UDC 519.21

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# EXISTENCE OF GENERALIZED LOCAL TIMES FOR GAUSSIAN RANDOM FIELDS

We consider a Gaussian centered random field that has values in the Euclidean space. We investigate the existence of local time for the random field as a generalized functional, an element of the Sobolev space constructed for our random field. We give the sufficient condition for such an existence in terms of the field covariation and apply it in a few examples: the Brownian motion with additional weight and the intersection local time of two Brownian motions.

## INTRODUCTION

This article was motivated by the following problem. Let  $W_1, W_2$  be two Brownian motions in the *d*-dimensional Euclidean space. Suppose that they are jointly Gaussian, and their covariation is as follows:  $EW_1(s)W_2(t)^T = \min(s,t)Q$  (Q is a  $d \times d$  matrix). Consider the intersection local time on the time interval [0,1] of these two processes as a limit of the approximations  $L_{\varepsilon} = \int_0^1 \int_0^1 f_{\varepsilon}(W_1(s) - W_2(t)) ds dt$ , where  $f_{\varepsilon}$  is an approximation of the measure with unit mass at zero:  $f_{\varepsilon}(x) = \frac{1}{\varepsilon^d} f(\frac{x}{\varepsilon}), f \in C_b^{\infty}(\mathbb{R}^d), f \ge 0, \int_{\mathbb{R}^d} f(x) dx = 1$  ( $C_b^{\infty}(\mathbb{R}^d)$  is the space of bounded infinitely differentiable functions on  $\mathbb{R}^d$ ). Find the conditions on d and Q such that these approximations converge in the Sobolev spaces  $D_{2,\alpha}$  constructed over an abstract Wiener space associated with the processes (for the exact definition, see below).

To answer this question, we derive the sufficient condition in a more general statement. The main problem here is how to deal with approximations in order to find the index of the Sobolev space, where we have the convergence. In work [1], this was done for the self-intersection local time of the Brownian motion by finding the explicit form for the Itô—Wiener expansion kernels of approximations. But it is enough, in fact, to find the scalar product of those kernels in the space of square integrable functions. Moreover, it is enough to find the asymptotic behavior of this product, as the kernel index tends to infinity. The latter can be done using a certain integral representation of kernel scalar products. We are able to generalize this approach and implement it for an arbitrary Gaussian random field. In our generalization, we use the fact that the intersection local time of two Brownian motions may be thought as the local time of a Gaussian random field on the two-dimensional parameter space.

Let  $(T, \mathfrak{B})$  be a measurable space with finite measure  $\nu$  on it. Consider a centered Gaussian random field  $\xi$  on  $(T, \mathfrak{B})$ , where each  $\xi(t)$  is a centered Gaussian random vector in  $\mathbb{R}^d$ . Denote  $K_{ij}(s,t) = E\xi_i(s)\xi_j(t)$ . We suppose that  $K_{ij}$  are measurable functions, and det K(t,t) > 0  $\nu$ -a.s. Additionally, we suppose that  $\xi$  is separable. In other words, it is determined by its values in a countable set of points. Then there exists an abstract Wiener space  $(B, H, \mu)$  (see [6]) such that  $\xi$  can be constructed on the probability space  $(B, \mathfrak{F}, \mu)$  ( $\mathfrak{F}$  is the  $\sigma$ -algebra of Borel sets in B) as a linear functional on

<sup>2000</sup> AMS Mathematics Subject Classification. Primary 60H07.

Key words and phrases. Local time, generalized functional, Gaussian random field.

it:  $\xi^k(t)(\omega) = l_t^k(\omega), \omega \in B, l_t^k \in H, k = 1, \ldots, d, t \in T$ . Here, *B* is a separable Banach space, and  $\mu$  is the centered Gaussian probability measure on it. Using the structure of an abstract Wiener space, we can introduce Sobolev spaces as in [5]. The space of square integrable functions can be represented as a direct sum of eigenspaces of the Ornstein— Uhlenbeck operator L:  $L_2(B,\mu) = \bigoplus_{n=0}^{\infty} M_n$ , where  $Lx = -nx, x \in M_n$ . Consequently, we can represent each square integrable function  $f \in L_2(B,\mu)$  as a sum of eigenvectors  $f = \sum_{n=0}^{\infty} f_n, f_n \in M_n$ . Such a decomposition is called the Itô—Wiener decomposition. Now we have  $L_2(\Omega) = L_2(B,\mu)$ , so every square integrable  $\xi$ -measurable random variable has this decomposition. Consider  $P(B,\mu) \subset L_2(B,\mu)$ , being a subspace of the space of square integrable functions which consists of the functions that have only finite non-zero members in the associated Itô—Wiener decomposition. Introduce the norm indexed by  $\alpha$  on the subspace  $P(B,\mu)$ :

$$\|f\|_{2,\alpha}^{2} = \left\|(I-L)^{\frac{\alpha}{2}}f\right\|_{2}^{2} = \sum_{n=0}^{N} (1+n)^{\alpha} \|f_{n}\|_{2},$$

where N is the maximum of indices corresponding to non-zero decomposition members, and  $\|\cdot\|_2$  is norm in  $L_2(B,\mu)$ . We denote the completion of  $P(B,\mu)$  equipped with this norm as  $D_{2,\alpha}$  and call it the Sobolev space with index 2,  $\alpha$ .

We want to have the convergence of approximations for the local time in these spaces. We consider the local time at x = 0 of  $\xi$  with regard to some finite measure  $\nu$  on  $(T, \mathfrak{B})$ and approximations having form  $L_{\varepsilon} = \int_{T} f_{\varepsilon}(\xi(t))\nu(dt)$ , where  $f_{\varepsilon}$  is the same as above.

We have to note that this approach to the local time generalization is not unique. There exists an approach which makes use of the white noise theory (see, e.g., [2]). Another approach introduces generalized homogeneous functionals for the Brownian motion [3].

In the first part, we give the sufficient condition for the existence of a square integrable local time (as a limit of approximations). In the second one, we get the integral representation for the scalar products of Itô—Wiener expansion members. In the third part, we study the asymptotic behavior of this representation and use it to prove the convergence of approximations in the appropriate Sobolev space. In examples, we show how our conditions for the local time existence can be used in some cases including the problems on the intersection and the self-intersection local time for the Brownian motion.

