

# The Interpolated Stochastic Heat and Wave Equation: Solvability and Exact Moment Asymptotics

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# The Interpolated Stochastic Heat and Wave Equation.

$$\begin{cases} (\partial_t^b + \frac{\nu}{2}(-\Delta)^{a/2}) u(t, x) = I_t^r \left[ \sqrt{\theta} u(t, x) \dot{W}(x) \right] & x \in \mathbb{R}^d, t > 0, \\ u(0, \cdot) = 1 & b \in (0, 1], \\ u(0, \cdot) = 1, \quad \partial_t u(0, \cdot) = 0 & b \in (1, 2), \end{cases}$$

- ▶  $\partial_t^b$  is the Caputo fractional derivative
- ▶  $(-\Delta)^{a/2}$  is the fractional Laplacian of order  $a \in (0, 2]$
- ▶  $I_t^r$  is the Riemann-Liouville fractional integral of order  $r \geq 0$
- ▶  $W = \{W(\phi) : \phi \in \mathcal{D}(\mathbb{R}^d)\}$  is a centered Gaussian and time-independent noise

Case of  $b = a = 2$  and  $r = 0$  and the Riesz kernel noise.

$$\begin{cases} \frac{\partial^2 u}{\partial t^2}(t, x) = \Delta u(t, x) + \sqrt{\theta} u(t, x) \dot{W}(x) & x \in \mathbb{R}^d, t > 0, \\ u(0, \cdot) = 1, \quad \partial_t u(0, \cdot) = 0 & b \in (1, 2), \end{cases}$$

Theorem (Balan, Chen, L. and Chen, X.)

Suppose  $\mu(dx) = |x|^{-d+\alpha} dx$  and that  $0 < \alpha < d \leq 3$ . Let  $t_p := (p-1)^{1/(4-\alpha)} t$ . Then the following moment asymptotic holds:

$$\lim_{t_p \rightarrow \infty} t_p^{-\frac{4-\alpha}{3-\alpha}} \log \|u(t, x)\|_p = \theta^{\frac{1}{3-\alpha}} 2^{\frac{-\alpha}{2(3-\alpha)}} \frac{3-\alpha}{2} \left( \frac{2\mathcal{M}^{1/2}}{4-\alpha} \right)^{\frac{4-\alpha}{3-\alpha}}.$$

In addition by fixing  $p \geq 2$ , we get that

$$\lim_{t \rightarrow \infty} t^{-\frac{4-\alpha}{3-\alpha}} \log \mathbb{E}|u(t, x)|^p = p(p-1)^{\frac{1}{3-\alpha}} \theta^{\frac{1}{3-\alpha}} 2^{\frac{-\alpha}{2(3-\alpha)}} \frac{3-\alpha}{2} \left( \frac{2\mathcal{M}^{1/2}}{4-\alpha} \right)^{\frac{4-\alpha}{3-\alpha}}.$$

# Gaussian Processes

A Gaussian Processes,  $G := \{G_t : t \in T\}$  is a stochastic process such that for any finite set of  $t_1, \dots, t_n \in T$ , the random vector  $(G(t_1), \dots, G(t_n))$  follows a  $n$ -dimensional Gaussian distribution. The processes  $G$  is uniquely determined by

1.  $\mu(t) = \mathbb{E}(G(t))$
2.  $C(s, t) = \mathbb{E}(G(t)G(s))$

## Theorem (Important)

*The space of all symmetric and non-negative definite functions,  $f(s, t)$ , on  $T \times T$  matches the space of all covariance functions of all Gaussian processes on  $T$ . Thus, given a non-negative definite function on  $T \times T$  then there exists a Gaussian Processes that possesses this covariance function.*

## The noise, $\dot{W}$ .

We start with a nonnegative and nonnegative definite tempered measure  $\Gamma$  with density  $\gamma$ ,

$$\Gamma(dx) = \gamma(x)dx.$$

Bochner's theorem guarantees the existence of a measure  $\mu$ , the spectral measure, defined through the following:

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

We now define a functional on  $C_0^\infty(\mathbb{R}^d)$ .

