

# Approximations of fractional Brownian motion

YUQIANG LI<sup>1</sup> and HONGSHUAI DAI<sup>2</sup>

<sup>1</sup>*School of Finance and Statistics, East China Normal University, Shanghai 200241, China.  
E-mail: yqli@stat.ecnu.edu.cn*

<sup>2</sup>*School of Mathematics, Central South University, Changsha 410075, China.  
E-mail: math\_dsh@yahoo.com.cn*

Approximations of fractional Brownian motion using Poisson processes whose parameter sets have the same dimensions as the approximated processes have been studied in the literature. In this paper, a special approximation to the one-parameter fractional Brownian motion is constructed using a two-parameter Poisson process. The proof involves the tightness and identification of finite-dimensional distributions.

*Keywords:* fractional Brownian motion; Poisson process; weak convergence

## 1. Introduction

The fractional Brownian motion with Hurst index  $H \in (0, 1)$  is a centered Gaussian process  $B^H = \{B^H(t), t \geq 0\}$  with covariance function

$$\mathbb{E}[B^H(t)B^H(s)] = \frac{1}{2}(|s|^{2H} + |t|^{2H} - |t-s|^{2H}). \quad (1.1)$$

It follows from (1.1) that  $B^H$  is self-similar with index  $H$  and has stationary increments. Unless  $H = 1/2$  (i.e.,  $B^H$  is Brownian motion),  $B^H$  is not Markovian. Moreover, it is known that  $B^H$  has long-range dependence if  $H \in (1/2, 1)$  and short-range dependence if  $H \in (0, 1/2)$  (see Samorodnitsky and Taqqu [14]). These properties have made  $B^H$  not only important theoretically, but also very popular as stochastic models in many areas including telecommunications, biology, hydrology and finance.

Weak convergence to fractional Brownian motion has been studied extensively since the works of Davydov [7] and Taqqu [16]. In recent years many new results on approximations of fractional Brownian motion have been established. For example, Enriquez [9] showed that fractional Brownian motion can be approximated in law by appropriately normalized correlated random walks. Meyer, Sellan and Taqqu [13] proved that the law of  $B^H$  can be approximated by those of a random wavelet series. By extending Stroock [15], Bardina *et al.* [4] and Delgado and Jolis [8] have established approximations in law to fractional Brownian motions by processes constructed using Poisson processes.

Let  $\{N(t), t \geq 0\}$  be a standard Poisson process, and for all  $\varepsilon > 0$ , define the processes  $X_\varepsilon = \{X_\varepsilon(t), t \geq 0\}$  by

$$X_\varepsilon(t) = \frac{1}{\varepsilon} \int_0^t (-1)^{N(r/\varepsilon^2)} dr, \quad t \geq 0.$$

Stroock [15] proved that as  $\varepsilon$  tends to zero, the laws of  $X_\varepsilon$  converge weakly in the Banach space  $\mathcal{C}[0, 1]$  (i.e., the space of continuous functions on  $[0, 1]$ ) to the law of Brownian motion. Delgado and Jolis [8] proved that every Gaussian process of the form

$$X_t = \int_0^1 K(t, s) dB_s,$$

where  $B$  is a one-dimensional Brownian motion and  $K$  a sufficiently regular deterministic kernel, can be weakly approximated by the family of processes

$$Y^\varepsilon(t) = \frac{1}{\varepsilon} \int_0^1 K(t, r)(-1)^{N(r/\varepsilon^2)} dr.$$

Their result can be applied to fractional Brownian motion. In addition, Bardina and Jolis [2] proved that as  $\varepsilon$  tends to 0, the family of two-parameter random fields  $Y_\varepsilon$  defined by

$$Y_\varepsilon(s, t) = \int_0^t \int_0^s \frac{1}{\varepsilon^2} \sqrt{xy}(-1)^{N(x/\varepsilon, y/\varepsilon)} dx dy \tag{1.2}$$

converges in law in the space of continuous functions on  $[0, 1]^2$  to the standard Brownian sheet. Bardina, Jolis and Tudor showed in [4] that as  $\varepsilon$  tends to 0, the family of two-parameter random fields

$$\hat{Y}_\varepsilon(s, t) = \int_0^1 \int_0^1 \frac{1}{\varepsilon^2} f(s, t, x, y) \sqrt{xy}(-1)^{N(x/\varepsilon, y/\varepsilon)} dx dy \tag{1.3}$$

converges in law to the two-parameter Gaussian process

$$W(s, t) = \int_0^1 \int_0^1 f(s, t, x, y) B(dx, dy),$$

where  $B$  is a standard Brownian sheet and the deterministic kernel

$$f(s, t, x, y) = f_1(s, x) f_2(t, y) \tag{1.4}$$

can be separated by the integration variables and satisfies certain conditions. As examples, the authors include the fractional Brownian sheet, among others. For more information, see Bardina, Jolis and Rovira [3], where an approximation to the  $d$ -parameter Wiener process by a  $d$ -parameter Poisson process was provided, and Bardina and Bascompte [1], where two independent Gaussian processes were constructed using a unique Poisson process.

We note that in the serial works [1–4,8], the dimension of the parameter set is always the same for the approximating and the approximated processes. Naturally, we will be interested in the problem of whether we can approximate the  $d$ -parameter fractional Brownian motions by  $r$ -parameter Poisson processes if  $d \neq r$ . The purpose of this paper is to study this problem in the case of  $d = 1$  and  $r = 2$ . We find that for a special deterministic kernel function which cannot be separated with respect to the integration variables, the answer is affirmative. Below, we introduce the deterministic kernel function.

In order to study a non-Gaussian and non-stable process arising as the limit of sums of rescaled renewal processes under the condition of intermediate growth, Gaigalas [11], page 451, introduced the function  $h(t, x, y)$ , defined as follows. For  $x, t \geq 0$  and  $y \in \mathbb{R}$ ,

$$\begin{aligned}
 h(t, x, y) &= ((t + y) \wedge 0 + x)_+ - (y \wedge 0 + x)_+ \\
 &= \int_{-t}^0 \mathbf{1}_{[0, x]}(u - y) \, du = \begin{cases} t, & -y < x, y < -t, \\ x + t + y, & -t - y < x \leq -y, y < -t, \\ -y, & -y < x, -t \leq y < 0, \\ x, & 0 < x \leq -y, -t \leq y < 0, \\ 0, & \text{otherwise.} \end{cases} \quad (1.5)
 \end{aligned}$$

Note that Kaj and Taqqu [12], page 388, interpreted the function

$$K_t(y, x) := (t - y)_+ \wedge x - (-y)_+ \wedge x = h(t, x, -x - y)$$

in the context of the infinite source Poisson model as a function of the starting time  $y$  and the duration  $x$  of a session that measures the length of the time interval contained in  $[0, t]$  during which the session is active. Therefore,  $h(t, x, y) = K_t(-x - y, x)$  measures the length of the time interval contained in  $[0, t]$  during which the session with duration  $x$  and finishing time  $-y$  is active. Obviously, by this definition,  $h(t, x, y) \neq 0$  if and only if  $y < 0$  and  $-y - x < t$ , that is,  $x + y > -t$ .

In this paper, we define

$$\begin{aligned}
 g_s(t, x, y) &:= h(t + s, x, y) - h(s, x, y) \\
 &= \int_{-t-s}^0 \mathbf{1}_{[0, x]}(u - y) \, du - \int_{-s}^0 \mathbf{1}_{[0, x]}(u - y) \, du \\
 &= \int_{-t-s}^{-s} \mathbf{1}_{[0, x]}(u - y) \, du = \int_{-t}^0 \mathbf{1}_{[0, x]}(u - y - s) \, du = h(t, x, y + s)
 \end{aligned} \quad (1.6)$$

for all  $t \geq 0, x \geq 0, y \in \mathbb{R}$  and any given  $s > 0$ . Then, according to the definition of an integral of random measure (see [14], Chapter 3), we can directly verify that for  $H \in (\frac{1}{2}, 1)$ ,

$$\begin{aligned}
 &\sqrt{C_H} \int_0^\infty \int_{\mathbb{R}_-} g_s(t, x, y) x^{H-2} W(\mathrm{d}x, \mathrm{d}y) \\
 &= \sqrt{C_H} \int_0^\infty \int_{\mathbb{R}_-} h(t, x, y + s) x^{H-2} W(\mathrm{d}x, \mathrm{d}y) \\
 &\stackrel{d}{=} \sqrt{C_H} \int_0^\infty \int_{\mathbb{R}_-} h(t, x, y) x^{H-2} W(\mathrm{d}x, \mathrm{d}y) \stackrel{d}{=} B^H(t),
 \end{aligned} \quad (1.7)$$

where  $W(\mathrm{d}x, \mathrm{d}y)$  is a Gaussian random measure on  $\mathbb{R}_+ \times \mathbb{R}$  with control measure  $\mathrm{d}x \, \mathrm{d}y$  (see Section 2 for its definition),  $C_H = H(2H - 1)(1 - H)(3 - 2H)$  and the notation  $\stackrel{d}{=}$  denotes

identification of finite-dimensional distributions. In fact, the last equality is taken from Gaigalas [11], page 454, although the constant  $C_H$  is omitted in Gaigalas' representation. Therefore,

$$C_H \int_0^\infty \int_{\mathbb{R}_-} \frac{g_s^2(t, x, y)}{x^{4-2H}} dx dy = \frac{1}{2} \mathbb{E}[(B^H(t))^2] = \frac{t^{2H}}{2}. \tag{1.8}$$

From the representations (1.2) and (1.7), inspired by (1.3), it seems reasonable that the law of the process  $B^H$  can be approximated by that of some process similar to  $\hat{Y}_\varepsilon$  with kernel  $g_s(t, x, y)$ .

