

The Renormalization of Self Intersection Local Times of Brownian Motion

Anis Rezgui

Universität Bielefeld, ZiF, Germany.

Faculté des sciences de Bizerte, Tunisia.

Ludwig Streit*

CCM, Universidade da Madeira, PT9000, Funchal.

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1 Introduction and statement of the results

Self intersection local time of Brownian motion have been the subject of numerous studies of half a century. This was, in part, motivated by the important role played by self intersection local time in the construction of certain Euclidean quantum fields, see for instance [8][22][23][27]. The expression of self intersection local time appears also in the study of polymers, see Edwards [10]. Using a chaos expansion in terms of Malliavin's calculus or, even more appropriately white noise calculus, many results were proven see for instance [4][11][14]. More precisely it was shown in [24] that the self-intersection local time of d -dimensional Brownian motion, renormalized by subtracting some terms in its chaos expansion, was a Hida distribution, we also point out [13][19].

*supported by FCT, Portugal.

An informal definition of the self intersection local time of the Brownian motion is

$$L = \int d^2t \delta(B(t_2) - B(t_1)),$$

where B is a d -dimensional Brownian motion. This could be understood as the amount of time, spent by the Brownian path intersecting itself during the time t . To make sense of this integral one can invoke an approximation

$$L^\epsilon(t) = \int_0^t dt_2 \int_0^{t_2} dt_1 \delta_\epsilon(B(t_2) - B(t_1)),$$

with, for x in \mathbb{R}^d

$$\delta_\epsilon(x) = \frac{1}{(2\pi\epsilon)^{d/2}} e^{-\frac{x^2}{2\epsilon}}.$$

The central problem is then the control of what happens when $\epsilon \rightarrow 0$. As is known, for $d \geq 2$, we need to center L_ϵ by

$$L^\epsilon - \mathbb{E}(L^\epsilon)$$

see [17][20][23].

A Tanaka formula method was initiated in [21], and used in [28] for $d=3$ with a regularization of type " $\delta_y(x) = \delta(x + y)$ " to show that

$$r_3(|y|)(L_y - \mathbb{E}(L_y)) \xrightarrow[y \rightarrow 0]{\mathcal{L}} c\beta$$

where β is a one dimensional Brownian motion, independent of the initial one and $r_3(|y|) = |\log |y||^{-1/2}$.

Using an explicit computation of the chaos of L_T^ϵ given in [11], it has been shown in [4] and for $d > 2$, that each chaos of the renormalized self-intersection local time, converges in law to a one dimensional Brownian motion, independent from the initial one.

Our aim in this work is to prove a similar result for the whole renormalized self-intersection local time. Again, using the explicit computation of the chaos of L_T^ϵ , we decompose the self-intersection local time into a martingale part M^ϵ and a negligible one N^ϵ , following in this what has been done in [4].

We prove that for $d > 2$ the renormalized martingale part $r_d(\epsilon)M^\epsilon$ converges in law to a one dimensional Brownian motion, independent of the initial one and that $r_d(\epsilon)N^\epsilon$ converges to zero in mean square.

This entails that the renormalized self-intersection local time $r_d(\epsilon)(L^\epsilon - \mathbb{E}(L^\epsilon))$ converges in finite dimensional distribution to a Brownian motion independent of the initial one.

Before announcing our results, let us give some notations and recall some results.

In [11], the kernels of the chaos expansion of the renormalized self-intersection local time were explicitly computed.

Let $\epsilon > 0$, $\vec{n} = (n_1, \dots, n_d) \in \mathbb{N}^d - \{0\}$, $n = \sum_i n_i$.

The kernels of $L_\epsilon - \mathbb{E}(L_\epsilon)$ in Fock space are, for any $d > 2$:

$$F_{\epsilon, \vec{n}}(s_1, \dots, s_n) = (-1)^{n/2} \left(\chi(\chi + 1)(2\pi)^{d/2} 2^{n/2} \frac{\vec{n}!}{2} \right)^{-1} \\ \theta(u)\theta(t-v) \left((v-u+\epsilon)^{-\chi} + (t+\epsilon)^{-\chi} - (v+\epsilon)^{-\chi} - (t-u+\epsilon)^{-\chi} \right)$$

if all n_i are even (and are zero otherwise) with $v = \max_i s_i$, $u = \min_i s_i$ and $\chi = \frac{n+d}{2} - 2$; θ is the Heaviside function.