$$C(\phi, \psi) = \int_{\mathbb{R}^d} \mathcal{F}\phi(\xi) \overline{\mathcal{F}\psi(\xi)} \mu(d\xi)$$

## The noise, $\dot{W}$ .

Recalling the theorem mentioned above, we have a zero mean Gaussian processes,  $W = \{W(\phi) : \phi \in C_0^\infty, \phi(x)\}$ , with covariance given by  $C(\cdot, \cdot)$ . In other words,

$$\begin{aligned}\mathbb{E}(W(\phi)W(\psi)) &= \int_{\mathbb{R}^d} \mathcal{F}\phi(\xi) \overline{\mathcal{F}\psi(\xi)} \mu(d\xi) \\ &= \int_{\mathbb{R}^{2d}} \phi(x) \psi(y) \gamma(x-y) dx dy.\end{aligned}$$

When  $\mu$  has a density, we denote it as  $\varphi(x)$ .

# Spatially Correlated Versus White Verses Independent

- ▶ Correlated:

$$\mathbb{E}(W(\phi)W(\psi)) = \int_{\mathbb{R}^{2d}} \phi(x)\psi(y)\gamma(x-y)dx dy.$$

- ▶ White:

$$\begin{aligned}\mathbb{E}(W(\phi)W(\psi)) &= \int_{\mathbb{R}^{2d}} \phi(x)\psi(y)\delta_0(x-y)dx dy \\ &= \int_{\mathbb{R}^{2d}} \phi(x)\psi(x)dx\end{aligned}$$

and informally we may think that  $\mathbb{E}(\dot{W}(x)\dot{W}(y)) = \delta_0(x-y)$ .

- ▶ Independent:

$$\mathbb{E}(W(\phi)W(\psi)) = \mathbb{E}(W(\phi))\mathbb{E}(W(\psi)) = 0 \quad \text{for all } \phi, \psi.$$

## The noise, $\dot{W}$ .

- ▶ Let  $k \in \{1, \dots, d\}$
- ▶ Partition the  $d$ -coordinates of  $x = (x_1, \dots, x_d)$  into  $k$  distinct groups of size  $d_i$  so that  $d_1 + \dots + d_k = d$
- ▶  $x_{(i)} = (x_{i_1}, \dots, x_{i_{d_i}})$  denotes the coordinates of the  $i^{th}$  partition.
- ▶ Choose  $\alpha_i \in (0, d_i)$  and denote  $\alpha = \sum_{i=1}^k \alpha_i$

We consider the following Riesz-type spatial correlation function and its spectral density:

$$\gamma(x) = \prod_{i=1}^k |x_{(i)}|^{-\alpha_i} \quad \varphi(\xi) = \prod_{i=1}^k C_{\alpha_i, d_i} |\xi_{(i)}|^{d_i - \alpha_i}.$$

Note that when  $k = 1$ , the above reduces to the Riesz kernel.



## The fundamental solution.

$$\begin{cases} (\partial_t^b + \frac{\nu}{2}(-\Delta)^{a/2}) u(t, x) = I_t^r [f(t, x)] & x \in \mathbb{R}^d, t > 0, \\ u(0, \cdot) = 1 & b \in (0, 1], \\ u(0, \cdot) = 1, \quad \partial_t u(0, \cdot) = 0 & b \in (1, 2), \end{cases}$$

When  $f(t, x)$  is a deterministic function, the fundamental solution given through the triple

$$\{Z_{a,b,r,\nu,d}(t, x), Z_{a,b,r,\nu,d}^*(t, x), Y_{a,b,r,\nu,d}(t, x)\}$$

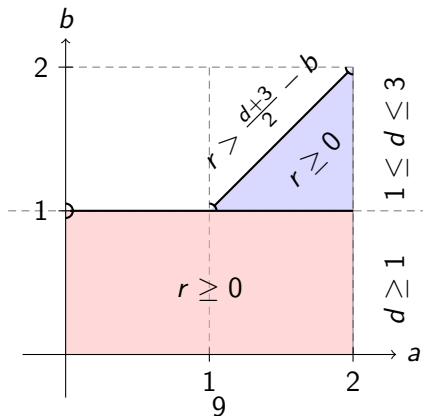
where each member is defined explicitly through the Fox-H function. We have for  $b \in (0, 1]$  and  $b \in (1, 2)$  respectively that

