

Space-time fractional stochastic partial differential equations

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Outline

- Fractional Diffusion
- Time Fractional SPDEs
- Intermittency
- Intermittency Fronts
- SPDEs in bounded domains
- SPDEs with space-colored noise
- Related Work in the literature
- Future work/directions

Acknowledgement

Joint work with

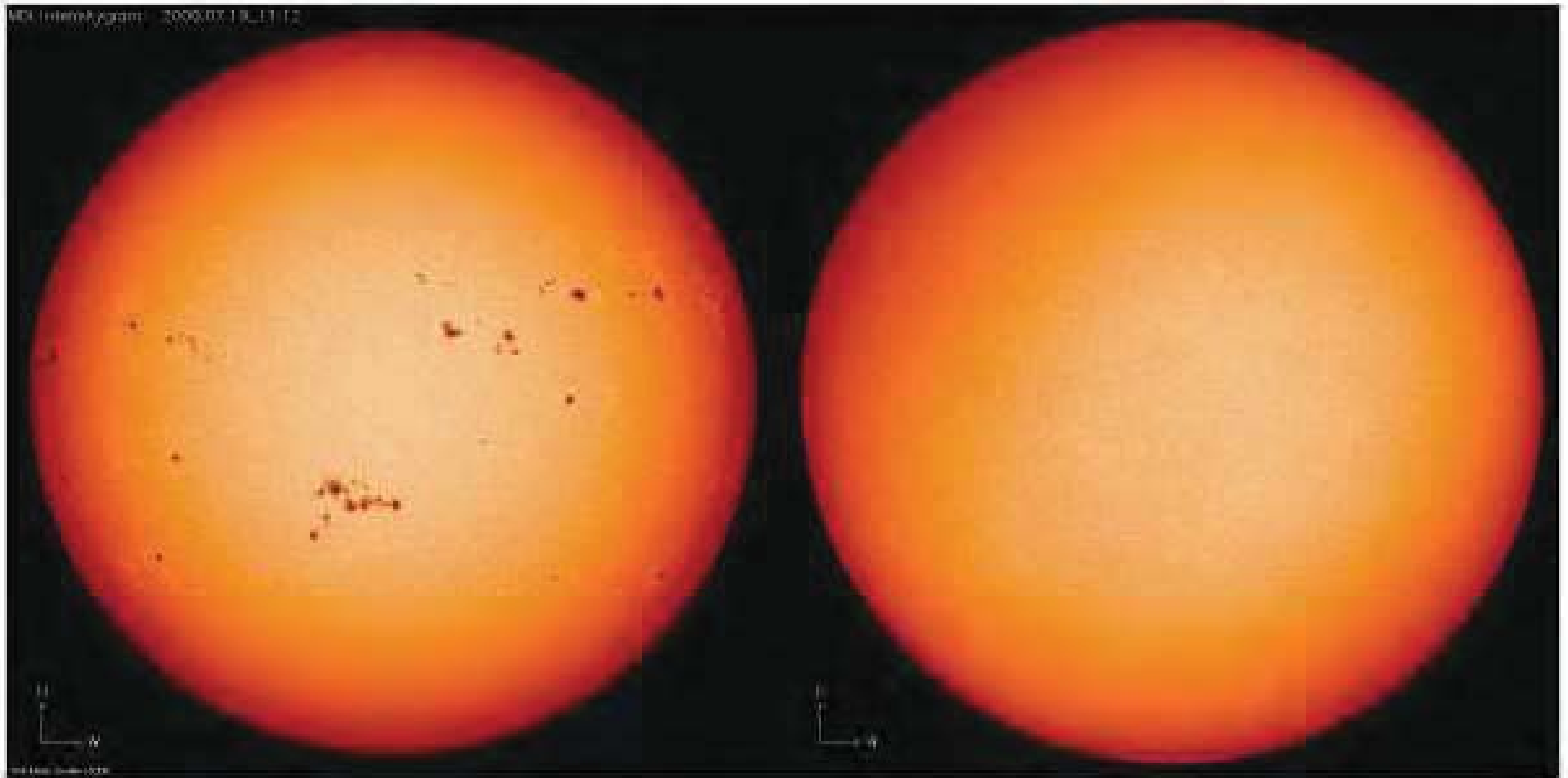
- Yimin Xiao, MSU
- Mark Meerschaert, MSU
- Boris Baeumer, University of Otago
- Jebessa Mijena, Georgia College and State University
- Mohammud Foondun, University of Strathclyde, Glasgow
- Current students: Sunday Asogwa, Ngartelbaye Guerngar, Auburn University

Pictures of the sun

URL: <https://www.youtube.com/watch?v=KL2FcyzagxM>

Nonlinear noise excitation (Lecture 1)

Is the Sun Missing Its Spots?



NASA

SUN GAZING These photos show sunspots near solar maximum on July 19, 2000, and near solar minimum on March 18, 2009. Some global warming skeptics speculate that the Sun may be on the verge of an extended slumber.

By KENNETH CHANG
Published: July 20, 2009

Nonlinear noise excitation (Lecture 1)

$\partial_t u = \frac{1}{2} \partial_x^2 u + \lambda u \xi$ on $[0, 1]$ with Dirichlet BC

$u_0(x) = \sin(\pi x)$ [K-Kim, 2013]

