

Some properties of non-linear fractional stochastic heat equations on bounded domains

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Abstract

Consider the following stochastic partial differential equation,

$$\partial_t u_t(x) = \mathcal{L}u_t(x) + \xi\sigma(u_t(x))\dot{F}(t, x),$$

where ξ is a positive parameter and σ is a globally Lipschitz continuous function. The stochastic forcing term $\dot{F}(t, x)$ is white in time but possibly colored in space. The operator \mathcal{L} is a non-local operator. We study the behaviour of the solution with respect to the parameter ξ , extending the results in [8] and [11].

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1 Introduction and main results

Stochastic Partial Differential Equations (SPDEs) have been used recently in many disciplines ranging from applied mathematics, statistical mechanics and theoretical physics to theoretical neuroscience, theory of complex chemical reactions (including polymer science), fluid dynamics and mathematical finance to quote only a few; see for example [9] and references therein.

In [8], the authors considered the following stochastic heat equation,

$$\partial_t u_t(x) = \mathcal{L}u_t(x) + \xi\sigma(u_t(x))\dot{F}(t, x), \tag{1.1}$$

where \mathcal{L} is the Dirichlet Laplacian on $B_R(0)$, the ball of radius R centered at the origin. Under some appropriate conditions, it was shown that the long time behaviour of the solution is dependent on the *noise level*, that is on the values of ξ . More precisely, it was shown that for large values of ξ , the moments of the solution grow exponentially with time while for small values of ξ , the moments decay exponentially. In this paper, we extend the results of [8] by taking \mathcal{L} to be a non-local operator, the generator of a killed stable process, namely $\mathcal{L} := -\nu(-\Delta)^{\alpha/2}$ for $0 < \alpha \leq 2$ with zero exterior boundary conditions. We also provide some clarification and simplification of the proofs in [8]. Non-local operators are becoming increasingly important due to their wide applicability for modeling purposes. The class of equations we study can for instance be used to model particles moving in a discontinuous fashion while being subject to some branching mechanism; see for example Walsh [10].

Throughout this paper, the initial condition u_0 is always assumed to be a non-negative bounded deterministic function such that for some set $K \subset B_R(0)$, the quantity

$$\int_K u_0(x) dx$$

is strictly positive. The function σ will be subjected to the following condition.

Assumption 1.1. *The function σ is assumed to be a globally Lipschitz function satisfying*

$$l_\sigma|x| \leq |\sigma(x)| \leq L_\sigma|x| \quad \text{for all } x \in \mathbf{R},$$

for some positive constants l_σ and L_σ .

Following Walsh [10], we look at the mild solution of (1.1) satisfying the following integral equation,

$$u_t(x) = (\mathcal{G}_D u_0)_t(x) + \xi \int_{B_R(0)} \int_0^t p_D(t-s, x, y) \sigma(u_s(y)) F(ds, dy), \quad (1.2)$$

where

$$(\mathcal{G}_D u_0)_t(x) = \int_{B_R(0)} u_0(y) p_D(t, x, y) dy,$$

and $p_D(t, x, y)$ denotes the heat kernel of the stable process. When the driving noise is white in space and time, existence-uniqueness considerations impose the conditions that $d = 1$ and $1 < \alpha < 2$. When the noise term is not space-time white noise, it will be spatially correlated that is,

$$\mathbb{E} \dot{F}(s, x) \dot{F}(t, y) = \delta_0(t-s) f(x, y),$$

where the correlation function f satisfies the inequality $f(x, y) \leq \tilde{f}(x-y)$, and \tilde{f} is a locally integrable positive continuous function on $\mathbf{R}^d \setminus \{0\}$ satisfying the following Dalang type condition,

$$\int_{\mathbf{R}^d} \frac{\hat{\tilde{f}}(\xi)}{1 + |\xi|^\alpha} d\xi < \infty,$$

where $\hat{\tilde{f}}$ denotes the Fourier transform of \tilde{f} ; see [4]. We will impose the following non-degeneracy condition on f ,

Assumption 1.2. *There exists a constant K_R such that*

$$\inf_{x, y \in B_R(0)} f(x, y) \geq K_R.$$

The above conditions on the correlation function are quite mild. Examples of correlation functions satisfying Assumption 1.2 include the Riesz kernel, Cauchy kernels and many more: See, for example, [6] and [7]. Our first set of results concerns equation (1.1) when the driving noise is space-time white noise which we denote by \dot{W} . In other words, we are looking at

$$\begin{cases} \partial_t u_t(x) = \mathcal{L}u_t(x) + \xi \sigma(u_t(x)) \dot{W}(t, x), & x \in B_R(0), \quad t > 0 \\ u_t(x) = 0, & x \in B_R(0)^c. \end{cases} \quad (1.3)$$

