continuous process $X : [0, \infty) \times [0, 1] \rightarrow \mathbb{R}$ is called a **weak solution** to the sticky-reflected SHE if for any $\varphi \in C^2[0, 1]$ with $\varphi(0) = \varphi(1) = 0$

$$M_t^\varphi := \langle X_t, \varphi \rangle - \langle X_0, \varphi \rangle - \frac{1}{2} \int_0^t \langle X_s, \varphi'' \rangle ds - \int_0^t \langle \lambda \mathbb{I}_{\{X_s=0\}}, \varphi \rangle ds$$

is a martingale with quadratic variation

$$[M^\varphi]_t = \int_0^t \|Q(\mathbb{I}_{\{X_s>0\}}\varphi)\|_{L_2}^2 ds.$$

Formulation of the main result

Sticky-reflected SHE:

$$\begin{aligned}\frac{\partial X_t}{\partial t} &= \frac{1}{2} \frac{\partial^2 X_t}{\partial u^2} + \lambda \mathbb{I}_{\{X_t=0\}} + \mathbb{I}_{\{X_t>0\}} Q \dot{W}_t \\ X_0 &= g \geq 0, \quad X_t(0) = X_t(1) = 0,\end{aligned}$$

where Q is non-negative definite self-adjoint Hilbert-Schmidt operator in $L_2[0, 1]$

Solution to sticky-reflected SHE

A continuous process $X : [0, \infty) \times [0, 1] \rightarrow \mathbb{R}$ is called a **weak solution** to the sticky-reflected SHE if for any $\varphi \in C^2[0, 1]$ with $\varphi(0) = \varphi(1) = 0$

$$M_t^\varphi := \langle X_t, \varphi \rangle - \langle X_0, \varphi \rangle - \frac{1}{2} \int_0^t \langle X_s, \varphi'' \rangle ds - \int_0^t \langle \lambda \mathbb{I}_{\{X_s=0\}}, \varphi \rangle ds$$

is a martingale with quadratic variation

$$[M^\varphi]_t = \int_0^t \|Q(\mathbb{I}_{\{X_s>0\}}\varphi)\|_{L_2}^2 ds.$$

Formulation of the main result

Let $\{e_k, k \geq 1\}$ and $\{\mu_k, k \geq 1\}$ be eigenvectors and eigenvalues of Q . Define

$$\chi^2 := \sum_{k=1}^{\infty} \mu_k^2 e_k^2.$$

Theorem K. 2021

If $\chi^2 > 0$ a.e., then the sticky-reflected SHE admits a weak solution.

Formulation of the main result

Let $\{e_k, k \geq 1\}$ and $\{\mu_k, k \geq 1\}$ be eigenvectors and eigenvalues of Q . Define

$$\chi^2 := \sum_{k=1}^{\infty} \mu_k^2 e_k^2.$$

Theorem K. 2021

If $\chi^2 > 0$ a.e., then the sticky-reflected SHE admits a weak solution.

Meaning of assumption $\chi^2 > 0$

The equation

$$\begin{aligned} dx(t) &= \lambda \mathbb{I}_{\{x(t)=0\}} dt + \mathbb{I}_{\{x(t)>0\}} dw(t) \\ x(0) &= 0 \end{aligned}$$

has **no solution**

$\chi^2 = \sum_{k=1}^{\infty} \mu_k^2 e_k^2 > 0$ means that the solution X to the equation

$$\frac{\partial X_t}{\partial t} = \frac{1}{2} \frac{\partial^2 X_t}{\partial u^2} + \lambda \mathbb{I}_{\{X_t=0\}} + \mathbb{I}_{\{X_t>0\}} Q \dot{W}_t$$

feels a noise at **any** point of $[0, 1]$.

Meaning of assumption $\chi^2 > 0$

The equation

$$\begin{aligned} dx(t) &= \lambda \mathbb{I}_{\{x(t)=0\}} dt + \mathbb{I}_{\{x(t)>0\}} dw(t) \\ x(0) &= 0 \end{aligned}$$

has ~~no solution~~ a weak solution

$\chi^2 = \sum_{k=1}^{\infty} \mu_k^2 e_k^2 > 0$ means that the solution X to the equation

$$\frac{\partial X_t}{\partial t} = \frac{1}{2} \frac{\partial^2 X_t}{\partial u^2} + \lambda \mathbb{I}_{\{X_t=0\}} + \mathbb{I}_{\{X_t>0\}} Q \dot{W}_t$$

feels a noise at any point of $[0, 1]$.

Meaning of assumption $\chi^2 > 0$

The equation

$$\begin{aligned} dx(t) &= \lambda \mathbb{I}_{\{x(t)=0\}} dt + \mathbb{I}_{\{x(t)>0\}} dw(t) \\ x(0) &= 0 \end{aligned}$$

has ~~no solution~~ a weak solution

$\chi^2 = \sum_{k=1}^{\infty} \mu_k^2 e_k^2 > 0$ means that the solution X to the equation

$$\frac{\partial X_t}{\partial t} = \frac{1}{2} \frac{\partial^2 X_t}{\partial u^2} + \lambda \mathbb{I}_{\{X_t=0\}} + \mathbb{I}_{\{X_t>0\}} Q \dot{W}_t$$

feels a noise at **any** point of $[0, 1]$.