In this paper we define a sequence of processes  $\{Y_n(t), t \in [0, 1]\}_{n \geq 1}$  as follows:

$$Y_n(t) = n\sqrt{2C_H} \int_0^n \int_{-n}^0 g_s(t, x, y) x^{H-2} \sqrt{|x|y|} (-1)^{N_n(x,y)} dx dy \tag{1.9}$$

for  $H \in (1/2, 1)$ . Here,  $\{N_n(x, y), (x, y) \in \mathbb{R}_+ \times \mathbb{R}_-\}$  is a Poisson process with intensity  $n$  (see Definition 2.1).

The main purpose of this paper is to show that the law of  $Y_n$  converges to the law of  $B^H$  for  $H \in (1/2, 1)$ . We note that the kernel  $g_s(t, x, y)$  in (1.9) cannot be separated by the arguments  $(x, y)$ , unlike the kernel function in (1.3). This difference is not trivial. As we will see in Remark 3.1, it causes many real technical difficulties.

The rest of the paper is organized as follows. Section 2 is devoted to introducing the necessary definitions, notation and the main result. In Section 3 we prove the family of processes  $\{Y_n(t)\}_{n \geq 1}$  given by (1.9) is tight in  $C[0, 1]$ . In Section 4 we prove that as  $n$  tends to infinity, the finite-dimensional distributions of  $\{Y_n(t)\}$  converge weakly to those of the fractional Brownian motion  $B^H$  with  $H \in (1/2, 1)$  and hence  $\{Y_n(t)\}$  converges weakly in  $C[0, 1]$  to the fractional Brownian motion  $B^H$ .

## 2. Preliminaries

We now give the definitions of the Brownian sheet and Poisson processes on  $\mathbb{R} \times \mathbb{R}$ . Let  $\mathcal{F}$  be the Borel algebra on  $\mathbb{R} \times \mathbb{R}$ .  $\nu$  and  $\mu$  denote a  $\sigma$ -finite measure and the Lebesgue measure on  $\mathbb{R} \times \mathbb{R}$ , respectively.

**Definition 2.1.** *Given a positive constant  $\beta > 0$ , a random set function  $N(\cdot)$  on the measure space  $(\mathbb{R} \times \mathbb{R}, \mathcal{F}, \nu)$  is called the Poisson random measure with density measure  $\beta\nu$  if it satisfies the following conditions:*

- (i) *for every  $A \in \mathcal{F}$  with  $\mu(A) < \infty$ ,  $N(A)$  is a Poisson random variable with parameter  $\beta\nu(A)$  defined on the same probability space;*
- (ii) *if  $A_1, \dots, A_n \in \mathcal{F}$  are disjoint and all have finite measure, then the random variables  $N(A_1), \dots, N(A_n)$  are independent;*
- (iii) *if  $A_1, A_2, \dots \in \mathcal{F}$  are disjoint and  $\nu(\bigcup_{i=1}^\infty A_i) < \infty$ , then  $N(\bigcup_{i=1}^\infty A_i) = \sum_{i=1}^\infty N(A_i)$  a.s.*

If  $N(\cdot)$  is a Poisson random measure with density measure  $\beta\mu$ , then we define

$$N(s, t) = \begin{cases} N([0, s] \times [0, t]), & s \geq 0, t \geq 0, \\ N([0, s] \times [t, 0]), & s \geq 0, t \leq 0, \\ N([s, 0] \times [0, t]), & s \leq 0, t \geq 0, \\ N([s, 0] \times [t, 0]), & s \leq 0, t \leq 0 \end{cases}$$

and call  $N = \{N(s, t), (s, t) \in \mathbb{R} \times \mathbb{R}\}$  the two-parameter Poisson process with intensity  $\beta$  in  $\mathbb{R} \times \mathbb{R}$ .

It is not hard to see that  $N = \{N(s, t), (s, t) \in \mathbb{R}_+ \times \mathbb{R}_-\}$  is independent with  $N_1 = \{N(s, t), (s, t) \in \mathbb{R}_+ \times \mathbb{R}_+\}$ , which is the ordinary two-parameter process in  $\mathbb{R}_+ \times \mathbb{R}_+$ , and for any  $(s, t) \in \mathbb{R}_+ \times \mathbb{R}_-$ ,  $\{N(s, t)\}$  has the same finite-dimensional distributions as those of  $\{N(s, |t|)\}$ .

**Definition 2.2.** Consider a random set function  $W(\cdot)$  on the measure space  $(\mathbb{R} \times \mathbb{R}, \mathcal{F}, \nu)$  such that:

- (i) for every  $A \in \mathcal{F}$  with  $\nu(A) < \infty$ ,  $W(A)$  is a centered Gaussian random variable defined on the same probability space with variance  $2\nu(A)$ ;
- (ii) if  $A_1, \dots, A_n \in \mathcal{F}$  are disjoint and have finite measure, then the random variables  $W(A_1), \dots, W(A_n)$  are independent;
- (iii) if  $A_1, A_2, \dots \in \mathcal{F}$  are disjoint and  $\nu(\bigcup_{i=1}^\infty A_i) < \infty$ , then  $W(\bigcup_{i=1}^\infty A_i) = \sum_{i=1}^\infty W(A_i)$  a.s.

We then call  $W(\cdot)$  a Gaussian random measure on  $\mathbb{R} \times \mathbb{R}$  with control measure  $\nu$ . In particular, if  $W(\cdot)$  is a Gaussian random measure on  $\mathbb{R} \times \mathbb{R}$  with control measure  $\frac{1}{2}\mu$ , then we define

$$B(s, t) = \begin{cases} W([0, s] \times [0, t]), & s \geq 0, t \geq 0, \\ W([0, s] \times [t, 0]), & s \geq 0, t \leq 0, \\ W([s, 0] \times [0, t]), & s \leq 0, t \geq 0, \\ W([s, 0] \times [t, 0]), & s \leq 0, t \leq 0 \end{cases}$$

and call  $B = \{B(s, t), (s, t) \in \mathbb{R} \times \mathbb{R}\}$  the two-parameter Brownian sheet in  $\mathbb{R} \times \mathbb{R}$ .

Similarly, we have that  $B = \{B(s, t), (s, t) \in \mathbb{R}_+ \times \mathbb{R}_-\}$  is independent of  $B_1 = \{B(s, t), (s, t) \in \mathbb{R}_+ \times \mathbb{R}_+\}$ , which is the ordinary Brownian sheet in  $\mathbb{R}_+ \times \mathbb{R}_+$ , and for any  $(s, t) \in \mathbb{R} \times \mathbb{R}$ , we have  $B(s, t) \stackrel{d}{=} B(|s|, |t|)$ . Hence, from (1.7) it is easy to check that

$$B^H(t) \stackrel{d}{=} \sqrt{2C_H} \int_0^\infty \int_{\mathbb{R}_-} g_s(t, x, y) x^{H-2} B(dx, dy). \tag{2.1}$$

Let  $\text{sgn}(x) = 1$  if  $x \geq 0$  and  $\text{sgn}(x) = -1$  if  $x < 0$ . We have the following conclusion, which essentially parallels Bardina and Jolis [2], Theorem 1.1. The proof is omitted.

**Lemma 2.1.** *Suppose that  $N = \{N(s, t), (s, t) \in \mathbb{R} \times \mathbb{R}\}$  is a two-parameter Poisson process with intensity 1 in  $\mathbb{R} \times \mathbb{R}$ . For any  $S > 0$  and  $T > 0$ , let*

$$B_n(u, v) = \text{sgn}(uv)n \int_0^u \int_0^v \sqrt{|xy|}(-1)^{N(x\sqrt{n}, y\sqrt{n})} \, dx \, dy \tag{2.2}$$

for any  $|u| \leq S, |v| \leq T$ . The finite-dimensional distributions of  $B_n$  then converge weakly to those of a two-parameter Brownian sheet  $B = \{B(u, v), |u| \leq S, |v| \leq T\}$ .

Naturally, (2.1) and (2.2) suggest that we consider the following approximation of  $B^H$  for  $H \in (1/2, 1)$ :

$$Y_n(t) = n\sqrt{2C_H} \int_0^n \int_{-n}^0 g_s(t, x, y)x^{H-2}\sqrt{|xy|}(-1)^{N_n(x, y)} \, dx \, dy \tag{2.3}$$

for  $n \in \mathbb{N} = \{1, 2, \dots\}$  and  $t \in [0, 1]$ , where  $N_n = \{N_n(x, y)\} = \{N(x\sqrt{n}, y\sqrt{n})\}$  is the two-parameter Poisson process with intensity  $n$ .