### EXISTENCE OF SQUARE INTEGRABLE LOCAL TIME

At first, we give the sufficient condition for the existence in  $L_2(\Omega)$ . The same condition for the processes mentioned in work [4]. By B(s,t), we denote a symmetric matrix  $2d \times 2d$ such that

$$\begin{split} B_{ij}(s,t) &= K_{ij}(s,s), i = 1..d, j = 1..d\\ B_{ij}(s,t) &= K_{i-d,j}(s,t), i = d + 1..2d, j = 1..d\\ B_{ij}(s,t) &= K_{i-d,j-d}(t,t), i = d + 1..2d, j = d + 1..2d. \end{split}$$

In other words, B(s,t) is the covariance matrix of a vector  $(\xi(s),\xi(t))$ .

**Theorem 1.** Suppose that

$$\int_T \int_T \frac{1}{\sqrt{\det B(s,t)}} \nu(ds) \nu(dt) < +\infty,$$

then the limit of  $L_{\varepsilon}$  exists in  $L_2(\Omega)$ .

*Proof.* It is sufficient to prove that  $EL_{\varepsilon_1}L_{\varepsilon_2} \to C < +\infty, \varepsilon_1, \varepsilon_2 \to 0+$ . Note that the condition of Theorem 1 gives: det B(s,t) > 0 a.e. with regard to the measure  $\nu \times \nu$ .

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Using the Fubini theorem and performing the change of variables  $\varepsilon_1 u = x, \varepsilon_2 w = y$ , we get

$$\begin{split} EL_{\varepsilon_1}L_{\varepsilon_2} \\ &= \int_T \int_T \int_{\mathbb{R}^d} \int_{\mathbb{R}^d} f_{\varepsilon_1}(x) f_{\varepsilon_2}(y) (2\pi)^{-d} (\det B(s,t))^{-\frac{1}{2}} \\ &\quad \times \exp\left(-\frac{1}{2} \begin{pmatrix} x \\ y \end{pmatrix}^T B^{-1}(s,t) \begin{pmatrix} x \\ y \end{pmatrix}\right) dx dy \nu(ds) \nu(dt) \\ &= \int_T \int_T \int_{\mathbb{R}^d} \int_{\mathbb{R}^d} f(u) f(w) (2\pi)^{-d} (\det B(s,t))^{-\frac{1}{2}} \\ &\quad \times \exp(-\frac{1}{2} \begin{pmatrix} \varepsilon_1 u \\ \varepsilon_2 w \end{pmatrix}^T B^{-1}(s,t) \begin{pmatrix} \varepsilon_1 u \\ \varepsilon_2 w \end{pmatrix}) du dw \nu(ds) \nu(dt). \end{split}$$

The function inside all integrals is monotonously converging to

$$f(u)f(w)(2\pi)^{-d}(\det B(s,t))^{-\frac{1}{2}}$$

when  $\varepsilon_1, \varepsilon_2 \to 0+$ . So, by applying the theorem on monotonous convergence, we conclude that  $EL_{\varepsilon_1}L_{\varepsilon_2} \to \int_T \int_T (2\pi)^{-d} (\det B(s,t))^{-\frac{1}{2}} \nu(ds) \nu(dt)$ , and Theorem 1 is proved.

# COVARIATION FORMULA

As we stated in Introduction, we can write an Itô—Wiener expansion for the following expression, because it is a bounded and therefore square integrable random variable:

$$\int_T f(\xi(t))\nu(dt) = \sum_{n=0}^\infty a_n(f), f \in C_b^\infty(\mathbb{R}^d), a_n(f) \in M_n$$

Note that this is the expression appeared in our local time approximations. We want to obtain a representation for  $Ea_n(f)a_n(g), f, g \in C_b^{\infty}(\mathbb{R}^d)$ .

**Theorem 2.** For any functions  $f, g \in C_b^{\infty}(\mathbb{R}^d)$ ,

$$Ea_{n}(f)a_{n}(g) = \int_{T} \int_{\mathbb{R}^{d}} \int_{\mathbb{R}^{d}} \int_{\mathbb{R}^{d}} \int_{\mathbb{R}^{d}} \int_{\mathbb{R}^{d}} f(x)g(y)e^{i(x,\gamma_{1})+i(y,\gamma_{2})}q_{n}(s,t,\gamma_{1},\gamma_{2})dxdyd\gamma_{1}d\gamma_{2}\nu(ds)\nu(dt),$$

where

$$q_n(s,t,x,y) = (2\pi)^{-2d} \frac{(-1)^n}{n!} (K(s,t)x,y)^n e^{-\frac{1}{2}(K(s,s)x,x)} e^{-\frac{1}{2}(K(t,t)y,y)}$$

Proof. Consider the Ornstein—Uhlenbeck semigroup  $\{T_{\lambda}, \lambda \geq 0\}$  on our abstract Wiener space. By definition (see [5]), it acts on any function  $h \in L_2(B, \mu)$  in the following way:  $T_{\lambda}h(x) = \int_B h(e^{-\lambda}x + \sqrt{1 - e^{-2\lambda}}y)\mu(dy)$ . It also has property that its action on the same function with the Itô—Wiener decomposion  $h = \sum_{n=0}^{\infty} h_n, h_n \in M_n$  may be written as  $T_{\lambda}h = \sum_{n=0}^{\infty} e^{-n\lambda}h_n$ . Using this property, we get

$$E(\int_T f(\xi(t))\nu(dt) \cdot T_\lambda \int_T g(\xi(t))\nu(dt)) = \sum_{n=0}^\infty e^{-n\lambda} Ea_n(f)a_n(g).$$

Here, we used the fact that  $Ea_n(f)a_m(g) = 0, m \neq n$ . On the other hand, by the definition of the Ornstein—Uhlenbeck semigroup and because  $\xi(t)(e^{-\lambda}x + \sqrt{1 - e^{-2\lambda}}y) = e^{-\lambda}\xi(t)(x) + \sqrt{1 - e^{-2\lambda}}\xi(t)(y), x, y \in B$ , we have

$$\begin{split} E(\int_T f(\xi(t))\nu(dt) \cdot T_\lambda \int_T g(\xi(t))\nu(dt)) &= \\ &= E(\int_T f(\xi(t))\nu(dt) \cdot E(\int_T g(e^{-\lambda}\xi(t) + \sqrt{1 - e^{-2\lambda}}\tilde{\xi}(t))\nu(dt)/\xi)) = \\ &= \int_T \int_T E(f(\xi(s))g(e^{-\lambda}\xi(t) + \sqrt{1 - e^{-2\lambda}}\tilde{\xi}(t)))\nu(ds)\nu(dt), \end{split}$$