Description of the idea of construction of solution

using the equation

$$dx(t) = \lambda \mathbb{I}_{\{x(t)=0\}} dt + \mathbb{I}_{\{x(t)>0\}} dw(t)$$

Approximating sequence

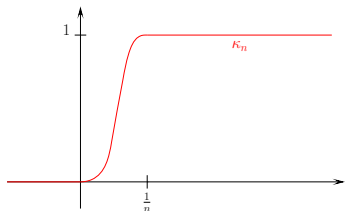
Consider the SDE for sticky-reflected BM:

$$\begin{aligned}dx(t) &= \lambda \mathbb{I}_{\{x(t)=0\}} dt + \mathbb{I}_{\{x(t)>0\}} dw(t), \\x(0) &= x_0 \geq 0.\end{aligned}$$

We approximate its solution by the solutions to the SDEs

$$\begin{aligned}dx_n(t) &= \lambda(1 - \kappa_n^2(x_n(t)))dt + \kappa_n(x_n(t))dw(t), \\x_n(0) &= x_0.\end{aligned}$$

which have non-negative strong solutions $x_n(t) \geq 0$.



$$\begin{aligned}\kappa_n(y) &\rightarrow \mathbb{I}_{\{y>0\}}, \\1 - \kappa_n^2(y) &\rightarrow 1 - \mathbb{I}_{\{y>0\}}^2 = \mathbb{I}_{\{y=0\}}\end{aligned}$$

for $y \geq 0$, as $n \rightarrow \infty$.

Approximating sequence

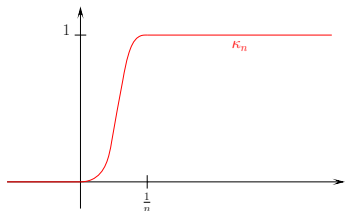
Consider the SDE for sticky-reflected BM:

$$\begin{aligned}dx(t) &= \lambda \mathbb{I}_{\{x(t)=0\}} dt + \mathbb{I}_{\{x(t)>0\}} dw(t), \\x(0) &= x_0 \geq 0.\end{aligned}$$

We approximate its solution by the solutions to the SDEs

$$\begin{aligned}dx_n(t) &= \lambda(1 - \kappa_n^2(x_n(t))) dt + \kappa_n(x_n(t)) dw(t), \\x_n(0) &= x_0.\end{aligned}$$

which have non-negative strong solutions $x_n(t) \geq 0$.



$$\begin{aligned}\kappa_n(y) &\rightarrow \mathbb{I}_{\{y>0\}}, \\1 - \kappa_n^2(y) &\rightarrow 1 - \mathbb{I}_{\{y>0\}}^2 = \mathbb{I}_{\{y=0\}}\end{aligned}$$

for $y \geq 0$, as $n \rightarrow \infty$.

Problem of approximation

One can show that $\{x_n, n \geq 1\}$ is tight in $C[0, T]$ and $\{\kappa_n^2(x_n), n \geq 1\}$ is tight in $L^2[0, T]$ in weak topology

↓

$x_n \rightarrow x$ in $C[0, T]$ and $\kappa_n^2(x_n) \rightarrow \rho^2$ in $L_2[0, T]$ in weak t.

along a subsequence.

But

$$M_n(t) := x_n(t) - x_0 + \int_0^t \lambda(1 - \kappa_n^2(x_n(s))) ds$$

↓

↓

$$M(t) := x(t) - x_0 + \int_0^t \lambda \mathbb{I}_{\{x(s)=0\}} ds$$

$$[M_n]_t = \int_0^t \kappa_n^2(x_n(s)) ds \not\rightarrow \int_0^t \mathbb{I}_{\{x(s)>0\}} ds$$

Why $\rho^2(s) = \mathbb{I}_{\{x(s)>0\}}$?

Problem of approximation

One can show that $\{x_n, n \geq 1\}$ is tight in $C[0, T]$ and $\{\kappa_n^2(x_n), n \geq 1\}$ is tight in $L^2[0, T]$ in weak topology

↓

$x_n \rightarrow x$ in $C[0, T]$ and $\kappa_n^2(x_n) \rightarrow \rho^2$ in $L_2[0, T]$ in weak t.

along a subsequence.

But

$$M_n(t) := x_n(t) - x_0 + \int_0^t \lambda(1 - \kappa_n^2(x_n(s))) ds$$

↓

↓

$$M(t) := x(t) - x_0 + \int_0^t \lambda \mathbb{I}_{\{x(s)=0\}} ds$$

$$[M_n]_t = \int_0^t \kappa_n^2(x_n(s)) ds \not\rightarrow \int_0^t \mathbb{I}_{\{x(s)>0\}} ds$$

Why $\rho^2(s) = \mathbb{I}_{\{x(s)>0\}}$?

Problem of approximation

One can show that $\{x_n, n \geq 1\}$ is tight in $C[0, T]$ and $\{\kappa_n^2(x_n), n \geq 1\}$ is tight in $L^2[0, T]$ in weak topology

↓

$x_n \rightarrow x$ in $C[0, T]$ and $\kappa_n^2(x_n) \rightarrow \rho^2$ in $L_2[0, T]$ in weak t.

along a subsequence.

But

$$M_n(t) := x_n(t) - x_0 + \int_0^t \lambda(1 - \kappa_n^2(x_n(s))) ds$$

↓

↓

↓

$$M(t) := x(t) - x_0 + \int_0^t \lambda(1 - \rho^2(s)) ds$$

$$[M_n]_t = \int_0^t \kappa_n^2(x_n(s)) ds \rightarrow \int_0^t \rho^2(s) ds = [M]_t$$

Why $\rho^2(s) = \mathbb{I}_{\{x(s) > 0\}}$?