Theorem 1.3. Let $u_t(x)$ be the unique solution of equation (1.3), then there exists $\xi_0 > 0$ such that for all $\xi < \xi_0$ and $x \in B_R(0)$,

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

Fix $\varepsilon > 0$, then there exists $\xi_1 > 0$ such that for all $\xi > \xi_1$ and $x \in B_{R-\varepsilon}(0)$,

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

As in [8], we define the energy of the solution by the following quantity,

$$\mathcal{E}_t(\xi) = \sqrt{\mathbb{E}\|u_t\|_{L^2(B_R(0))}^2}. \quad (1.4)$$

The next corollary now follows easily from the above theorem.

Corollary 1.4. With ξ_0 and ξ_1 as in Theorem 1.3, we have

$$-\infty < \limsup_{t \rightarrow \infty} \frac{1}{t} \log \mathcal{E}_t(\xi) < 0 \quad \text{for all } \xi < \xi_0$$

and

$$0 < \liminf_{t \rightarrow \infty} \frac{1}{t} \log \mathcal{E}_t(\xi) < \infty \quad \text{for all } \xi > \xi_1.$$

Our next set of results concerns equation (1.1) with colored noise satisfying the conditions above. That is, we consider

$$\begin{cases} \partial_t u_t(x) = \mathcal{L}u_t(x) + \xi \sigma(u_t(x)) \dot{F}(t, x), & x \in B_R(0), \quad t > 0 \\ u_t(x) = 0, & x \in B_R(0)^c. \end{cases} \quad (1.5)$$

Theorem 1.5. Assume that u_t is the unique solution to equation (1.5). Then there exists $\xi_2 > 0$ such that for all $\xi < \xi_2$ and $x \in B_R(0)$

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

Fix $\varepsilon > 0$, then there exists $\xi_3 > 0$ such that for all $\xi > \xi_3$ and $x \in B_{R-\varepsilon}(0)$,

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

We then have the following easy consequence.

Corollary 1.6. Let ξ_2 and ξ_3 be as in Theorem 1.5, then

$$-\infty < \limsup_{t \rightarrow \infty} \frac{1}{t} \log \mathcal{E}_t(\xi) < 0 \quad \text{for all } \xi < \xi_2$$

and

$$0 < \liminf_{t \rightarrow \infty} \frac{1}{t} \log \mathcal{E}_t(\xi) < \infty \quad \text{for all } \xi > \xi_3.$$

We end this introduction with a plan of the article. In section 2, we provide some estimates needed for the proofs of our main results which are presented in section 3. Finally section 4 contains some extensions of our main results to higher moments and to some other non-local operators instead of the fractional Laplacian. Throughout this paper, the letter c with or without subscript(s) will denote a constant whose value is not important and can vary from place to place.

2 Some estimates

We begin this section with some estimates on heat kernel of the Dirichlet fractional Laplacian. For more information on these, see [1] and references therein.

•

$$p_D(t, x, y) \leq c_1(t^{-\frac{d}{\alpha}} \wedge \frac{t}{|x-y|^{d+\alpha}}). \quad (2.1)$$

We will often use the above inequality in the form of $p_D(t, x, y) \leq c_1 p(t, x-y)$, where $p_t(\cdot)$ is the heat kernel of the (unkilled) stable process.

- Fix $\epsilon > 0$ and let $x, y \in B_{R-\epsilon}(0)$, then for all $t \leq \epsilon^\alpha$,

$$p_D(t, x, y) \geq c_2(t^{-\frac{d}{\alpha}} \wedge \frac{t}{|x-y|^{d+\alpha}}). \quad (2.2)$$

- There exist $t_0 > 0$ and $\mu_1 > 0$ such that,

$$c_1 e^{-\mu_1 t} \leq p_D(t, x, y) \leq c_2 e^{-\mu_1 t} \quad \text{for } t \geq t_0. \quad (2.3)$$

The upper bound is valid for any $x, y \in B_R(0)$ while the lower bound is valid for $x, y \in B_{R-\epsilon}(0)$ with $\epsilon > 0$.