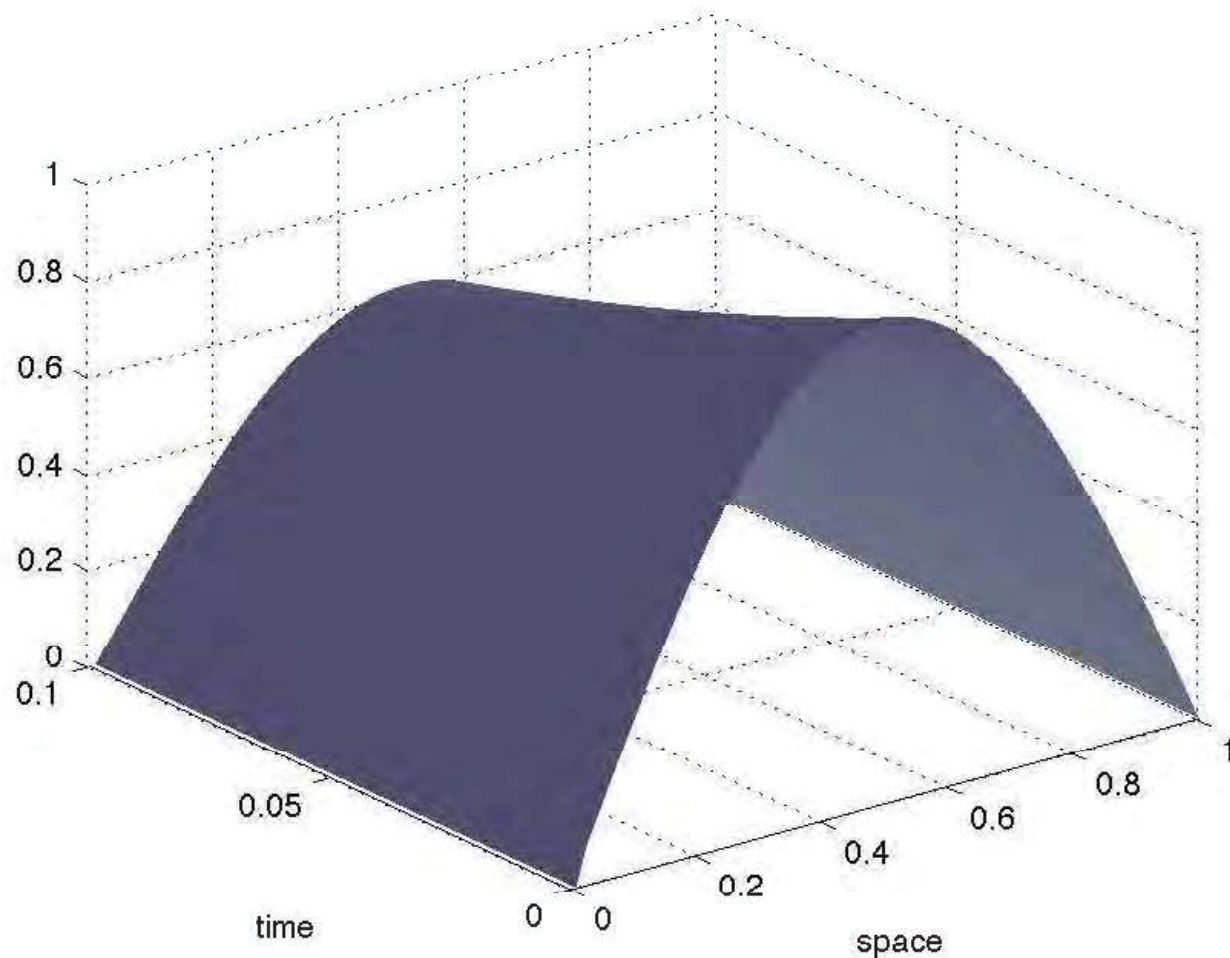


Figure: $\lambda = 0$; $u_t(x) = \sin(\pi x) \exp(-\pi^2 t/2)$

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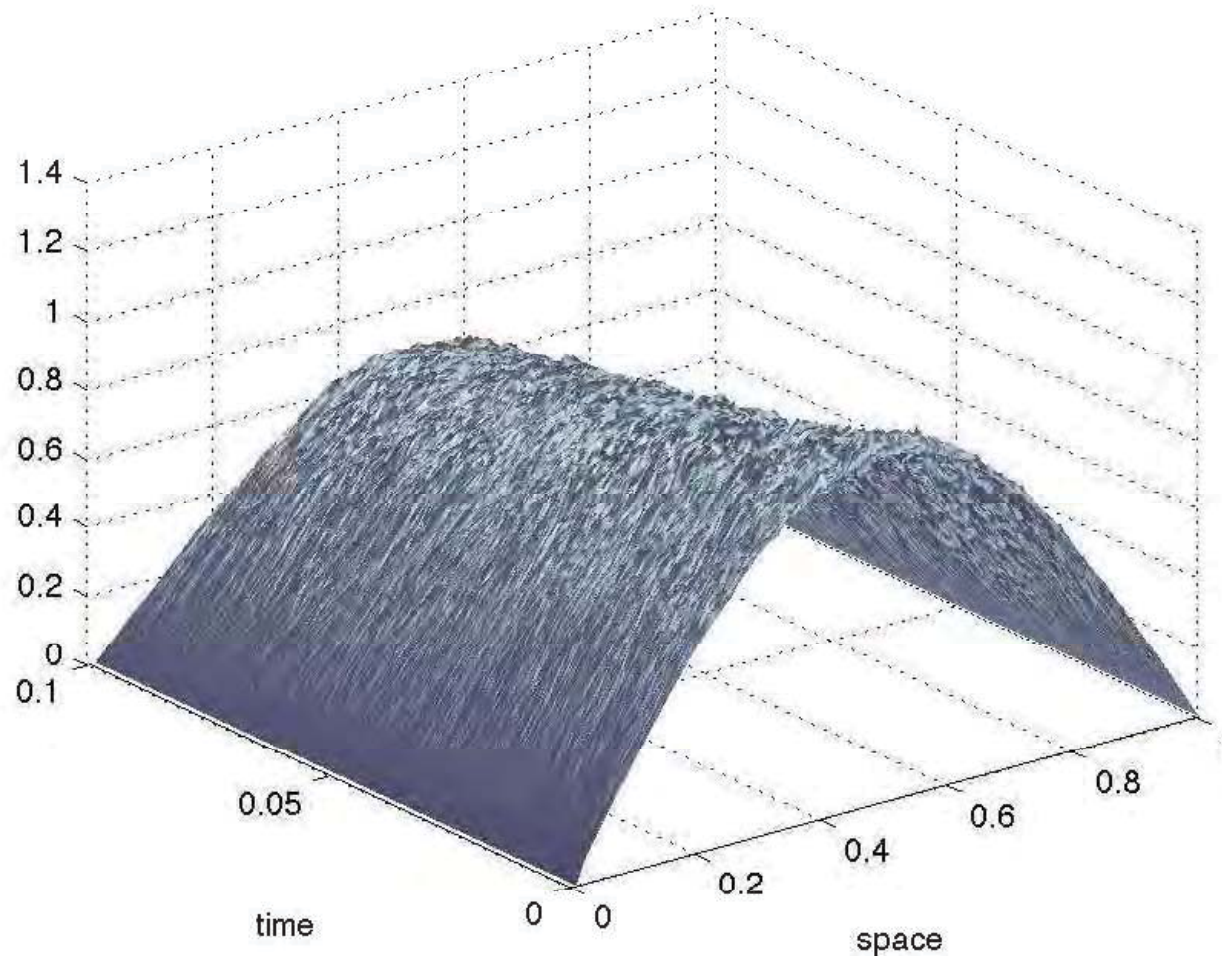


Figure: $\lambda = 0.1$; max. peak ≈ 1.4

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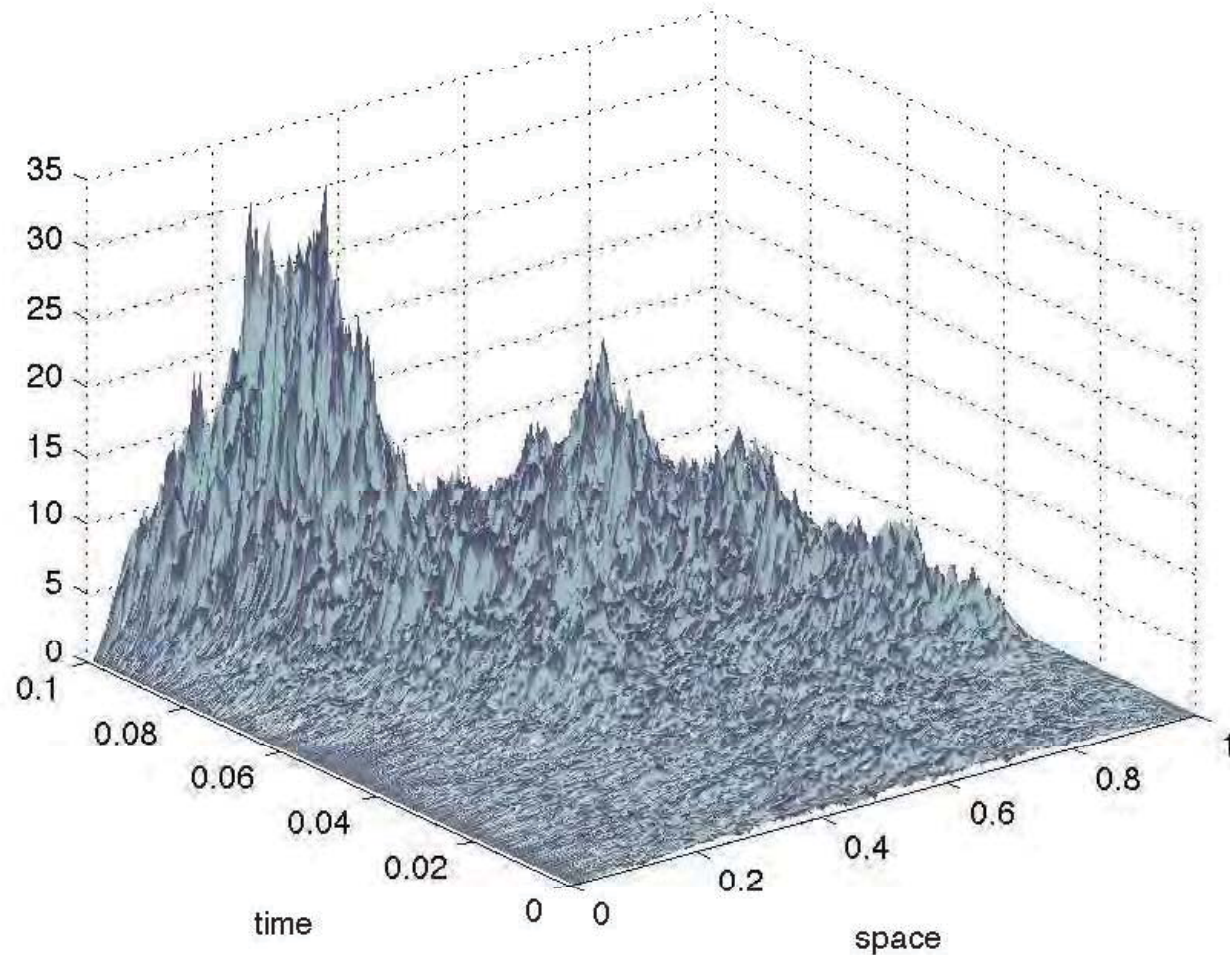