The main result of this paper is as follows.

**Theorem 2.1.** *For any fixed  $s > 0$ , the law of the process  $\{Y_n(t), t \in [0, 1]\}$  given by (2.3) converges weakly to the law of  $\{B^H(t), t \in [0, 1]\}$  for  $H \in (\frac{1}{2}, 1)$  in  $\mathcal{C}[0, 1]$ .*

**Remark 2.1.** If we define  $\bar{h}(t, x, y) := h(t, x, -y)$ , then  $\bar{h}(t, x, y)$  has a more natural physical interpretation. It measures the length of the time interval contained in  $[0, t]$  during which the session with duration  $x$  and finishing time  $y$  is active. Define

$$\bar{Y}_n(t) := n\sqrt{2C_H} \int_0^n \int_0^n \bar{g}_s(t, x, y)x^{H-2}\sqrt{xy}(-1)^{N_n(x, y)} \, dx \, dy,$$

where  $\bar{g}_s(t, x, y) = \bar{h}(t + s, x, y) - \bar{h}(s, x, y) = \bar{h}(t, x, y - s)$ . It is then easy to see that Theorem 2.1 holds for  $\bar{Y}_n$ .

Note that  $s$  is a given positive number. In the sequel we will treat it as a constant. In addition, since the parameter  $C_H$  cannot affect our discussion, we will take it to be 1 in order to simplify matters.

We now introduce some auxiliary notation. In the sequel,  $\infty$  and  $-\infty$  will denote positive infinity and negative infinity, respectively. For all  $x' \geq x \geq 0, y' \geq y \geq 0$ , by Cairoli and Walsh [6], we define

$$\Delta_{(x, y)}N(x', y') = N((x, x'] \times (y, y']). \tag{2.4}$$

For any  $t \in [0, 1], x > 0$  and  $y \leq 0$ , we let

$$\phi_t(x, y) := g_s(t, x, y)/x^{2-H} \geq 0.$$

Define a function  $F_{n,f}(x, y)$  as follows:

$$F_{n,f}(x, y) = \sqrt{x|y|}n^2 \int_x^\infty \int_y^0 f(x_2, y_2)\sqrt{x_2|y_2|}e^{-2n[x(y_2-y)-(x_2-x)y_2]} dx_2 dy_2, \tag{2.5}$$

where  $n > 0, x \geq 0, y \leq 0$  and  $f$  is a measurable function such that the integral is meaningful. Obviously, if  $0 \leq f \leq g$ , then  $0 \leq F_{n,f}(x, y) \leq F_{n,g}(x, y)$ .

### 3. Tightness of $Y_n$ in $\mathcal{C}[0, 1]$

Let  $Y_n = \{Y_n(t), t \in [0, 1]\}$  be the process defined by (2.3). The purpose of this section is to prove the tightness of the processes  $\{Y_n\}_{n \geq 1}$ .

**Proposition 3.1.** *The family of laws of the processes  $\{Y_n\}_{n \geq 1}$  is tight in  $\mathcal{C}[0, 1]$ .*

To prove the proposition, we need the following lemmas.

**Lemma 3.1.** *Let  $\Omega_1 = \{0 < x_1 \leq x_2, y_2 \leq y_1 \leq 0\}$  and  $\Omega_2 = \{0 < x_1 \leq x_2, y_1 < y_2 \leq 0\}$ . For a non-negative function  $f(x, y)$ , if  $\int_0^\infty \int_{-\infty}^0 f^2(x, y) dx dy < \infty$ , then for any  $u, v > 0$ ,*

$$\begin{aligned} & \mathbb{E} \left[ \left( n \int_0^u \int_{-v}^0 f(x, y)\sqrt{x|y|}(-1)^{N_n(x,y)} dx dy \right)^2 \right] \\ & \leq 2(I_1(n, f) + I_2(n, f)), \end{aligned} \tag{3.1}$$

where

$$I_1(n, f) = n^2 \int_{\Omega_1} \prod_{i=1}^2 (f(x_i, y_i)\sqrt{x_i|y_i|})e^{-2n(x_1y_1-x_2y_2)} dx_1 dy_1 dx_2 dy_2, \tag{3.2}$$

$$I_2(n, f) = n^2 \int_{\Omega_2} \prod_{i=1}^2 (f(x_i, y_i)\sqrt{x_i|y_i|})e^{-2n[x_1(y_2-y_1)-(x_2-x_1)y_2]} dx_1 dy_1 dx_2 dy_2. \tag{3.3}$$

**Proof.** Let  $I(n, x_{1,2}, y_{1,2}) = \mathbb{E}[(-1)^{N_n(x_1, y_1)+N(x_2, y_2)}]$ . By Fubini’s theorem, the left-hand side of (3.1) is equal to

$$n^2 \int_0^u \int_{-v}^0 \int_0^u \int_{-v}^0 f(x_1, y_1)f(x_2, y_2)\sqrt{x_1|y_1|}\sqrt{x_2|y_2|}I(n, x_{1,2}, y_{1,2}) dx_1 dy_1 dx_2 dy_2. \tag{3.4}$$

Define  $\Omega_3 = \{x_1 > x_2 > 0, 0 \geq y_1 \geq y_2\}$  and  $\Omega_4 = \{x_1 > x_2 > 0, y_1 < y_2 \leq 0\}$ . Then, by (2.4),  $I(n, x_{1,2}, y_{1,2})$  equals

$$\begin{aligned} & \mathbb{E}[(-1)^{N_n(x_1, |y_1|)+N_n(x_2, |y_2|)}] \\ & = \mathbb{E}[(-1)^{\Delta_{(0,0)}N_n(x_1, |y_1|)+\Delta_{(0,0)}N_n(x_2, |y_2|)}]. \end{aligned} \tag{3.5}$$

Note that  $\sum_{i=1}^2 \Delta_{(0,0)} N_n(x_i, |y_i|)$  is equal to the sum of the increments of the Poisson process over some disjoint rectangles, and the rectangles which contribute to the value of  $I(n, x_{1,2}, y_{1,2})$  are those which appear only once. Since two-parameter Poisson processes have independent increments, after some simple calculation, we obtain that on  $\Omega_1$ ,

$$\begin{aligned} I(n, x_{1,2}, y_{1,2}) &= \mathbb{E}\left[(-1)^{\Delta_{(0,|y_1|)}N_n(x_1,|y_2|)+\Delta_{(x_1,|y_1|)}N_n(x_2,|y_2|)+\Delta_{(x_1,0)}N_n(x_2,|y_1|)}\right] \\ &= \mathbb{E}\left[(-1)^{\Delta_{(0,|y_1|)}N_n(x_1,|y_2|)}\right]\mathbb{E}\left[(-1)^{\Delta_{(x_1,|y_1|)}N_n(x_2,|y_2|)}\right]\mathbb{E}\left[(-1)^{\Delta_{(x_1,0)}N_n(x_2,|y_1|)}\right] \\ &= \exp\{-2n(x_1y_1 - x_2y_2)\}. \end{aligned} \tag{3.6}$$

Using the same method as above, we obtain that

$$I(n, x_{1,2}, y_{1,2}) = \exp\{-2n[x_1(y_2 - y_1) - (x_2 - x_1)y_2]\} \quad \text{on } \Omega_2, \tag{3.7}$$

$$I(n, x_{1,2}, y_{1,2}) = \exp\{-2n[x_2(y_1 - y_2) - (x_1 - x_2)y_1]\} \quad \text{on } \Omega_3, \tag{3.8}$$

$$I(n, x_{1,2}, y_{1,2}) = \exp\{-2n(x_2y_2 - x_1y_1)\} \quad \text{on } \Omega_4. \tag{3.9}$$

Substituting (3.6)–(3.9) into (3.4) and using a change of the integration variables if necessary, we can easily obtain (3.1). □

**Lemma 3.2.** *If  $\int_0^\infty \int_{-\infty}^0 f^2(x, y) \, dx \, dy < \infty$ , then  $I_1(n, f)$  defined by (3.2) is such that*

$$I_1(n, f) \leq \frac{37}{8} \int_0^\infty \int_{-\infty}^0 f^2(x, y) \, dx \, dy. \tag{3.10}$$