Our first lemma will be important for the white noise driven equation. The spatial dimension is restricted to $d = 1$.

Lemma 2.1. *There exists a constant $K_{\beta, \mu_1, \alpha}$ depending only on β , μ_1 and α such that for all $\beta \in (0, \mu_1)$ and $x \in B_R(0)$, we have*

$$\int_0^\infty e^{\beta t} p_D(t, x, x) dt \leq K_{\beta, \mu_1, \alpha}.$$

Proof. We begin by writing

$$\int_0^\infty e^{\beta t} p_D(t, x, x) dt = \int_0^{t_0} e^{\beta t} p_D(t, x, x) dt + \int_{t_0}^\infty e^{\beta t} p_D(t, x, x) dt,$$

where t_0 is as in (2.3). Now using (2.1), we have

$$\int_0^{t_0} e^{\beta t} p_D(t, x, x) dt \leq c_3 \int_0^{t_0} e^{\beta t} t^{-\frac{1}{\alpha}} dt,$$

where we have used the fact that $d = 1$. It is now clear that the above integral has an upper bound depending on β . Since $\beta < \mu_1$, we can use (2.3) to write

$$\begin{aligned} \int_{t_0}^{\infty} e^{\beta t} p_D(t, x, x) dt \\ \leq c_5 \int_{t_0}^{\infty} e^{-(\mu_1 - \beta)t} dt \\ \leq \frac{c_6}{\mu_1 - \beta}. \end{aligned}$$

Combining the estimates, we obtain the result. \square

Lemma 2.2. *Let $\beta \in (0, \mu_1)$ and $x \in B_R(0)$. Then there exists a constant $c_{R, \alpha}$ depending on R and α such that for all $t > 0$*

$$\int_{B_R(0)} e^{\beta t} p_D(t, x, y) dy \leq c_{R, \alpha}.$$

Proof. Fix t_0 as in (2.3). For $0 < t < t_0$, we have

$$\begin{aligned} \int_{B_R(0)} e^{\beta t} p_D(t, x, y) dy \\ \leq e^{\beta t} \int_{\mathbb{R}^d} p(t, x, y) dy \\ \leq e^{\beta t_0}, \end{aligned}$$

and for $t > t_0$ we use (2.3) to get

$$\begin{aligned} \int_{B_R(0)} e^{\beta t} p_D(t, x, y) dy \\ \leq c_2 e^{-(\mu_1 - \beta)t_0}. \end{aligned}$$

The result now easily follows from the two inequalities above. \square

Lemma 2.3. *Let $\beta \in (0, 2\mu_1)$. Then there exists a constant c_{β, μ_1} depending on β and μ_1 such that*

$$\int_0^{\infty} e^{\beta t} \int_{B_R(0) \times B_R(0)} p_D(t, x_1, y_1) p_D(t, x_2, y_2) f(y_1, y_1) dt dy_1 dy_2 \leq c_{\beta, \mu_1}, \quad (2.4)$$

for all $x_1, x_2, y_1, y_2 \in B_R(0)$.

Proof. We again use (2.3) so we fix t_0 accordingly. We begin by splitting the integral as follows,

$$\begin{aligned} \int_0^{\infty} e^{\beta t} \int_{B_R(0) \times B_R(0)} p_D(t, x_1, y_1) p_D(t, x_2, y_2) f(y_1, y_1) dt dy_1 dy_2 \\ = \int_0^{t_0} e^{\beta t} \int_{B_R(0) \times B_R(0)} p_D(t, x_1, y_1) p_D(t, x_2, y_2) f(y_1, y_1) dt dy_1 dy_2 \\ + \int_{t_0}^{\infty} e^{\beta t} \int_{B_R(0) \times B_R(0)} p_D(t, x_1, y_1) p_D(t, x_2, y_2) f(y_1, y_1) dt dy_1 dy_2 \\ := I_1 + I_2. \end{aligned}$$

I_1 can be bounded as follows: we use (2.1) to obtain

$$\begin{aligned} I_1 &\leq e^{\beta t_0} \int_0^{t_0} e^{-\beta t} e^{\beta t} \int_{B_R(0) \times B_R(0)} p_D(t, x_1, y_1) p_D(t, x_2, y_2) f(y_1, y_2) dt dy_1 dy_2 \\ &\leq e^{2\beta t_0} \int_0^\infty e^{-\beta t} \int_{\mathbb{R}^d \times \mathbb{R}^d} p(t, x_1, y_1) p(t, x_2, y_2) \tilde{f}(y_1 - y_2) dt dy_1 dy_2 \\ &\leq c_1 e^{2\beta t_0}. \end{aligned}$$