Let us set up some notations, for $\vec{n} \in \mathbb{N}^d - \{0\}$ and $\epsilon > 0$

$$M_{\vec{n}}^\epsilon = \int_{[0,t]^n} d^n s (v-u+\epsilon)^{-\chi} : w^{\otimes \vec{n}}(s) : \quad (1)$$

$$N_{\vec{n}}^\epsilon = \int_{[0,t]^n} d^n s \left((t+\epsilon)^{-\chi} - (v+\epsilon)^{-\chi} - (t-u+\epsilon)^{-\chi} \right) : w^{\otimes \vec{n}}(s) : . \quad (2)$$

Note that for each \vec{n} and ϵ the above processes are continuous in time.

For the \vec{n}^{th} order chaos of the renormalized self-intersection local time we set

$$K_{\vec{n}}^\epsilon = \alpha_{\vec{n}} (M_{\vec{n}}^\epsilon + N_{\vec{n}}^\epsilon) .$$

It was shown in [4], that for $d \geq 3$ and $\vec{n} \in \mathbb{N}^d - \{0\}$,

$$r_d(\epsilon) K_{\vec{n}}^\epsilon \xrightarrow[\epsilon \rightarrow 0^+]{\mathcal{L}} c_{\vec{n}} \beta_{\vec{n}} \quad (3)$$

where for $\vec{n} \neq 0$, $\beta_{\vec{n}}$ is a one dimensional Brownian motion independent of the initial one and among each other, with

$$c_{\vec{n}}^2 = k_n^2 \alpha_{\vec{n}}^2 ,$$

$$k_n^2 = \begin{cases} n(n-1) & d = 3 \\ \frac{n!(d-4)!}{(n+d-5)!} & d > 3, \end{cases}$$

$$\alpha_{\vec{n}} = (-1)^{n/2} \left(\chi(\chi+1)(2\pi)^{d/2} 2^{n/2} \frac{\vec{n}!}{2} \right)^{-1},$$

$$r_d(\epsilon) = \begin{cases} |\log \epsilon|^{-1/2} & d = 3 \\ \epsilon^{\frac{d-3}{2}} & d > 3. \end{cases}$$

Let us denote the dominant (martingale) part of the renormalized self-intersection local time by

$$M^\epsilon = \sum_{\vec{n} \neq 0} \alpha_{\vec{n}} M_{\vec{n}}^\epsilon.$$

Our main result is as follows

Theorem 1 For $d \geq 3$,

$$r_d(\epsilon) M^\epsilon \xrightarrow[\epsilon \rightarrow 0^+]{\mathcal{L}} c\beta, \quad (4)$$

where β is a one dimensional standard Brownian motion independent of the initial one and $c = \sqrt{\sum_{\vec{n} \neq 0} c_{\vec{n}}^2}$.

Remark

It was shown in [4], by explicit computation, that for all $\vec{n} \neq 0$ and uniformly in finite t-intervals that

$$r_d(\epsilon) N_{t, \vec{n}}^\epsilon \xrightarrow{(L^2)} 0.$$

Considering the summed up process

$$N_t^\epsilon = \sum_{\vec{n}, \vec{n} \neq 0} \alpha_{\vec{n}} N_{t, \vec{n}}^\epsilon,$$

we prove

Proposition 1 For every $t \geq 0$ and $d > 2$

$$r_d(\epsilon) N_t^\epsilon \xrightarrow{(L^2)} 0,$$

and the convergence is uniform in any finite t-interval.

We set

$$\begin{aligned} K^\epsilon &= L^\epsilon - \mathbb{E}(L^\epsilon) \\ &= M^\epsilon + N^\epsilon. \end{aligned}$$

Then using the theorem 1 and the proposition 1 we obtain

Theorem 2 For $d \geq 3$,

$$r_d(\epsilon)K^\epsilon \xrightarrow{\mathcal{L}_f} c\beta,$$

where \mathcal{L}_f means convergence of finite dimensional distributions.