Figure: $\lambda = 2$; max. peak ≈ 35

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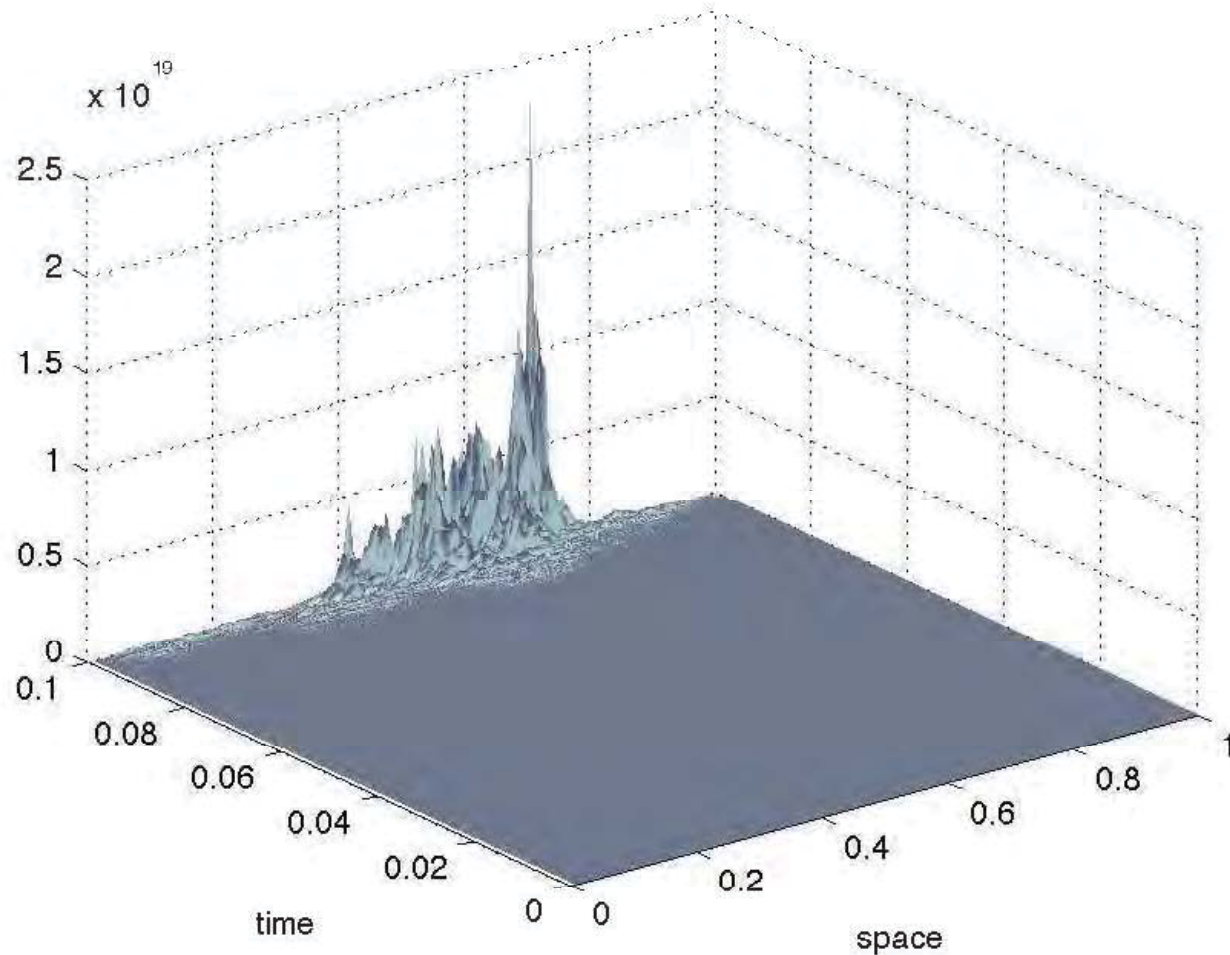


Figure: $\lambda = 5$; max. peak $\approx 2.5 \times 10^{19}$

Nonlinear noise excitation (Lecture 1)

Intermittency

- ▶ **In•ter•mit•tent** (Dictionary.com):
 - ▶ *stopping or ceasing for a time; alternately ceasing and beginning again: an intermittent pain;*
 - ▶ *alternately functioning and not functioning or alternately functioning properly and improperly.*
- ▶ Deep relations to fluid dynamics (Baxendale-Rozovskiĭ, 1993), turbulence (Mandelbrot, 1983; Majda, 1993; Gibbon and Titi, 2005), complex chemical reactions and the large-scale structure of galaxies (Molchanov, 1991; Shandarin-Zel'dovich, 1989; Zel'dovich et al, 1987, 1988, 1990) ...
- ▶ Complex problems in random media are associated to intermittency: As the systems feels more noise, it can begin to act erratically.
- ▶ **Many field theories (SPDEs) yield intermittent solutions.**

Curse of Dimensionality for SPDEs

- Consider the SPDE,

$$\partial_t Z(t, x) = \frac{\nu}{2} \Delta Z(t, x) + \dot{W}(t, x) \quad (t > 0, x \in \mathbb{R}^d, d \geq 2)$$

subject to $Z(0, x) = 0$.

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subject to $Z(0, x) = 0$.

- The weak solution is

$$Z(t, x) = \int_{(0,t) \times \mathbb{R}} p_{t-s}(y-x) W(dsdy),$$

where $p_t(x) = (2\nu\pi t)^{-d/2} \exp\{-\|x\|^2/(2\nu t)\}$.

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- Not a random function; If it were then it would be a GRF with

$$\begin{aligned} E(|Z(t, x)|^2) &= \int_0^t ds \int_{\mathbb{R}^d} dy [p_s(y)]^2 = \int_0^t p_{2s}(0) ds \\ &\approx \int_0^t s^{-d/2} ds = \infty. \end{aligned} \tag{1}$$

Fractional time derivative: Two approaches

- **Riemann-Liouville** fractional derivative of order $0 < \beta < 1$;

$$\mathbb{D}_t^\beta g(t) = \frac{1}{\Gamma(1-\beta)} \frac{\partial}{\partial t} \left[\int_0^t g(s) \frac{ds}{(t-s)^\beta} \right]$$

with Laplace transform $s^\beta \tilde{g}(s)$, $\tilde{g}(s) = \int_0^\infty e^{-st} g(t) dt$ denotes the usual Laplace transform of g .

- **Caputo** fractional derivative of order $0 < \beta < 1$;

$$D_t^\beta g(t) = \frac{1}{\Gamma(1-\beta)} \int_0^t \frac{dg(s)}{ds} \frac{ds}{(t-s)^\beta} \quad (2)$$

was invented to properly handle initial values (Caputo 1967). Laplace transform of $D_t^\beta g(t)$ is $s^\beta \tilde{g}(s) - s^{\beta-1} g(0)$ incorporates the initial value in the same way as the first derivative.

examples

- $$D_t^\beta(t^p) = \frac{\Gamma(1+p)}{\Gamma(p+1-\beta)} t^{p-\beta}$$

- $$D_t^\beta(e^{\lambda t}) = \lambda^\beta e^{\lambda t} - \frac{t^{-\beta}}{\Gamma(1-\beta)}?$$

- $$D_t^\beta(\sin t) = \sin\left(t + \frac{\pi\beta}{2}\right)$$

Space-Time fractional PDE

The solution to the equation

$$\partial_t^\beta u(t, x) = -\nu(-\Delta)^{\alpha/2} u(t, x); \quad u(0, x) = u_0(x), \quad (3)$$

where ∂_t^β is the Caputo fractional derivative of index $\beta \in (0, 1)$ and $\alpha \in (0, 2]$ is given by

$$\begin{aligned} u(t, x) &= \mathbb{E}_x(u_0(Y(E_t))) = \int_0^\infty P(s, x) f_{E_t}(s) ds \\ &= \int_{\mathbb{R}^d} \left(\int_0^\infty p(s, x - y) f_{E_t}(s) ds \right) u_0(y) dy \end{aligned} \quad (4)$$

where $f_{E_t}(s)$ is the density of inverse stable subordinator of index $\beta \in (0, 1)$, and Y is α -stable process. Here $P(t, x) = \mathbb{E}_x(u_0(Y(t)))$ is the semigroup of α -stable process.

