

## Existence and regularization of the local times of a Gaussian process

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### Abstract

We study an existence result in the mean square sense of the local times of a one-dimensional Gaussian process defined by an indefinite Wiener integral. For any spatial dimension, we prove that the local times of a Gaussian process, after appropriately renormalized, exist as white noise distributions. We also present a regularization of the local times and show a convergence result in Hida distributions space.

**Keywords:** Local times, b-Gaussian process, white noise analysis.

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### 1. Introduction

The present paper concerns the investigation of white noise analysis approach to the local times of a certain class of Gaussian processes defined by indefinite Wiener integrals. The first idea of analyzing local times using white noise approach goes back at least to the work of Watanabe [12]. A study of local times of Brownian motion using white noise approach without renormalization was briefly mentioned in [9]. White noise technique has been further applied to the problem of local times and self-intersection local times, see e.g. [1, 4, 3, 7] just to mention a few. White noise approach to self-intersection local times has been applied also to problem in physics, see for example [2] and [5].

First of all let us fix a strictly positive real number  $T$ . The space of real-valued square-integrable function with respect to the Lebesgue measure on the interval  $[0, T]$  will be denoted by  $L^2([0, T])$ . Let  $f \in L^2([0, T])$  and  $B = (B_t)_{t \in [0, T]}$  be a standard Brownian motion defined on some probability space  $(\Omega, \mathcal{F}, \mathbb{P})$ . It is a well-known result from Itô's stochastic calculus that the stochastic process  $X = (X_t)_{t \in [0, T]}$  defined by the indefinite Wiener integral  $X_t := \int_0^t f(u) dB_u$  is a centered Gaussian process with covariance function  $\mathbb{E}(X_s X_t) = \int_0^{s \wedge t} f(u)^2 du$ ,  $s, t \geq 0$ . Here  $\mathbb{E}$  denotes the expectation with respect to the probability measure  $\mathbb{P}$  and  $s \wedge t$  denotes the minimum between  $s$  and  $t$ . In fact,  $X$  is also an  $L^2(\mathbb{P})$ -continuous martingale with respect to the natural

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filtration of  $B$ . In this work we further assume that  $f$  is bounded on  $[0, T]$ . We call the corresponding stochastic process as *b-Gaussian process*. By choosing  $f$  to be the constant function 1, we see that the class of b-Gaussian processes contains Brownian motion as an example. Moreover, by  $d$ -dimensional b-Gaussian process we mean the random vector  $(X^1, \dots, X^d)$  where  $X^1, \dots, X^d$  are  $d$  independent copies of one-dimensional b-Gaussian process. In [7] b-Gaussian process has been studied in the context of self-intersection local times. The main object of study in the present paper will be the *local time* of a b-Gaussian process  $X$  at a point  $c \in \mathbb{R}$ , which is informally defined as

$$(1.1) \quad \int_0^T \delta(X_t - c) dt,$$

where  $\delta$  denotes the Dirac-delta function at 0. The (generalized) random variable (1.1) is intended to measure the amount of time in which the sample path of a b-Gaussian process  $X$  spends at a given point  $c \in \mathbb{R}$  within the time interval  $[0, T]$ . A priori the expression (1.1) has no mathematical meaning. One common way to give such expression a sense is via an approximation using a Dirac sequence. More precisely, we interpret (1.1) as the limiting object of the approximated local time  $\mathcal{L}_{X,\varepsilon}(T)$  of b-Gaussian process  $X$  defined by  $\mathcal{L}_{X,\varepsilon}(T) := \int_0^T p_\varepsilon(X_t - c) dt$ ,  $\varepsilon > 0$ , as  $\varepsilon \rightarrow 0$ , where  $p_\varepsilon$  is the heat kernel  $p_\varepsilon(x) := \frac{1}{\sqrt{2\pi\varepsilon}} \exp\left(-\frac{x^2}{2\varepsilon}\right)$ ,  $x \in \mathbb{R}$ . This approximation procedure makes the limiting object, which we denote by  $\mathcal{L}_X(T)$ , more and more singular as the dimension of the process  $X$  increases. Hence, we need to do a *renormalization*, i.e. removal of the divergent terms, to obtain a well-defined object.

Now we describe briefly our main results. First, we investigate the existence of the local times of a b-Gaussian process as the density of the occupation measure. This density does exist in dimension one and in that case we show that the local times is a well-defined object as a limit in the mean square sense. In the white noise analysis framework we investigate the local times of b-Gaussian process for any spatial dimension. Under some conditions on the dimension of the b-Gaussian process  $X$  and the number of subtracted terms in the truncated Donsker's delta function, we are able to show the existence of the (truncated) local time  $\mathcal{L}_X(T)$  as a well-defined object in some white noise distribution space. We also analyze a regularization corresponding to the Gaussian approximation described above and prove a convergence result. The organization of the paper is as follows. In section 2 we summarize some of the standard facts from the theory of white noise analysis used throughout this paper. Section 3 contains a detailed exposition of the main results and their proofs. Some concluding remarks are given in the last section.