**Proof.** Define  $B = \{0 < x_1 \leq x_2 \leq 2x_1\}$ ,  $C = \{2y_1 \leq y_2 \leq y_1 \leq 0\}$  and  $A = B \cap C$ . Let  $I_1^A(n)$ ,  $I_1^B(n)$  and  $I_1^C(n)$  denote the integral (3.2), where  $\Omega_1$  is replaced by  $A$ ,  $\Omega_1 \setminus B$  and  $\Omega_1 \setminus C$ , respectively. Then,

$$I_1(n, f) \leq I_1^A(n, f) + I_1^B(n, f) + I_1^C(n, f). \tag{3.11}$$

Using the elementary inequality  $2ab \leq a^2 + b^2$ , from (3.2), we have that

$$I_1^A(n, f) \leq \frac{1}{2}(I_{11}(n) + I_{12}(n)), \tag{3.12}$$

where  $I_{11}(n)$  is

$$\begin{aligned} &n^2 \int_A f^2(x_1, y_1)x_1|y_1|e^{-2n(x_1y_1-x_2y_2)} \, dx_1 \, dy_1 \, dx_2 \, dy_2 \\ &= n^2 \int_0^\infty \int_{-\infty}^0 \int_{x_1}^{2x_1} \int_{2y_1}^{y_1} f^2(x_1, y_1)x_1|y_1|e^{-2n(x_1y_1-x_2y_2)} \, dx_1 \, dy_1 \, dx_2 \, dy_2 \end{aligned}$$



and  $I_{12}(n)$  is

$$\begin{aligned} n^2 \int_A f^2(x_2, y_2)x_2|y_2|e^{-2n(x_1y_1-x_2y_2)} dx_1 dy_1 dx_2 dy_2 \\ = n^2 \int_0^\infty \int_{-\infty}^0 \int_{x_1}^{2x_1} \int_{2y_1}^{y_1} f^2(x_2, y_2)x_2|y_2|e^{-2n(x_1y_1-x_2y_2)} dx_1 dy_1 dx_2 dy_2. \end{aligned}$$

Since, for any  $(x_1, y_1, x_2, y_2) \in \Omega_1$ ,

$$x_1y_1 - x_2y_2 \geq (x_2 - x_1)|y_1| + (|y_2| - |y_1|)x_1, \tag{3.13}$$

$I_{11}(n)$  is bounded by

$$n^2 \int_0^\infty \int_{-\infty}^0 \int_{x_1}^{2x_1} \int_{2y_1}^{y_1} f^2(x_1, y_1)x_1|y_1|e^{-2n[(x_2-x_1)|y_1|+(|y_2|-|y_1|)x_1]} dx_1 dy_1 dx_2 dy_2.$$

Integrating with respect to  $x_2$  and then with respect to  $y_2$  in the above integral, we obtain the following bound:

$$I_{11}(n) \leq \frac{1}{4} \int_0^\infty \int_{-\infty}^0 f^2(x_1, y_1) dx_1 dy_1. \tag{3.14}$$

Furthermore, in the region A, it follows from (3.13) that

$$x_1y_1 - x_2y_2 \geq (x_2 - x_1)|y_2|/2 + (|y_2| - |y_1|)x_2/2.$$

By the same argument as above, we then have that

$$I_{12}(n) \leq \int_0^\infty \int_{-\infty}^0 f^2(x_2, y_2) dx_2 dy_2. \tag{3.15}$$

Therefore, from (3.12), (3.14) and (3.15), it follows that

$$I_1^A(n, f) \leq \frac{5}{8} \int_0^\infty \int_{-\infty}^0 f^2(x, y) dx dy. \tag{3.16}$$

We now consider  $I_1^B(n, f)$ . Note that

$$\begin{aligned} 2(x_1y_1 - x_2y_2) &= 2(x_2 - x_1)|y_1| + 2(|y_2| - |y_1|)x_2 \\ &\geq (x_2 - x_1)|y_1| + (|y_2| - |y_1|)x_1 + x_1|y_1| \\ &= (x_2 - x_1)|y_1| + |y_2|x_1 \\ &\geq \frac{1}{2}x_2|y_1| + |y_2|x_1 \end{aligned}$$

for any  $(x_1, y_1, x_2, y_2) \in \Omega_1 \setminus B = \{x_2 > 2x_1 > 0, y_2 \leq y_1 \leq 0\}$ . Define  $\tilde{I}_{11}(n)$  and  $\tilde{I}_{12}(n)$  as

$I_{11}(n)$  and  $I_{12}(n)$ , respectively, with  $\Omega_1 \setminus B$  instead of  $A$ . Then,

$$\begin{aligned} \tilde{I}_{11}(n) &\leq n^2 \int_0^\infty \int_{-\infty}^0 \int_{2x_1}^\infty \int_{-\infty}^{y_1} f^2(x_1, y_1)x_1|y_1|e^{-n(|y_2|x_1+(1/2)x_2|y_1|)} dx_1 dy_1 dx_2 dy_2, \\ \tilde{I}_{12}(n) &\leq n^2 \int_0^\infty \int_{-\infty}^0 \int_{2x_1}^\infty \int_{-\infty}^{y_1} f^2(x_2, y_2)x_2|y_2|e^{-n(|y_2|x_1+(1/2)x_2|y_1|)} dx_1 dy_1 dx_2 dy_2. \end{aligned}$$

Integrating in  $\tilde{I}_{11}(n)$  with respect to  $x_2$  and  $y_2$ , and in  $\tilde{I}_{12}(n)$  with respect to  $x_1$  and  $y_1$ , respectively, we obtain that

$$\tilde{I}_{11}(n) + \tilde{I}_{12}(n) \leq 4 \int_0^\infty \int_{-\infty}^0 f^2(x, y) dx dy.$$

Since  $I_1^B(n, f) \leq \frac{1}{2}(\tilde{I}_{11}(n) + \tilde{I}_{12}(n))$ , we have

$$I_1^B(n, f) \leq 2 \int_0^\infty \int_{-\infty}^0 f^2(x, y) dx dy. \tag{3.17}$$

Similarly, for any  $(x_1, y_1, x_2, y_2) \in \Omega_1 \setminus C = \{x_2 \geq x_1 > 0, y_2 < 2y_1 \leq 0\}$ , we have that

$$\begin{aligned} 2(x_1y_1 - x_2y_2) &= 2(x_2|y_2| - |y_1|x_1) \geq 2(|y_2| - |y_1|)x_2 \\ &\geq (|y_2| - |y_1|)x_2 + |y_1|x_2 \geq (|y_2| - |y_1|)x_1 + |y_1|x_2 \\ &\geq \frac{1}{2}|y_2|x_1 + x_2|y_1|. \end{aligned}$$

Therefore, using a similar approach to the one above, we have

$$I_1^C(n, f) \leq 2 \int_0^\infty \int_{-\infty}^0 f^2(x, y) dx dy. \tag{3.18}$$

Combining (3.11) with (3.16)–(3.18), we get (3.10). □

From (1.8) and Lemma 3.2, we can immediately get the following corollary.

**Corollary 3.1.** *For each  $n > 0$ ,  $I_1(n, \phi_t) \leq \frac{37}{16}t^{2H}$ .*

Observe that for every non-negative measurable function  $f(x, y)$ , by (2.5) and (3.3),

$$I_2(n, f) = \int_0^\infty dx_1 \int_{-\infty}^0 f(x_1, y_1)F_{n,f}(x_1, y_1) dy_1. \tag{3.19}$$

The following lemmas concern  $I_2(n, f)$ . We focus on the case where  $f = \phi_t$ .

**Lemma 3.3.** For any  $0 \leq t \leq 1$ , let  $S = \{x > 0, -t - s < y < -s\}$  and  $\tilde{\phi}_t(x, y) = \phi_t(x, y)\mathbf{1}_S(x, y)$ . There then exist a non-negative function  $\tilde{\Psi}_t(x, y)$  on  $\{x > 0, y \leq 0\}$  such that for all  $n > 0$ ,

$$F_{n, \tilde{\phi}_t}(x_1, y_1)\mathbf{1}_S(x_1, y_1) \leq \tilde{\Psi}_t(x_1, y_1), \tag{3.20}$$

and a positive constant  $K_1$  which depends only on  $H$ , such that for all  $n > 0$ ,

$$I_2(n, \tilde{\phi}_t) \leq \int_S \phi_t(x_1, y_1)\tilde{\Psi}_t(x_1, y_1) dx_1 dy_1 \leq K_1 \sqrt{\frac{s+1}{s}} t^{2H}. \tag{3.21}$$

**Proof.** By (1.5) and (1.6),

$$\phi_t(x, y) = \begin{cases} t/x^{2-H}, & -y - s < x, y < -t - s, \\ (x + t + s + y)/x^{2-H}, & -t - s - y < x \leq -y - s, y < -t - s, \\ -(y + s)/x^{2-H}, & -y - s < x, -t - s \leq y < -s, \\ 1/x^{1-H}, & 0 < x \leq -y - s, -t - s \leq y < -s, \\ 0, & \text{otherwise.} \end{cases} \tag{3.22}$$

Let  $S_1 = S \cap \{0 < x < -y - s\}$  and  $S_2 = S \cap \{x > -y - s\}$ . Then,

$$F_{n, \tilde{\phi}_t}(x_1, y_1)\mathbf{1}_{S_1}(x_1, y_1) = \mathbf{1}_{S_1}(x_1, y_1)\sqrt{x_1}(I_{21} + I_{22} + I_{23}), \tag{3.23}$$

where

$$\begin{aligned} I_{21} &= n^2 \int_{y_1}^{-s-x_1} \int_{x_1}^{-s-y_2} \phi_t(x_2, y_2)\sqrt{x_2|y_1y_2|}e^{-2n[x_1(y_2-y_1)-(x_2-x_1)y_2]} dx_2 dy_2, \\ I_{22} &= n^2 \int_{y_1}^{-s-x_1} \int_{-s-y_2}^{\infty} \phi_t(x_2, y_2)\sqrt{x_2|y_1y_2|}e^{-2n[x_1(y_2-y_1)-(x_2-x_1)y_2]} dx_2 dy_2, \\ I_{23} &= n^2 \int_{-s-x_1}^{-s} \int_{x_1}^{\infty} \phi_t(x_2, y_2)\sqrt{x_2|y_1y_2|}e^{-2n[x_1(y_2-y_1)-(x_2-x_1)y_2]} dx_2 dy_2. \end{aligned}$$