The last inequality needs some justifications which are quite straightforward under the current conditions; see [6] for details. For I_2 , we use (2.3) to write

$$\begin{aligned} I_2 &\leq \int_{t_0}^\infty e^{\beta t} \sup_{y_1, y_2 \in B_R(0)} p_D(t, x_1, y_1) p_D(t, x_2, y_2) dt \int_{B_R(0) \times B_R(0)} f(y_1 - y_2) dy_1 dy_2 \\ &\leq c_2 \int_0^\infty e^{-(2\mu_1 - \beta)t} dt. \end{aligned}$$

Combining the above estimates yields the result. \square

Lemma 2.4. *Fix $\varepsilon > 0$. Then, there exist $t_0 > 0$ and a constant c_{β, μ_1, t_0} such that for all $\beta > 0$,*

$$\int_0^\infty e^{-\beta t} p_D(t, x_1, y_1) p_D(t, x_2, y_2) dt \geq c_{\beta, \mu_1, t_0},$$

whenever $x_1, x_2, y_1, y_2 \in B_{R-\varepsilon}(0)$. The constant c_{β, μ_1, t_0} depends on β, μ_1 and t_0 .

Proof. Using (2.3), we have

$$\begin{aligned} &\int_0^\infty e^{-\beta t} p_D(t, x_1, y_1) p_D(t, x_2, y_2) dt \\ &\geq \int_{t_0}^\infty e^{-\beta t} p_D(t, x_1, y_1) p_D(t, x_2, y_2) dt \\ &\geq c_1 \int_{t_0}^\infty e^{-\beta t} e^{-2\mu_1 t} dt \\ &= \frac{c_2 e^{-(\beta + 2\mu_1)t_0}}{\beta + 2\mu_1}. \end{aligned}$$

\square

3 Proofs of main results.

3.1 Proof of Theorem 1.3

Proof of Theorem 1.3. Using (1.2) and the Walsh isometry, we have

$$\mathbb{E}|u_t(x)|^2 = |(\mathcal{G}_D u_0)_t(x)|^2 + \xi^2 \int_0^t \int_{B_R(0)} p_D^2(t-s, x, y) \mathbb{E}|\sigma(u_s(y))|^2 dy ds. \quad (3.1)$$

from which we obtain

$$\mathbb{E}|u_t(x)|^2 \geq |(\mathcal{G}_D u_0)_t(x)|^2. \quad (3.2)$$

Using the assumption on u_0 , we have for $\epsilon > 0$ small enough,

$$\begin{aligned} (\mathcal{G}_D u_0)_t(x) &= \int_{B_R(0)} u_0(y) p_D(t, x, y) \, dy \\ &\geq \int_{B_{R-\epsilon}(0)} u_0(y) p_D(t, x, y) \, dy \\ &\geq c_1 e^{-\mu_1 t}, \end{aligned}$$

whenever $t \geq t_0$ with t_0 as in (2.3) and $x \in B_{R-\epsilon}(0)$. This immediately gives

$$\liminf_{t \rightarrow \infty} \frac{1}{t} \log \mathbb{E}|u_t(x)|^2 > -\infty \quad \text{for } x \in B_{R-\epsilon}(0).$$

We now look at the upper bound. We will assume that $\beta \in (0, 2\mu_1)$. From (3.1) and the assumption on σ , we have

$$\begin{aligned} \mathbb{E}|u_t(x)|^2 &\leq |(\mathcal{G}_D u_0)_t(x)|^2 + \xi^2 L_\sigma^2 \int_0^t \int_{B_R(0)} p_D^2(t-s, x, y) \mathbb{E}|u_s(y)|^2 \, dy \, ds \\ &:= I_1 + I_2. \end{aligned}$$

Using Lemma 2.2, we have

$$\begin{aligned} I_1 &\leq c_2 e^{-\beta t} \left| \int_{B_R(0)} e^{\frac{\beta t}{2}} p_D(t, x, y) \, dy \right|^2 \\ &\leq c_2 e^{-\beta t}. \end{aligned}$$