## 2. Elements of white noise analysis

We briefly recall some pertinent results and notions from white noise analysis. For a more comprehensive discussion we refer to [6, 11], among others. A survey on white noise analysis and its application to Feynman integral is given in [8]. Let  $(S'_d(\mathbb{R}), \mathcal{C}, \mu)$  be the  $\mathbb{R}^d$ -valued white noise space, i.e.,  $S'_d(\mathbb{R})$  is the space of  $\mathbb{R}^d$ -valued tempered distributions,  $\mathcal{C}$  is the Borel  $\sigma$ -algebra generated by cylinder sets in  $S'_d(\mathbb{R})$ , and  $\mu$  is the so-called white noise measure. The probability measure  $\mu$  is uniquely determined through the Bochner-Minlos theorem by fixing the characteristic function

$$C(\vec{f}) := \int_{S'_d(\mathbb{R})} \exp\left(i\langle \vec{\omega}, \vec{f} \rangle\right) d\mu(\vec{\omega}) = \exp\left(-\frac{1}{2}|\vec{f}|_0^2\right)$$

for all  $\mathbb{R}^d$ -valued Schwartz test function  $\vec{f} \in S_d(\mathbb{R})$ . Here  $|\cdot|_0$  denotes the usual norm in the real Hilbert space  $L^2_d(\mathbb{R})$  of all  $\mathbb{R}^d$ -valued Lebesgue square-integrable functions, and  $\langle \cdot, \cdot \rangle$  denotes the dual pairing between  $S'_d(\mathbb{R})$  and  $S_d(\mathbb{R})$ . We also have the Gel'fand

triple, i.e. the continuous and dense embedding  $\mathcal{S}_d(\mathbb{R}) \hookrightarrow L^2_d(\mathbb{R}) \hookrightarrow \mathcal{S}'_d(\mathbb{R})$ . Let  $f$  be a function in the subset of  $L^2([0, T])$  consisting all real-valued bounded functions on  $[0, T]$ . In the frame of white noise analysis, a  $d$ -dimensional b-Gaussian process can be represented by a continuous version of the stochastic process  $X = (X_t)_{t \in [0, T]}$  with  $X_t := (\langle \cdot, \mathbf{1}_{[0, t]} f \rangle, \dots, \langle \cdot, \mathbf{1}_{[0, t]} f \rangle)$ , such that for independent  $d$ -tuples of Gaussian white noise  $\vec{\omega} = (\omega_1, \dots, \omega_d) \in \mathcal{S}'_d(\mathbb{R})$  it holds that  $X_t(\vec{\omega}) = (\langle \omega_1, \mathbf{1}_{[0, t]} f \rangle, \dots, \langle \omega_d, \mathbf{1}_{[0, t]} f \rangle)$ ,  $\vec{\omega} = (\omega_1, \dots, \omega_d) \in \mathcal{S}'_d(\mathbb{R})$ . Here  $\mathbf{1}_A$  denotes the indicator function of a set  $A \subset \mathbb{R}$ .

Let us denote the complex Hilbert space  $L^2(\mathcal{S}'_d(\mathbb{R}), \mathcal{C}, \mu)$  by  $L^2(\mu)$ . There are several ways to construct space of white noise test functions and distributions. For example, starting from  $L^2(\mu)$  and making use of the Wiener-Itô-Segal isomorphism and the second quantization operator of the Hamiltonian of a harmonic oscillator we can obtain the Gel'fand triple  $(\mathcal{S}) \hookrightarrow L^2(\mu) \hookrightarrow (\mathcal{S})^*$ . The space  $(\mathcal{S})$  of white noise test functions is obtained by taking the intersection of a family of Hilbert subspaces of  $L^2(\mu)$ . It is equipped with the projective limit topology and has the structure of nuclear Fréchet space. The space of generalized white noise functionals  $(\mathcal{S})^*$  is defined as the topological dual space of  $(\mathcal{S})$ . Elements of  $(\mathcal{S})$  and  $(\mathcal{S})^*$  are also known as *Hida test functions* and *Hida distributions*, respectively. The main example of element of  $(\mathcal{S})^*$  is the  $d$ -dimensional white noise process  $W_t := (\langle \cdot, \delta_t \rangle, \dots, \langle \cdot, \delta_t \rangle)$ , where  $\delta_t$  is the Dirac-delta function at  $t \in \mathbb{R}$ . It can be considered as the (componentwise) time-derivative of the  $d$ -dimensional Brownian motion. The rest of this section is devoted to a characterization of Hida distributions. The *S-transform* of an element  $\Phi \in (\mathcal{S})^*$  is defined as  $(S\Phi)(\vec{f}) := \langle \langle \Phi, : \exp(\langle \cdot, \vec{f} \rangle) : \rangle \rangle$ ,  $\vec{f} \in \mathcal{S}_d(\mathbb{R})$ , where  $: \exp(\langle \cdot, \vec{f} \rangle) := C(\vec{f}) \exp(\langle \cdot, \vec{f} \rangle) \in (\mathcal{S})$ , is the so-called Wick exponential and  $\langle \langle \cdot, \cdot \rangle \rangle$  denotes the dual pairing between  $(\mathcal{S})^*$  and  $(\mathcal{S})$ . The S-transform provides a quite useful way to identify a Hida distribution  $\Phi \in (\mathcal{S})^*$ , in particular, when it is very hard or impossible to find the explicit form for the Wiener-Itô chaos decomposition of  $\Phi$ .