For  $(x_1, y_1) \in S_1$ , from (3.22), we have that

$$\begin{aligned} I_{21} &= n^2 \int_{y_1}^{-s-x_1} \int_{x_1}^{-s-y_2} \frac{1}{x_2^{1/2-H}} \sqrt{|y_1y_2|}e^{-2n[x_1(y_2-y_1)-(x_2-x_1)y_2]} dx_2 dy_2 \\ &\leq \sqrt{\frac{t+s}{s}} n^2 \int_{y_1}^{-s-x_1} \int_{x_1}^{-s-y_2} \frac{1}{x_2^{1/2-H}} |y_2|e^{-2n[x_1(y_2-y_1)-(x_2-x_1)y_2]} dx_2 dy_2 \\ &\leq t^{H-1/2} \sqrt{\frac{t+s}{s}} \frac{n}{2} \int_{y_1}^{-s-x_1} e^{-2nx_1(y_2-y_1)} dy_2 \\ &\leq \frac{t^{H-1/2}}{4x_1} C_s =: \psi_1(x_1, y_1), \end{aligned} \tag{3.24}$$

where  $C_s = \sqrt{(1+s)/s}$  is a constant. Similarly, for  $(x_1, y_1) \in S_1$ ,

$$\begin{aligned}
 I_{22} &= n^2 \int_{y_1}^{-s-x_1} \int_{-s-y_2}^{\infty} \frac{-s-y_2}{x_2^{3/2-H}} \sqrt{|y_1 y_2|} e^{-2n[x_1(y_2-y_1)-(x_2-x_1)y_2]} dx_2 dy_2 \\
 &\leq C_s n^2 \int_{y_1}^{-s-x_1} \int_{-s-y_2}^{\infty} (-s-y_2)^{H-1/2} |y_2| e^{-2n[x_1(y_2-y_1)-(x_2-x_1)y_2]} dx_2 dy_2 \\
 &\leq (-s-y_1)^{H-1/2} C_s \frac{n}{2} \int_{y_1}^{-s-x_1} e^{-2nx_1(y_2-y_1)} dy_2 \\
 &\leq \frac{(-s-y_1)^{H-1/2}}{4x_1} C_s =: \psi_2(x_1, y_1)
 \end{aligned} \tag{3.25}$$

and

$$\begin{aligned}
 I_{23} &= n^2 \int_{-s-x_1}^{-s} \int_{x_1}^{\infty} \frac{-s-y_2}{x_2^{3/2-H}} \sqrt{|y_1||y_2|} e^{-2n[x_1(y_2-y_1)-(x_2-x_1)y_2]} dx_2 dy_2 \\
 &\leq C_s n^2 \int_{-s-x_1}^{-s} \int_{x_1}^{\infty} \frac{x_1}{x_1^{3/2-H}} |y_2| e^{-2n[x_1(y_2-y_1)-(x_2-x_1)y_2]} dx_2 dy_2 \\
 &\leq C_s \frac{1}{4x_1^{3/2-H}} =: \psi_3(x_1, y_1).
 \end{aligned} \tag{3.26}$$

In addition, for  $(x_1, y_1) \in S_2$ ,

$$\begin{aligned}
 F_{n, \tilde{\phi}_t}(x_1, y_1) &= n^2 \int_{y_1}^{-s} \int_{x_1}^{\infty} \frac{-s-y_2}{x_2^{3/2-H}} \sqrt{x_1|y_1 y_2|} e^{-2n[x_1(y_2-y_1)-(x_2-x_1)y_2]} dx_2 dy_2 \\
 &\leq C_s \frac{-s-y_1}{4x_1^{2-H}} =: \psi_4(x_1, y_1).
 \end{aligned} \tag{3.27}$$

Define  $\tilde{\Psi}_t(x_1, y_1)$  as

$$\sqrt{x_1} (\psi_1(x_1, y_1) + \psi_2(x_1, y_1) + \psi_3(x_1, y_1)) \mathbf{1}_{S_1}(x_1, y_1) + \psi_4(x_1, y_1) \mathbf{1}_{S_2}(x_1, y_1).$$

Obviously,  $\tilde{\Psi}_t(x_1, y_1)$  is positive and (3.20) follows from (3.23)–(3.27). Note that

$$\begin{aligned}
 I_2(n, \tilde{\phi}_t) &= \int_S \tilde{\phi}_t(x_1, y_1) F_{n, \tilde{\phi}_t}(x_1, y_1) dx_1 dy_1 \\
 &\leq \int_S \tilde{\phi}_t(x_1, y_1) \tilde{\Psi}_t(x_1, y_1) dx_1 dy_1.
 \end{aligned}$$

With some basic calculations, we obtain the following results:

$$\begin{aligned} \int_{S_1} \phi_t(x, y) \sqrt{x_1} \psi_1(x_1, y_1) dx_1 dy_1 &= \frac{C_s}{4} \int_{-t-s}^{-s} \int_0^{-y_1-s} \frac{t^{H-1/2}}{x_1^{3/2-H}} dx_1 dy_1 = \frac{C_s t^{2H}}{4H^2 - 1}; \\ \int_{S_1} \phi_t(x, y) \sqrt{x_1} \psi_2(x_1, y_1) dx_1 dy_1 &= \frac{C_s}{4} \int_{-t-s}^{-s} \int_0^{-y_1-s} \frac{(-s - y_1)^{H-1/2}}{x_1^{3/2-H}} dx_1 dy_1 \\ &= \frac{C_s t^{2H}}{4(2H - 1)H}; \\ \int_{S_1} \phi_t(x, y) \sqrt{x_1} \psi_3(x_1, y_1) dx_1 dy_1 &= \frac{C_s}{4} \int_{-t-s}^{-s} \int_0^{-y_1-s} \frac{1}{x_1^{2-2H}} dx_1 dy_1 = \frac{C_s t^{2H}}{8(2H - 1)H}; \\ \int_{S_2} \phi_t(x, y) \psi_4(x_1, y_1) dx_1 dy_1 &= \frac{C_s}{4} \int_{-t-s}^{-s} \int_{-y_1-s}^{\infty} \frac{(-y_1 - s)^2}{x_1^{4-2H}} dx_1 dy_1 \\ &= \frac{C_s t^{2H}}{8(3 - 2H)H}. \end{aligned}$$

From the above integrals, (3.21) follows with  $K_1 = \frac{3}{8H(2H-1)} + \frac{1}{8H(3-2H)} + \frac{1}{4H^2-1}$ . □

**Lemma 3.4.** For  $0 \leq t \leq 1$  and each constant  $M < -t - s$ , let  $G(M) = \{-t - s - y < x, M < y < -t - s\}$  and  $\hat{\phi}_t(x, y) = \phi_t(x, y) \mathbf{1}_{G(M)}(x, y)$ . There then exist a non-negative function  $\hat{\Psi}_t(x, y)$  on  $\{x > 0, y \leq 0\}$  such that for all  $n > 0$ ,

$$F_{n, \hat{\phi}_t}(x_1, y_1) \mathbf{1}_{G(M)}(x_1, y_1) \leq \hat{\Psi}_t(x_1, y_1), \tag{3.28}$$

and a positive constant  $K_2$  which only depends on  $H$ , such that

$$I_2(n, \hat{\phi}_t) \leq \int_{G(M)} \phi_t(x_1, y_1) \hat{\Psi}_t(x_1, y_1) dx_1 dy_1 \leq K_2 \sqrt{\frac{|M|}{s}} t^{2H}. \tag{3.29}$$

**Proof.** From (3.22), we have that

$$F_{n, \hat{\phi}_t}(x_1, y_1) = \hat{I}_{21} + \hat{I}_{22}, \tag{3.30}$$

where

$$\begin{aligned} \hat{I}_{21} &= n^2 \int_{y_1}^{-t-s} \int_{-s-y_2}^{\infty} \frac{t}{x_2^{3/2-H}} \sqrt{x_1 |y_1 y_2|} e^{-2n[x_1(y_2-y_1)-(x_2-x_1)y_2]} dx_2 dy_2, \\ \hat{I}_{22} &= n^2 \int_{y_1}^{-t-s} \int_{-x_1}^{-s-y_2} \frac{x_2 + t + s + y_2}{x_2^{3/2-H}} \sqrt{x_1 |y_1 y_2|} e^{-2n[x_1(y_2-y_1)-(x_2-x_1)y_2]} dx_2 dy_2. \end{aligned}$$

For  $(x_1, y_1) \in G(M)$ , since  $|y_1| < |M|$ ,

$$\begin{aligned} \hat{I}_{21} \vee \hat{I}_{22} &\leq \sqrt{\frac{|M|}{s}} \frac{t}{x_1^{1-H}} \int_{y_1}^{-t-s} dy_1 \int_{-s-y_2}^{\infty} n^2 |y_2| e^{-2n[x_1(y_2-y_1)-(x_2-x_1)y_2]} dx_2 dy_2 \\ &\leq \sqrt{\frac{|M|}{s}} \frac{t}{4x_1^{2-H}}. \end{aligned} \tag{3.31}$$