Problem of approximation

One can show that $\{x_n, n \geq 1\}$ is tight in $C[0, T]$ and $\{\kappa_n^2(x_n), n \geq 1\}$ is tight in $L^2[0, T]$ in weak topology

↓

$x_n \rightarrow x$ in $C[0, T]$ and $\kappa_n^2(x_n) \rightarrow \rho^2$ in $L_2[0, T]$ in weak t.

along a subsequence.

But

$$M_n(t) := x_n(t) - x_0 + \int_0^t \lambda(1 - \kappa_n^2(x_n(s))) ds$$

↓ ↓ ↓

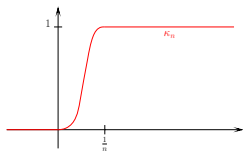
$$M(t) := x(t) - x_0 + \int_0^t \lambda(1 - \rho^2(s)) ds$$

$$[M_n]_t = \int_0^t \kappa_n^2(x_n(s)) ds \rightarrow \int_0^t \rho^2(s) ds = [M]_t$$

Why $\rho^2(s) = \mathbb{I}_{\{x(s) > 0\}}$?

Two observations

- $\kappa_n^2(y_n) \not\rightarrow \mathbb{I}_{\{y>0\}}$ as $y_n \rightarrow y$.



- If x is a continuous non-negative semimartingale with q.v.

$$[x]_t = \int_0^t \sigma^2(s) ds,$$

then $[x]_t = \int_0^t \mathbb{I}_{\{x(s)>0\}} \sigma^2(s) ds$.

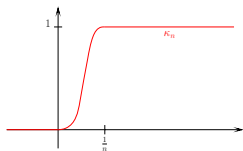
Proof.

$$\begin{aligned} \int_0^t \sigma^2(s) \mathbb{I}_{\{x(s)=0\}} ds &= \int_0^t \mathbb{I}_{\{0\}}(x(s)) d[x]_s \\ &= \int_{-\infty}^{+\infty} \mathbb{I}_{\{0\}}(y) L_t^y dx = 0, \end{aligned}$$

where L_t^y is the local time of x at y .

Two observations

- $\mathbb{I}_{\{y>0\}} \kappa_n^2(y_n) \rightarrow \mathbb{I}_{\{y>0\}}$ as $y_n \rightarrow y$.



- If x is a continuous non-negative semimartingale with q.v.

$$[x]_t = \int_0^t \sigma^2(s) ds,$$

then $[x]_t = \int_0^t \mathbb{I}_{\{x(s)>0\}} \sigma^2(s) ds$.

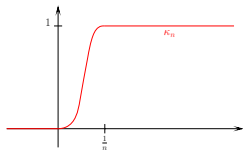
Proof.

$$\begin{aligned} \int_0^t \sigma^2(s) \mathbb{I}_{\{x(s)=0\}} ds &= \int_0^t \mathbb{I}_{\{0\}}(x(s)) d[x]_s \\ &= \int_{-\infty}^{+\infty} \mathbb{I}_{\{0\}}(y) L_t^y dx = 0, \end{aligned}$$

where L_t^y is the local time of x at y .

Two observations

- $\mathbb{I}_{\{y>0\}} \kappa_n^2(y_n) \rightarrow \mathbb{I}_{\{y>0\}}$ as $y_n \rightarrow y$.



- If x is a continuous non-negative semimartingale with q.v.

$$[x]_t = \int_0^t \sigma^2(s) ds,$$

then $[x]_t = \int_0^t \mathbb{I}_{\{x(s)>0\}} \sigma^2(s) ds$.

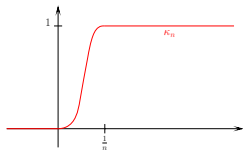
Proof.

$$\begin{aligned} \int_0^t \sigma^2(s) \mathbb{I}_{\{x(s)=0\}} ds &= \int_0^t \mathbb{I}_{\{0\}}(x(s)) d[x]_s \\ &= \int_{-\infty}^{+\infty} \mathbb{I}_{\{0\}}(y) L_t^y dx = 0, \end{aligned}$$

where L_t^y is the local time of x at y .

Two observations

- $\mathbb{I}_{\{y>0\}} \kappa_n^2(y_n) \rightarrow \mathbb{I}_{\{y>0\}}$ as $y_n \rightarrow y$.



- If x is a continuous non-negative semimartingale with q.v.

$$[x]_t = \int_0^t \sigma^2(s) ds,$$

then $[x]_t = \int_0^t \mathbb{I}_{\{x(s)>0\}} \sigma^2(s) ds$.

Proof.

$$\begin{aligned} \int_0^t \sigma^2(s) \mathbb{I}_{\{x(s)=0\}} ds &= \int_0^t \mathbb{I}_{\{0\}}(x(s)) d[x]_s \\ &= \int_{-\infty}^{+\infty} \mathbb{I}_{\{0\}}(y) L_t^y dx = 0, \end{aligned}$$

where L_t^y is the local time of x at y .