We then look at the second term I_2 . Using the semigroup property and Lemma 2.1, we have

$$\begin{aligned} I_2 &= \xi^2 L_\sigma^2 e^{-\beta t} \int_0^t \int_{B_R(0)} e^{\beta(t-s)} p_D^2(t-s, x, y) e^{\beta s} \mathbb{E}|u_s(y)|^2 \, dy \, ds \\ &\leq \xi^2 L_\sigma^2 e^{-\beta t} \sup_{t>0, x \in B_R(0)} e^{\beta t} \mathbb{E}|u_t(x)|^2 \int_0^t \int_{B_R(0)} e^{\beta(t-s)} p_D^2(t-s, x, y) \, dy \, ds \\ &\leq \xi^2 L_\sigma^2 e^{-\beta t} \sup_{t>0, x \in B_R(0)} e^{\beta t} \mathbb{E}|u_t(x)|^2 \int_0^t e^{\beta s} p_D(2s, x, x) \, ds \\ &\leq K_{\beta, \mu_1, \alpha} \xi^2 L_\sigma^2 e^{-\beta t} \sup_{t>0, x \in B_R(0)} e^{\beta t} \mathbb{E}|u_t(x)|^2. \end{aligned}$$

Combining the above inequalities, we have

$$\sup_{t>0, x \in B_R(0)} e^{\beta t} \mathbb{E}|u_t(x)|^2 \leq c_2 + K_{\beta, \mu_1, \alpha} \xi^2 L_\sigma^2 \sup_{t>0, x \in B_R(0)} e^{\beta t} \mathbb{E}|u_t(x)|^2.$$

We now choose ξ_0 such that for $\xi \leq \xi_0$, we have $K_{\beta, \mu_1, \alpha} \xi^2 L_\sigma^2 < \frac{1}{2}$. This immediately gives

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

We have thus proved the first half of the theorem. For the second half, we look at the following ‘Laplace transform’,

$$I_\beta := \int_0^\infty e^{-\beta t} \inf_{x \in B_{R-\epsilon}(0)} \mathbb{E}|u_t(x)|^2 dx.$$

Using the mild formulation and the condition on σ , we have

$$\mathbb{E}|u_t(x)|^2 \geq |(\mathcal{G}_D u_0)_t(x)|^2 + \xi^2 l_\sigma^2 \int_0^t \int_{B_R(0)} p_D^2(t-s, x, y) \mathbb{E}|u_s(y)|^2 dy ds.$$

From the above, we have $I_\beta \geq I_1 + I_2$, where I_1 and I_2 are Laplace transforms of the first and second term of the above display respectively. We look at I_1 first. Note that for fixed $\epsilon > 0$,

$$\begin{aligned} \inf_{x \in B_{R-\epsilon}(0)} (\mathcal{G}_D u_0)_t(x) &\geq \int_{B_{R-\epsilon}(0)} u_0(y) p_D(t, x, y) dy \\ &\geq c_4 \inf_{x, y \in B_{R-\epsilon}(0)} p_D(t, x, y). \end{aligned}$$

Using (2.3), for $t \geq t_0$, we have

$$\begin{aligned} I_1 &\geq \int_{t_0}^\infty e^{-\beta t} \inf_{x \in B_{R-\epsilon}(0)} |(\mathcal{G}_D u_0)_t(x)|^2 dt \\ &\geq \frac{c_3 e^{-(\beta+2\mu_1)t_0}}{\beta + 2\mu_1}. \end{aligned}$$

For the second term, we obtain

$$\begin{aligned} I_2 &\geq \xi^2 l_\sigma^2 I_\beta \int_{t_0}^\infty e^{-\beta s} \inf_{x \in B_{R-\epsilon}(0)} p_D^2(s, x, y) dy \\ &\geq c_5 \xi^2 l_\sigma^2 I_\beta \frac{e^{-(\beta+2\mu_1)t_0}}{\beta + 2\mu_2}. \end{aligned}$$

Combining the above inequalities yields

$$I_\beta \geq \frac{c_3 e^{-(\beta+2\mu_1)t_0}}{\beta + 2\mu_1} + c_5 \xi^2 l_\sigma^2 I_\beta \frac{e^{-(\beta+2\mu_1)t_0}}{\beta + 2\mu_2}.$$

We can now choose ξ_1 large enough so that for $\xi \geq \xi_1$, we have

$$I_\beta \geq \frac{c_3 e^{-(\beta+2\mu_1)t_0}}{\beta + 2\mu_1} + 2I_\beta,$$

which gives us $I_\beta = \infty$. This proves

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

The fact that

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

easily follows from the ideas in [5]. We leave it to the reader to fill in the details. \square