**2.1. Theorem.** [10] *A function  $F : \mathcal{S}_d(\mathbb{R}) \rightarrow \mathbb{C}$  is the S-transform of a unique Hida distribution in  $(\mathcal{S})^*$  if and only if it satisfies the conditions:*

- (1)  *$F$  is ray analytic, i.e., for every  $\vec{f}, \vec{g} \in \mathcal{S}_d(\mathbb{R})$  the mapping  $\mathbb{R} \ni \gamma \mapsto F(\gamma \vec{f} + \vec{g})$  has an entire extension to  $\gamma \in \mathbb{C}$ , and*
- (2)  *$F$  has growth of second order, i.e., there exist constants  $K_1, K_2 > 0$  and a continuous norm  $\|\cdot\|$  on  $\mathcal{S}_d(\mathbb{R})$  such that  $|F(z\vec{f})| \leq K_1 \exp(K_2 |z|^2 \|\vec{f}\|^2)$ , for all  $z \in \mathbb{C}$ ,  $\vec{f} \in \mathcal{S}_d(\mathbb{R})$ .*

There are two important consequences of the above characterization theorem. For details and proofs see [10].

**2.2. Corollary.** *Let  $(\Omega, \mathcal{A}, \nu)$  be a measure space and  $\gamma \mapsto \Phi_\gamma$  be a mapping from  $\Omega$  to  $(\mathcal{S})^*$ . If the S-transform of  $\Phi_\gamma$  fulfils the following two conditions:*

- (1) *the mapping  $\gamma \mapsto S(\Phi_\gamma)(\vec{f})$  is measurable for all  $\vec{f} \in \mathcal{S}_d(\mathbb{R})$ , and*
- (2) *there exist  $C_1(\gamma) \in L^1(\Omega, \mathcal{A}, \nu)$ ,  $C_2(\gamma) \in L^\infty(\Omega, \mathcal{A}, \nu)$  and a continuous norm  $\|\cdot\|$  on  $\mathcal{S}_d(\mathbb{R})$  such that  $|S(\Phi_\gamma)(z\vec{f})| \leq C_1(\gamma) \exp(C_2(\gamma) |z|^2 \|\vec{f}\|^2)$ , for all  $z \in \mathbb{C}$ ,  $\vec{f} \in \mathcal{S}_d(\mathbb{R})$ ,*

*then  $\Phi_\gamma$  is Bochner integrable with respect to some Hilbertian norm which topologizing  $(\mathcal{S})^*$ . Hence  $\int_\Omega \Phi_\gamma d\nu(\gamma) \in (\mathcal{S})^*$ , and furthermore*

$$S\left(\int_\Omega \Phi_\gamma d\nu(\gamma)\right) = \int_\Omega S(\Phi_\gamma) d\nu(\gamma).$$

**2.3. Corollary.** Let  $(\Phi_n)_{n \in \mathbb{N}}$  be a sequence in  $(S)^*$  such that

- (1) for all  $\vec{f} \in \mathcal{S}_d(\mathbb{R})$ ,  $(S(\Phi_n)(\vec{f}))_{n \in \mathbb{N}}$  is a Cauchy sequence in  $\mathbb{C}$ , and
- (2) there exist constants  $K_1, K_2 > 0$  and a continuous norm  $\|\cdot\|$  on  $\mathcal{S}_d(\mathbb{R})$  such that
 
$$|S(\Phi_n)(z\vec{f})| \leq K_1 \exp\left(K_2|z|^2 \|\vec{f}\|^2\right), \text{ for all } z \in \mathbb{C}, \vec{f} \in \mathcal{S}_d(\mathbb{R}), n \in \mathbb{N}.$$

Then  $(\Phi_n)_{n \in \mathbb{N}}$  converges strongly in  $(S)^*$  to a unique Hida distribution  $\Phi \in (S)^*$ .

### 3. Local times of b-Gaussian processes

Let  $f : [0, T] \rightarrow \mathbb{R}$  be a (nonrandom) Borel measurable function and  $\lambda$  be the Lebesgue measure in  $[0, T]$ . The *occupation measure* of  $f$  up to "time"  $T$  is defined by  $\mu_T(A) := \lambda(\{t \in [0, T] : f(t) \in A\})$ , where  $A$  is a Borel set in  $\mathbb{R}$ . Thus,  $\mu_T(A)$  describes the amount of time spent by  $f$  in the set  $A$  during  $[0, T]$ . If  $[0, T] \ni t \mapsto X_t(\omega) \in \mathbb{R}$  is a sample path of a stochastic process, then its occupation measure is defined in the same way, but now  $\mu_T(A)$  will depend also on the sample point  $\omega$  from the underlying probability space. The following theorem gives condition on which the occupation measure of a b-Gaussian process possesses a density. In that case we call the density as the local time of the b-Gaussian process.

**3.1. Proposition.** Let  $X = (X_t)_{t \in [0, T]}$  be a  $d$ -dimensional b-Gaussian process defined on a probability space  $(\Omega, \mathcal{F}, \mathbb{P})$ . For  $d = 1$ , the occupation measure  $\mu_T$  of  $X$ , i.e.  $\mu_T(A) := \int_0^T \mathbf{1}_A(X_t) dt = \lambda(\{t \in [0, T] : X_t \in A\})$ ,  $A \in \mathcal{B}(\mathbb{R}^d)$ , where  $\mathcal{B}(\mathbb{R}^d)$  is the Borel  $\sigma$ -algebra on  $\mathbb{R}^d$ , is  $\mathbb{P}$ -almost surely absolutely continuous with respect to the Lebesgue measure  $\lambda_d$  on  $\mathbb{R}^d$ .