Equivalence to Higher order PDE's

Let $L_x g(x) = -(-\Delta)^{\alpha/2} g(x)$, be the fractional Laplacian of g .

- For any $m = 2, 3, 4, \dots$ both the Cauchy problem

$$\partial_t u(t, x) = \sum_{j=1}^{m-1} \frac{t^{j/m-1}}{\Gamma(j/m)} L_x^j u_0(x) + L_x^m u(t, x); \quad u(0, x) = u_0(x) \quad (5)$$

and the fractional Cauchy problem:

$$\partial_t^{1/m} u(t, x) = L_x u(t, x); \quad u(0, x) = u_0(x), \quad (6)$$

have the same unique solution given by

$$u(t, x) = \mathbb{E}(u_0(Y(E_t))) = \int_0^\infty P(s, x) f_{E_t}(s) ds \quad (7)$$

- Due to Baeumer, Meerschaert, and Nane TAMS(2009).

Time fractional spde model?

We want to study the equations of the following type

$$\partial_t^\beta u(t, x) = -\nu(-\Delta)^{\alpha/2} u(t, x) + \lambda\sigma(u) \dot{W}(t, x); \quad u(0, x) = u_0(x), \quad (8)$$

where $\dot{W}(t, x)$ is a space-time white noise with $x \in \mathbb{R}^d$.

Assume that $\sigma(\cdot)$ satisfies the following global Lipschitz condition, i.e. there exists a generic positive constant Lip such that :

$$|\sigma(x) - \sigma(y)| \leq Lip|x - y| \quad \text{for all } x, y \in \mathbb{R}. \quad (9)$$

Clearly, (9) implies the uniform linear growth condition of $\sigma(\cdot)$.

Assume also that the initial datum is $L^p(\Omega)$ bounded ($p \geq 2$), that is

$$\sup_{x \in \mathbb{R}^d} \mathbb{E}|u_0(x)|^p < \infty. \quad (10)$$

Time fractional Duhamel's principle

Let $G(t, x)$ be the fundamental solution of the time fractional PDE $\partial_t^\beta u = L_x u$. The solution to the time-fractional PDE with force term $f(t, x)$

$$\partial_t^\beta u(t, x) = L_x u(t, x) + f(t, x); \quad u(0, x) = u_0(x), \quad (11)$$

is given by Duhamel's principle (Umarov and Saydmatov, 2006), the influence of the external force $f(t, x)$ to the output can be count as

$$\partial_t^\beta V(\tau, t, x) = L_x V(\tau, t, x); \quad V(\tau, \tau, x) = \partial_t^{1-\beta} f(t, x)|_{t=\tau}, \quad (12)$$

which has solution

$$V(t, \tau, x) = \int_{\mathbb{R}^d} G(t - \tau, x - y) \partial_\tau^{1-\beta} f(\tau, x) dx$$

Hence solution to (11) is given by

$$u(t, x) = \int_{\mathbb{R}^d} G(t, x-y) u_0(y) dy + \int_0^t \int_{\mathbb{R}^d} G(t-r, x-y) \partial_r^{1-\beta} f(r, x) dx dr.$$

Model?

Hence if we use this approach we will get the solution of

$$\partial_t^\beta u(t, x) = L_x u(t, x) + \dot{W}(t, x); \quad u(0, x) = u_0(x), \quad (13)$$

to be of the form (informally):

$$\begin{aligned} u(t, x) = & \int_{\mathbb{R}^d} G(t, x - y) u_0(y) dy \\ & + \int_0^t \int_{\mathbb{R}^d} G(t - r, x - y) \partial_r^{1-\beta} [\dot{W}(r, y)] dy dr. \end{aligned} \quad (14)$$

here I am not sure what the fractional derivative in the Walsh-Dalang integral mean?

Another point is that the stochastic integral maybe non-Gaussian!?

Let $\gamma > 0$, define the fractional integral by

$$I_t^\gamma f(t) := \frac{1}{\Gamma(\gamma)} \int_0^t (t - \tau)^{\gamma-1} f(\tau) d\tau.$$

For every $\beta \in (0, 1)$, and $g \in L^\infty(\mathbb{R}_+)$ or $g \in C(\mathbb{R}_+)$

$$\partial_t^\beta I_t^\beta g(t) = g(t).$$

Therefore if we consider the time fractional PDE with a force given by $f(t, x) = I_t^{1-\beta} g(t, x)$, then by the Duhamel's principle the solution to (11) will be given by

$$\begin{aligned} u(t, x) &= \int_{\mathbb{R}^d} G(t, x - y) u^0(y) dy \\ &+ \int_0^t \int_{\mathbb{R}^d} G(t - r, x - y) \partial_r^{1-\beta} [I_r^{1-\beta} g(r, x)] dx dr \\ &= \int_{\mathbb{R}^d} G(t, x - y) u^0(y) dy + \int_0^t \int_{\mathbb{R}^d} G(t - r, x - y) g(r, x) dx dr. \end{aligned} \tag{15}$$

“Correct” TFSPDE Model

We should consider the following model problem (Here $L_x = -(-\Delta)^{\alpha/2}$):

$$\partial_t^\beta u(t, x) = L_x u(t, x) + \lambda I_t^{1-\beta} [\sigma(u) \dot{W}(t, x)]; \quad u(0, x) = u_0(x), \quad (16)$$

By the Duhamel's principle, mentioned above, (16) will have solution

$$\begin{aligned} u(t, x) = & \int_{\mathbb{R}^d} G(t, x - y) u_0(y) dy \\ & + \lambda \int_0^t \int_{\mathbb{R}^d} G(t - r, x - y) \sigma(u(r, y)) W(dy dr). \end{aligned} \quad (17)$$

These time-fractional SPDEs may arise naturally by considering the heat equation in a material with thermal memory, Chen et al. (2015).

The fractional integral above in equation (16) for functions $\phi \in L^2(\mathbb{R}^d)$ is defined as

$$\int_{\mathbb{R}^d} \phi(x) I_t^{1-\beta} [\dot{W}(t, x)] dx = \frac{1}{\Gamma(1-\beta)} \int_{\mathbb{R}^d} \int_0^t (t-\tau)^{-\beta} \phi(x) W(d\tau dx),$$

is well defined only when $0 < \beta < 1/2!$

An important reason to take the fractional integral of the noise in equation (16): Apply the fractional derivative of order $1 - \beta$ to both sides of the equation (16) to see the forcing function, in the traditional units x/t (Baeumer et al. (2005)).

Physical explanation/motivation of the model!

The following discussion is adapted from Chen et al.(2015). Let $u(t, x)$, $e(t, x)$ and $\vec{F}(t, x)$ denote the body temperature, internal energy and flux density, reps. the the relations

$$\begin{aligned}\partial_t e(t, x) &= -\operatorname{div} \vec{F}(t, x) \\ e(t, x) &= \beta u(t, x), \quad \vec{F}(t, x) = -\lambda \nabla u(t, x)\end{aligned}\tag{18}$$

yields the classical heat equation $\beta \partial_t u = \lambda \Delta u$.

According to the law of classical heat equation, the speed of heat flow is infinite. However in real modeling, the propagation speed can be finite because the heat flow can be disrupted by the response of the material.

In a material with thermal memory

$$e(t, x) = \bar{\beta}u(t, x) + \int_0^t n(t-s)u(s, x)ds$$

holds with some appropriate constant $\bar{\beta}$ and kernel n . Typically $n(t) = \Gamma(1 - \beta)^{-1}t^{-\beta_1}$. The convolution implies that the nearer past affects the present more! If in addition the internal energy also depends on past random effects, then

$$e(t, x) = \bar{\beta}u(t, x) + \int_0^t n(t-s)u(s, x)ds + \int_0^t l(t-s)h(s, u) \dot{W}(s, x)ds \quad (19)$$

Where \dot{W} is the space time white noise, modeling the random effects.