Therefore, from (3.30) and (3.31), it follows that

$$F_{n, \hat{\phi}_t}(x_1, y_1) \mathbf{1}_{G(M)}(x_1, y_1) \leq \sqrt{\frac{|M|}{s}} \frac{t}{4x_1^{2-H}}.$$

Let  $\hat{\Psi}_t(x, y) = \sqrt{\frac{|M|}{s}} t x^{H-2} / 2$ . Then, (3.28) holds. Furthermore, by some basic calculations,

$$\begin{aligned} &\int_{G(M)} \phi_t(x_1, y_1) \hat{\Psi}_t(x_1, y_1) dx_1 dy_1 \\ &\leq \sqrt{\frac{|M|}{s}} \int_M \int_{-t-s-y_1}^{-t-s} \phi_t(x_1, y_1) \frac{t}{2x_1^{2-H}} dx_1 dy_1 \\ &\leq \sqrt{\frac{|M|}{s}} \int_{-\infty}^{-t-s} \left[ \int_{-s-y_1}^{\infty} \frac{t^2}{2x_1^{4-2H}} dx_1 + \int_{-t-s-y_1}^{-s-y_1} \frac{t(x_1+t+s+y_1)}{2x_1^{4-2H}} dx_1 \right] dy_1 \tag{3.32} \\ &= \sqrt{\frac{|M|}{s}} \left[ \frac{t^{2H}}{4(1-H)(3-2H)} + \int_0^t dx_1 \int_{-t-s-x_1}^{-t-s} \frac{t(x_1+t+s+y_1)}{2x_1^{4-2H}} dy_1 \right. \\ &\quad \left. + \int_t^{\infty} dx_1 \int_{-t-s-x_1}^{-s-x_1} \frac{t(x_1+t+s+y_1)}{2x_1^{4-2H}} dy_1 \right] \\ &= \frac{1}{4(1-H)(3-2H)(2H-1)} \sqrt{\frac{|M|}{s}} t^{2H}. \end{aligned}$$

Hence, (3.29) holds for  $K_2 = \frac{1}{4(1-H)(3-2H)(2H-1)}$ . □

**Lemma 3.5.** For  $0 \leq t \leq 1$ ,  $M \leq -2 - s$ , let  $\bar{G}(M) = \{-t - s - y < x, y \leq M\}$  and  $\bar{\phi}_t(x, y) = \phi_t(x, y) \mathbf{1}_{\bar{G}(M)}(x, y)$ . There then exist a non-negative function  $\bar{\Psi}_{s,t}(x, y)$  on  $\{x > 0, y \leq 0\}$  such that for all  $n > 0$ ,

$$F_{n, \bar{\phi}_t}(x_1, y_1) \mathbf{1}_{\bar{G}(M)}(x_1, y_1) \leq \bar{\Psi}_t(x_1, y_1), \tag{3.33}$$

and a constant  $C > 0$  which is independent of  $t, M$  and  $H$ , such that

$$I_2(n, \bar{\phi}_t) \leq \int_{\bar{G}(M)} \phi_t(x_1, y_1) \bar{\Psi}_t(x_1, y_1) dx_1 dy_1 \leq K_2 C t^{2H}, \tag{3.34}$$

where  $K_2$  is the constant in Lemma 3.4.

**Proof.** From (3.22), it follows that

$$\begin{aligned}
 F_{n, \bar{\phi}_t}(x_1, y_1) &= n^2 \int_{y_1}^M \int_{x_1}^\infty \bar{\phi}_t(x_2, y_2) \prod_{i=1}^2 \sqrt{x_i |y_i|} e^{-2n[x_1(y_2-y_1)-(x_2-x_1)y_2]} dx_2 dy_2 \\
 &= \frac{tn^2}{x_1^{3/2-H}} \int_{y_1}^M \sqrt{x_1 |y_1|} \int_{x_1}^\infty e^{-2n[x_1(y_2-y_1)-(x_2-x_1)y_2]} dx_2 dy_2 \\
 &= \frac{t}{2x_1^{2-H}} \int_{|M|}^{|y_1|} \frac{nx_1 |y_1|}{|y_2|} e^{2nx_1 |y_2|} d|y_2| e^{-2nx_1 |y_1|} \\
 &= \frac{t}{2x_1^{2-H}} e^{-2nx_1 |y_1|} nx_1 |y_1| \int_{nx_1 |M|}^{nx_1 |y_1|} \frac{1}{w} e^{2w} dw.
 \end{aligned} \tag{3.35}$$

Let  $Q(z) = e^{-2z} z \int_1^z \frac{1}{w} e^{2w} dw$  for  $z \geq 1$ . Then,  $Q(z)$  is continuous on  $[1, \infty)$  and  $Q(1) = 0$ ,  $\lim_{z \rightarrow \infty} Q(z) = 1/2$ . Hence, there is a constant  $C > 0$ , which is independent of  $M, t$  and  $H$ , such that  $Q(z)$  is bounded by  $C$ . Note that on  $\bar{G}(M)$ , because  $|M| > 1$  and  $x_1 \geq -t - s - M > 1$ , we have  $nx_1 |M| > 1$  for all  $n > 0$ . Therefore, (3.35) yields

$$F_{n, \bar{\phi}_t}(x_1, y_1) \mathbf{1}_{\bar{G}(M)}(x_1, y_1) \leq \frac{t}{2x_1^{2-H}} C.$$

Let  $\bar{\Psi}_t(x, y) = Ctx^{H-2}/2$ . By calculations similar to those in (3.32), (3.34) holds for  $K_2 = \frac{1}{4(1-H)(3-2H)(2H-1)}$ . □

**Proposition 3.2.** For any  $0 \leq t \leq 1$ , there exists a constant  $K$ , independent of  $t$  but dependent on  $s$ , such that

$$\mathbb{E} \left[ \left( n \int_0^n \int_{-n}^0 \phi_t(x, y) \sqrt{x|y|} (-1)^{N_n(x,y)} dx dy \right)^2 \right] \leq Kt^{2H}. \tag{3.36}$$

**Proof.** Using the same notation as in Lemmas 3.1–3.5, taking  $M = -2 - s$  and observing that from (3.22),  $\phi_t(x, y) = 0$  if  $(x, y)$  is not in the set

$$\{(x, y) : y < -s, x > 0 \text{ and } x + y > -t - s\} = S \cup G(M) \cup \bar{G}(M),$$

we can rewrite the left-hand side of (3.36) by

$$\begin{aligned}
 I(n, \phi_t) &:= \mathbb{E} \left[ \left( n \int_0^n \int_{-n}^0 \phi_t(x, y) \sqrt{x|y|} (-1)^{N_n(x,y)} dx dy \right)^2 \right] \\
 &= \mathbb{E} \left[ \left( n \int_0^n \int_{-n}^0 (\tilde{\phi}_t(x, y) + \hat{\phi}_t(x, y) + \bar{\phi}_t(x, y)) \sqrt{x|y|} (-1)^{N_n(x,y)} dx dy \right)^2 \right],
 \end{aligned} \tag{3.37}$$

which is bounded by

$$3I(n, \tilde{\phi}_t) + 3I(n, \hat{\phi}_t) + 3I(n, \bar{\phi}_t). \tag{3.38}$$

Note that  $0 \leq \tilde{\phi}_t, \hat{\phi}_t, \bar{\phi}_t \leq \phi_t$ . Lemma 3.1, Corollary 3.1 and Lemma 3.3 imply that

$$\begin{aligned} I(n, \tilde{\phi}_t) &\leq 2I_1(n, \tilde{\phi}_t) + 2I_2(n, \tilde{\phi}_t) \\ &\leq \left( \frac{37}{8} + 2K_1 \sqrt{\frac{1+s}{s}} \right) t^{2H}, \end{aligned} \tag{3.39}$$

and Lemma 3.1, Corollary 3.1 and Lemma 3.4 imply that

$$I(n, \hat{\phi}_t) \leq 2I_1(n, \hat{\phi}_t) + 2I_2(n, \hat{\phi}_t) \leq \left( \frac{37}{8} + 2K_2 \sqrt{\frac{2+s}{s}} \right) t^{2H}. \tag{3.40}$$

Furthermore, from Lemma 3.1, Corollary 3.1 and Lemma 3.5, we have

$$I(n, \bar{\phi}_t) \leq 2I_1(n, \bar{\phi}_t) + 2I_2(n, \bar{\phi}_t) \leq \left( \frac{37}{8} + 2K_2C \right) t^{2H}. \tag{3.41}$$

Therefore, (3.37)–(3.41) yield that

$$I(n, \phi_t) \leq 3 \left[ \frac{111}{8} + 2K_1 \sqrt{(1+s)/s} + 2K_2 \sqrt{(2+s)/s} + 2K_2C \right] t^{2H}.$$

Taking  $K = 3[\frac{111}{8} + 2K_1 \sqrt{(1+s)/s} + 2K_2 \sqrt{(2+s)/s} + 2K_2C]$  then completes the proof of Proposition 3.2. □

Finally, we prove Proposition 3.1.