*Proof.* A standart result from geometric measure theory states that absolute continuity of  $\mu_T$  with respect to  $\lambda_d$  holds if for  $\mu_T$ -a.e.  $x \in \mathbb{R}^d$  we have

$$\liminf_{r \rightarrow 0} \frac{\mu_T(B(x, r))}{\lambda_d(B(x, r))} < \infty,$$

where  $B(x, r)$  is the closed ball around  $x$  with radius  $r$ . We apply Fatou's lemma and Fubini's theorem to obtain

$$\begin{aligned} & \mathbb{E} \int_{\mathbb{R}^d} \liminf_{r \rightarrow 0} \frac{\mu_T(B(x, r))}{\lambda_d(B(x, r))} d\mu_T(x) \\ & \leq \frac{\Gamma(1 + d/2)}{\pi^{d/2}} \liminf_{r \rightarrow 0} \frac{1}{r^d} \mathbb{E} \int_{\mathbb{R}^d} \mu_T(B(x, r)) d\mu_T(x) \\ & = \frac{\Gamma(1 + d/2)}{\pi^{d/2}} \liminf_{r \rightarrow 0} \frac{1}{r^d} \mathbb{E} \int_0^T \int_{\mathbb{R}^d} \mathbf{1}_{B(x, r)}(X_t) d\mu_T(x) dt \\ & = \frac{\Gamma(1 + d/2)}{\pi^{d/2}} \liminf_{r \rightarrow 0} \frac{1}{r^d} \mathbb{E} \int_0^T \int_0^T \mathbf{1}_{B(X_t, r)}(X_s) ds dt \\ & = \frac{\Gamma(1 + d/2)}{\pi^{d/2}} \liminf_{r \rightarrow 0} \frac{1}{r^d} \int_0^T \int_0^T \mathbb{P}(|X_t - X_s| \leq r) ds dt \\ & \leq \frac{\Gamma(1 + d/2)}{\pi^{d/2}} \liminf_{r \rightarrow 0} \frac{1}{r^d} \int_0^T \int_0^T \left( \frac{1}{2\pi \int_s^t f(u)^2 du} \right)^{d/2} \lambda_d(B(x, r)) ds dt \\ & \leq \frac{2}{\alpha^d (2\pi)^{d/2}} \int_0^T \int_0^t (t-s)^{-d/2} ds dt. \end{aligned}$$

The positive constant  $\alpha$  exists by the assumption on the function  $f$ . Notice that in the last integral we have assumed, without loss of generality, that  $s < t$ . Furthermore, the last integral is finite if and only if  $d = 1$ .  $\square$

By the Radon-Nikodym theorem, the occupation measure  $\mu_T$  of  $X$  has a density function and it is a feasible measure for the time spent at a given point during the time interval  $[0, T]$ . Hence, it is reasonable to define local time of b-Gaussian process at a point  $c \in \mathbb{R}$  during  $[0, T]$  by (1.1). Now we proceed to establish the existence of (1.1) as the limiting object of a sequence of square-integrable functions.

**3.2. Theorem.** *The approximated local time*

$$\mathcal{L}_{X,\varepsilon}(T) := \int_0^T \frac{1}{\sqrt{2\pi\varepsilon}} \exp\left(-\frac{(X_t - c)^2}{2\varepsilon}\right) dt, \quad \varepsilon > 0$$

of one-dimensional b-Gaussian process  $X$  converges in  $L^2(\mathbb{P})$  as  $\varepsilon$  tends to zero.

*Proof.* We observe that  $\mathcal{L}_{X,\varepsilon}(T) = \frac{1}{2\pi} \int_0^T \int_{\mathbb{R}} \exp(i\xi(X_t - c)) \exp(-\frac{\varepsilon}{2}\xi^2) d\xi dt$ . Let us denote  $D := \{(t_1, t_2) : 0 < t_1 < t_2 < T\}$ . Hence,

$$\begin{aligned} & \mathbb{E}(\mathcal{L}_{X,\varepsilon}(T)^2) \\ &= \mathbb{E}\left(\frac{1}{4\pi^2} \int_D \int_{\mathbb{R}^2} \exp\left(i \sum_{j=1}^2 \xi_j(X_{t_j} - c)\right) \exp\left(-\frac{\varepsilon}{2} \sum_{j=1}^2 \xi_j^2\right) d\xi dt\right) \\ &= \frac{1}{4\pi^2} \int_D \int_{\mathbb{R}^2} \mathbb{E}\left(\exp\left(i \sum_{j=1}^2 \xi_j(X_{t_j} - c)\right)\right) \exp\left(-\frac{\varepsilon}{2} \sum_{j=1}^2 \xi_j^2\right) d\xi dt \\ &= \frac{1}{4\pi^2} \int_D \int_{\mathbb{R}^2} \exp\left(-\frac{1}{2} \text{var}\left(\sum_{j=1}^2 \xi_j(X_{t_j} - c)\right)\right) \exp\left(-ic \sum_{j=1}^2 \xi_j\right) \\ & \quad \times \exp\left(-\frac{\varepsilon}{2} \sum_{j=1}^2 \xi_j^2\right) d\xi dt, \end{aligned}$$

where  $\text{var}(X)$  denotes the variance of the random variable  $X$ . Note that by Lebesgue's dominated convergence theorem  $\mathbb{E}(\mathcal{L}_{X,\varepsilon}(T)^2)$  converges to

$$L := \frac{1}{4\pi^2} \int_D \int_{\mathbb{R}^2} \exp\left(-\frac{1}{2} \text{var}\left(\sum_{j=1}^2 \xi_j(X_{t_j} - c)\right)\right) \exp\left(-ic \sum_{j=1}^2 \xi_j\right) d\xi dt$$

as  $\varepsilon$  tends to zero, provided

$$l := \int_D \int_{\mathbb{R}^2} \exp\left(-\frac{1}{2} \text{var}\left(\sum_{j=1}^2 \xi_j(X_{t_j} - c)\right)\right) d\xi dt < \infty.$$