Take $\bar{\beta} = 0$, $l(t) = \Gamma(2 - \beta_2)^{-1}t^{1-\beta_2}$, then after differentiation (19)

gives $\partial_t^{\beta_1} u = \operatorname{div} \vec{F} + \frac{1}{\Gamma(1-\beta_2)} \int_0^t (t-s)^{-\beta_2} h(s, u(s, x)) \dot{W}(s, x)ds$

Walsh-Dalang Integral

Need to make sense of the stochastic integral in the mild solution (17). We use the Brownian Filtration $\{\mathcal{F}_t\}$ and the Walsh-Dalang integrals:

- $(t, x) \rightarrow \Phi_t(x)$ is an elementary random field when $\exists 0 \leq a < b$ and an \mathcal{F}_a -meas. $X \in L^2(\Omega)$ and $\phi \in L^2(\mathbb{R}^d)$ such that

$$\Phi_t(x) = X 1_{[a,b]}(t) \phi(x) \quad (t > 0, x \in \mathbb{R}^d).$$

- If $h = h_t(x)$ is non-random and Φ is elementary, then

$$\int h \Phi dW := X \int_{(a,b) \times \mathbb{R}^d} h_t(x) \phi(x) W(dt dx).$$

- The stochastic integral is Wiener's; well defined iff $h_t(x) \phi(x) \in L^2([a, b] \times \mathbb{R}^d)$.
- We have Walsh isometry,

$$\mathbb{E} \left(\left| \int h \Phi dW \right|^2 \right) = \int_a^b ds \int dy [h_s(y)]^2 \mathbb{E}(|\Phi_s(y)|^2)$$

$G(t, x)$ is the density function of $Y(Q_t)$, where Y is an isotropic α -stable Lévy process in \mathbb{R}^d and Q_t is the first passage time of a β -stable subordinator $D = \{D_r, r \geq 0\}$. Let $p_{Y(s)}(x)$ and $f_{Q_t}(s)$ be the density of $Y(s)$ and Q_t , respectively. Then the Fourier transform of $p_{Y(s)}(x)$ is given by

$$\widehat{p_{Y(s)}}(\xi) = e^{-s\nu|\xi|^\alpha}, f_{Q_t}(x) = t\beta^{-1}x^{-1-1/\beta}g_\beta(tx^{-1/\beta}), \quad (20)$$

where $g_\beta(\cdot)$ is the density function of D_1 . By conditioning, we have

$$G(t, x) = \int_0^\infty p_{Y(s)}(x)f_{Q_t}(s)ds. \quad (21)$$

Lemma 1. Let $d < \min\{2, \beta^{-1}\}\alpha$, then

$$\int_{\mathbb{R}^d} G^2(t, x)dx = t^{-\beta d/\alpha} \frac{(\nu)^{-d/\alpha} 2\pi^{d/2}}{\alpha\Gamma(\frac{d}{2})} \frac{1}{(2\pi)^d} \int_0^\infty z^{d/\alpha-1} (E_\beta(-z))^2 dz \quad (22)$$

Stochastic Convolutions

Given a random field $\Phi := \{\Phi_t(x)\}_{t \geq 0, x \in \mathbb{R}^d}$ and space-time noise W , we define the [space-time] *stochastic convolution* $G \circledast \Phi$ to be the random field that is defined as

$$(G \circledast \Phi)_t(x) := \int_{(0,t) \times \mathbb{R}^d} G(t-s, y-x) \Phi_s(y) W(ds dy),$$

for $t > 0$ and $x \in \mathbb{R}^d$, [as a Walsh-Dalang integral] and

$(G \circledast W)_0(x) := 0$. Let Φ be a random field, and for every $\gamma > 0$ and $k \in [2, \infty)$ define

$$\mathcal{N}_{\gamma,k}(\Phi) := \sup_{t \geq 0} \sup_{x \in \mathbb{R}^d} (e^{-\gamma t} \|\Phi_t(x)\|_k) = \sup_{t \geq 0} \sup_{x \in \mathbb{R}^d} \left(e^{-\gamma t} \left[\mathbb{E} |\Phi_t(x)|^k \right]^{1/k} \right) \quad (23)$$

We denote by $\mathcal{L}^{\gamma,2}$ the completion of the space of all simple random fields in the norm $\mathcal{N}_{\gamma,2}$.

Theorem (Mijena and N., 2015)

Let $d < \min\{2, \beta^{-1}\}\alpha$. If σ is Lipschitz continuous and u_0 is measurable and bounded, then there exists a continuous random variable $u \in \cup_{\gamma>0} \mathcal{L}^{\gamma,2}$ that solves (16) with initial function u_0 . Moreover, u is a.s.-unique among all random fields that satisfy the following: There exists a positive and finite constant L -depending only on Lip , and $\sup_{z \in \mathbb{R}^d} |u_0(z)|$ - such that

$$\sup_{x \in \mathbb{R}^d} \mathbb{E} \left(|u_t(x)|^k \right) \leq L^k \exp(Lk^{1+\alpha/(\alpha-\beta d)} t) \quad (24)$$

In contrast to SPDEs, TFSPDEs have random field (function) solutions for $d < \min\{2, \beta^{-1}\}\alpha$.

We use Picard iteration and the stochastic Young inequality below to prove this Theorem.

Picard iteration

Define $u_t^{(0)}(x) := u_0(x)$, and iteratively define $u_t^{(n+1)}$ from $u_t^{(n)}$ as follows:

$$u_t^{(n+1)}(x) := (G_t * u_0)(x) + \lambda \int_{(0,t) \times \mathbb{R}^d} G(t-r, x-y) \sigma(u^{(n)}(r, y)) W(dr dy) \quad (25)$$

for all $n \geq 0$, $t > 0$, and $x \in \mathbb{R}^d$. Moreover, we set $u_0^{(k)}(x) := u_0(x)$ for every $k \geq 1$ and $x \in \mathbb{R}^d$.

Proposition (A stochastic Young inequality)

For all $\gamma > 0$, $k \in [2, \infty)$, $d < \min\{2, \beta^{-1}\}\alpha$, and $\Phi \in \mathcal{L}^{\gamma,2}$,

$$\mathcal{N}_{\gamma,k}(G \circledast \Phi) \leq c_0 k^{1/2} \cdot \mathcal{N}_{\gamma,k}(\Phi).$$

$$\begin{aligned} \|(G \circledast \Phi)_t(x)\|_k^2 &\leq 4k \int_0^t ds \int_{\mathbb{R}^d} dy [G(t-s, y-x)]^2 \|\Phi(y)\|_k^2 \\ &\leq 4k [\mathcal{N}_{\gamma,k}(\Phi)]^2 \int_0^t e^{2\gamma s} ds \int_{\mathbb{R}^d} [G(t-s, y-x)]^2 dy \\ &= 4k C^* [\mathcal{N}_{\gamma,k}(\Phi)]^2 \int_0^t e^{2\gamma s} (t-s)^{-\beta d/\alpha} ds \\ &= 4k C^* [\mathcal{N}_{\gamma,k}(\Phi)]^2 e^{2\gamma t} \int_0^t e^{-2\gamma u} u^{-\beta d/\alpha} du \\ &\leq k C_{\alpha,\beta,d} [\mathcal{N}_{\gamma,k}(\Phi)]^2 e^{2\gamma t} (\gamma)^{-(1-\beta d/\alpha)}. \end{aligned}$$

Finite energy solution

Random field u is a finite energy solution to the stochastic heat equation (16) when $u \in \cup_{\gamma>0} \mathcal{L}^{\gamma,2}$ and there exists $\rho_* > 0$ such that

$$\int_0^\infty e^{-\rho_* t} \mathbb{E}(|u_t(x)|^2) dt < \infty \quad \text{for all } x \in \mathbb{R}^d.$$

If $\rho \in (0, \infty)$, then

$$\int_0^\infty e^{-\rho t} \mathbb{E}(|u_t(x)|^2) dt \leq [\mathcal{N}_{\gamma,2}(u)]^2 \cdot \int_0^\infty e^{-(\rho-2\gamma)t} dt.$$

Therefore if $\rho > 2\gamma$ and $\mathcal{N}_{\gamma,2}(u) < \infty$, then the preceding integral is finite. When σ is Lipschitz-continuous function and u_0 is bounded and measurable, then there exists a finite energy solution to the time fractional stochastic type equation (16).