$$u(t, x) = \begin{cases} (Z(t, \cdot) * u(0, \cdot))(x) + (Y \star f)(t, x) \\ (Z^*(t, \cdot) * u(0, \cdot))(x) + (Z(t, \cdot) * \partial_t u(0, \cdot))(x) + (Y \star f)(t, x). \end{cases}$$

## Assumption on $G$

We need to make an assumption that  $G(t, x) = Y_{a,b,r,v,d}(t, x)$  is nonnegative. This is the case under any of the following:

- ▶  $d \geq 1$ ,  $b \in (0, 1]$ ,  $a \in (0, 2]$ ,  $r \geq 0$ ;
- ▶  $1 \leq d \leq 3$ ,  $1 < b < a \leq 2$ ,  $r > 0$ ;
- ▶  $1 \leq d \leq 3$ ,  $1 < b = a < 2$ ,  $r > \frac{d+3}{2} - b$ .



## Mild solution

Considering the constant 1 initial condition, it can be shown the solution to the homogeneous equation is equal to 1.

### Definition

Let  $G(t, x) = Y_{a,b,r,\nu,d}(t, x)$ . For  $T > 0$ , a random field  $u = \{u(t, x) : t \in (0, T), x \in \mathbb{R}^d\}$  is called a *mild solution* if  $G(t - s, x - \cdot)u(s, \cdot)1_{\{s < t\}}$  is Skorohod integrable and the following holds almost surely:

$$u(t, x) = 1 + \sqrt{\theta} \int_0^t \left( \int_{\mathbb{R}^d} G(t - s, x - y) u(s, y) W(\delta y) \right) ds.$$

- ▶ *Global Solution:*  $\|u(t, x)\|_p < \infty$  for any  $t > 0$  and  $x \in \mathbb{R}^d$ .
- ▶ *Local Solution:* If there exists  $0 < T_1 \leq T_2 < \infty$  such that  $\|u(t, x)\|_2 < \infty$  when  $0 < t < T_1$  and  $\|u(t, x)\|_2$  D.N.E. for  $t > T_2$ .

## Wiener Chaos Expansion:

If  $t = t_{n+1}$  and  $x = x_{n+1}$ , then a Picard iteration scheme will suggest that for

$$f_n(x_1, \dots, x_n; x, t) = \int_0^t \int_0^{t_n} \cdots \int_0^{t_2} \prod_{k=1}^n G(t_{k+1} - t_k, x_{k+1} - x_k) dt_1 \cdots dt_n,$$

the solution can be written as

$$u(t, x) = 1 + \sum_{k=1}^{\infty} \theta^{k/2} I_k(f_k(\cdot, x, t)), \quad (t, x) \in (0, T) \times \mathbb{R}^d. \quad (0.1)$$

This further suggests

$$\mathbb{E}(u(t, x)^2) = \sum_{k=0}^{\infty} \theta^k \|f_k(\cdot, x, t)\|_{\mathcal{H}^{\otimes n}}^2, \quad (t, x) \in (0, T) \times \mathbb{R}^d. \quad (0.2)$$

## Theorem 3.3 (Chen and Eisenberg)

### Theorem

Fix any  $T \in (0, \infty]$ . Suppose that  $f_n(\cdot, x; t) \in \mathcal{H}^{\otimes n}$  for any  $t \in (0, T)$  and  $x \in \mathbb{R}^d$ . Then the SPDE in question has a unique  $L^2(\Omega)$  solution on  $(0, T) \times \mathbb{R}^d$  if and only if the series

$$\sum_{k=0}^{\infty} \theta^k \|f_k(\cdot, x, t)\|_{\mathcal{H}^{\otimes n}}^2, \quad (t, x) \in (0, T) \times \mathbb{R}^d.$$

converges in  $L^2(\Omega)$  for any  $(t, x) \in (0, T) \times \mathbb{R}^d$ . In this case, the second moment of the solution is given by this series.