We also consider

$$\begin{aligned} & \mathbb{E}(\mathcal{L}_{X,\varepsilon}(T)\mathcal{L}_{X,\delta}(T)) \\ &= \frac{1}{4\pi^2} \int_D \int_{\mathbb{R}^2} \mathbb{E}\left(\exp\left(i \sum_{j=1}^2 \xi_j(X_{t_j} - c)\right)\right) \exp\left(-\frac{\varepsilon}{2}\xi_1^2 - \frac{\delta}{2}\xi_2^2\right) d\xi dt. \end{aligned}$$

If  $l < \infty$ , then we also have that  $\lim_{\varepsilon, \delta \rightarrow 0} \mathbb{E}(\mathcal{L}_{X,\varepsilon}(T)\mathcal{L}_{X,\delta}(T)) = L$ . This implies that  $\mathcal{L}_{X,\varepsilon}(T)$ ,  $\varepsilon > 0$  is Cauchy in  $L^2(\mathbb{P})$ , that is  $\mathbb{E}((\mathcal{L}_{X,\varepsilon}(T) - \mathcal{L}_{X,\delta}(T))^2) = \mathbb{E}(\mathcal{L}_{X,\varepsilon}(T)^2) + \mathbb{E}(\mathcal{L}_{X,\delta}(T)^2) - 2\mathbb{E}(\mathcal{L}_{X,\varepsilon}(T)\mathcal{L}_{X,\delta}(T))$  converges to 0 as  $\varepsilon, \delta \rightarrow 0$ . As a consequence,  $\mathcal{L}_{X,\varepsilon}(T)$  converges in  $L^2(\mathbb{P})$  as  $\varepsilon$  tends to zero. Therefore, if we can show that  $l < \infty$ , the proof is finished. Indeed,

$$l = \int_0^T \int_0^{t_2} \int_{\mathbb{R}^2} \exp\left(-\frac{1}{2} \text{var}\left(\sum_{j=1}^2 \xi_j(X_{t_j} - c)\right)\right) d\xi dt_1 dt_2$$

$$\begin{aligned}
&= 2\pi \int_0^T \int_0^{t_2} \frac{1}{\sqrt{\text{var}(X_{t_1})\text{var}(X_{t_2}) - (\text{cov}(X_{t_1}, X_{t_2}))^2}} dt_1 dt_2 \\
&= 2\pi \int_0^T \int_0^{t_2} \frac{1}{\sqrt{\int_0^{t_1} f(u)^2 du \int_{t_1}^{t_2} f(u)^2 du}} dt_1 dt_2 \\
&\leq \frac{2\pi}{\alpha^2} \int_0^T \int_0^{t_2} \frac{1}{\sqrt{t_1(t_2 - t_1)}} dt_1 dt_2 \\
&< \infty,
\end{aligned}$$

where  $\text{cov}(X, Y)$  denotes the covariance between random variables  $X$  and  $Y$ .  $\square$

Up to this point we are able to give a meaning to the local time  $\mathcal{L}_X(T)$  as a square-integrable function with respect to the probability space on which the one-dimensional b-Gaussian process is defined. Below we establish a mathematically rigorous meaning to the random variable  $\mathcal{L}_X(T)$  for any  $d$ -dimensional b-Gaussian process,  $d \in \mathbb{N}$ . This can be done using the theory of white noise analysis. For this purpose we consider the *Donsker delta function* of b-Gaussian process which is defined as the formal composition of the Dirac-delta function  $\delta_d \in \mathcal{S}'(\mathbb{R}^d)$  with a  $d$ -dimensional b-Gaussian process  $(X_t)_{t \in [0, T]}$ , i.e.,  $\delta_d(X_t - c)$ , with  $c \in \mathbb{R}^d$ . We can give a precise meaning to the Donsker delta function as a Hida distribution.

**3.3. Proposition.** *Let  $X = (X_t)_{t \in [0, T]}$  be a  $d$ -dimensional b-Gaussian process and  $c \in \mathbb{R}^d$ . The Bochner integral*

$$\delta_d(X_t - c) := \left(\frac{1}{2\pi}\right)^d \int_{\mathbb{R}^d} \exp(ix(X_t - c)) d\lambda_d(x),$$

is a Hida distribution with  $S$ -transform  $S(\delta_d(X_t - c))(\vec{f})$  given by

$$\left(\frac{1}{2\pi \int_0^t f(u)^2 du}\right)^{d/2} \exp\left(-\frac{1}{2 \int_0^t f(u)^2 du} \sum_{j=1}^d \left(\int_0^t f_j(u) f(u) du - c_j\right)^2\right),$$

for all  $\vec{f} = (f_1, \dots, f_d) \in \mathcal{S}_d(\mathbb{R})$ .