**Proof of Proposition 3.1.** To prove the tightness of  $\{Y_n\}_{n \geq 1}$  in  $\mathcal{C}[0, 1]$ , it suffices to show that for some  $r > 0$  there exist two constants  $\tilde{M} > 0$  and  $\eta > 1$  such that for any  $t, t' \in [0, 1]$ ,

$$\begin{aligned} &\mathbb{E} \left[ \left( n \int_0^n \int_{-n}^0 \frac{g_s(t, x, y) - g_s(t', x, y)}{x^{2-H}} \sqrt{|x|y|} (-1)^{N_n(x,y)} dx dy \right)^r \right] \\ &\leq \tilde{M}(t - t')^\eta, \end{aligned}$$

which follows from the criterion given by Billingsley (see [5], Theorem 12.3) and the fact that our processes are null at the origin.

Without loss of generality, let  $t > t'$ . Note that from (1.6), we have

$$\begin{aligned} g_s(t, x, y) - g_s(t', x, y) &= h(t + s, x, y) - h(t' + s, x, y) \\ &= g_{t'+s}(t - t', x, y). \end{aligned}$$



By Proposition 3.2, it is easy to check that the above inequality holds for  $r = 2$ ,  $\eta = 2H > 1$  and  $\tilde{M} = 3[\frac{111}{8} + 2K_1\sqrt{(1+s)/s} + 2K_2\sqrt{(2+s)/s} + 2K_2C]$ , which completes the proof of Proposition 3.1.  $\square$

**Remark 3.1.** Compared with the proofs of tightness in Bardina et al. [2,4], we have found that under the condition that the kernel  $f$  can be separated by its arguments  $(x, y)$ , the calculation of  $I(n, f)$  is transformed to the calculation of  $I_1(n, f)$ , which is relatively simple; see the proofs of Lemmas 3.1 and 3.2 in [2] and the proof of Lemma 3.1 in [4]. However, in our case, the kernel  $f$  cannot be separated by the arguments  $(x, y)$ , so we need to discuss  $I_2(n, f)$ . From our proof, we can see that the calculation of  $I_2(n, f)$  is more complicated and delicate than that of  $I_1(n, f)$ . In addition, the fact that the kernel  $f$  cannot be separated by the arguments  $(x, y)$  also creates some difficulties in the identification of the limit law; see the proof of (4.7) in the next section.

**Remark 3.2.** By Proposition 10.3 in [10], page 149, Proposition 3.1 also shows that  $\mathbb{P}(Y_n \in \mathcal{C}[0, 1], n \in \mathbb{N}) = 1$ .

### 4. Limit law of $Y_n$

In this section, we proceed with the identification of the limit law. We will prove the following proposition.

**Proposition 4.1.** *The finite-dimensional distributions of  $Y_n = \{Y_n(t), t \in [0, 1]\}$  defined by (2.3) converge weakly, as  $n$  tends to  $\infty$ , to those of a fractional Brownian motion  $B^H = \{B^H(t), t \in [0, 1]\}$  with Hurst index  $H \in (1/2, 1)$ .*

**Proof.** For each  $k \in \mathbb{N}$ ,  $a_1, \dots, a_k \in \mathbb{R}$  and  $0 \leq t_1 < t_2 < \dots < t_k \leq 1$ , we define

$$L_n = \sum_{j=1}^k a_j Y_n(t_j) \quad \text{and} \quad U = \sum_{j=1}^k a_j B^H(t_j).$$

It suffices to prove that for any  $\xi \in \mathbb{R}$ , as  $n \rightarrow \infty$ ,

$$J(n) := |\mathbb{E}[\exp(i\xi L_n)] - \mathbb{E}[\exp(i\xi U)]| \rightarrow 0. \tag{4.1}$$

For  $T > 1 + s$ , let

$$L_n(T) := n \sum_{j=1}^k a_j \int_0^T \int_{-T}^0 \phi_{t_j}(x, y) \sqrt{|x| |y|} (-1)^{N_n(x, y)} \, dx \, dy$$

and

$$U(T) := \sum_{j=1}^k a_j \int_0^T \int_{-T}^0 \phi_{t_j}(x, y) B(dx, dy),$$

where  $B(x, y)$  is given by Lemma 2.1. Let

$$\begin{aligned} J_1(n, T) &= |\mathbb{E}[\exp(i\xi L_n(T))] - \mathbb{E}[\exp(i\xi U(T))]|, \\ J_2(n, T) &= |\mathbb{E}[\exp(i\xi L_n)] - \mathbb{E}[\exp(i\xi L_n(T))]|, \\ J_3(T) &= |\mathbb{E}[\exp(i\xi U(T))] - \mathbb{E}[\exp(i\xi U)]|. \end{aligned}$$

Then,

$$J(n) \leq J_1(n, T) + J_2(n, T) + J_3(T). \tag{4.2}$$

Below, we estimate  $J_1(n, T)$ ,  $J_2(n, T)$  and  $J_3(T)$ , respectively.

(1) We estimate  $J_1(n, T)$ .

Noting that  $\phi_{t_j}(x, y)$  is a non-negative measurable function on  $\{x > 0, y \leq 0\}$ , we can find a sequence of elementary functions  $q^{m,j}(x, y)$  such that

$$0 \leq q^{m,j}(x, y) \leq \phi_{t_j}(x, y) \quad \text{and} \quad q^{m,j}(x, y) \rightarrow \phi_{t_j}(x, y) \quad \text{a.e. as } m \rightarrow \infty. \tag{4.3}$$

Then, by the dominated convergence theorem, it follows from the fact that  $\int_0^\infty \int_{-\infty}^0 [\phi_{t_j}(x, y)]^2 dx dy < \infty$  that as  $m \rightarrow \infty$ ,

$$\int_0^T \int_{-T}^0 (\phi_{t_j}(x, y) - q^{m,j}(x, y))^2 dx dy \rightarrow 0. \tag{4.4}$$

For any  $m, j, n > 0$ , define

$$\begin{aligned} Y_n^{m,j} &= n \int_0^T \int_{-T}^0 q^{m,j}(x, y) \sqrt{x|y|} (-1)^{N_n(x,y)} dx dy, \\ B^{m,j} &= \int_0^T \int_{-T}^0 q^{m,j}(x, y) B(dx, dy). \end{aligned}$$

By Lemma 2.1, we can readily verify that for fixed  $m \in \mathbb{N}$ , as  $n \rightarrow \infty$ ,

$$J_{11}(n, T, m) := \left| \mathbb{E} \left[ \exp \left( i\xi \sum_{j=1}^k a_j Y_n^{m,j} \right) \right] - \mathbb{E} \left[ \exp \left( i\xi \sum_{j=1}^k a_j B^{m,j} \right) \right] \right| \rightarrow 0 \tag{4.5}$$

because  $Y_n^{m,j}$  is essentially a linear combination of increments of  $B_n$  defined by (2.2), and  $B^{m,j}$  is the same linear combination of the corresponding limits of increments of  $B_n$ .

We further define

$$\begin{aligned} J_{12}(n, T, m) &:= \left| \mathbb{E}[\exp(i\xi L_n(T))] - \mathbb{E} \left[ \exp \left( i\xi \sum_{j=1}^k a_j Y_n^{m,j} \right) \right] \right|, \\ J_{13}(T, m) &:= \left| \mathbb{E}[\exp(i\xi U(T))] - \mathbb{E} \left[ \exp \left( i\xi \sum_{j=1}^k a_j B^{m,j} \right) \right] \right|. \end{aligned}$$

Then, for any  $n, m$ ,

$$J_1(n, T) \leq J_{11}(n, T, m) + J_{12}(n, T, m) + J_{13}(n, T, m). \quad (4.6)$$

Below, we will show that for all fixed  $T$ , there exists some  $\gamma_T(m) > 0$  such that for any  $n > 0$ ,

$$J_{12}(n, T, m) \leq \gamma_T(m) \rightarrow 0 \quad (4.7)$$

as  $m \rightarrow \infty$ . To this end, let  $f_{m,j}(x, y) = \phi_{t_j}(x, y) - q^{m,j}(x, y)$ . Define

$$\begin{aligned} \tilde{f}_{m,j}(x, y) &:= f_{m,j}(x, y) \mathbf{1}_{[0, T] \times [-t_j - s, 0]}(x, y), \\ \hat{f}_{m,j}(x, y) &:= f_{m,j}(x, y) \mathbf{1}_{[0, T] \times [-T, -t_j - s]}(x, y). \end{aligned}$$

By (4.3), as  $m \rightarrow \infty$ ,

$$\phi_{t_j}(x, y) \geq f_{m,j}(x, y) \rightarrow 0 \quad \text{a.e. in } [0, T] \times [-T, 0]. \quad (4.8)$$

Define

$$R_j(n, m, T) = \mathbb{E} \left[ \left| n \int_0^T \int_{-T}^0 f_{m,j}(x, y) \sqrt{|x| |y|} (-1)^{N_n(x, y)} dx dy \right|^2 \right].$$