## Theorem 1.6 (Chen and Eisenberg)

Assuming  $G(t, x)$  is nonnegative and the noise has spatial correlation  $\gamma(x) = \prod_{i=1}^k |x_{(i)}|^{-\alpha_i}$ :

- ▶ A *global solution* exists provided

$$0 < \alpha < \min \left( \frac{a}{b} [2(b+r) - 1], 2a, d \right)$$

- ▶ Otherwise, if

$$r \in [0, 1/2] \quad \text{and} \quad 0 < \alpha = \frac{a}{b} [2(b+r) - 1] \leq d,$$

then a *local solution* exists for  $t \in (0, T_p)$  where

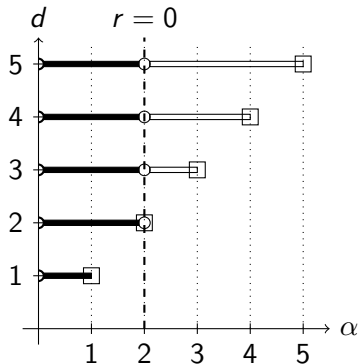
$$T_p := \frac{\nu^{\alpha/a}}{2\theta(p-1)\mathcal{M}_a^{(2a-\alpha)/a}}$$

and the solution D.N.E. for  $t > T_2$ .



## Example: Solvability for the SHE ( $a = 2$ and $b = 1$ )

By setting  $a = 2$ ,  $b = 1$  and  $r = 0$ , we obtain the following condition for existence of a local solution:  $\alpha = 2 \leq d$ .



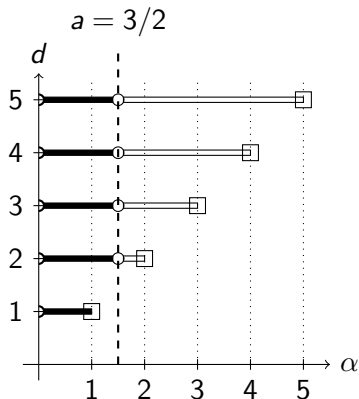
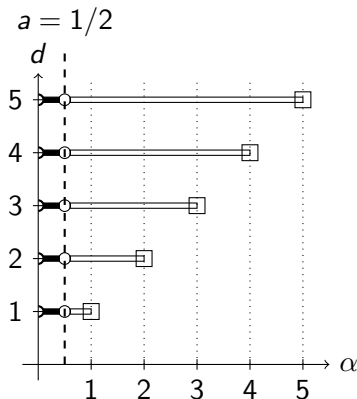
- ▶ When  $\alpha = d = 2$ , the critical time becomes 
$$T_p = \frac{\nu}{2\theta(p-1)\mathcal{M}_{2,2}(\delta_0)}.$$
- ▶ Theorem 4.1 (Hu) proves that an  $L^2(\Omega)$  solution exists for  $t < 2$  but not for  $t > 2\pi$ .
- ▶  $T_2$  being precise implies 
$$2 \leq T_2 = \frac{1}{2\mathcal{M}_{2,2}(\delta_0)} \leq 2\pi$$



## Example: Solvability For The SHE With Fractional Laplace ( $b = 1$ , $r = 0$ and $0 < a \leq 2$ )

A local solution exists when  $\alpha = a \leq d$ . The critical time is given by

$$T_p = \frac{\nu}{2\theta(p-1)\mathcal{M}_{2,2}(\delta_0)}.$$



# The moment asymptotic

## Theorem (Chen and Eisenberg)

*Suppose a global solution to the SPDE in question exists and  $\dot{W}$  is given through the generalized Riesz kernel defined above. Then,*

$$\begin{aligned} & \lim_{t_p \rightarrow \infty} t_p^{-\beta} \log \|u(t, x)\|_p \\ &= \left(\frac{1}{2}\right) \left(\frac{2a}{2a(b+r) - b\alpha}\right)^\beta \\ & \quad \times \left(\theta \nu^{-\alpha/a} \mathcal{M}_a^{\frac{2a-\alpha}{a}}\right)^{\frac{a}{2a(b+r) - b\alpha - a}} \left(2(b+r) - \frac{b\alpha}{a} - 1\right), \end{aligned}$$

where

$$\beta := \frac{2(b+r) - \frac{b\alpha}{a}}{2(b+r) - \frac{b\alpha}{a} - 1} \quad \text{and} \quad t_p := (p-1)^{1-1/\beta} t. \quad (0.3)$$