Identification of quadratic variation

Remind

$$M_n(t) := x_n(t) - x_0 + \int_0^t \lambda(1 - \kappa_n^2(x_n(s)))ds$$

$$\downarrow \quad \downarrow \quad \downarrow$$

$$M(t) := x(t) - x_0 + \int_0^t \lambda(1 - \rho^2(s))ds$$

$$[M_n]_t = \int_0^t \kappa_n^2(x_n(s))ds \rightarrow \int_0^t \rho^2(s)ds = [M]_t$$

Therefore,

$$\begin{aligned} [x]_t &= [M]_t = \int_0^t \rho^2(s)ds = \int_0^t \mathbb{I}_{\{x(s)>0\}} \rho^2(s)ds \\ &= \lim_n \int_0^t \mathbb{I}_{\{x(s)>0\}} \kappa_n^2(x_n(s))ds = \int_0^t \mathbb{I}_{\{x(s)>0\}} ds \end{aligned}$$

Identification of quadratic variation

Remind

$$M_n(t) := x_n(t) - x_0 + \int_0^t \lambda(1 - \kappa_n^2(x_n(s)))ds$$

↓ ↓ ↓

$$M(t) := x(t) - x_0 + \int_0^t \lambda(1 - \rho^2(s))ds$$

$$[M_n]_t = \int_0^t \kappa_n^2(x_n(s))ds \rightarrow \int_0^t \rho^2(s)ds = [M]_t$$

Therefore,

$$\begin{aligned} [x]_t &= [M]_t = \int_0^t \rho^2(s)ds = \int_0^t \mathbb{I}_{\{x(s)>0\}} \rho^2(s)ds \\ &= \lim_n \int_0^t \mathbb{I}_{\{x(s)>0\}} \kappa_n^2(x_n(s))ds = \int_0^t \mathbb{I}_{\{x(s)>0\}} ds \end{aligned}$$

Proof of existence of solution to
sticky-reflected SHE

$$\frac{\partial X_t}{\partial t} = \frac{1}{2} \frac{\partial^2 X_t}{\partial u^2} + \lambda \mathbb{I}_{\{X_t=0\}} + \mathbb{I}_{\{X_t>0\}} Q \dot{W}_t$$

Discrete equation

We discretize only the space variable $u \in [0, 1]$ by $\frac{k}{n}$, $k = 1, \dots, n$.

Set $\pi_k^n = \mathbb{I}_{[\frac{k-1}{n}, \frac{k}{n})}$ and define

$$w_k^n(t) := \sqrt{n} \int_0^t \int_0^1 (Q\pi_k^n)(u) W(du, ds)$$

Consider the following SDE

$$dx_k(t) = \frac{1}{2} \Delta^n x_k(t) dt + \mathbb{I}_{\{x_k(t)=0\}} dt + \sqrt{n} \mathbb{I}_{\{x_k(t)>0\}} dw_k^n(t), \quad k = 1, \dots, n,$$

$$\text{with } x_0(t) = x_{n+1}(t) = 0 \quad \text{and} \quad \Delta^n x_k = n^2 (x_{k+1} + x_{k-1} - 2x_k)$$

Tightness of linear approximation

Define

$$\tilde{X}_t^n(u) = (un - k + 1)x_k^n(t) + (k - nu)x_{k-1}^n(t), \quad t \in [0, T], \quad u \in \pi_k^n.$$

Then $\tilde{X}^n \geq 0$ and $X^n \in C([0, T], C[0, 1])$ a.s.

Theorem. Funaki '83

The family of processes $\{\tilde{X}^n, n \geq 1\}$ is tight in $C([0, T], C[0, 1])$

Idea of Proof.

$$\begin{aligned} \tilde{X}_t^n(u) &= \int_0^1 p^n(t, u, v)g(v)dv + \lambda \int_0^t \int_0^1 p^n(t-s, u, v)\mathbb{I}_{\{\tilde{X}_s^n(\lceil v \rceil)=0\}} dsdv \\ &\quad + \int_0^t \int_0^1 p^n(t-s, u, v)\mathbb{I}_{\{\tilde{X}_s^n(\lceil v \rceil)>0\}} QdW_s du \end{aligned}$$

By properties of the discrete heat semigroup p^n , for every $\gamma > 0$, $T > 0$,

$$\mathbb{E} \left[(\tilde{X}_t^n(u))^\gamma \right] \leq C \quad \text{and} \quad \mathbb{E} \left[|\tilde{X}_t^n(u) - \tilde{X}_s^n(v)|^{2\gamma} \right] \leq C \left(|t-s|^{\frac{\gamma}{2}} + |u-v|^{\frac{\gamma}{2}} \right)$$

Tightness of linear approximation

Define

$$\tilde{X}_t^n(u) = (un - k + 1)x_k^n(t) + (k - nu)x_{k-1}^n(t), \quad t \in [0, T], \quad u \in \pi_k^n.$$

Then $\tilde{X}^n \geq 0$ and $X^n \in C([0, T], C[0, 1])$ a.s.