Then, by Lemma 3.1,

$$\begin{aligned} [R_j(n, m, T)]^2 &\leq n^2 E \left[ \left( \int_0^T \int_{-T}^0 (\hat{f}_{m,j}(x, y) + \tilde{f}_{m,j}(x, y)) \sqrt{|x| |y|} (-1)^{N_n(x, y)} dx dy \right)^2 \right] \\ &\leq 2(I(n, \hat{f}_{m,j}) + I(n, \tilde{f}_{m,j})) \\ &\leq 4(I_1(n, \hat{f}_{m,j}) + I_2(n, \hat{f}_{m,j}) + I_1(n, \tilde{f}_{m,j}) + I_2(n, \tilde{f}_{m,j})). \end{aligned} \quad (4.9)$$

Lemma 3.2 and (4.4) show that for any  $n > 0$ , as  $m \rightarrow \infty$ ,

$$I_1(n, \tilde{f}_{m,j}) \leq \frac{37}{8} \int_0^\infty \int_{-\infty}^0 \tilde{f}_{m,j}^2(x, y) dx dy =: \tilde{\alpha}_j(m, T) \rightarrow 0. \quad (4.10)$$

Note that  $f_{m,j}(x_1, y_1) \tilde{\Psi}_{t_j}(x_1, y_1) \rightarrow 0$  a.e. as  $m \rightarrow \infty$ . From Lemma 3.3 we know that

$$F_{n, \tilde{f}_{m,j}}(x_1, y_1) \leq F_{n, \phi_{t_j}}(x_1, y_1) \leq \tilde{\Psi}_{t_j}(x_1, y_1)$$

and that

$$\begin{aligned} &\int_0^T \int_{-t_j - s}^0 f_{m,j}(x_1, y_1) F_{n, \tilde{f}_{m,j}}(x_1, y_1) dx_1 dy_1 \\ &\leq \int_0^T \int_{-t_j - s}^0 \phi_{t_j}(x_1, y_1) \tilde{\Psi}_{t_j}(x_1, y_1) dx_1 dy_1 < \infty. \end{aligned}$$

By the dominated convergence theorem, as  $m \rightarrow \infty$ ,

$$I_2(n, \tilde{f}_{m,j}) \leq \tilde{\beta}_j(m, T) := \int_0^T \int_{-t_j-s}^0 f_{m,j}(x_1, y_1) \tilde{\Psi}_{t_j}(x_1, y_1) dx_1 dy_1 \rightarrow 0. \tag{4.11}$$

In a similar way, we know there are  $\hat{\alpha}_j(m, T), \hat{\beta}_j(m, T)$  such that for all  $n > 0$ , as  $m \rightarrow \infty$ ,

$$I_1(n, \hat{f}_{m,j}) \leq \hat{\alpha}_j(m, T) \rightarrow 0 \quad \text{and} \quad I_2(n, \hat{f}_{m,j}) \leq \hat{\beta}_j(m, T) \rightarrow 0. \tag{4.12}$$

On the other hand, using the mean value theorem, we obtain that

$$\begin{aligned} J_{12}(n, T, m) &\leq k|\xi| \max_{1 \leq j \leq k} \mathbb{E} \left[ \left| a_j n \int_0^T \int_{-T}^0 f_{m,j}(x, y) \sqrt{x|y|} (-1)^{N_n(x,y)} dx dy \right| \right] \\ &= k|\xi| \max_{1 \leq j \leq k} (|a_j| R_j(n, m, T)). \end{aligned} \tag{4.13}$$

Hence, (4.7) follows from (4.9)–(4.13) with

$$\gamma_T(m) = 2k|\xi| \max_{1 \leq j \leq k} (|a_j| \sqrt{\tilde{\alpha}_j(m, T) + \tilde{\beta}_j(m, T) + \hat{\alpha}_j(m, T) + \hat{\beta}_j(m, T)}).$$

For  $J_{13}(m, T)$ , we apply the mean value theorem again. Then, as  $m \rightarrow \infty$ ,

$$\begin{aligned} J_{13}(m, T) &\leq \xi \mathbb{E} \left[ \left| a_j \sum_{j=1}^k \int_0^T \int_{-T}^0 (\phi_{t_j}(x, y) - q^{m,j}(x, j)) B(dx, dy) \right| \right] \\ &\leq k\xi \max_{1 \leq j \leq k} \mathbb{E} \left[ \left| a_j \int_0^T \int_{-T}^0 (\phi_{t_j}(x, y) - q^{m,j}(x, j)) B(dx, dy) \right| \right] \\ &\leq k\xi \max_{1 \leq j \leq k} \left\{ \left[ \int_0^T \int_{-T}^0 (\phi_{t_j}(x, y) - q^{m,j}(x, y))^2 dx dy \right]^{1/2} \right\} \rightarrow 0. \end{aligned} \tag{4.14}$$

From (4.5)–(4.7) and (4.14), we obtain that for any fixed  $T$ , as  $n \rightarrow \infty$ ,

$$J_1(n, T) \rightarrow 0. \tag{4.15}$$

(2) In a similar way as was used to prove (4.7), there exists some  $\zeta(T) > 0$  such that for  $n > T$ , as  $T \rightarrow \infty$ ,

$$J_2(n, T) \leq \zeta(T) \rightarrow 0. \tag{4.16}$$

(3) In a similar way as was used to prove (4.14), we obtain that there exists some  $\theta(T) > 0$  such that as  $T \rightarrow \infty$ ,

$$J_3(T) \leq \theta(T) \rightarrow 0. \tag{4.17}$$

Therefore, combining (4.2) and (4.15)–(4.17), we can obtain that  $J(n) \rightarrow 0$  as  $n \rightarrow \infty$ , completing the proof of Proposition 4.1.  $\square$

Note that Theorem 2.1 is an immediate result of Propositions 3.1 and 4.1. Therefore, the proof of Theorem 2.1 is complete.

## Acknowledgments

This work was done during our visit to the Department of Statistics and Probability at Michigan State University. The authors thank Professor Yimin Xiao for stimulating discussions and the department for its good working conditions. This research was supported in part by the National Natural Science Foundation of China (No. 10901054). Thanks is also due to the anonymous referees for their careful reading and detailed comments which improved the quality of the paper.

## References

- [1] Bardina, X. and Bascompte, D. (2010). Weak convergence towards two independent Gaussian processes form a unique Poisson process. *Collect. Math.* **61** 191–204. [MR2666230](#)
- [2] Bardina, X. and Jolis, M. (2000). Weak approximation of the Brownian sheet from a Poisson process in the plane. *Bernoulli* **6** 653–665. [MR1777689](#)
- [3] Bardina, X., Jolis, M. and Rovira, C. (2000). Weak approximation of the Wiener process from a Poisson process: The multidimensional parameter set case. *Statist. Probab. Lett.* **50** 245–255. [MR1792303](#)
- [4] Bardina, X., Jolis, M. and Tudor, C.A. (2003). Weak convergence to the fractional Brownian sheet and other two-parameter Gaussian processes. *Statist. Probab. Lett.* **65** 317–329. [MR2039877](#)
- [5] Billingsley, P. (1968). *Convergence of Probability Measures*. New York: Wiley. [MR0233396](#)
- [6] Cairoli, R. and Walsh, J.B. (1975). Stochastic integrals in the plane. *Acta Math.* **134** 111–183. [MR0420845](#)
- [7] Davydov, Y. (1970). The invariance principle for stationary processes. *Teor. Veroyatn. Primen.* **15** 498–509. [MR0283872](#)
- [8] Delgado, R. and Jolis, M. (2000). Weak approximation for a class of Gaussian process. *J. Appl. Probab.* **37** 400–407. [MR1780999](#)
- [9] Enriquez, N. (2004). A simple construction of the fractional Brownian motion. *Stochastic Process. Appl.* **109** 203–223. [MR2031768](#)
- [10] Ethier, S. and Kurtz, T. (1986). *Markov Processes: Characterization and Convergence*. New York: Wiley. [MR0838085](#)
- [11] Gaigalas, R. (2006). A Poisson bridge between fractional Brownian motion and stable Levy motion. *Stochastic Process. Appl.* **116** 447–462. [MR2199558](#)
- [12] Kaj, I. and Taqqu, M.S. (2008). Convergence to fractional Brownian motion and to the Telecom process: The integral representation approach. In *In and Out of Equilibrium 2*. *Progr. Probab.* **60** 383–427. Basel: Birkhäuser. [MR2477392](#)
- [13] Meyer, Y., Sellan, F. and Taqqu, M.S. (1999). Wavelets, generalized white noise and fractional integration: The synthesis of fractional Brownian motion. *J. Fourier Anal. Appl.* **5** 465–494. [MR1755100](#)
- [14] Samorodnitsky, G. and Taqqu, M.S. (1994). *Stable Non-Gaussian Random Processes*. New York: Chapman and Hall. [MR1280932](#)

- [15] Stroock, D. (1982). *Topics in Stochastic Differential Equations*. Bombay: Tata Institute of Fundamental Research. [MR0685758](#)
- [16] Taqqu, M.S. (1975). Weak convergence to fractional Brownian motion and to the Rosenblatt process. *Z. Wahrsch. Verw. Gebiete* **31** 287–302. [MR0400329](#)

*Received January 2010 and revised August 2010*