If we drop the assumption of linear growth for σ , then we have the next theorem that extends the result of Foondun and Parshad (2014).

Theorem (Mijena and N., 2015)

Suppose $\inf_{z \in \mathbb{R}^d} u_0(z) > 0$ and $\inf_{y \in \mathbb{R}^d} |\sigma(y)|/|y|^{1+\epsilon} > 0$. Then, there is no finite-energy solution to the time fractional stochastic heat equation (16).

Hence there is no solution that satisfies

$$\sup_{x \in \mathbb{R}^d} \mathbb{E}(|u_t(x)|^2) \leq L \exp(Lt) \quad \text{for all } t > 0.$$

Intermittency

Definition: The random field $u(t, x)$ is called weakly intermittent if $\inf_{z \in \mathbb{R}^d} |u_0(z)| > 0$, and $\gamma_k(x)/k$ is strictly increasing for $k \geq 2$ for all $x \in \mathbb{R}^d$, where

$$\gamma_k(x) := \liminf_{t \rightarrow \infty} \frac{1}{t} \log \mathbb{E}(|u(t, x)|^k).$$

Fact: If $\gamma_2(x) > 0$ for all $x \in \mathbb{R}^d$, then $\gamma_k(x)/k$ is strictly increasing for $k \geq 2$ for all $x \in \mathbb{R}^d$.

Theorem (Mijena and N., 2016)

Let $d < \min\{2, \beta^{-1}\}\alpha$. If $\inf_{z \in \mathbb{R}^d} |u_0(z)| > 0$, then

$$\inf_{x \in \mathbb{R}^d} \gamma_2(x) \geq [C^*(L_\sigma)^2 \Gamma(1 - \beta d/\alpha)]^{\frac{1}{(1-\beta d/\alpha)}}$$

where $L_\sigma := \inf_{z \in \mathbb{R}^d} |\sigma(z)/z|$. Therefore, the solution $u(t, x)$ of (16) is weakly intermittent when $\inf_{z \in \mathbb{R}^d} |u_0(z)| > 0$ and $L_\sigma > 0$.

This theorem extends the results of Foondun and Khoshnevisan (2009) to the time fractional stochastic heat type equations. Recall the constant $C^* = \text{const} \cdot \nu^{-d/\alpha}$. Hence Theorem above implies the so-called “very fast dynamo property,” $\lim_{\nu \rightarrow \infty} \inf_{x \in \mathbb{R}^d} \eta_2(x) = \infty$. This property has been studied in fluid dynamics: Arponen and Horvai (2007), Baxendale and Rozovskii (1993), Galloway (2003). This theorem is proved by using an application of non-linear renewal theory.

What if the initial function satisfies the following assumptions:

- The initial function u_0 is non-negative on a set of positive measure.
- The function σ satisfies $\sigma(x) \geq L_\sigma|x|$ with L_σ being a positive number.

Theorem (Foondun and N. 2015)

Under Assumptions above, there exists a $T > 0$, such that

$$\inf_{x \in B(0, t^{\beta/\alpha})} \mathbb{E}|u_t(x)|^2 \geq c_3 e^{c_4 \lambda \frac{2\alpha}{\alpha-d\beta} t} \quad \text{for all } t > T. \quad (26)$$

Here c_3 and c_4 are positive constants.

Remark

The two theorems above imply that for any fixed $x \in \mathbb{R}^d$.

$$\begin{aligned}
 c_4 \lambda^{2\alpha/(\alpha-\beta d)} &\leq \liminf_{t \rightarrow \infty} \frac{1}{t} \log \mathbb{E} |u_t(x)|^2 \\
 &\leq \limsup_{t \rightarrow \infty} \frac{1}{t} \log \mathbb{E} |u_t(x)|^2 \leq c_2 \lambda^{2\alpha/(\alpha-\beta d)},
 \end{aligned} \tag{27}$$

for any fixed $x \in \mathbb{R}^d$.

Ideas of the proof:

- We know from Walsh isometry that the second moment of the solution satisfies

$$\begin{aligned}\mathbb{E}|u(t, x)|^2 &= |(\mathcal{G}u_0)_t(x)|^2 \\ &+ \lambda^2 \int_0^t \int_{\mathbb{R}^d} G_{t-s}^2(x-y) \mathbb{E}|\sigma(u(s, y))|^2 dy ds.\end{aligned}$$

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- Renewal theoretic ideas are used to prove the upper bound.
- When the initial function u_0 is bounded uniformly from below, renewal theoretical ideas work well.
- When the initial function u_0 is non-negative on a set of positive measure, then the methods used include Localization and heat kernel estimates for TFPDE:
- There exists some finite positive constants C_1, C_2 such that for all $x \in \mathbb{R}$,

$$C_1 \left(t^{-\beta d/\alpha} \wedge \frac{t^\beta}{|x|^{d+\alpha}} \right) \leq G_t(x) \leq C_2 \left(t^{-\beta d/\alpha} \wedge \frac{t^\beta}{|x|^{d+\alpha}} \right). \quad (28)$$

- the lower bound holds for all $x \in \mathbb{R}^d$.

Non-linear noise excitation

We set

$$\mathcal{E}_t(\lambda) := \sqrt{\int_{\mathbb{R}^d} \mathbb{E}|u_t(x)|^2 dx}.$$

and define the nonlinear excitation index by

$$e(t) := \lim_{\lambda \rightarrow \infty} \frac{\log \log \mathcal{E}_t(\lambda)}{\log \lambda}$$

This was first studied by Khoshnevisan and Kim (2013). This measures in some sense the effect of non-linear noise on the solution of the SPDE.

The following theorem shows that as the value of λ increases, the solution rapidly develops tall peaks that are distributed over relatively small islands!

Theorem (Foondun and N. 2015.)

Fix $t > 0$ and $x \in \mathbb{R}$, we then have

$$\lim_{\lambda \rightarrow \infty} \frac{\log \log \mathbb{E}|u_t(x)|^2}{\log \lambda} = \frac{2\alpha}{\alpha - \beta d}.$$

Moreover, if the energy of the solution exists, then the excitation index, $e(t)$ is also equal to $\frac{2\alpha}{\alpha - \beta d}$.

We now give a relationship between the excitation index and the Hölder exponent of the solution.

Theorem (Mijena and N., 2014)

Let $\eta < (\alpha - \beta d)/2\alpha$ then for every $x \in \mathbb{R}$, $\{u_t(x), t > 0\}$, the solution to (16) has Hölder continuous trajectories with exponent η .

Intermittency fronts

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- We consider $\alpha = 2$.
- $\sigma : \mathbb{R} \rightarrow \mathbb{R}$ Lipschitz and $\sigma(0) = 0$.
- A kind of weak intermittency occur. Roughly, tall peaks arise as $t \rightarrow \infty$, but the farthest peaks move roughly linearly with time away from the origin—intermittency fronts.

- Define, for all $p \geq 2$ and for all $\theta \geq 0$,

$$\mathcal{L}_p(\theta) := \limsup_{t \rightarrow \infty} \frac{1}{t} \sup_{|x| > \theta t} \log \mathbb{E} (|u_t(x)|^p). \quad (29)$$

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- of $\theta_{U_p} > 0$ as an *intermittency upper front* if $\mathcal{L}_p(\theta) > 0$ whenever $\theta < \theta_{U_p}$.
- If there exists θ_* that is both a lower front and an upper front then θ_* is the intermittency front–Phase transition.

Theorem (Mijena, N. (2016), Asogwa, N. (2016?))