Theorem. Funaki '83

The family of processes $\{\tilde{X}^n, n \geq 1\}$ is tight in $C([0, T], C[0, 1])$

Idea of Proof.

$$\begin{aligned} \tilde{X}_t^n(u) &= \int_0^1 p^n(t, u, v)g(v)dv + \lambda \int_0^t \int_0^1 p^n(t-s, u, v)\mathbb{I}_{\{\tilde{X}_s^n(\lceil v \rceil)=0\}} dsdv \\ &\quad + \int_0^t \int_0^1 p^n(t-s, u, v)\mathbb{I}_{\{\tilde{X}_s^n(\lceil v \rceil)>0\}} QdW_s du \end{aligned}$$

By properties of the discrete heat semigroup p^n , for every $\gamma > 0$, $T > 0$,

$$\mathbb{E} \left[(\tilde{X}_t^n(u))^\gamma \right] \leq C \quad \text{and} \quad \mathbb{E} \left[|\tilde{X}_t^n(u) - \tilde{X}_s^n(v)|^{2\gamma} \right] \leq C \left(|t-s|^{\frac{\gamma}{2}} + |u-v|^{\frac{\gamma}{2}} \right)$$

Tightness of linear approximation

Define

$$\tilde{X}_t^n(u) = (un - k + 1)x_k^n(t) + (k - nu)x_{k-1}^n(t), \quad t \in [0, T], \quad u \in \pi_k^n.$$

Then $\tilde{X}^n \geq 0$ and $X^n \in C([0, T], C[0, 1])$ a.s.

Theorem. Funaki '83

The family of processes $\{\tilde{X}^n, n \geq 1\}$ is tight in $C([0, T], C[0, 1])$

Idea of Proof.

$$\begin{aligned} \tilde{X}_t^n(u) &= \int_0^1 p^n(t, u, v)g(v)dv + \lambda \int_0^t \int_0^1 p^n(t-s, u, v)\mathbb{I}_{\{\tilde{X}_s^n(\lceil v \rceil)=0\}} dsdv \\ &\quad + \int_0^t \int_0^1 p^n(t-s, u, v)\mathbb{I}_{\{\tilde{X}_s^n(\lceil v \rceil)>0\}} QdW_s du \end{aligned}$$

By properties of the discrete heat semigroup p^n , for every $\gamma > 0$, $T > 0$,

$$\mathbb{E} \left[(\tilde{X}_t^n(u))^\gamma \right] \leq C \quad \text{and} \quad \mathbb{E} \left[|\tilde{X}_t^n(u) - \tilde{X}_s^n(v)|^{2\gamma} \right] \leq C \left(|t-s|^{\frac{\gamma}{2}} + |u-v|^{\frac{\gamma}{2}} \right)$$

Martingale problem for X^n

Set

$$X_t^n(u) = x_k(t), \quad \frac{k-1}{n} \leq u < \frac{k}{n}, \quad u \in [0, 1].$$

Remark that $X^n \geq 0$.

For every $\varphi \in C^2[0, 1]$ with $\varphi(0) = \varphi(1) = 0$,

$$M_t^{n,\varphi} := \langle X_t^n, \varphi \rangle - \langle X_0^n, \varphi \rangle - \frac{1}{2} \int_0^t \langle X_s^n, \tilde{\Delta}^n \varphi \rangle ds - \lambda \int_0^t \langle \mathbb{I}_{\{X_s^n=0\}}, \varphi \rangle ds$$

is a martingale with q.v.

$$[M^{n,\varphi}]_t = \int_0^t \|Q(\mathbb{I}_{\{X_s^n > 0\}} \varphi)\|^2 ds$$

Martingale problem for X^n

Set

$$X_t^n(u) = x_k(t), \quad \frac{k-1}{n} \leq u < \frac{k}{n}, \quad u \in [0, 1].$$

Remark that $X^n \geq 0$.

For every $\varphi \in C^2[0, 1]$ with $\varphi(0) = \varphi(1) = 0$,

$$M_t^{n,\varphi} := \langle X_t^n, \varphi \rangle - \langle X_0^n, \varphi \rangle - \frac{1}{2} \int_0^t \langle X_s^n, \tilde{\Delta}^n \varphi \rangle ds - \lambda \int_0^t \langle \mathbb{I}_{\{X_s^n=0\}}, \varphi \rangle ds$$

is a martingale with q.v.

$$[M^{n,\varphi}]_t = \int_0^t \|Q(\mathbb{I}_{\{X_s^n > 0\}} \varphi)\|^2 ds$$

Martingale problem for X^n

Set

$$X_t^n(u) = x_k(t), \quad \frac{k-1}{n} \leq u < \frac{k}{n}, \quad u \in [0, 1].$$

Remark that $X^n \geq 0$.

For every $\varphi \in \mathcal{C}^2[0, 1]$ with $\varphi(0) = \varphi(1) = 0$,

$$M_t^{n,\varphi} := \langle X_t^n, \varphi \rangle - \langle X_0^n, \varphi \rangle - \frac{1}{2} \int_0^t \langle X_s^n, \tilde{\Delta}^n \varphi \rangle ds - \lambda \int_0^t \langle \mathbb{I}_{\{X_s^n=0\}}, \varphi \rangle ds$$

is a martingale with q.v.

$$[M^{n,\varphi}]_t = \int_0^t \|Q(\mathbb{I}_{\{X_s^n > 0\}} \varphi)\|^2 ds$$

Convergence result

For every $\varphi \in \mathcal{C}^2[0, 1]$ with $\varphi(0) = \varphi(1) = 0$, along a subsequence

$$\begin{array}{ccccccccc} M_t^{n,\varphi} & := & \langle X_t^n, \varphi \rangle & - & \langle X_0^n, \varphi \rangle & - & \frac{1}{2} \int_0^t \langle X_s^n, \tilde{\Delta}^n \varphi \rangle ds & - & \lambda \int_0^t \langle \mathbb{I}_{\{X_s^n=0\}}, \varphi \rangle ds \\ \downarrow & & \downarrow & & \downarrow & & \downarrow & & \downarrow \\ M_t^\varphi & := & \langle X_t, \varphi \rangle & - & \langle X_0, \varphi \rangle & - & \frac{1}{2} \int_0^t \langle X_s, \varphi'' \rangle ds & - & \lambda \int_0^t \langle 1 - \sigma_s, \varphi \rangle ds \end{array}$$

is a continuous martingale with quadratic variation

$$[M^{n,\varphi}]_t = \int_0^t \|Q(\mathbb{I}_{\{X_s^n > 0\}} \varphi)\|^2 ds \rightarrow \int_0^t \|Q(\sigma_s \varphi)\|^2 ds = [M^\varphi]_t$$