3.2 Proof of Corollary 1.4

Proof of Corollary 1.4. The proof follows essentially from Theorem 1.3 and the definition of the energy of the solution together with the following estimate

$$|B_{R-\epsilon}(0)| \inf_{x \in B_{R-\epsilon}(0)} \mathbb{E}|u_t(x)|^2 \leq \int_{B_R(0)} \mathbb{E}|u_t(x)|^2 dx \leq |B_R(0)| \sup_{x \in B_R(0)} \mathbb{E}|u_t(x)|^2.$$

□

3.3 Proof of Theorem 1.5

While one can expect the proof of Theorem 1.5 to follow a similar pattern to that of Theorem 1.3, the noise term is now colored thus the proof is harder and requires a new idea. We provide the details of the proof of Theorem 1.5. The proof of Corollary 1.6 is omitted since it is similar to that of Corollary 1.4.

Proof of Theorem 1.5. Using the mild formulation of the solution and the assumption on σ , we obtain

$$\begin{aligned} \mathbb{E}|u_t(x)|^2 &= |(\mathcal{G}_D u_0)_t(x)|^2 + \xi^2 \int_0^t \int_{B_R(0) \times B_R(0)} p_D(t-s, x, y_1) p_D(t-s, x, y_2) f(y_1, y_2) \\ &\quad \times \mathbb{E}|\sigma(u_s(y_1))\sigma(u_s(y_2))| dy_1 dy_2 ds \\ &\leq |(\mathcal{G}_D u_0)_t(x)|^2 + \xi^2 L_\sigma^2 \int_0^t \int_{B_R(0) \times B_R(0)} p_D(t-s, x, y_1) p_D(t-s, x, y_2) f(y_1, y_2) \\ &\quad \times \mathbb{E}|u_s(y_1)u_s(y_2)| dy_1 dy_2 ds \\ &= I_1 + I_2. \end{aligned}$$

Set $\beta \in (0, 2\mu_1)$. Take t_0 as in (2.3). As is the proof of Theorem 1.3, we have

$$I_1 \leq c_1 e^{-\beta t} \quad \text{whenever } t > t_0.$$

We now bound I_2 by using Lemma 2.3.

$$\begin{aligned} I_2 &\leq \xi^2 L_\sigma^2 e^{-\beta t} \sup_{t>0, x \in B_R(0)} e^{\beta t} \mathbb{E}|u_t(x)|^2 \\ &\quad \times \int_0^\infty e^{\beta t} \int_{B_R(0) \times B_R(0)} p_D(t, x, y_1) p_D(t, x, y_2) f(y_1, y_2) dy_1 dy_2 dt \\ &\leq c_2 \xi^2 L_\sigma^2 e^{-\beta t} \sup_{t>0, x \in B_R(0)} e^{\beta t} \mathbb{E}|u_t(x)|^2. \end{aligned}$$

Using the two bounds above, we can use the arguments of the first part of the proof of Theorem 1.3 to show that

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

The first part of our first theorem also gives us

$$\liminf_{t \rightarrow \infty} \frac{1}{t} \log \mathbb{E}|u_t(x)|^2 > -\infty \quad \text{for } x \in B_{R-\epsilon}(0).$$

We now turn our attention to the final part of the proof. Fix $\beta, \epsilon > 0$ and consider the following 'Laplace transform',

$$J_\beta := \int_0^\infty e^{-\beta t} \inf_{x, y \in B_{R-\epsilon}(0)} \mathbb{E}|u_t(x)u_t(y)| dt.$$

From the mild solution, we have

$$\begin{aligned} \mathbb{E}(u_t(x_1)u_t(x_2)) &= (\mathcal{G}_D u_0)_t(x_1)(\mathcal{G}_D u_0)(x_2) + \xi^2 \int_0^t \int_{B_R(0) \times B_R(0)} p_D(t-s, x_1, y_1) \\ &\quad \times p_D(t-s, x_2, y_2) f(y_1, y_2) \mathbb{E}(\sigma(u_s(y_1))\sigma(u_s(y_2))) dy_1 dy_2 ds. \end{aligned} \quad (3.3)$$

Using the condition on σ , we have

$$\begin{aligned} \mathbb{E}(|u_t(x_1)u_t(x_2)|) &\geq |(\mathcal{G}_D u_0)_t(x_1)(\mathcal{G}_D u_0)(x_2)| \\ &\quad + \xi^2 l_\sigma^2 \int_0^t \int_{B_R(0) \times B_R(0)} p_D(t-s, x_1, y_1) p_D(t-s, x_2, y_2) f(y_1, y_2) \mathbb{E}|u_s(y_1)u_s(y_2)| dy_1 dy_2 ds \\ &:= J_1 + J_2. \end{aligned}$$