Under the above conditions, the time fractional stochastic heat equation (16) has a positive intermittency lower front. In fact,

$$\mathcal{L}_p(\theta) < 0 \text{ if } \theta > \frac{p^2}{4} \left(\frac{4\nu}{p} \right)^{1/\beta} (\text{Lip}_\sigma c_0)^{2\left(\frac{2-\beta}{2-\beta d}\right)}. \quad (30)$$

In addition, under the cone condition $L_\sigma = \inf_{z \in \mathbb{R}} |\sigma(z)/z| > 0$, there exists $\theta_0 > 0$ such that

$$\mathcal{L}_p(\theta) > 0 \text{ if } \theta \in (0, \theta_0). \quad (31)$$

That is, in this case, the stochastic heat equation has a finite intermittency upper front.

This theorem in the case of the stochastic heat equation (for $p = 2$, and $d = 1$) was proved by Conus and Khoshnevisan (2012).

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$$\frac{2^{1/\beta} (Ac_0)^{4/(2-\beta)}}{(Ac_0)^{2\beta/(2-\beta)}}.$$

- When $\beta = 1$. The existence of an intermittency front has been proved recently by Le Chen and Dalang (2012); in fact, they proved that the intermittency front is at $A^2/2$.

TFSPDE in bounded domains with Dirichlet boundary conditions.

How about the moment estimates and growth of the solution of the following TFSPDEs with Dirichlet boundary conditions? (t fixed large λ and λ fixed, large t .)

$$\begin{aligned} \partial_t^\beta u(t, x) &= L_x u(t, x) + \lambda I_t^{1-\beta} [\sigma(u) \dot{W}(t, x)], \quad x \in B(0, R); \\ u(0, x) &= u^0(x), \end{aligned} \quad (32)$$

Following Walsh (1986) and using the time fractional Duhamel's principle, (32) will have (mild/integral) solution

$$\begin{aligned} u(t, x) &= \int_B G_B(t, x - y) u^0(y) dy \\ &+ \lambda \int_0^t \int_B G_B(t - r, x - y) \sigma(u(s, y)) W(dy dr). \end{aligned} \quad (33)$$

Theorem (Foondun, Mijena, N. (2016?))

Suppose that $d < (2 \wedge \beta^{-1})\alpha$. Then under Lipschitz condition on σ , there exists a unique random-field solution to (32) satisfying

$$\sup_{x \in B(0,R)} \mathbb{E}|u_t(x)|^2 \leq c_1 e^{c_2 \lambda^{\frac{2\alpha}{\alpha-d\beta}} t} \quad \text{for all } t > 0.$$

Here c_1 and c_2 are positive constants.

Theorem (Foondun, Mijena, N. (2016?))

Fix $\epsilon > 0$ and let $x \in B(0, R - \epsilon)$, then for any $t > 0$,

$$\lim_{\lambda \rightarrow \infty} \frac{\log \log \mathbb{E}|u_t(x)|^2}{\log \lambda} = \frac{2\alpha}{\alpha - d\beta},$$

where u_t is the mild solution to (32). The excitation index of the solution to (32), $e(t)$ is equal to $\frac{2\alpha}{\alpha - d\beta}$.

This result for $\beta = 1, \alpha \in (1, 2)$ is established by Foondun et al (2015), and for $\beta = 1, \alpha = 2$ it is established by Khoshnevisan and Kim (2015) and Foondun and Joseph (2015).

Large time fixed λ behavior of the solution

Recently Foondun and Nualart (2015) studied the long time ($t \rightarrow \infty$) behavior of the the second moment of the solution to (32) when $\alpha = 2$, $\beta = 1$ $d = 1$. We have extended their results to $\alpha \in (1, 2)$

Theorem (Foondun, Guerngrar, and N. (2016?))

There exists $\lambda_0 > 0$ such that for all $\lambda < \lambda_0$ and $x \in (0, 1)$

$$-\infty < \limsup_{t \rightarrow \infty} \frac{1}{t} \log \mathbb{E}|u_t(x)|^2 < 0.$$

On the other hand, for all $\epsilon > 0$, there exists $\lambda_1 > 0$ such that for all $\lambda > \lambda_1$ and $x \in [\epsilon, 1 - \epsilon]$,

$$0 < \liminf_{t \rightarrow \infty} \frac{1}{t} \log \mathbb{E}|u_t(x)|^2 < \infty.$$

Theorem says that the solution grows exponentially when λ is large.

When λ is small the solution decays exponentially.
 We define the energy of the solution u as

$$\mathcal{E}_t(\lambda) := \sqrt{\mathbb{E} \|u_t\|^2}.$$

We have also the following

$$-\infty < \limsup_{t \rightarrow \infty} \frac{1}{t} \log \mathcal{E}_t(\lambda) < 0 \quad \text{for all } \lambda < \lambda_0.$$

$$0 < \liminf_{t \rightarrow \infty} \frac{1}{t} \log \mathcal{E}_t(\lambda) < \infty \quad \text{for all } \lambda > \lambda_1.$$

Intuitively these results show that if the amount of noise is small, then the heat lost in the system modeled by the Dirichlet equation is not enough to increase the energy in the long run.

On the other hand, if the amount of noise is large enough, then the energy will increase

Space-colored noise

Look at the equation with colored noise.

$$\partial_t^\beta u(t, x) = L_x u(t, x) + \lambda |t|^{1-\beta} [\sigma(u) \dot{F}(t, x)]; \quad u(0, x) = u^0(x), \quad (34)$$

in $(d + 1)$ dimensions, where $\nu > 0$, $\beta \in (0, 1)$, $\alpha \in (0, 2]$,

$-(-\Delta)^{\alpha/2}$ is the generator of an isotropic stable process, $\dot{F}(t, x)$ is white noise in time and colored in space, and $\sigma : \mathbb{R} \rightarrow \mathbb{R}$ is Lipschitz continuous, and satisfies $L_\sigma |x| \leq \sigma(x) \leq Lip_\sigma |x|$ with L_σ and Lip_σ being positive constants.

\dot{F} denotes the Gaussian colored noise satisfying the following property, $\mathbb{E}[\dot{F}(t, x)\dot{F}(s, y)] = \delta_0(t - s)f(x, y)$. This can be interpreted more formally as

$$\text{Cov}\left(\int \phi dF, \int \psi dF\right) = \int_0^\infty \int_{\mathbb{R}^d} dx \int_{\mathbb{R}^d} dy \phi_s(x) \psi_s(y) f(x - y) \quad (35)$$

where we use the notation $\int \phi dF$ to denote the wiener integral of ϕ

We will assume that the spatial correlation of the noise term is given by the following function for $\gamma < d$,

$$f(x, y) := \frac{1}{|x - y|^\gamma}.$$

Following Walsh (1986), we define the mild solution of (34) as the predictable solution to the following integral equation

$$u_t(x) = (\mathcal{G}u_0)_t(x) + \lambda \int_{\mathbb{R}^d} \int_0^t G_{t-s}(x, y) \sigma(u_s(y)) F(ds dy). \quad (36)$$

where

$$(\mathcal{G}u_0)_t(x) := \int_{\mathbb{R}^d} G_t(x, y) u_0(y) dy,$$

and $G_t(x, y)$ is the space-time fractional heat kernel.

Recently Chen et al. (2016) has studied the following space-time fractional SPDE:

$$\partial_t^\beta u = -\nu(-\Delta)^{\alpha/2}u + \lambda I_t^{1-\beta}[u\dot{H}(t, x)], \quad (37)$$

in $(d + 1)$ dimensions, where $\nu > 0$, $\beta \in (0, 1)$, $\alpha \in (0, 2]$,

$-(-\Delta)^{\alpha/2}$ is the generator of an isotropic stable process, $\dot{H}(t, x)$ is space-time colored Gaussian noise:

\dot{H} denotes the Gaussian colored noise satisfying the following property,

$$\mathbb{E}[\dot{H}(t, x)\dot{H}(s, y)] = \gamma(t - s)f(x, y).$$

This can be interpreted more formally as

$$\text{Cov}\left(\int \phi dH, \int \psi dH\right) = \int_{\mathbb{R}_+^2} \int_{\mathbb{R}^{2d}} \phi_s(x)\psi_t(y)f(x-y)\gamma(s-t)dx dy ds dt \quad (38)$$

where we use the notation $\int \phi dH$ to denote the wiener integral of ϕ with respect to H , and the right-most integral converges absolutely.