2 Tools from white noise analysis

We quote some white noise analysis concepts as introduced in [4], referring to [12] for a systematic presentation.

Consider a white noise space $(S'(\mathbb{R})^d, \mathcal{B}, \mu)$, where \mathcal{B} is the weak Borel σ -algebra of $S'(\mathbb{R})^d$, and μ is the centered Gaussian measure with covariance given by the inner product of $L^2(\mathbb{R})^d$.

Then a realization of a vector of independent Brownian motions $B_i, i = 1, \dots, d$, is obtained by

$$B_i(t) = \langle \omega_i, \mathbf{1}_{[0,t]} \rangle = \int_0^t \omega_i(s) ds.$$

Hence we consider independent d-tuples of Gaussian white noise $\omega = (w_1, \dots, w_d)$ and correspondingly, d-tuples of test functions $\mathbf{f} = (f_1, \dots, f_d) \in S(\mathbb{R}, \mathbb{R}^d)$, and use the following multi-index notation:

$$\vec{n}! = \prod_1^d n_i! \quad n = \sum_{i=1}^d n_i$$

$$\langle \mathbf{f}, \mathbf{f} \rangle = \sum_{i=1}^d \int dt f_i^2(t)$$

$$\langle F_{\vec{n}}, \mathbf{f}^{\otimes \vec{n}} \rangle = \int d^n t F_{\vec{n}}(t_1, \dots, t_n) \bigotimes_{i=1}^d f_i^{\otimes n_i}(t_1, \dots, t_n)$$

and similarly for $\langle : \omega^{\otimes \vec{n}} :, F_{\vec{n}} \rangle$ where for d-tuples of white noise the usual Wick product $: \cdots :$ (see [12]) generalises to

$$: \omega^{\otimes \vec{n}} := \bigotimes_{i=1}^d : \omega_i^{\otimes n_i} : .$$

The vector valued white noise has the characteristic function

$$C(\mathbf{f}) = \mathbb{E}(e^{i\langle \omega, \mathbf{f} \rangle}) = \int_{S^*(\mathbb{R}, \mathbb{R}^d)} d\mu[\omega] e^{i\langle \omega, \mathbf{f} \rangle} = e^{-\frac{1}{2}\langle \mathbf{f}, \mathbf{f} \rangle},$$

where $\langle \omega, \mathbf{f} \rangle = \sum_{i=1}^d \langle \omega_i, f_i \rangle$ and $f_i \in S(\mathbb{R}, \mathbb{R})$.

The Hilbert space

$$(L^2) = L^2(d\mu)$$

is canonically isomorphic to the d-fold tensor product of Fock spaces of symmetric square integrable functions:

$$(L^2) \simeq \left(\bigoplus_{k=0}^{\infty} \text{Sym}L^2(\mathbb{R}^k, k!d^k t) \right)^{\otimes d} := \mathcal{F}$$

for a general element of (L^2) this implies the chaos expansion

$$\varphi(\omega) = \sum_{\vec{n}=0}^{\infty} \langle : \omega^{\otimes \vec{n}} :, F_{\vec{n}} \rangle,$$

the norm of φ is given by

$$\|\varphi\|_{(L^2)}^2 = \sum_{\vec{n}} \vec{n}! |F_{\vec{n}}|_{2,n}^2$$

with kernel functions F in \mathcal{F} .

3 Proof of the results

Our strategy to prove the theorem 1 is a classical one. In fact we first prove the tightness of the family of processes $\{r_d(\epsilon)M^\epsilon\}_{\epsilon>0}$, and then we prove the appropriate convergence of the associated sequence of finite dimensional distributions.

3.1 Tightness of $\{r_d(\epsilon)M^\epsilon\}_{\epsilon>0}$

We recall that to prove the tightness of $\{M_{\vec{n}}^\epsilon; \epsilon > 0\}$ for each $\vec{n} \neq 0$, the authors in [4] have used the hypercontractivity of the Ornstein-Uhlenbeck semigroup.