*Proof.* By direct computation we have

$$\begin{aligned}
&S(\exp(ix(X_t - c)))(\vec{f}) \\
&= \left\langle \left\langle \exp(ix(\langle \cdot, \mathbf{1}_{[0, t]} f \rangle - c)), \cdot \right\rangle, \exp(\langle \cdot, \vec{f} \rangle) \right\rangle \\
&= \exp\left(-\frac{1}{2} |\vec{f}|_0^2\right) \exp(-ixc) \int_{\mathcal{S}'_d(\mathbb{R})} \exp(\langle \vec{\omega}, ix\mathbf{1}_{[0, t]} f + \vec{f} \rangle) d\mu(\vec{\omega}) \\
&= \exp\left(-\frac{1}{2} |\vec{f}|_0^2\right) \exp(-ixc) \exp\left(\frac{1}{2} |\vec{f} + ix\mathbf{1}_{[0, t]} f|_0^2\right) \\
&= \exp\left(-\frac{1}{2} |x|^2 \int_0^t f(u)^2 du\right) \exp\left(ix(\langle \vec{f}, \mathbf{1}_{[0, t]} f \rangle - c)\right),
\end{aligned}$$

which is a measurable function of  $x \in \mathbb{R}^d$  for each  $\vec{f} \in \mathcal{S}_d(\mathbb{R})$ . Furthermore, let  $z \in \mathbb{C}$  and  $\vec{f} \in \mathcal{S}_d(\mathbb{R})$ . Then

$$\begin{aligned}
&\left|S(\exp(ix(X_t - c)))(z\vec{f})\right| \\
&\leq \exp\left(-\frac{1}{2} |x|^2 \int_0^t f(u)^2 du\right) \exp(|x||z| |\langle \vec{f}, \mathbf{1}_{[0, t]} f \rangle|)
\end{aligned}$$

$$\begin{aligned} &\leq \exp\left(-\frac{1}{2}|x|^2 \int_0^t f(u)^2 du\right) \exp\left(|x||z|\beta t \sum_{j=1}^d \sup_{u \in \mathbb{R}} |f_j(u)|\right) \\ &\leq \exp\left(-\frac{1}{4}|x|^2 \int_0^t f(u)^2 du\right) \exp\left(\frac{\beta^2 t^2}{\int_0^t f(u)^2 du} |z|^2 \|\vec{f}\|_{\infty,1}^2\right) \\ &\leq \exp\left(-\frac{1}{4}|x|^2 \alpha^2 t\right) \exp\left(\frac{\beta^2}{\alpha^2} T |z|^2 \|\vec{f}\|_{\infty,1}^2\right), \end{aligned}$$

where  $\|\cdot\|_{\infty,1}$  is a continuous norm on  $\mathcal{S}_d(\mathbb{R})$  defined by

$$\|\vec{f}\|_{\infty,1} := \sum_{j=1}^d \sup_{u \in \mathbb{R}} |f_j(u)|,$$

and for some positive constants  $\alpha$  and  $\beta$ . The first factor is an integrable function of  $\lambda_d$ , and the second factor is constant. Hence, according to the Corollary 2.2  $\delta_d(X_t - c) \in (S)^*$ . We may now interchange the S-transform and integration to obtain

$$\begin{aligned} &S(\delta_d(X_t - c))(\vec{f}) \\ &= \left(\frac{1}{2\pi}\right)^d \int_{\mathbb{R}^d} S(\exp(ix(X_t - c))) (\vec{f}) d\lambda_d(x) \\ &= \left(\frac{1}{2\pi}\right)^d \int_{\mathbb{R}^d} \exp\left(-\frac{1}{2}|x|^2 \int_0^t f(u)^2 du\right) \\ &\quad \times \exp\left(ix \left(\langle \vec{f}, \mathbf{1}_{[0,t]} f \rangle - c\right)\right) d\lambda_d(x) \\ &= \left(\frac{1}{2\pi}\right)^d \left(\frac{2\pi}{\int_0^t f(u)^2 du}\right)^{d/2} \prod_{j=1}^d \exp\left(\frac{\left(i \left(\int_0^t f_j(u) f(u) du - c_j\right)\right)^2}{2 \int_0^t f(u)^2 du}\right) \\ &= \left(\frac{1}{2\pi \int_0^t f(u)^2 du}\right)^{d/2} \\ &\quad \times \exp\left(-\frac{1}{2 \int_0^t f(u)^2 du} \sum_{j=1}^d \left(\int_0^t f_j(u) f(u) du - c_j\right)^2\right). \quad \square \end{aligned}$$

In the following, in order to simplify the notation, we denote by  $p(f)$  the prefactor  $\left(\frac{1}{2\pi \int_0^t f(u)^2 du}\right)^{d/2}$ . We are now in the position to prove our main results on local times  $\mathcal{L}_X(T)$  and their subtracted counterparts  $\mathcal{L}_X^{(N)}(T)$ . First, we introduce the notion of the truncated Donsker's delta function which is well-defined due to Theorem 2.1.

**3.4. Definition.** The truncated Donsker delta function  $\delta_d^{(N)}(X_t - c)$  is defined as the Hida distribution such that for every  $\vec{f} \in \mathcal{S}_d(\mathbb{R})$  its S-transform is given by

$$\begin{aligned} &S(\delta_d^{(N)}(X_t - c))(\vec{f}) \\ &= p(f) \exp^{(N)}\left(-\frac{1}{2 \int_0^t f(u)^2 du} \sum_{j=1}^d \left(\int_0^t f_j(u) f(u) du - c_j\right)^2\right), \end{aligned}$$

where the truncated exponential series  $\exp^{(N)}$  is given by

$$\exp^{(N)}(x) := \sum_{m=N}^{\infty} \frac{x^m}{m!}.$$

**3.5. Theorem.** Let  $X = (X_t)_{t \in [0, T]}$  be a  $d$ -dimensional b-Gaussian process and  $c \in \mathbb{R}^d$ . For any pair of integers  $d \geq 1$  and  $N \geq 0$  such that  $2N > d - 2$ , the Bochner integral  $\mathcal{L}_X^{(N)}(T) := \int_0^T \delta_d^{(N)}(X_t - c) dt$  is a Hida distribution.