We bound J_2 first by using the condition on the correlation function.

$$\begin{aligned} J_2 &\geq \xi^2 l_\sigma^2 K_R \int_0^t \int_{B_{R-\epsilon}(0) \times B_{R-\epsilon}(0)} p_D(t-s, x_1, y_1) p_D(t-s, x_2, y_2) \mathbb{E}(|u_s(y_1)u_s(y_2)|) dy_1 dy_2 ds \\ &\geq c_3 \xi^2 l_\sigma^2 K_R \int_0^t \inf_{y_1, y_2 \in B_{R-\epsilon}(0)} p_D(t-s, x_1, y_1) p_D(t-s, x_2, y_2) \mathbb{E}(|u_s(y_1)u_s(y_2)|) ds. \end{aligned}$$

Using these estimates, we have

$$J_\beta \geq \tilde{J}_1 + \tilde{J}_2,$$

where \tilde{J}_1 and \tilde{J}_2 are the Laplace transforms of J_1 and J_2 respectively. As in the proof of Theorem 1.3, we have

$$\begin{aligned} \tilde{J}_1 &\geq \int_0^\infty e^{-\beta t} |(\mathcal{G}_D u_0)_t(x_1)(\mathcal{G}_D u_0)(x_2)| dt \\ &\geq \frac{c_4 e^{-(\beta+2\mu_1)t_0}}{\beta + 2\mu_1} \quad \text{for } x_1, x_2 \in B_{R-\epsilon}(0). \end{aligned}$$

\tilde{J}_2 can be estimated using Lemma 2.4 as follows.

$$\tilde{J}_2 \geq c_4 \xi^2 l_\sigma^2 K_R J_\beta \frac{e^{-(\beta+2\mu_1)t_0}}{\beta + 2\mu_1}.$$

We therefore have

$$J_\beta \geq \frac{c_4 e^{-(\beta+2\mu_1)t_0}}{\beta + \mu_1} + c_4 \xi^2 l_\sigma^2 K_R J_\beta \frac{e^{-(\beta+2\mu_1)t_0}}{\beta + 2\mu_1}.$$

Therefore there exists a $\xi_3 > 0$ such that we have $J_\beta = \infty$ for $\xi \geq \xi_3$. Using the ideas above, we have

$$\int_0^\infty e^{-\beta t} \mathbb{E}|u_t(x)|^2 dt \geq c_5 K_R J_\beta \frac{e^{-(\beta+2\mu_1)t_0}}{\beta + 2\mu_1} \quad \text{for } x \in B_{R-\epsilon}(0).$$

Therefore for $\xi \geq \xi_3$, we obtain

$$\int_0^\infty e^{-\beta t} \mathbb{E}|u_t(x)|^2 dt = \infty,$$

which implies that

$$\liminf_{t \rightarrow \infty} \frac{1}{t} \log \mathbb{E}|u_t(x)|^2 > 0 \quad \text{for } x \in B_{R-\epsilon}(0)..$$

Again the ideas of [6] give

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

The theorem is therefore proved. □

4 Some extensions

We conclude this paper with some extensions that can be proved using the methods developed in our paper. Since our main theorems (Theorem 1.3 & Theorem 1.5) are about second moments of the solution to the corresponding equation, one may naturally ask if they also hold for higher moments. This is actually answered in the following theorems.

Theorem 4.1. *If u_t is the unique solution to (1.3), then for all $p \geq 2$, there exists $\xi_0(p) > 0$ such that for all $\xi < \xi_0(p)$ and $x \in B_R(0)$,*

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

On the other hand, for all $\varepsilon > 0$, there exists $\xi_1(p) > 0$ such that for all $\xi > \xi_1(p)$ and $x \in B_{R-\epsilon}(0)$,

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

The proof of this theorem follows from the Burkholder-Davis-Gundy inequality for the upper bound and Jensen's inequality for the lower bound. We do not provide a proof here. The reader can refer to [8] for details. Next we state a result similar to Theorem 4.1 in higher dimension.