They obtained for $0 \leq t, s \leq T$ and $\alpha > 2$ the following inequality

$$\mathbb{E} [r_d(\epsilon) |M_{\vec{n}}^\epsilon(t) - M_{\vec{n}}^\epsilon(s)|]^\alpha \leq C_{T, \vec{n}} |t - s|^{\alpha/2}$$

which essentially implies tightness [15].

Such hypercontractivity argument cannot readily be extended to the process $\{r_d(\epsilon)M_T^\epsilon; \epsilon > 0\}$. To circumvent this difficulty, we prove a technical lemma.

Lemma 1 *Let $(X^n)_{n \in \mathbb{N}}$, $(X_k^n)_{n, k \in \mathbb{N}}$ be two families of processes with continuous paths and starting from the origin. Suppose that :*

i) For each $k \in \mathbb{N}$, $(X_k^n)_{n \in \mathbb{N}}$ is tight.

ii) For each $T > 0$ and $\epsilon > 0$,

$$\sup_{n \in \mathbb{N}} \mathbb{P} \left(\max_{0 \leq t \leq T} |X_k^n(t) - X^n(t)| > \epsilon \right) \xrightarrow[k \rightarrow +\infty]{} 0, \quad (5)$$

then the family $(X^n)_{n \in \mathbb{N}}$ is tight.

Proof of lemma 1

Following [2][15], we have to prove for each $\epsilon > 0$ and $T > 0$, that

$$\sup_{n \in \mathbb{N}} \mathbb{P} \left(\max_{|t-s| \leq \delta, 0 \leq t, s \leq T} |X^n(t) - X^n(s)| > \epsilon \right) \xrightarrow[\delta \downarrow 0]{} 0.$$

Let $\epsilon > 0$, $T > 0$ and $n \in \mathbb{N}$

$$|X^n(t) - X^n(s)| \leq |X^n(t) - X_k^n(t)| + |X_k^n(t) - X_k^n(s)| + |X_k^n(s) - X^n(s)|$$

so

$$\begin{aligned} \mathbb{P} \left(\max_{|t-s| \leq \delta, 0 \leq t, s \leq T} |X^n(t) - X^n(s)| > \epsilon \right) &\leq \mathbb{P} (2I_{1,k} + I_{2,k} > \epsilon) \\ &\leq \mathbb{P} \left(I_{1,k} > \frac{\epsilon}{4} \right) + \mathbb{P} \left(I_{2,k} > \frac{\epsilon}{2} \right) \end{aligned}$$

where

$$I_{1,k} = \max_{0 \leq t \leq T} |X_k^n(t) - X^n(t)|$$

$$I_{2,k} = \max_{|t-s| \leq \delta, 0 \leq t, s \leq T} |X_k^n(t) - X_k^n(s)|$$

Let $\alpha > 0$. By the second hypothesis, there exists $k_0 \in \mathbb{N}$ such that

$$\sup_{n \in \mathbb{N}} \mathbb{P} \left(I_{1,k_0} > \frac{\epsilon}{4} \right) \leq \frac{\alpha}{2}$$

Choose $k = k_0$; because of the tightness of $(X_{k_0}^n)_{n \in \mathbb{N}}$ there exists $\delta_0 > 0$ such that for every $\delta > \delta_0$

$$\sup_{n \in \mathbb{N}} \mathbb{P} \left(I_{2,k_0} > \frac{\epsilon}{2} \right) < \frac{\alpha}{2},$$

so the lemma is proved.

Let us now prove the tightness of $(r_d(\epsilon)M^\epsilon)_{\epsilon > 0}$. For each $k \in \mathbb{N}^*$, denote by

$$M_k^\epsilon = \sum_{\bar{n} \neq 0, n \leq k} \alpha_{\bar{n}} M_{\bar{n}}^\epsilon$$

so that we have, for $t > 0$

$$\|M^\epsilon(t) - M_k^\epsilon(t)\|_2^2 = \sum_{\bar{n} \neq 0, n > k} \alpha_{\bar{n}}^2 \|M_{\bar{n}}^\epsilon(t)\|_2^2.$$