Equivalently

$$\frac{\partial X_t}{\partial t} = \frac{1}{2} \frac{\partial^2 X_t}{\partial u^2} + \lambda(1 - \sigma_t) + \sigma_t Q \dot{W},$$

$$X_t(0) = X_t(1) = 0, \quad X_0(u) = g(u).$$

Convergence result

For every $\varphi \in \mathcal{C}^2[0, 1]$ with $\varphi(0) = \varphi(1) = 0$, along a subsequence

$$\begin{array}{ccccccc} M_t^{n,\varphi} := \langle X_t^n, \varphi \rangle - \langle X_0^n, \varphi \rangle - \frac{1}{2} \int_0^t \langle X_s^n, \tilde{\Delta}^n \varphi \rangle ds - \lambda \int_0^t \langle \mathbb{I}_{\{X_s^n=0\}}, \varphi \rangle ds \\ \downarrow \qquad \qquad \downarrow \qquad \qquad \downarrow \qquad \qquad \qquad \downarrow \qquad \qquad \qquad \qquad \downarrow \\ M_t^\varphi := \langle X_t, \varphi \rangle - \langle X_0, \varphi \rangle - \frac{1}{2} \int_0^t \langle X_s, \varphi'' \rangle ds - \lambda \int_0^t \langle 1 - \sigma_s, \varphi \rangle ds \end{array}$$

is a continuous martingale with quadratic variation

$$[M^{n,\varphi}]_t = \int_0^t \|Q(\mathbb{I}_{\{X_s^n > 0\}} \varphi)\|^2 ds \rightarrow \int_0^t \|Q(\sigma_s \varphi)\|^2 ds = [M^\varphi]_t$$

Equivalently

$$\frac{\partial X_t}{\partial t} = \frac{1}{2} \frac{\partial^2 X_t}{\partial u^2} + \lambda(1 - \sigma_t) + \sigma_t Q \dot{W},$$

$$X_t(0) = X_t(1) = 0, \quad X_0(u) = g(u).$$

Convergence result

For every $\varphi \in \mathcal{C}^2[0, 1]$ with $\varphi(0) = \varphi(1) = 0$, along a subsequence

$$\begin{array}{ccccccc} M_t^{n,\varphi} := \langle X_t^n, \varphi \rangle - \langle X_0^n, \varphi \rangle - \frac{1}{2} \int_0^t \langle X_s^n, \tilde{\Delta}^n \varphi \rangle ds - \lambda \int_0^t \langle \mathbb{I}_{\{X_s^n=0\}}, \varphi \rangle ds \\ \downarrow \qquad \qquad \downarrow \qquad \qquad \downarrow \qquad \qquad \qquad \downarrow \qquad \qquad \qquad \qquad \downarrow \\ M_t^\varphi := \langle X_t, \varphi \rangle - \langle X_0, \varphi \rangle - \frac{1}{2} \int_0^t \langle X_s, \varphi'' \rangle ds - \lambda \int_0^t \langle 1 - \sigma_s, \varphi \rangle ds \end{array}$$

is a continuous martingale with quadratic variation

$$[M^{n,\varphi}]_t = \int_0^t \|Q(\mathbb{I}_{\{X_s^n > 0\}} \varphi)\|^2 ds \rightarrow \int_0^t \|Q(\sigma_s \varphi)\|^2 ds = [M^\varphi]_t$$

Equivalently

$$\frac{\partial X_t}{\partial t} = \frac{1}{2} \frac{\partial^2 X_t}{\partial u^2} + \lambda(1 - \sigma_t) + \sigma_t Q \dot{W},$$

$$X_t(0) = X_t(1) = 0, \quad X_0(u) = g(u).$$

Identification of coefficient σ

Proposition (K., 2021)

Let X solves the equation

$$\frac{\partial X_t}{\partial t} = \frac{1}{2} \frac{\partial^2 X_t}{\partial u^2} + a_t + \sigma_t Q \dot{W},$$

$$X_t(0) = X_t(1) = 0, \quad X_0(u) = g(u)$$

and $X \geq 0$. Then $\sigma_t = \mathbb{I}_{\{X_t > 0\}} \sigma_t$.