where  $\tilde{\xi}$  is an independent copy of  $\xi$ . We can swap integrals and expectations, because the expression inside all integrals is bounded. Denote  $B_{\beta}(s,t) = \begin{pmatrix} K(s,s) & \beta K(t,s) \\ \beta K(s,t) & K(t,t) \end{pmatrix}$ . The distribution of  $(\xi(s), e^{-\lambda}\xi(t) + \sqrt{1 - e^{-2\lambda}}\tilde{\xi}(t))$  is centered Gaussian with the covariation matrix  $B_{e^{-\lambda}}(s,t)$ . By Lemma 1 and our assumptions on the covariation: if  $\lambda > 0$ , then det  $B_{e^{-\lambda}}(s,t) > 0$  for all  $(s,t) \in T^2$  except a set of the zero measure  $\nu \times \nu$ . This yields

$$\begin{split} E(\int_{T} f(\xi(t))\nu(dt) \cdot T_{\lambda} \int_{T} g(\xi(t))\nu(dt)) &= \\ &= \int_{T} \int_{T} \int_{\mathbb{R}^{d}} \int_{\mathbb{R}^{d}} f(x)g(y) \frac{1}{(2\pi)^{d}\sqrt{\det B_{e^{-\lambda}}(s,t)}} \cdot \\ &\cdot \exp(-\frac{1}{2} \begin{pmatrix} x \\ y \end{pmatrix}^{T} B_{e^{-\lambda}}^{-1}(s,t) \begin{pmatrix} x \\ y \end{pmatrix}) dx dy \nu(ds)\nu(dt) = \\ &= \int_{T} \int_{T} \int_{\mathbb{R}^{d}} \int_{\mathbb{R}^{d}} \int_{\mathbb{R}^{d}} \int_{\mathbb{R}^{d}} \int_{\mathbb{R}^{d}} f(x)g(y)e^{i(x,\gamma_{1})+i(y,\gamma_{2})}q^{\lambda}(s,t,\gamma_{1},\gamma_{2})d\gamma_{1}d\gamma_{2}dx dy\nu(ds)\nu(dt), \end{split}$$

where

$$\begin{aligned} q^{\lambda}(s,t,x,y) &= (2\pi)^{-2d} \exp\left(-\frac{1}{2} \begin{pmatrix} x \\ y \end{pmatrix}^T B_{e^{-\lambda}}(s,t) \begin{pmatrix} x \\ y \end{pmatrix}\right) = \\ &= (2\pi)^{-2d} \exp\left(-\frac{1}{2}((K(s,s)x,x) + (K(t,t)y,y) + 2e^{-\lambda}(K(s,t)x,y))) \right) \end{aligned}$$

Obviously:

$$q^{\lambda}(s,t,x,y) = \sum_{n=0}^{\infty} e^{-n\lambda} q_n(s,t,x,y), s, t \in T, x, y \in \mathbb{R}^d.$$

We substitute q in the integral with this sum. By the Lebesgue dominated convergence theorem, we can exchange this sum with the integral by  $\lambda_1, \lambda_2$ . If we can show that it can be exchanged with other integrals for all  $\lambda > 0$ , we get two different representations for the starting expression in the form of a power series of  $e^{-\lambda}$ . Then we match members of these series and get the desired integral representation. Fix  $(s, t) \in T^2$  such that det K(s, s) > 0, det K(t, t) > 0, and  $\lambda > 0$ . Denote:

$$\begin{split} p(s,t,x,y) &= \frac{1}{(2\pi)^d \sqrt{\det B_0(s,t)}} \exp(-\frac{1}{2} \begin{pmatrix} x \\ y \end{pmatrix}^T B_0^{-1}(s,t) \begin{pmatrix} x \\ y \end{pmatrix} \\ &= \frac{1}{(2\pi)^d \sqrt{\det K(s,s) \det K(t,t)}} \exp(-\frac{1}{2}((K^{-1}(s,s)x,x) + (K^{-1}(t,t)y,y)))), \\ r_n(s,t,x,y) &= \frac{1}{p(s,t,x,y)} \int_{\mathbb{R}^d} \int_{\mathbb{R}^d} e^{i(x,\gamma_1) + i(y,\gamma_2)} q_n(s,t,\gamma_1,\gamma_2) d\gamma_1 d\gamma_2, \end{split}$$

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$$\begin{split} R_n(s,t,x,y) &= \sum_{k=0}^n e^{-k\lambda} r_k(s,t,\cdot,\cdot), \\ R(s,t,x,y) &= \sum_{k=0}^\infty e^{-k\lambda} r_k(s,t,\cdot,\cdot) \\ &= \frac{1}{p(s,t,x,y)} \int_{\mathbb{R}^d} \int_{\mathbb{R}^d} e^{i(x,\gamma_1)+i(y,\gamma_2)} q^\lambda(s,t,\gamma_1,\gamma_2) d\gamma_1 d\gamma_2 \\ &= \frac{\sqrt{\det B_0(s,t)}}{\sqrt{\det B_{e^{-\lambda}}(s,t)}} \exp\left(-\frac{1}{2} \begin{pmatrix} x \\ y \end{pmatrix}^T (B_{e^{-\lambda}}^{-1} - B_0^{-1})(s,t) \begin{pmatrix} x \\ y \end{pmatrix}\right) \end{split}$$

We want to show that the integrals can be exchanged with the sum. Now it can be formulated as

$$\lim_{n \to \infty} \int_T \int_T \int_{\mathbb{R}^d} \int_{\mathbb{R}^d} f(x)g(y)R_n(s,t,x,y)p(s,t,x,y)dxdy\nu(ds)\nu(dt) = \\ = \int_T \int_T \int_{\mathbb{R}^d} \int_{\mathbb{R}^d} f(x)g(y)R(s,t,x,y)p(s,t,x,y)dxdy\nu(ds)\nu(dt).$$

Consider the space of functions on  $\mathbb{R}^{2d}$  which are square integrable by a Gaussian measure with density p(s, t, x, y) and denote it as  $L_2(\mathbb{R}^{2d}, p(s, t, x, y)dxdy)$  with the corresponding scalar product  $\langle \cdot, \cdot \rangle$ . It is easy to check that  $\langle r_n(s, t, \cdot, \cdot), r_m(s, t, \cdot, \cdot) \rangle = 0$  if  $m \neq n$ (because  $r_n$  is a linear combination of multidimensional Hermitian polynomials of degree 2n). Therefore,

$$\begin{split} \sup_{n} &< R_{n}(s,t,\cdot,\cdot), R_{n}(s,t,\cdot,\cdot) > = \\ &= \sup_{n} \sum_{k=0}^{n} e^{-2k\lambda} < r_{k}(s,t,\cdot,\cdot), r_{k}(s,t,\cdot,\cdot) > \leq \\ &\leq < R(s,t,\cdot,\cdot), R(s,t,\cdot,\cdot) > . \end{split}$$