It was shown in [4] (proof of lemma 3.2), that

$$\|M_{\bar{n}}^\epsilon(t)\|_2^2 = \bar{n}! n(n-1) \epsilon^{3-d} \int_0^t dv \int_0^{v/\epsilon} dx \frac{x^{n-2}}{(x+1)^{n+d-4}}. \quad (6)$$

For $d=3$

$$r_3(\epsilon)^2 \|M_{\bar{n}}^\epsilon(t)\|_2^2 \leq \bar{n}! n(n-1) \left\{ \frac{t|\log \epsilon| + (t+\epsilon) \log(t+\epsilon) - t}{|\log \epsilon|} \right\}$$

the function of t and ϵ in the right hand side of the last inequality is uniformly bounded on bounded sets of $(t, \epsilon) \in \mathbb{R}_+^2$.

If $d > 3$

$$\int_0^{v/\epsilon} dx \frac{x^{n-2}}{(x+1)^{n+d-4}} \leq \int_0^{+\infty} dx \frac{x^{n-2}}{(x+1)^{n+d-4}} \leq \int_0^{+\infty} dx \frac{1}{(x+1)^{d-2}} = \text{const.}$$

so in view of (6), if $d > 3$, we obtain

$$r_d(\epsilon)^2 \|M_{\vec{n}}^\epsilon(t)\|_2^2 \leq \text{const} \vec{n}! n(n-1)t.$$

It was proved in [5] that

$$\sum_{\vec{n} \neq 0} \alpha_n^2 \vec{n}! n(n-1) < +\infty.$$

Then we have proved that for every $d \geq 3$,

$$r_d^2(\epsilon) \|M^\epsilon(t) - M_k^\epsilon(t)\|_2^2 \xrightarrow[k \rightarrow +\infty]{} 0 \quad (7)$$

uniformly in bounded sets of $(t, \epsilon) \in \mathbb{R}_+^2$.

On the other hand, using the same technique as in [4] it has been proved that $\{r_d(\epsilon)M_{\vec{n}}^\epsilon; n \leq k\}$ is also tight when $\epsilon \rightarrow 0$.

Hence, also for each $k \geq 1$ $r_d(\epsilon)M_k^\epsilon$ is tight when $\epsilon \rightarrow 0$.

Let $\eta > 0$ and $T > 0$ fixed,

$$\mathbb{P} \left(\max_{0 \leq t \leq T} (r_d(\epsilon) |M^\epsilon(t) - M_k^\epsilon(t)|) > \eta \right) \leq \frac{\mathbb{E} (r_d(\epsilon) |M^\epsilon(T) - M_k^\epsilon(T)|)}{\eta}$$

because $M^\epsilon(\cdot) - M_k^\epsilon(\cdot)$ is a martingale, see e.g. [15].

Using (7) we find (5) and we prove that $\{r_d(\epsilon)M^\epsilon; \epsilon > 0\}$ is tight.

3.2 Convergence of the finite dimensional distributions

We intend to prove that, for every $0 \leq t_1 < \dots < t_m$, $m \in \mathbb{N} - \{0\}$,

$$r_d(\epsilon) (M^\epsilon(t_1); \dots; M^\epsilon(t_m)) \xrightarrow[\epsilon \rightarrow 0]{\mathcal{L}} c(\beta(t_1); \dots; \beta(t_m)),$$

where β is a standard one dimensional Brownian motion, and $c^2 := \lim_{k \rightarrow \infty} c_k^2 = \sum_{\vec{n} \neq 0} c_{\vec{n}}^2$.

Using the same technique as in [4] one finds that for every $k \in \mathbb{N} - \{0\}$

$$r_d(\epsilon) M_k^\epsilon \xrightarrow[\epsilon \rightarrow 0]{\mathcal{L}} c_k \beta \quad (8)$$

where $c_k^2 = \sum_{\vec{n} \neq 0; n \leq k} c_{\vec{n}}^2$.