We define

$$\begin{aligned}\mathcal{M}_{a,d}(\gamma, \theta) &:= \sup_{g \in \mathcal{F}_a} \left\{ \left( \iint_{\mathbb{R}^{2d}} g^2(x) g^2(y) \gamma(x+y) dx dy \right)^{1/2} - \frac{\theta}{2} \mathcal{E}_a(g, g) \right\} \\ &= \sup_{g \in \mathcal{F}_a} \left\{ \langle g^2 * g^2, \gamma \rangle_{L^2(\mathbb{R}^d)}^{1/2} - \frac{\theta}{2} \mathcal{E}_a(g, g) \right\},\end{aligned}$$

where

$$\mathcal{E}_a(g, g) := (2\pi)^{-d} \int_{\mathbb{R}^d} |\xi|^a |\mathcal{F}g(\xi)|^2 d\xi \quad \text{and} \quad (0.4)$$

$$\mathcal{F}_a := \left\{ f \in L^2(\mathbb{R}^d) : \|f\|_{L^2(\mathbb{R}^d)} = 1, \mathcal{E}_a(f, f) < \infty \right\}. \quad (0.5)$$

$\rho$

We define

$$\rho_{\nu,a}(\gamma) = \sup_{\|f\|_{L^2(\mathbb{R}^d)}=1} \int_{\mathbb{R}^d} \left[ \int_{\mathbb{R}^d} \frac{f(x+y)f(y)}{\sqrt{1+\frac{\nu}{2}|x+y|^a} \sqrt{1+\frac{\nu}{2}|y|^a}} dy \right]^2 \mu(dx)$$

### Theorem (Chen and Eisenberg)

When  $\mu(dx) = \varphi(x)dx$  where  $\varphi$  is the generalized Riesz kernel introduced above then,

$$\begin{aligned} \lim_{n \rightarrow \infty} \frac{1}{n} \log \left[ \frac{1}{(n!)^2} \int_{(\mathbb{R}^d)^n} \left( \sum_{\sigma \in \Sigma_n} \prod_{k=1}^n \frac{1}{1 + \frac{\nu}{2} |\sum_{j=k}^n \xi_{\sigma(j)}|^a} \right)^2 \mu(d\vec{\xi}) \right] \\ = \log(\rho_{\nu,a}(\gamma)). \end{aligned}$$

where  $\gamma$  is the spatial correlation function.

# The Connection between $\rho$ and $\mathcal{M}$

## Theorem (Chen and Eisenberg)

When  $\mu(dx) = \varphi(x)dx$  where  $\varphi$  is the generalized Riesz kernel introduced above then,

$$\rho_{\nu,a}(\gamma) = \nu^{-\alpha/a} \mathcal{M}_a^{2-(\alpha/a)}(\gamma) < \infty$$

## $\rho$ and the SWE

$$\zeta([0, t_1] \times [0, t_2]) = \int_0^{t_1} \int_0^{t_2} |X_1(s_1) + X_2(s_2)|^{-\sigma} ds_1 ds_2$$

where  $X_1(t)$  and  $X_2(t)$  are i.i.d.  $d$ -dimensional symmetric stable processes of index  $\beta = 2$ ,

$$\mathbb{E} e^{i\lambda X_i(t)} = e^{-t|\lambda|^\beta}.$$

### Theorem (X. Chen)

$$\lim_{n \rightarrow \infty} \frac{1}{n} \log \frac{1}{(n!)^2} \mathbb{E} \left[ \zeta([0, \tau_1] \times [0, \tau_2])^n \right] = \log [\rho_{2,2}(\gamma)]$$