Using Lemmas 2 and 3 (see below), we obtain

$$< R(s,t,\cdot,\cdot), R(s,t,\cdot,\cdot) > = \\ = \int_{\mathbb{R}^d} \int_{\mathbb{R}^d} \frac{1}{(2\pi)^d} \frac{\sqrt{\det B_0(s,t)(s,t)}}{\det B_{e^{-\lambda}}(s,t)} \exp(-\frac{1}{2} \begin{pmatrix} x \\ y \end{pmatrix}^T (2B_{e^{-\lambda}}^{-1} - B_0^{-1})(s,t) \begin{pmatrix} x \\ y \end{pmatrix}) dxdy = \\ = \sqrt{\det B_0^{-1} B_{e^{-\lambda}} B_{-e^{-\lambda}}^{-1}(s,t)} \frac{\sqrt{\det B_0(s,t)(s,t)}}{\det B_{e^{-\lambda}}(s,t)} = \frac{\det B_0(s,t)}{\sqrt{\det B_{e^{-\lambda}}(s,t) \det B_{-e^{-\lambda}}(s,t)}} \le \\ \le (1 - e^{-\lambda})^{-2d}.$$

Applying the inequalities given above, we get

$$\begin{split} &\int_{\mathbb{R}^d} \int_{\mathbb{R}^d} |f(x)g(y)R_n(s,t,x,y)| p(s,t,x,y) dx dy \leq \\ &\leq \sqrt{\langle f(x)g(y), f(x)g(y) \rangle \langle R_n(s,t,\cdot,\cdot), R_n(s,t,\cdot,\cdot) \rangle \rangle} \leq \\ &\leq \sup_{x,y \in \mathbb{R}^d} |f(x)g(y)| (1-e^{-\lambda})^{-d}. \end{split}$$

We can see that the bound is independent of (s,t). So, by the Lebesgue dominated convergence theorem, we proved the equality and Theorem 2.

Now we prove a few technical results. Let m, n be positive integers, and let A be an  $(m + n) \times (m + n)$  symmetric non-negative definite real matrix of the form

 $A = \begin{pmatrix} A_{11} & A_{12} \\ A_{12}^T & A_{22} \end{pmatrix}, \text{ where } A_{11}, A_{12}, A_{22} \text{ are matrices with sizes } m \times m, m \times n, n \times$ 

**Lemma 1.** If  $|\alpha| < 1$ , then  $A_{\alpha}$  is positive definite.

*Proof.* Take the n + m-vector  $\begin{pmatrix} u \\ v \end{pmatrix}$  as a concatenation of the m-vector u and the n-vector v. Then we have to prove that  $(A_{\alpha} \begin{pmatrix} u \\ v \end{pmatrix}, \begin{pmatrix} u \\ v \end{pmatrix}) = (\begin{pmatrix} A_{11}u + \alpha A_{12}v \\ \alpha A_{12}^Tu + A_{22}v \end{pmatrix}, \begin{pmatrix} u \\ v \end{pmatrix}) = (A_{11}u, u) + 2\alpha(A_{12}v, u) + (A_{22}v, v)$  is always positive. Let  $u_1 = \alpha u, v_1 = v$ , and let us use that  $A = A_1$  is non-negative definite. We have

$$0 \leq (A_1 \begin{pmatrix} u_1 \\ v_1 \end{pmatrix}, \begin{pmatrix} u_1 \\ v_1 \end{pmatrix}) = \alpha^2 (A_{11}u, u) + 2\alpha (A_{12}v, u) + (A_{22}v, v) =$$
$$= (\alpha^2 - 1)(A_{11}u, u) + (A_\alpha \begin{pmatrix} u \\ v \end{pmatrix}, \begin{pmatrix} u \\ v \end{pmatrix}) \leq (A_\alpha \begin{pmatrix} u \\ v \end{pmatrix}, \begin{pmatrix} u \\ v \end{pmatrix}).$$

So we have that  $A_{\alpha}$  is non-negative definite. Moreover, if

$$(A_{\alpha}\begin{pmatrix} u\\v \end{pmatrix}, \begin{pmatrix} u\\v \end{pmatrix}) = 0,$$

then the equality holds in the inequality above, and we get  $(A_{11}u, u) = 0$ . But  $A_{11}$  is positive definite, and so u = 0. Similarly using that  $A_{22}$  is positive definite, we get v = 0. Hence,  $A_{\alpha}$  is positive definite.

**Lemma 2.** If  $|\alpha| < 1$ , then  $2A_{\alpha}^{-1} - A_0^{-1} = A_0^{-1}A_{-\alpha}A_{\alpha}^{-1}$ , and the matrix on the righthand side is positive definite.

*Proof.* Note that  $A_{\alpha}^{-1}$  and  $A_{0}^{-1}$  exist by the previous lemma. Consider the obvious relation  $2A_0 - A_\alpha = A_{-\alpha}$ . Multiplying it by  $A_{\alpha}^{-1}$  from the right and by  $A_{0}^{-1}$  from the left, we get the first part of the assertion. From Lemma 1, we also have that  $A_{0}^{-1}$ ,  $A_{\alpha}^{-1}$ ,  $andA_{-\alpha}$  are positive definite. Then  $A_{0}^{-1}A_{-\alpha}A_{\alpha}^{-1}$  is positive definite as a product of positive definite symmetric matrices.

**Lemma 3.** If  $|\alpha| < 1$ , then there exists a constant  $C(\alpha) > 0$  such that, for any matrix A, det  $A_0 \leq C(\alpha)$  det  $A_{\alpha}$  with  $C(\alpha) = (1 - |\alpha|)^{-(m+n)}$ .

*Proof.* We will prove that if B and C are symmetric  $l \times l$  non-negative definite matrices, then  $\det(B+C) \ge \det B$ . There exists the basis in  $\mathbb{R}^l$  such that C is a diagonal matrix with diagonal elements  $\{c_i \ge 0, i = 1, \ldots, l\}$ . Convert our matrices to this basis. By  $C_i$ , we denote the matrix with only one non-zero element  $c_i$  at the place (i, i) such that  $C = \sum_{i=1}^{l} C_i$ . By  $B_i$ , we denote the matrix obtained from B by removing the row i and the column i. Then we have  $\det B_i \ge 0$ , because B is non-negative definite, and  $\det(B+C_1) = \det B+c_1 \det B_1 \ge \det B$ . The matrix  $B+C_1$  is again non-negative definite and, by the same argument,  $\det(B+C_1+C_2) \ge \det(B+C_1) \ge \det B$ . Proceeding by induction, we get  $\det(B+C) = \det(B+\sum_{i=1}^{l} C_i) \ge \det B$ .