(37) will have solution

$$\begin{aligned} u(t, x) = & \int_{\mathbb{R}^d} G(t, x - y) u^0(y) dy \\ & + \lambda \int_0^t \int_{\mathbb{R}^d} G(t - r, x - y) u(s, y) H(dy dr). \end{aligned} \tag{39}$$

Chen et al. (2016) have proved bounds for the moments of any order and a lower bound for the second moment under some assumptions on the correlation functions.

Theorem

Under some condition on α, β and the initial function and $f(x, y) = |x - y|^\kappa$ where $0 < \kappa < \min(\alpha/\beta, d)$, the solution to equation (37) satisfies for all $p \geq 1$

$$\mathbb{E}[u(t, x)^p] \leq C_{1,t} \exp\left(t C_{2,t} p^{\frac{2\alpha - \beta\kappa}{\alpha - \beta\kappa}}\right)$$

for some constants $C_{1,t}, C_{2,t} > 0$. $C_{1,t}$ is related to finiteness of the solution of the equation without the noise. $C_{2,t} = 2C \int_0^t \gamma(s) ds$ is related to the correlation function in time.

If the noise is white in time, and the initial function is a constant then

$$\mathbb{E}[u(t, x)^2] \geq cu_0^2 \exp(tC).$$

Chen et al (2016) also considered

$$\partial_t^\beta u = -\nu(-\Delta)^{\alpha/2}u + u\dot{H}(t, x), \quad (40)$$

in $(d + 1)$ dimensions, where $\nu > 0$, $\beta \in (0, 1)$, $\alpha \in (0, 2]$, $(-\Delta)^{\alpha/2}$ is the generator of an isotropic stable process, $\dot{H}(t, x)$ is space-time colored Gaussian noise:

(40) will have solution

$$\begin{aligned}
 u(t, x) = & \int_{\mathbb{R}^d} G(t, x - y)u^0(y)dy \\
 & + \lambda \int_0^t \int_{\mathbb{R}^d} G^*(t - r, x - y)u(s, y)H(dydr).
 \end{aligned} \tag{41}$$

Where G^* is obtained from G by fractional differentiation in time!

To prove existence of solutions and moment bounds the methods used are:

Wiener-Chaos expansion of the solution.

Explicit integral representation of the kernels G and G^* by Fox H-functions.

Plancharel type identities...

Future work/Open problems

- ① Large space behavior of STF-SPDEs
- ② Employing the STF-SPDEs for modelling data!
- ③ Inverse problems for SPDEs
- ④ Existence-non-existence of solutions, blow up of solutions in finite time.
- ⑤ Comparison of solutions for different initial values and/or σ s!
- ⑥ Approximations of SPDEs: what happens when you have time fractional derivatives? For regular diffusions and stable process, one can have random walk approximations and come up with a system of SDEs. Can one follow the same strategy here? We can approximate time fractional diffusions by CTRWs.!

Thank You!

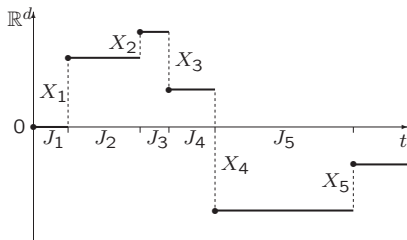
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Continuous time random walks



The CTRW is a random walk with jumps X_n separated by random waiting times J_n . The random vectors (X_n, J_n) are i.i.d.

Heavy tailed waiting times

Random wait J_n between jumps, n th jump time given by a random walk

$$T(n) = J_1 + \cdots + J_n$$

Number of jumps by time t is inverse $N(t) \geq n \iff T(n) \leq t$

For heavy tail waiting times $P(J_n > t) \approx Ct^{-\beta}$ ($0 < \beta < 1$)

$$c^{-1/\beta} T(ct) \Rightarrow P(t) \iff c^{-\beta} N(ct) \Rightarrow Q(t)$$

Inverse processes have inverse scaling

$$P(ct) \approx c^{1/\beta} P(t) \iff Q(ct) \approx c^\beta Q(t)$$

Continuous time random walks (CTRW)

Particle jump random walk has scaling limit $c^{-1/2}S([ct]) \implies W(t)$.

Number of jumps has scaling limit $c^{-\beta}N(ct) \implies Q(t)$.

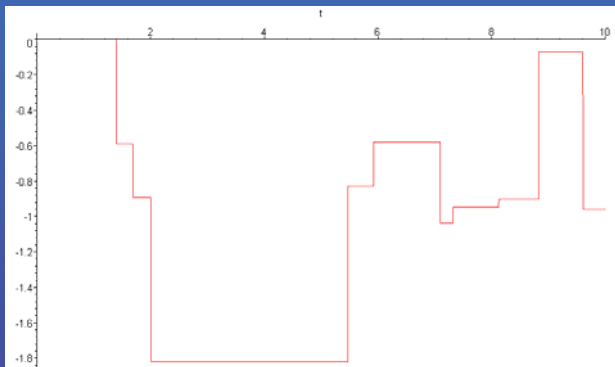
CTRW is a random walk subordinated to (a renewal process) $N(t)$

$$S(N(t)) = X_1 + X_2 + \cdots + X_{N(t)}$$

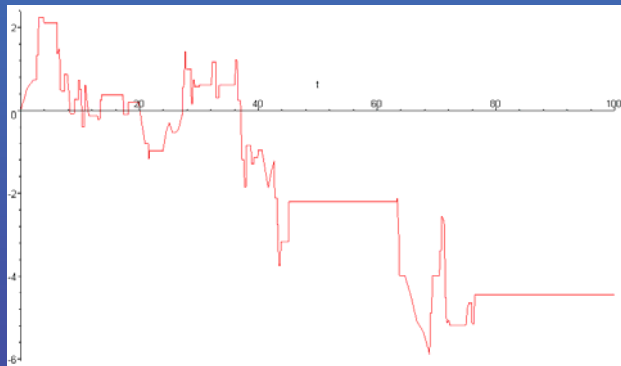
CTRW scaling limit is a subordinated process:

$$\begin{aligned} c^{-\beta/2}S(N(ct)) &= (c^\beta)^{-1/2}S(c^\beta \cdot c^{-\beta}N(ct)) \\ &\approx (c^\beta)^{-1/2}S(c^\beta Q(t)) \implies W(Q(t)). \end{aligned}$$

CTRW simulation with heavy tail waiting times



Longer time scale



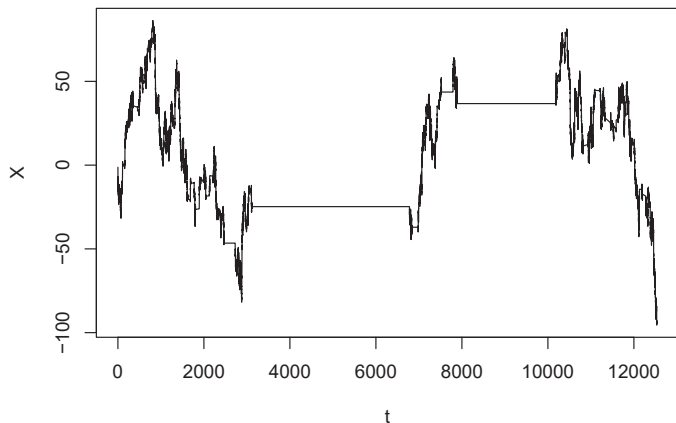


Figure: Typical sample path of the iterated process $W(Q(t))$. Here $W(t)$ is a Brownian motion and $Q(t)$ is the inverse of a $\beta = 0.8$ -stable subordinator. Graph has dimension $1 + \beta/2 = 1 + 0.4$. The limit process retains long resting times

Power law waiting times

- Wait between solar flares $1 < \beta < 2$
- Wait between raindrops $\beta = 0.68$
- Wait between money transactions $\beta = 0.6$
- Wait between emails $\beta \approx 1.0$
- Wait between doctor visits $\beta \approx 1.4$
- Wait between earthquakes $\beta = 1.6$
- Wait between trades of German bond futures $\beta \approx 0.95$
- Wait between Irish stock trades $\beta = 0.4$ (truncated)