## $\rho$ and the SWE

### Theorem (X. Chen)

Suppose  $\tau_1$  and  $\tau_2$  are exponential random variables with mean 1. Then,

$$\mathbb{E} [\zeta ([0, \tau_1] \times [0, \tau_2])^n] = \int_{(\mathbb{R}^d)^n} \left[ \sum_{\sigma \in \Sigma_n} \prod_{k=1}^n \frac{1}{1 + \left| \sum_{j=k}^n \xi_{\sigma(j)} \right|^2} \right]^2 \mu(d\vec{\xi})$$

It can be shown that

$$\begin{aligned} & \frac{1}{(n!)^2} \mathbb{E} [\zeta ([0, \tau_1] \times [0, \tau_2])^n] \\ &= \int_{(\mathbb{R}^d)^2} \mathcal{FL} \left[ \tilde{f}_n(\circ_1, \dots \circ_n, 0; \cdot) \right]^2 (\xi_1, \dots, \xi_n, 1) \mu(d\vec{\xi}) \end{aligned}$$

# Exponential Growth of the Moments in Time

## Theorem (Chen and Eisenberg)

For  $p \geq 2$  fixed, we have that

$$\begin{aligned} & \lim_{t \rightarrow \infty} t^{-\beta} \log \mathbb{E}(|u(t, x)|^p) \\ &= p(p-1)^{\frac{1}{2(b+r) - \frac{b\alpha}{a} - 1}} \left(\frac{1}{2}\right) \left(\frac{2a}{2a(b+r) - b\alpha}\right)^{\beta} \\ & \quad \times \left(\theta \nu^{-\alpha/a} \mathcal{M}_a^{\frac{2a-\alpha}{a}}\right)^{\frac{a}{2a(b+r) - b\alpha - a}} \left(2(b+r) - \frac{b\alpha}{a} - 1\right). \end{aligned}$$

From this we can deduce that  $t \mapsto \mathbb{E}(|u(t, x)|^p)$  grows as fast as  $\exp(K_1 t^\beta)$ , in other words, the moments will grow exponentially in time.



# Exponential Growth of the moments for Fixed Time

## Theorem (Chen and Eisenberg)

For  $t > 0$  fixed , we have that

$$\begin{aligned} & \lim_{p \rightarrow \infty} p^{-\beta} \log \mathbb{E} (|u(t, x)|^p) \\ &= t^\beta \left( \frac{1}{2} \right) \left( \frac{2a}{2a(b+r) - b\alpha} \right)^\beta \\ & \quad \times \left( \theta \nu^{-\alpha/a} \mathcal{M}_a^{\frac{2a-\alpha}{a}} \right)^{\frac{a}{2a(b+r) - b\alpha - a}} \left( 2(b+r) - \frac{b\alpha}{a} - 1 \right). \end{aligned}$$

From this we can deduce that  $p \mapsto \mathbb{E}(|u(t, x)|^p)$  grows as fast as  $\exp(K_2 p^\beta)$ , in other words, the moments will grow exponentially in  $p$  as well.

## Final Comments

Two possible future works.

1. Consider a time and spatially dependent noise. I.e,  
 $W = \{W(\phi), \phi \in \mathcal{D}(\mathbb{R} \times \mathbb{R}^d)\}$

$$\mathbb{E}(W(\phi)W(\psi)) = \int_{\mathbb{R}^2} \int_{\mathbb{R}^{2d}} \gamma(t-s)f(x-y)\phi(t,x)\psi(s,y)dx dy dt ds$$

where  $\gamma$  is the time correlation and  $f$  is the spatial correlation.

2. Under what type of convergence do we have as we vary the fractional parameters? How is the following limit taken:

$$\lim_{a \rightarrow a_0, b \rightarrow b_0, r \rightarrow r_0, \nu \rightarrow \nu_0, \theta \rightarrow \theta_0} u_{a,b,r,\nu,\theta}(t,x) \rightarrow u_{a_0,b_0,r_0,\nu_0,\theta_0}(t,x).$$