Identification of the coefficients

We come back to our equation with undefined coefficients:

$$\frac{\partial X_t}{\partial t} = \frac{1}{2} \frac{\partial^2 X_t}{\partial u^2} + \lambda(1 - \sigma_t) + \sigma_t Q\dot{W},$$

$$X_t(0) = X_t(1) = 0, \quad X_0(u) = g(u), \quad X_t \geq 0.$$

By the previous proposition,

$$\sigma_t(u) = \mathbb{I}_{\{X_t(u) > 0\}} \sigma_t(u) = \lim_n \mathbb{I}_{\{X_t(u) > 0\}} \mathbb{I}_{\{X_t^n(u) > 0\}} = \mathbb{I}_{\{X_t(u) > 0\}}$$

Identification of the coefficients

We come back to our equation with undefined coefficients:

$$\frac{\partial X_t}{\partial t} = \frac{1}{2} \frac{\partial^2 X_t}{\partial u^2} + \lambda \mathbb{I}_{\{X_t=0\}} + \mathbb{I}_{\{X_t>0\}} Q \dot{W},$$

$$X_t(0) = X_t(1) = 0, \quad X_0(u) = g(u), \quad X_t \geq 0.$$

By the previous proposition,

$$\sigma_t(u) = \mathbb{I}_{\{X_t(u)>0\}} \sigma_t(u) = \lim_n \mathbb{I}_{\{X_t(u)>0\}} \mathbb{I}_{\{X_t^n(u)>0\}} = \mathbb{I}_{\{X_t(u)>0\}}$$

Idea of proof of the key proposition

Proposition

Let X solves the equation

$$\frac{\partial X_t}{\partial t} = \frac{1}{2} \frac{\partial^2 X_t}{\partial u^2} + a_t + \sigma_t Q \dot{W}_t,$$

$$X_t(0) = X_t(1) = 0, \quad X_0(u) = g(u).$$

and $X \geq 0$. Then a.s. $\sigma_t = \mathbb{I}_{\{X_t > 0\}} \sigma_t$ for t -a.e.

Ito's formula

Lemma. (Ito's formula)

Assume that $F \in C^2(\mathbb{R})$ has a bounded second derivative. Then

$$\begin{aligned}\langle F(X_t), 1 \rangle &= \langle F(X_0), 1 \rangle - \frac{1}{2} \int_0^t \langle F''(X_s) \dot{X}_s, \dot{X}_s \rangle ds + \int_0^t \langle F'(X_s), a_s \rangle ds \\ &\quad + \frac{1}{2} \int_0^t \langle Q[\sigma_s F''(X_s) \cdot], Q[\sigma_s \cdot] \rangle_{HS} ds + M_F(t),\end{aligned}$$

where

$$[M_F]_t = \int_0^t \|Q[\sigma_s F'(X_s)]\|^2 ds$$

and

$$\dot{X}_t = \sum_{k=1}^{\infty} \langle X_t, e_k \rangle e'_k, \quad e_k(u) = \sqrt{2} \sin \pi k u$$

Idea of Proof. Idea of proof: Apply usual Ito's formula to $\langle F(Z_t^n), 1 \rangle$ for $Z_t^n = \sum_{k=1}^n \langle X_t, e_k \rangle e_k$

Ito's formula

Lemma. (Ito's formula)

Assume that $F \in C^2(\mathbb{R})$ has a bounded second derivative. Then

$$\begin{aligned}\langle F(X_t), 1 \rangle &= \langle F(X_0), 1 \rangle - \frac{1}{2} \int_0^t \langle F''(X_s) \dot{X}_s, \dot{X}_s \rangle ds + \int_0^t \langle F'(X_s), a_s \rangle ds \\ &\quad + \frac{1}{2} \int_0^t \langle Q[\sigma_s F''(X_s) \cdot], Q[\sigma_s \cdot] \rangle_{HS} ds + M_F(t),\end{aligned}$$

where

$$[M_F]_t = \int_0^t \|Q[\sigma_s F'(X_s)]\|^2 ds$$

and

$$\dot{X}_t = \sum_{k=1}^{\infty} \langle X_t, e_k \rangle e'_k, \quad e_k(u) = \sqrt{2} \sin \pi k u$$

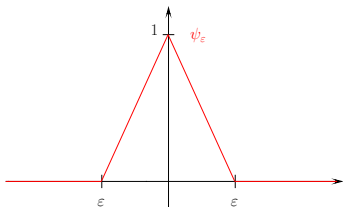
Idea of Proof. Idea of proof: Apply usual Ito's formula to $\langle F(Z_t^n), 1 \rangle$ for $Z_t^n = \sum_{k=1}^n \langle X_t, e_k \rangle e_k$

Proof of the key proposition

Take

$$F_\varepsilon(x) := \int_{-\infty}^x \int_{-\infty}^y \psi_\varepsilon(r) dy dr,$$

$$0 \leq F'_\varepsilon(x) \leq 2\varepsilon, \quad F''_\varepsilon(x) \rightarrow \mathbb{I}_{\{0\}}(x)$$



Apply Ito's formula to F_ε :

$$\begin{aligned} \langle F_\varepsilon(X_t) - F_\varepsilon(X_0), 1 \rangle &= -\frac{1}{2} \int_0^t \langle F''_\varepsilon(X_s) \dot{X}_s, \dot{X}_s \rangle ds + \int_0^t \langle F'_\varepsilon(X_s), a_s \rangle ds \\ &\quad + \frac{1}{2} \int_0^t \langle Q[\sigma_s F''_\varepsilon(X_s) \cdot], Q[\sigma_s \cdot] \rangle_{HS} ds + M_{F_\varepsilon}(t), \end{aligned}$$

Hence all green terms $\rightarrow 0$ and red term $\rightarrow \int_0^t \langle Q[\sigma_s \mathbb{I}_{\{X_s=0\}} \cdot], Q[\sigma_s \cdot] \rangle_{HS} ds$