*Proof.* From the definition of truncated Donsker's delta function we see immediately that  $S\left(\delta_d^{(N)}(X_t - c)\right)(\vec{f})$  is a measurable function of  $t$  for every  $\vec{f} \in \mathcal{S}_d(\mathbb{R})$ . Furthermore, for every  $z \in \mathbb{C}$  and  $\vec{f} \in \mathcal{S}_d(\mathbb{R})$ , by using Proposition 3.3 it follows that

$$\begin{aligned} & \left| S\left(\delta_d^{(N)}(X_t - c)\right)(z\vec{f}) \right| \\ & \leq p(f) \exp^{(N)} \left( \frac{1}{2 \int_0^t f(u)^2 du} |Re(z^2)| \sum_{j=1}^d \left( \int_0^t f_j(u) f(u) du \right)^2 \right) \\ & \quad \times \exp^{(N)} \left( \frac{1}{\int_0^t f(u)^2 du} \sum_{j=1}^d |c_j| |Re(z)| \left| \int_0^t f_j(u) f(u) du \right| \right) \\ & \quad \times \exp^{(N)} \left( -\frac{1}{2 \int_0^t f(u)^2 du} \sum_{j=1}^d c_j^2 \right) \\ & \leq p(f) \exp^{(N)} \left( \frac{1}{2 \int_0^t f(u)^2 du} |Re(z^2)| \sum_{j=1}^d \left( \int_0^t f_j(u) f(u) du \right)^2 \right) \\ & \quad \times \exp^{(N)} \left( \frac{1}{2 \int_0^t f(u)^2 du} Re(z)^2 \sum_{j=1}^d \left( \int_0^t f_j(u) f(u) du \right)^2 \right) \\ & \leq p(f) \exp^{(N)} \left( \frac{1}{\int_0^t f(u)^2 du} |z|^2 \sum_{j=1}^d \left( \int_0^t f_j(u) f(u) du \right)^2 \right) \\ & \leq \left( \frac{1}{2\pi\alpha^2 t} \right)^{d/2} \exp^{(N)} \left( \frac{\beta^2}{\alpha^2} t |z|^2 \|\vec{f}\|_{\infty, 2}^2 \right) \\ & \leq \left( \frac{1}{2\pi\alpha^2} \right)^{d/2} \left( \frac{1}{T} \right)^N t^{N-d/2} \exp \left( \frac{\beta^2 T}{\alpha^2} |z|^2 \|\vec{f}\|_{\infty, 2}^2 \right), \end{aligned}$$

where  $\|\vec{f}\|_{\infty, 2}^2 := \sum_{j=1}^d (\sup_{u \in \mathbb{R}} |f_j(u)|)^2$  is a continuous norm on  $\mathcal{S}_d(\mathbb{R})$ . Note that  $t^{N-d/2}$  is integrable with respect to the Lebesgue measure on  $[0, T]$  if and only if  $N - d/2 > -1$ . Therefore we can conclude using Corollary 2.2 that  $\mathcal{L}_X^{(N)}(T) \in (\mathcal{S})^*$ .  $\square$

Theorem 3.5 asserts that for one-dimensional b-Gaussian process all local times  $\mathcal{L}_X^{(N)}(T)$  are well-defined as Hida distributions. This fact is not really surprising since we have already known from Theorem 3.2 that  $\mathcal{L}_X(T) \in L^2(\mathbb{P})$ , and in particular,  $\mathcal{L}_X(T) \in L^2(\mu)$ . We should emphasize the result for higher dimension. For  $d \geq 2$ , local times only become well-defined after omission of the divergent terms which occur in the low order terms in the truncated Donsker delta function. For example, for  $d = 2$  and  $d = 3$  it is sufficient to take  $N = 1$ , which means we only need to throw away the first lower term to have  $\mathcal{L}_X(T)$  as a member of  $(\mathcal{S})^*$ . As an immediate consequence of Theorem 3.5 we can also compute the expectation of local times  $\mathcal{L}_X^{(N)}(T)$ , that is  $\mathbb{E}_\mu(\mathcal{L}_X^{(N)}(T)) = \int_0^T p(f) \exp^{(N)} \left( -\frac{|c|^2}{2 \int_0^t f(u)^2 du} \right) dt$ . It is also clear that the expectation is finite only in dimension one, and for higher dimension ( $d \geq 2$ ) the expectation blows up.

The renormalization procedure, i.e. dropping the divergent terms, as described in Theorem 3.5 above, motivates the study of a regularization. We define the regularized



Donsker's delta function of b-Gaussian process as

$$\delta_{d,\varepsilon}(X_t - c) := \left(\frac{1}{2\pi\varepsilon}\right)^{d/2} \exp\left(-\frac{|X_t - c|^2}{2\varepsilon}\right)$$

and the corresponding regularized local time of b-Gaussian process  $\mathcal{L}_{X,\varepsilon}^d(T) := \int_0^T \delta_{d,\varepsilon}(X_t - c) dt$ .