Theorem 4.2. *Let $u_t(x)$ be the unique solution to (1.5), then for all $p \geq 2$ there exists $\xi_2(p) > 0$ such that for all $\xi < \xi_2(p)$ and $x \in B_R(0)$*

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

On the other hand, for all $\varepsilon > 0$, there exists $\xi_3(p) > 0$ such that for all $\xi > \xi_3(p)$ and $x \in B_{R-\varepsilon}(0)$

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

In the remainder of this section, we show that the method developed in this paper can be used to study problems with operators other than the fractional Laplacian.

Example 4.3. Consider the following stochastic heat equation with linear drift

$$\begin{cases} \partial_t u_t(x) = -\nu(-\Delta)^{\frac{\alpha}{2}} u_t(x) + \lambda u_t(x) + \xi \sigma(u_t(x)) \dot{W}(t, x), & x \in B_R(0), \quad t > 0 \\ u_t(x) = 0, & x \in B_R(0)^c, \quad t > 0 \end{cases} \quad (4.1)$$

where λ is a real number and all the other conditions are the same as in (1.3). The mild solution is given by

$$u_t(x) = (\mathcal{G}_D^* u_0)_t(x) + \xi \int_{B_R(0)} \int_0^t p_D^*(t-s, x, y) \sigma(u_s(y)) F(ds, dy) \quad (4.2)$$

where

$$p_D^*(t, x, y) = e^{\lambda t} p_D(t, x, y)$$

and

$$(\mathcal{G}_D^* u_0)_t(x) = \int_{B_R(0)} u_0(y) p_D^*(t, x, y) dy.$$

Then there exists $\xi_0(\lambda) > 0$ such that for all $\xi < \xi_0(\lambda)$ and $x \in B_R(0)$

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

while for $\varepsilon > 0$ there exists $\xi_1(\lambda) > 0$ such that for all $\xi > \xi_1(\lambda)$ and $x \in B_{R-\varepsilon}(0)$

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

The proof is very similar to that of Theorem 1.3, we only need to adjust Lemma 2.1 as follows:

$$\int_0^\infty e^{\beta t} p_D^*(t, x, x) dt \leq c_1 \left[(\lambda + \beta)^{-\frac{1}{\alpha}} + \frac{1}{\mu_1 - (\lambda + \beta)} \right].$$

for all $\beta > 0$, $x \in B_R(0)$ provided $0 < \lambda + \beta < \mu_1$. Now if $x, y \in B_{R-\varepsilon}(0)$ and $2(\mu_1 - \lambda) + \beta > 0$, we have

$$\int_0^\infty e^{-\beta t} (p_D^*(t, x, y))^2 dt \geq c_2 \frac{e^{-(2\mu_1 - 2\lambda + \beta)t_0}}{2\mu_1 + 2\lambda - \beta},$$

where c_1 and c_2 are some positive constants. With some modifications of the proofs above, Theorems 1.3, 1.5, 4.1 and 4.2 hold for the solution of equation (4.1).

Example 4.4. In this example we consider the generator of a relativistic stable process killed upon exiting $B_R(0)$ instead of the fractional Laplacian. We therefore look at

$$\begin{cases} \partial_t u_t(x) = m u_t(x) - (m^{\frac{2}{\alpha}} - \Delta)^{\frac{\alpha}{2}} u_t(x) + \xi \sigma(u_t(x)) \dot{F}(t, x), & x \in B_R(0), \quad t > 0 \\ u_t(x) = 0, & x \in B_R(0)^c. \end{cases} \quad (4.3)$$

Here m is some fixed positive number and all the other conditions are the same as in (1.5). We refer the reader to [3] for the needed heat kernel bounds to prove appropriate versions of Theorems 1.3, 1.5, 4.1 and 4.2. We leave it for the reader to fill in the details.

Example 4.5. We conclude this section with this interesting problem. Let $1 < \beta < \alpha < 2$ and consider the following:

$$\begin{cases} \partial_t u_t(x) = -\nu(-\Delta)^{\frac{\alpha}{2}} u_t(x) - a^\beta (-\Delta)^{\frac{\beta}{2}} u_t(x) + \xi \sigma(u_t(x)) \dot{F}(t, x), & x \in B_R(0), \quad t > 0 \\ u_t(x) = 0, & x \in B_R(0)^c. \end{cases} \quad (4.4)$$

Here we refer the reader to [2] for the heat kernel bounds needed to prove suitable versions of Theorems 1.3, 1.5, 4.1 and 4.2 for the solution of the above equation. We leave it for the reader to fill in the details.

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