Suppose  $\alpha \ge 0$ . From the statement above, we obtain

$$\det A_{\alpha} = \det((1-\alpha)A_0 + \alpha A) \ge \det((1-\alpha)A_0) = (1-\alpha)^{m+n} \det A_0.$$

The case  $\alpha < 0$  is similar:

$$\det A_{\alpha} = \det((1+\alpha)A_0 - \alpha A_{-1}) \ge \det((1+\alpha)A_0) = (1+\alpha)^{m+n} \det A_0$$

 $(A_{-1}$  is non-negative definite by the same arguments as in Lemma 1).

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## MAIN RESULT

We want to know the asymptotic behavior of  $Ea_n(f_{\varepsilon_1})a_n(f_{\varepsilon_2})$  as  $n \to \infty$ , where  $f_{\varepsilon}$  approximates the delta-measure as defined in Introduction. Since

$$\begin{split} |Ea_{n}(f_{\varepsilon_{1}})a_{n}(f_{\varepsilon_{2}})| \\ &= \left| \int_{T} \int_{\mathbb{R}^{d}} \int_{\mathbb{R}^{d}} \int_{\mathbb{R}^{d}} \int_{\mathbb{R}^{d}} \int_{\mathbb{R}^{d}} f_{\varepsilon_{1}}(x) f_{\varepsilon_{2}}(y) e^{i(x,\gamma_{1})+i(y,\gamma_{2})} q_{n}(s,t,\gamma_{1},\gamma_{2}) dx dy d\gamma_{1} d\gamma_{2} \nu(ds) \nu(dt) \right| \\ &\leq \int_{T} \int_{T} \int_{\mathbb{R}^{d}} \int_{\mathbb{R}^{d}} \int_{\mathbb{R}^{d}} \int_{\mathbb{R}^{d}} \int_{\mathbb{R}^{d}} f_{\varepsilon_{1}}(x) f_{\varepsilon_{2}}(y) |q_{n}(s,t,\gamma_{1},\gamma_{2})| dx dy d\gamma_{1} d\gamma_{2} \nu(ds) \nu(dt) \\ &= \int_{T} \int_{T} \int_{\mathbb{R}^{d}} \int_{\mathbb{R}^{d}} \int_{\mathbb{R}^{d}} |q_{n}(s,t,\gamma_{1},\gamma_{2})| d\gamma_{1} d\gamma_{2} \nu(ds) \nu(dt), \end{split}$$

we are interested in the asymptotic behavior of  $\int_{\mathbb{R}^d} \int_{\mathbb{R}^d} |q_n(s,t,x,y)| dx dy$ . Fix  $(s,t) \in T^2$  such that det K(s,s) > 0 and det K(t,t) > 0. Denote

$$G(s,t) = K^{-1/2}(t,t)K(s,t)K^{-1/2}(s,s)$$
$$\|G(s,t)\| = \sup_{x \in \mathbb{R}^d, |x|=1} \|G(s,t)x\|$$

In other words, we have the correlation matrix and its operator norm.

**Theorem 3.** There exists a constant C > 0 such that, for any  $(s,t) \in T^2$  satisfying det K(s,s) > 0 and det K(t,t) > 0, the following inequality holds for all integers  $n \ge 0$ :

$$\int_{\mathbb{R}^d} \int_{\mathbb{R}^d} |q_n(s,t,x,y)| dx dy \le C \left( \det K(s,s) \det K(t,t) \right)^{-1/2} \|G(s,t)\|^n (n+1)^{d/2-1}.$$

Proof. We have

$$\begin{split} &\int_{\mathbb{R}^d} \int_{\mathbb{R}^d} |q_n(s,t,x,y)| dx dy = \\ &= \int_{\mathbb{R}^d} \int_{\mathbb{R}^d} \frac{1}{(2\pi)^{2d}} \frac{|(K(s,t)x,y)|^n}{n!} e^{-\frac{1}{2}(K(s,s)x,x)} e^{-\frac{1}{2}(K(t,t)y,y)} dx dy = \\ &= \frac{1}{(2\pi)^{2d}} \left( \det K(s,s) \det K(t,t) \right)^{-1/2} \int_{\mathbb{R}^d} \int_{\mathbb{R}^d} \frac{|(G(s,t)x,y)|^n}{n!} e^{-\frac{||x||^2}{2}} e^{-\frac{||y||^2}{2}} dx dy. \end{split}$$

Note that if  $n = 2k + 1, k \ge 0$ , then, by the Cauchy inequality,

$$\begin{split} &\int_{\mathbb{R}^d} \int_{\mathbb{R}^d} \frac{|(G(s,t)x,y)|^{2k+1}}{(2k+1)!} \frac{1}{(2\pi)^d} e^{-\frac{\|x\|^2}{2}} e^{-\frac{\|y\|^2}{2}} dx dy \leq \\ &\leq \sqrt{\frac{2k+2}{2k+1}} \sqrt{\int_{\mathbb{R}^d} \int_{\mathbb{R}^d} \frac{|(G(s,t)x,y)|^{2k}}{(2k)!} \frac{1}{(2\pi)^d} e^{-\frac{\|x\|^2}{2}} e^{-\frac{\|y\|^2}{2}} dx dy} \times \\ &\times \sqrt{\int_{\mathbb{R}^d} \int_{\mathbb{R}^d} \frac{|(G(s,t)x,y)|^{2k+2}}{(2k+2)!} \frac{1}{(2\pi)^d} e^{-\frac{\|x\|^2}{2}} e^{-\frac{\|y\|^2}{2}} dx dy}. \end{split}$$

So, it is enough to consider the case n = 2k. Using Lemma 4 (see below), we can establish a change of variables in the integral associated with the rotation around the origin in  $\mathbb{R}^{2d}$  such that  $(G(s,t)x,y) = \sum_{n=0}^{d} \lambda_n u_n v_n$ , where u, v are the new variables, and  $\lambda_i^2, i = 1, \ldots, n$ , are the eigenvalues of  $G(s,t)G^T(s,t)$ . Note that  $\lambda_i$  can be chosen non-negative, and  $\max_{i=1,\ldots,n} \lambda_i = \sqrt{\|G(s,t)G^T(s,t)\|} = \|G(s,t)\|$ . We note that

$$(\sum_{i=0}^{d} \lambda_{i} u_{i} v_{i})^{2n} = (2n)! \sum_{\sum_{i=0}^{d} k_{i} = 2n} \prod_{i=0}^{d} \frac{(\lambda_{i} u_{i} v_{i})^{k_{i}}}{k_{i}!},$$

and all members of the sum with one of  $k_i$  being odd became zero after the integrating with a Gaussian measure. All other members are non-negative, so we can increase the sum by replacing all  $\lambda_i$  by ||G(s,t)||. We get

$$\begin{split} &\int_{\mathbb{R}^d} \int_{\mathbb{R}^d} \frac{(G(s,t)x,y)^{2n}}{(2n)!} e^{-\frac{\|x\|^2}{2}} e^{-\frac{\|y\|^2}{2}} dx dy = \\ &\int_{\mathbb{R}^d} \int_{\mathbb{R}^d} \frac{(\sum_{i=0}^d \lambda_i u_i v_i)^{2n}}{(2n)!} e^{-\frac{\|u\|^2}{2}} e^{-\frac{\|v\|^2}{2}} du dv \le \\ &\leq \|G(s,t)\|^{2n} \int_{\mathbb{R}^d} \int_{\mathbb{R}^d} \frac{(\sum_{i=0}^d u_i v_i)^{2n}}{(2n)!} e^{-\frac{\|u\|^2}{2}} e^{-\frac{\|v\|^2}{2}} du dv. \end{split}$$