**3.6. Theorem.** *Let  $X = (X_t)_{t \in [0,T]}$  be a  $d$ -dimensional b-Gaussian process and  $c \in \mathbb{R}^d$ . For all  $\varepsilon > 0$  and  $d \geq 1$  the regularized local time  $\mathcal{L}_{X,\varepsilon}^d(T)$  is a Hida distribution. Moreover, if  $2N > d - 2$ , then the (truncated) regularized local times  $\mathcal{L}_{X,\varepsilon}^{(N)}(T) := \int_0^T \delta_{d,\varepsilon}^{(N)}(X_t - c) dt$  converges strongly as  $\varepsilon \rightarrow 0$  in  $(\mathcal{S})^*$  to the (truncated) local times  $\mathcal{L}_X^{(N)}(T)$ .*

*Proof.* The first part of the proof follows again by an application of Corollary 2.2 with respect to the Lebesgue measure on  $[0, T]$ . For all  $\vec{f} \in \mathcal{S}_d(\mathbb{R})$  we obtain

$$\begin{aligned} & S(\delta_{d,\varepsilon}(X_t - c))(\vec{f}) \\ &= \left(\frac{1}{2\pi(\varepsilon + \int_0^t f(u)^2 du)}\right)^{d/2} \\ &\quad \times \exp\left(-\frac{1}{2(\varepsilon + \int_0^t f(u)^2 du)} \sum_{j=1}^d \left(\int_s^t f_j(u)f(u) du - c_j\right)^2\right), \end{aligned}$$

which is evidently measurable. Hence for all  $z \in \mathbb{C}$  we have

$$\begin{aligned} & \left| S(\delta_{d,\varepsilon}(X_t - c))(z\vec{f}) \right| \\ & \leq \left(\frac{1}{2\pi(\varepsilon + \int_0^t f(u)^2 du)}\right)^{d/2} \exp\left(\frac{\beta^2}{2(\varepsilon + \int_0^t f(u)^2 du)} t^2 |z|^2 \|\vec{f}\|_{\infty,2}^2\right). \end{aligned}$$

We observe that  $\frac{t^2}{\varepsilon + \int_0^t f(u)^2 du}$  is bounded on  $[0, T]$  and  $\left(\frac{1}{2\pi(\varepsilon + \int_0^t f(u)^2 du)}\right)^{d/2}$  is integrable on  $[0, T]$ . By Corollary 2.2 we may then conclude that  $\mathcal{L}_{X,\varepsilon}^d(T) \in (\mathcal{S})^*$ , for every  $\varepsilon > 0$  and  $d \geq 1$ . Now we have to verify the convergence of  $\mathcal{L}_{X,\varepsilon}^{(N)}(T)$  as  $\varepsilon \rightarrow 0$ . To this end we shall use Corollary 2.3. Since for every  $\vec{f} \in \mathcal{S}_d(\mathbb{R})$

$$S(\mathcal{L}_{X,\varepsilon}^{(N)}(T))(\vec{f}) = \int_0^T S(\delta_{d,\varepsilon}^{(N)}(X_t - c))(\vec{f}) dt,$$

then for all  $z \in \mathbb{C}$  we have, by using similar computations as in the proof of Theorem 3.5,

$$\begin{aligned} & \left| S(\mathcal{L}_{X,\varepsilon}^{(N)}(T))(z\vec{f}) \right| \\ & \leq \int_0^T \left| S(\delta_{d,\varepsilon}^{(N)}(X_t - c))(z\vec{f}) \right| dt \\ & \leq \int_0^T \left(\frac{1}{2\pi \int_0^t f(u)^2 du}\right)^{d/2} \exp^{(N)}\left(\frac{\beta^2}{2 \int_0^t f(u)^2 du} t^2 |z|^2 \|\vec{f}\|_{\infty,2}^2\right) dt \\ & \leq \left(\frac{1}{2\pi\alpha^2}\right)^{d/2} \left(\frac{1}{T}\right)^N \left(\int_0^T t^{N-d/2} dt\right) \exp\left(\frac{\beta^2 T}{\alpha^2} |z|^2 \|\vec{f}\|_{\infty,2}^2\right). \end{aligned}$$

This shows the uniform boundedness condition. In particular, we have

$$\left| S \left( \delta_{d,\varepsilon}^{(N)} (X_t - c) \right) (\bar{f}) \right| \leq \left( \frac{1}{2\pi\alpha^2} \right)^{d/2} \left( \frac{1}{T} \right)^N t^{N-d/2} \exp \left( \frac{\beta^2 T}{\alpha^2} \|\bar{f}\|_{\infty,2}^2 \right).$$

The latter upper bound is an integrable function on  $[0, T]$ . Finally, Lebesgue's dominated convergence theorem and Corollary 2.3 deliver the assertion of the theorem.  $\square$

#### 4. Conclusion

In this article, we have showed the existence of the local times of a Gaussian process  $X$  defined by an indefinite Wiener integral. In dimension one the local times of  $X$  exist as square integrable functions. In higher dimensions, by using a white noise analysis method, we proved that the renormalized local times of  $X$  are Hida distributions. Furthermore, a convergence result to the renormalized local times was also established. We observe that renormalization method (removal of divergent terms) works since in the Wiener-Itô chaos decomposition the kernel functions of increasing order are less and less singular in the  $L^1$ -sense.

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