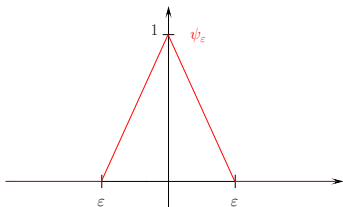
\implies We can replace σ_s by $\mathbb{I}_{\{X_s>0\}} \sigma_s$

Proof of the key proposition

Take

$$F_\varepsilon(x) := \int_{-\infty}^x \int_{-\infty}^y \psi_\varepsilon(r) dy dr,$$

$$0 \leq F'_\varepsilon(x) \leq 2\varepsilon, \quad F''_\varepsilon(x) \rightarrow \mathbb{I}_{\{0\}}(x)$$



Apply Ito's formula to F_ε :

$$\begin{aligned} \langle F_\varepsilon(X_t) - F_\varepsilon(X_0), 1 \rangle &= -\frac{1}{2} \int_0^t \langle F''_\varepsilon(X_s) \dot{X}_s, \dot{X}_s \rangle ds + \int_0^t \langle F'_\varepsilon(X_s), a_s \rangle ds \\ &\quad + \frac{1}{2} \int_0^t \langle Q[\sigma_s F''_\varepsilon(X_s) \cdot], Q[\sigma_s \cdot] \rangle_{HS} ds + M_{F_\varepsilon}(t), \end{aligned}$$

Hence all green terms $\rightarrow 0$ and red term $\rightarrow \int_0^t \langle Q[\sigma_s \mathbb{I}_{\{X_s=0\}} \cdot], Q[\sigma_s \cdot] \rangle_{HS} ds$

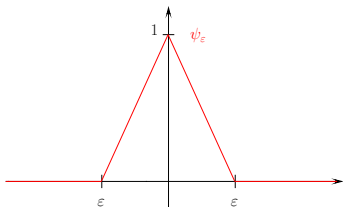
\implies We can replace σ_s by $\mathbb{I}_{\{X_s>0\}} \sigma_s$

Proof of the key proposition

Take

$$F_\varepsilon(x) := \int_{-\infty}^x \int_{-\infty}^y \psi_\varepsilon(r) dy dr,$$

$$0 \leq F'_\varepsilon(x) \leq 2\varepsilon, \quad F''_\varepsilon(x) \rightarrow \mathbb{I}_{\{0\}}(x)$$



Apply Ito's formula to F_ε :

$$\begin{aligned} \langle F_\varepsilon(X_t) - F_\varepsilon(X_0), 1 \rangle &= -\frac{1}{2} \int_0^t \langle F''_\varepsilon(X_s) \dot{X}_s, \dot{X}_s \rangle ds + \int_0^t \langle F'_\varepsilon(X_s), a_s \rangle ds \\ &\quad + \frac{1}{2} \int_0^t \langle Q[\sigma_s F''_\varepsilon(X_s) \cdot], Q[\sigma_s \cdot] \rangle_{HS} ds + M_{F_\varepsilon}(t), \end{aligned}$$

Hence all green terms $\rightarrow 0$ and red term $\rightarrow \int_0^t \langle Q[\sigma_s \mathbb{I}_{\{X_s=0\}} \cdot], Q[\sigma_s \cdot] \rangle_{HS} ds = 0$

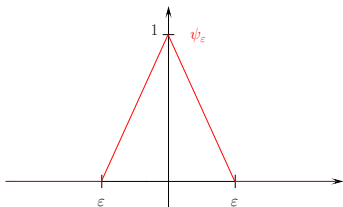
\implies We can replace σ_s by $\mathbb{I}_{\{X_s>0\}} \sigma_s$

Proof of the key proposition

Take

$$F_\varepsilon(x) := \int_{-\infty}^x \int_{-\infty}^y \psi_\varepsilon(r) dy dr,$$

$$0 \leq F'_\varepsilon(x) \leq 2\varepsilon, \quad F''_\varepsilon(x) \rightarrow \mathbb{I}_{\{0\}}(x)$$



Apply Ito's formula to F_ε :

$$\begin{aligned} \langle F_\varepsilon(X_t) - F_\varepsilon(X_0), 1 \rangle &= -\frac{1}{2} \int_0^t \langle F''_\varepsilon(X_s) \dot{X}_s, \dot{X}_s \rangle ds + \int_0^t \langle F'_\varepsilon(X_s), a_s \rangle ds \\ &\quad + \frac{1}{2} \int_0^t \langle Q[\sigma_s F''_\varepsilon(X_s) \cdot], Q[\sigma_s \cdot] \rangle_{HS} ds + M_{F_\varepsilon}(t), \end{aligned}$$

Hence all green terms $\rightarrow 0$ and red term $\rightarrow \int_0^t \langle Q[\sigma_s \mathbb{I}_{\{X_s=0\}} \cdot], Q[\sigma_s \cdot] \rangle_{HS} ds = 0$

\implies We can replace σ_s by $\mathbb{I}_{\{X_s>0\}} \sigma_s$

Open problem and references

Open problems:

- Is a solution to the equation unique?
- Does the solution of the equation with the identity operator Q exists?
- What is the invariant measure for the dynamics?
- How much time does the equation spend at zero?



Vitalii Konarovskiy,
Sticky-Reflected Stochastic Heat Equation Driven by Colored Noise
Ukrain. Math. J., Vol. 72, no. 9, 2021
(arXiv:2005.11773)



Vitalii Konarovskiy,
Coalescing-Fragmentating Wasserstein Dynamics: particle approach
(arXiv:1711.03011)

Thank you for your attention!