The last integral can be calculated precisely and satisfies the inequality

$$\int_{\mathbb{R}^d} \int_{\mathbb{R}^d} \frac{(\sum_{i=0}^d u_i v_i)^{2n}}{(2n)!} e^{-\frac{\|u\|^2}{2}} e^{-\frac{\|v\|^2}{2}} du dv \le \tilde{C} (2n+1)^{d/2-1}.$$

Hence we proved Theorem 3.

The following lemma was used in Theorem 3.

**Lemma 4.** For any  $d \times d$  matrix A, there exists the orthogonal  $2d \times 2d$  matrix U ( $U^T U = I$ ) such that  $(Ax, y) = \sum_{n=0}^{d} \lambda_n u_n v_n$ , where  $x, y, u, v \in \mathbb{R}^d$ ,  $\begin{pmatrix} u \\ v \end{pmatrix} = U \begin{pmatrix} x \\ y \end{pmatrix}$  and  $\lambda_i^2, i = 1, \ldots, n$ , are the eigenvalues of  $AA^T$ .

Proof. Denote  $z = \begin{pmatrix} x \\ y \end{pmatrix} \in \mathbb{R}^{2d}, x, y \in \mathbb{R}^d$ , and  $Q = \begin{pmatrix} 0 & A^T \\ A & 0 \end{pmatrix}$ . Note that (Qz, z) = 2(Ax, y), and Q is symmetric. All eigenvalues of Q are the set  $\{\lambda_i, -\lambda_i, i = 1, \dots, n\}$ , where  $\lambda_i$  are from the statement of Lemma 4. This is true, because if  $\lambda^2 > 0$ , then the equations  $Qz = \lambda z$  and  $AA^Ty = \lambda^2 y, \lambda x = A^Ty$  are equivalent. So we can find the orthogonal  $2d \times 2d$  matrix  $U_1$  such that  $2(Ax, y) = \sum_{n=0}^d \lambda_n (\tilde{u}_n^2 - \tilde{v}_n^2)$  and  $\begin{pmatrix} \tilde{u} \\ \tilde{v} \end{pmatrix} =$ 

 $U_1\begin{pmatrix} x\\ y \end{pmatrix}$ . This easily yields the statement of Lemma 4.

We are ready to establish the convergence of approximations. Denote

$$J_n = \int_T \int_T \|G(s,t)\|^n \left(\det K(s,s) \det K(t,t)\right)^{-1/2} \nu(ds)\nu(dt)$$

**Theorem 4.** 1) If  $J_n < +\infty$ , then  $\lim_{\varepsilon \to 0+} a_n(f_{\varepsilon})$  exists in  $L_2(\Omega)$ . For odd n, the limit is equal to zero.

 $\hat{2}$  If  $\sum_{n=0}^{\infty} J_n(n+1)^{\alpha+\frac{d}{2}-1} < +\infty$ , then  $\lim_{\varepsilon \to 0+} L_{\varepsilon}$  exists in  $D_{2,\alpha}$ .

Proof. From Theorem 2, we know that

$$\begin{aligned} Ea_n(f_{\varepsilon_1})a_n(f_{\varepsilon_2}) \\ &= \int_T \int_T \int_{\mathbb{R}^d} \int_{\mathbb{R}^d} \int_{\mathbb{R}^d} \int_{\mathbb{R}^d} f_{\varepsilon_1}(x) f_{\varepsilon_2}(y) e^{i(x,\gamma_1) + i(y,\gamma_2)} q_n(s,t,\gamma_1,\gamma_2) dx dy d\gamma_1 d\gamma_2 \nu(ds) \nu(dt) \\ &= \int_T \int_T \int_{\mathbb{R}^d} \int_{\mathbb{R}^d} \int_{\mathbb{R}^d} \int_{\mathbb{R}^d} f(x) f(y) e^{i(\varepsilon_1 x,\gamma_1) + i(\varepsilon_2 y,\gamma_2)} q_n(s,t,\gamma_1,\gamma_2) dx dy d\gamma_1 d\gamma_2 \nu(ds) \nu(dt). \end{aligned}$$

From Theorem 3, we have

$$\begin{aligned} |Ea_n(f_{\varepsilon_1})a_n(f_{\varepsilon_2})| &\leq \\ &\leq \int_T \int_{\mathbb{R}^d} \int_{\mathbb{R}^d} \int_{\mathbb{R}^d} \int_{\mathbb{R}^d} \int_{\mathbb{R}^d} f(x)f(y)|q_n(s,t,\gamma_1,\gamma_2)|dxdyd\gamma_1d\gamma_2\nu(ds)\nu(dt) \leq . \\ &\leq CJ_n(n+1)^{d/2-1} \end{aligned}$$

By the Lebesgue dominated convergence theorem

$$Ea_n(f_{\varepsilon_1})a_n(f_{\varepsilon_2}) \to \int_T \int_T \int_{\mathbb{R}^d} \int_{\mathbb{R}^d} q_n(s,t,x,x) dx dy \nu(ds) \nu(dt), \varepsilon_1, \varepsilon_2 \to 0 + .$$

We conclude that the limit of  $a_n(f_{\varepsilon})$  as  $\varepsilon \to 0+$  exists in  $L_2$ . For odd n, the limit is obviously zero so we proved the first part of Theorem 4. The second part easily follows from the first part because each  $a_n(f_{\varepsilon})$  satisfies (as noted above)  $||a_n(f_{\varepsilon})||_2^2 \leq CJ_n(n+1)^{d/2-1}$ .

## EXAMPLES

All conditions we imposed on  $\xi$  in Introduction are obviously satisfied when  $T \subset \mathbb{R}^d$ and  $\xi$  is a continuous centered Gaussian random field. So, without a further notice, we will use our results for such a kind of random fields.