Then for every $k, m \in \mathbb{N}^*$ and $0 \leq t_1 < \dots < t_m$

$$r_d(\epsilon) (M_k^\epsilon(t_1); \dots; M_k^\epsilon(t_m)) \xrightarrow[\epsilon \rightarrow 0]{\mathcal{L}} c_k (\beta(t_1); \dots; \beta(t_m)),$$

Now we vectorize

$$\begin{aligned}\vec{M}_k^\epsilon &= r_d(\epsilon) (M_k^\epsilon(t_1); \dots; M_k^\epsilon(t_m)), \\ \vec{M}^\epsilon &= r_d(\epsilon) (M_{t_1}^\epsilon; \dots; M_{t_m}^\epsilon), \\ \vec{\beta} &= (\beta(t_1); \dots; \beta(t_m)).\end{aligned}$$

Let X, Y be two random vectors in \mathbb{R}^m (not necessarily defined on the same probability space), we define the following distance

$$d(X, Y) = \sup_{\|f\|_1 \leq 1} |\mathbb{E}(f(X)) - \mathbb{E}(f(Y))|$$

with $\|f\|_1 = \|f\|_\infty + \|f'\|_\infty$ for f in $C_b^1(\mathbb{R}^m)$. With this distance the topology of the convergence in law of random vectors on Euclidean space becomes metrisable, see [1]. Then we have

$$d(\vec{M}^\epsilon, c\vec{\beta}) \leq d(\vec{M}^\epsilon, \vec{M}_k^\epsilon) + d(\vec{M}_k^\epsilon, c_k\vec{\beta}) + d(c_k\vec{\beta}, c\vec{\beta}).$$

In view of (7) the first term of the rhs can be made arbitrarily small for k large enough, uniformly in $\epsilon \geq 0$. The second distance is small when $\epsilon \rightarrow 0$ because of (8), and finally the third becomes small when k is large by the definition of c .

Note that this argument does not require us to distinguish the β arising in (8) for different k since they all have the same law.

Proof of proposition 1

In view of (2) we have

$$\|N_{t, \vec{n}}^\epsilon\|_{L^2}^2 \leq 3\vec{n}! \left(\|(t+\epsilon)^{-x}\|_{L^2([0, t]^n)}^2 + \|(v+\epsilon)^{-x}\|_{L^2([0, t]^n)}^2 + \|(t-u+\epsilon)^{-x}\|_{L^2([0, t]^n)}^2 \right),$$

with elementary computations we find

$$r_d(\epsilon)^2 \|(v+\epsilon)^{-x}\|_{L^2([0, t]^n)}^2 \leq \begin{cases} \frac{t}{|\log \epsilon|} n & \text{for } d = 3 \\ \epsilon \log \frac{t+\epsilon}{\epsilon} n & \text{for } d = 4 \\ \frac{\epsilon}{d-4} n & \text{for } d > 4 \end{cases}$$

also

$$\|(t+\epsilon)^{-x}\|_{L^2([0, t]^n)}^2 \leq \|(t-u+\epsilon)^{-x}\|_{L^2([0, t]^n)}^2 = \|(v+\epsilon)^{-x}\|_{L^2([0, t]^n)}^2 = o(1)n$$

so we have the same estimation for all three terms. Thus for

$$\|r_d(\epsilon)N_t^\epsilon\|_{(L^2)}^2 = r_d(\epsilon)^2 \sum_{\vec{n} \neq 0} \alpha_{\vec{n}}^2 \|N_{t,\vec{n}}^\epsilon\|_{(L^2)}^2,$$

in view of the last estimation and for t in a finite interval, we have for every $d \geq 3$

$$\|r_d(\epsilon)N_t^\epsilon\|_{(L^2)}^2 \leq o(1) \sum_{\vec{n} \neq 0} \alpha_{\vec{n}}^2 \vec{n}! n,$$

the last series converges and so the proposition is proved.

Proof of theorem 2

Recall that $K^\epsilon = M^\epsilon + N^\epsilon$, so in view of theorem 4.1 in [2] we only need to show that for every $m \in \mathbb{N}^*$ and every $0 \leq t_1 < \dots < t_m$

$$r_d(\epsilon) (N^\epsilon(t_1); \dots; N^\epsilon(t_m)) \xrightarrow{\epsilon \rightarrow 0} 0$$

in probability. This follows immediately from proposition 1.

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