**Example 1.** Let  $T = [0,1], v(dt) = dt, \xi(t) = t^r W(t)$ , where W(t) is the *d*-dimensional Brownian motion. In this case, we have

$$\begin{split} K_{ii}(s,t) &= (st)^r \min(s,t), K_{ij}(s,t) = 0, \\ \det K(s,s) &= s^{(2r+1)d}, \\ \|G(s,t)\| &= \frac{\min(s,t)}{\sqrt{st}}, \\ J_n &= \int_0^1 \int_0^1 (st)^{-\frac{n}{2} - \frac{(2r+1)d}{2}} (\min(s,t))^n ds dt \end{split}$$

It easy to see that the condition  $J_n < +\infty$  is satisfied for  $r < \frac{1}{d} - \frac{1}{2}$  for all  $n \ge 0$  (and failed for all n in either case), and

$$J_n = \frac{4}{(n+2-(2r+1)d)(2-(2r+1)d)} = O(\frac{1}{n})$$

Using Theorem 4, we conclude that the limit of  $L_{\varepsilon}$  exists in  $D_{2,\alpha}$  for  $\alpha < 1 - \frac{d}{2}$ .

**Example 2.** Let  $T = [0, 1], v(dt) = dt, \xi(t) = \int_0^t s^r dW(s)$ . The integral exists if  $r > -\frac{1}{2}$ . As in the previous example, we can easily compute  $J_n$ :

$$\begin{split} K_{ii}(s,t) &= \frac{1}{2r+1} (\min(s,t))^{2r+1}, K_{ij}(s,t) = 0, \\ \det K(s,s) &= (2r+1)^{-d} s^{(2r+1)d}, \\ \|G(s,t)\| &= \left(\frac{\min(s,t)}{\sqrt{st}}\right)^{2r+1}, \\ J_n &= (2r+1)^{-d} \int_0^1 \int_0^1 (st)^{-\frac{n(2r+1)}{2} - \frac{d(2r+1)}{2}} (\min(s,t))^{n(2r+1)} ds dt = \\ &= (2r+1)^{-d} \frac{4}{(n(2r+1)+2 - d(2r+1))(2 - d(2r+1))} = O(\frac{1}{n}). \end{split}$$

As in the previous example,  $J_n < +\infty$  is satisfied for  $r < \frac{1}{d} - \frac{1}{2}$ , and  $L_{\varepsilon}$  converges in  $D_{2,\alpha}$  for  $\alpha < 1 - \frac{d}{2}$ . This example and the previous one are the generalizations of the classical result about the existence of a local time for the Brownian motion. The additional multiplier  $t^r, r \in (-\frac{1}{2}, -\frac{1}{2} + \frac{1}{d})$  prevents the explosion of Itô—Wiener expansion kernels. It is interesting to note that the explosion occurs near t = 0. So if we take  $T = [\frac{1}{2}, 1]$  for this example and the previous one, then we may drop the restriction on r (and we may

set r = 0 to get the usual Brownian motion). The convergence still takes place under same conditions on  $\alpha$ .

Now we consider two special cases of the problem stated in Introduction for Q = 0and Q = I. The second case can also be found in [1].

**Example 3.** Let  $T = [0,1]^2$ ,  $v(dt) = dt_1 dt_2$ ,  $\xi(t_1,t_2) = W_1(t_1) - W_2(t_2)$ , where  $W_i(t)$ , i = 1, 2, are two independent *d*-dimensional Brownian motions. We have

$$\begin{split} K_{ii}(s,t) &= \min(s_1,t_1) + \min(s_2,t_2), K_{ij}(s,t) = 0, \\ \det K(s,s) &= (s_1 + s_2)^d, \\ \|G(s,t)\| &= \frac{\min(s_1,t_1) + \min(s_2,t_2)}{\sqrt{(s_1 + s_2)(t_1 + t_2)}}, \\ J_n &= \int_0^1 \int_0^1 \int_0^1 \int_0^1 \frac{(\min(s_1,t_1) + \min(s_2,t_2))^n}{((s_1 + s_2)(t_1 + t_2))^{\frac{n}{2} + \frac{d}{2}}} ds_1 ds_2 dt_1 dt_2. \end{split}$$

To find the asymptotic behaviour for  $J_n$ , we divide the region of integration  $T^2 = [0, 1]^4$  into subregions  $T_1 = \{0 \le s_1 \le t_1 \le 1; 0 \le s_2 \le t_2 \le 1\}, T_2 = \{0 \le s_1 \le t_1 \le 1; 0 \le t_2 \le s_2 \le 1\}, T_3 = \{0 \le t_1 \le s_1 \le 1; 0 \le s_2 \le t_2 \le 1\}, T_4 = \{0 \le t_1 \le s_1 \le 1; 0 \le t_2 \le s_2 \le 1\}.$ 

$$\begin{split} J_n^1 &= \int_{T_1} \frac{(\min(s_1, t_1) + \min(s_2, t_2))^n}{((s_1 + s_2)(t_1 + t_2))^{\frac{n}{2} + \frac{d}{2}}} ds_1 ds_2 dt_1 dt_2 \\ &= \int_0^1 \int_0^1 \int_0^{t_2} \int_0^{t_1} (s_1 + s_2)^{\frac{n-d}{2}} (t_1 + t_2)^{\frac{-n-d}{2}} ds_1 ds_2 dt_1 dt_2 \\ &= \frac{1}{(1 + \frac{n-d}{2})(2 + \frac{n-d}{2})} \int_0^1 \int_0^1 (t_1 + t_2)^{\frac{-n-d}{2}} ((t_1 + t_2)^{\frac{n-d}{2} + 2} - t_1^{\frac{n-d}{2} + 2}) dt_1 dt_2 \\ &\leqslant \tilde{C} \frac{1}{(n+1)^2} \int_0^1 \int_0^1 (t_1 + t_2)^{2-d} dt_1 dt_2 \leqslant \tilde{\tilde{C}} \frac{1}{(n+1)^2} \int_0^1 u^{3-d} du. \end{split}$$

We can see that the sufficient condition for the integral to exist is d < 4 (the special cases d = 2, n = 0; d = 3, n = 1 behave themselves in a slightly different way: logarithms appear after the partial integration, but the corresponding integrals are still finite by same arguments).

$$\begin{split} J_n^2 &= \int_{T_2} \frac{(\min(s_1, t_1) + \min(s_2, t_2))^n}{((s_1 + s_2)(t_1 + t_2))^{\frac{n}{2} + \frac{d}{2}}} ds_1 ds_2 dt_1 dt_2 = \\ &= \int_0^1 \int_0^1 \int_{s_1}^1 \int_{t_2}^1 (s_1 + t_2)^n (s_1 + s_2)^{\frac{-n-d}{2}} (t_1 + t_2)^{\frac{-n-d}{2}} ds_2 dt_1 ds_1 dt_2 = \\ &= \frac{1}{(1 - \frac{n+d}{2})^2} \int_0^1 \int_0^1 (s_1 + t_2)^n ((1 + s_1)^{1 - \frac{n+d}{2}} - (s_1 + t_2)^{1 - \frac{n+d}{2}}) \times \\ &\times ((1 + t_2)^{1 - \frac{n+d}{2}} - (s_1 + t_2)^{1 - \frac{n+d}{2}}) ds_1 dt_2 \leqslant \\ &\leqslant C \frac{1}{(n+1)^2} \int_0^1 \int_0^1 (s_1 + t_2)^{2-d} ds_1 dt_2 \leqslant \\ &\leqslant \tilde{C} \frac{1}{(n+1)^2} \int_0^1 u^{3-d} du. \end{split}$$

We assumed above that n + d > 2. In all other cases, d = 1, n = 0, 1; d = 2, n = 0, we can check that the integral is finite. Again we have the sufficient condition d < 4. Two

last subregions can be treated similarly, and we omit them. We conclude:  $J_n < +\infty$  is satisfied for d < 4 and all  $n \ge 0$ . In this case,  $J_n = O(\frac{1}{n^2})$ , so the limit exists in  $D_{2,\alpha}$ for  $\alpha < 2 - \frac{d}{2}$ . As a consequence, we have the existence of the intersection local time for two independent *d*-dimensional Brownian motions for d = 1, 2, 3.

**Example 4.** Let 
$$T = [0, 1]^2$$
,  $v(dt) = dt_1 dt_2$ ,  $\xi(t_1, t_2) = W(t_1) - W(t_2)$ . We have

$$\begin{split} K_{ii}(s,t) &= \min(s_1,t_1) + \min(s_2,t_2) - \min(s_2,t_1) - \min(s_1,t_2), K_{ij}(s,t) = 0, \\ \det K(s,s) &= (s_1 + s_2 - 2\min(s_1,s_2))^d, \\ \|G(s,t)\| &= \frac{\min(s_1,t_1) + \min(s_2,t_2) - \min(s_2,t_1) - \min(s_1,t_2)}{\sqrt{(s_1 + s_2 - 2\min(s_1,s_2))(t_1 + t_2 - 2\min(t_1,t_2))}}, \\ J_n &= \int_0^1 \int_0^1 \int_0^1 \int_0^1 \frac{(\min(s_1,t_1) + \min(s_2,t_2) - \min(s_2,t_1) - \min(s_1,t_2))^n}{((s_1 + s_2 - 2\min(s_1,s_2))(t_1 + t_2 - 2\min(t_1,t_2)))^{\frac{n}{2} + \frac{d}{2}}} ds_1 ds_2 dt_1 dt_2 \end{split}$$

As in the previous example, we have to split the region  $[0,1]^4$  into a few subregions. We consider only three of them, others are similar:  $T_1 = \{0 \leq s_1 \leq s_2 \leq t_1 \leq t_2 \leq 1\}$ ,  $T_2 = \{0 \leq s_1 \leq t_1 \leq s_2 \leq t_2 \leq 1\}$ ,  $T_3 = \{0 \leq s_1 \leq t_1 \leq t_2 \leq s_2 \leq 1\}$ . In the first subregion,  $K_{ii}(s,t) = 0$  and  $J_n^1 = 0$ . Consider the second subregion:

$$\begin{split} J_n^2 &= \int_0^1 \int_0^{s_2} \int_0^{t_1} \int_{s_2}^1 \frac{(s_2 - t_1)^n}{(s_2 - s_1)^{\frac{n+d}{2}} (t_2 - t_1)^{\frac{n+d}{2}}} dt_2 ds_1 dt_1 ds_2 = \frac{1}{(1 - \frac{n+d}{2})^2} \\ &\times \int_0^1 \int_0^{s_2} (s_2 - t_1)^n ((1 - t_1)^{1 - \frac{n+d}{2}} - (s_2 - t_1)^{1 - \frac{n+d}{2}}) (s_2^{1 - \frac{n+d}{2}} - (s_2 - t_1)^{1 - \frac{n+d}{2}}) dt_1 ds_2 \\ &\leqslant C \frac{1}{(n+1)^2} \int_0^1 \int_0^{s_2} (s_2 - t_1)^{2 - d} dt_1 ds_2 \leqslant C \frac{1}{(n+1)^2} \int_0^1 u^{2 - d} du. \end{split}$$

We assumed that n + d > 2. In that case, the sufficient condition for the integral to exist is d < 3. In the cases d = 1, n = 0, 1 and d = 2, n = 0, the integral is also finite.

$$\begin{split} J_n^3 &= \int_0^1 \int_0^{s_2} \int_{s_1}^{s_2} \int_{s_1}^{t_2} \frac{(t_2 - t_1)^{\frac{n-d}{2}}}{(s_2 - s_1)^{\frac{n+d}{2}}} dt_1 dt_2 ds_1 ds_2 = \\ &= \frac{1}{(1 + \frac{n-d}{2})(2 + \frac{n-d}{2})} \int_0^1 \int_0^{s_2} (s_2 - s_1)^{2-d} ds_1 ds_2 \leqslant \\ &\leqslant C \frac{1}{(n+1)^2} \int_0^1 u^{2-d} du. \end{split}$$

The sufficient conditions for the integral to exist: d < 3 and n > 2 - d. Again we have to treat the cases d = 1, n = 0, 1 and d = 2, n = 0 separately. But, in the last case, the integral is infinite:

$$J_0^3 = \int_0^1 \int_0^{s_2} \int_{s_1}^{s_2} \int_{s_1}^{t_2} \frac{1}{(s_2 - s_1)(t_2 - t_1)} dt_1 dt_2 ds_1 ds_2 =$$
  
=  $\int_0^1 \int_0^{s_2} \int_{s_1}^{s_2} (\int_0^{t_2 - s_1} \frac{1}{u} du) \frac{1}{(s_2 - s_1)} dt_2 ds_1 ds_2 = +\infty.$ 

Finally,  $J_n < +\infty$  is satisfied for d < 3 and all  $n \ge 0$  except n = 0, d = 2. The limit exists in  $D_{2,\alpha}$ ,  $\alpha < 2 - \frac{d}{2}$  for d = 1. If d = 2, then as our calculations suggest, it is possible to subtract the first term of the expansion and prove the convergence of the rest. It is known as the renormalization of the self-intersection local time for the Brownian motion in two dimensions (see [1]).

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