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Non-central limit theorem for the cubic variation of a class of
selfsimilar stochastic processes

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Abstract

By using multiple Wiener-Itô stochastic integrals, we study the cubic variation of a class of selfsimilar stochastic processes with stationary increments (the Rosenblatt process with selfsimilarity order $H \in (\frac{1}{2}, 1)$). This study is motivated by statistical purposes. We prove that this renormalized cubic variation satisfies a non-central limit theorem and its limit is (in the $L^2(\Omega)$ sense) still the Rosenblatt process.

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1 Introduction

The self-similarity property for a stochastic process means that scaling of time is equivalent to an appropriate scaling of space. That is, a process $(Y_t)_{t \geq 0}$ is selfsimilar of order $H > 0$ if for all $c > 0$ the processes $(Y_{ct})_{t \geq 0}$ and $(c^H Y_t)_{t \geq 0}$ have the same finite dimensional distributions. The selfsimilar processes are of interest for various applications, such as economics, internet traffic of hydrology. The fractional Brownian motion (fBm) is the usual candidate to model phenomena in which the selfsimilarity property can be observed from the empirical data. Recall that the fractional Brownian motion is a centered Gaussian process with covariance function $R^H(t, s) = \frac{1}{2}(t^{2H} + s^{2H} - |t - s|^{2H})$. The parameter $H \in (0, 1)$ characterizes almost all the important properties of the process. The fBm can be also defined as the only Gaussian process which is selfsimilar with stationary increments. In some models the gaussianity assumption
could be not plausible and in this case one needs to use a different self-similar process with stationary increments to model the phenomena. Natural candidates are the Hermite processes: these stochastic processes appear as limits in the so-called Non-Central Limit Theorem (see [3], [7], [18], [10]). In contrast with the classical Central Limit Theorem, the non-central limit theorem deals with sequences of dependent random variable whose renormalized sum converges in some situations to a non gaussian distribution. For a complete exposition of limit theorems in probability theory, we refer to [9] or [16]. Except the Gaussian character, these Hermite processes have the same property as the fBm with Hurst parameter $H > \frac{1}{2}$: self-similarity, stationarity of increments, Hölder continuous path, long-range dependence. While the fractional Brownian motion can be expressed as a Wiener integral with respect to the standard Wiener process, the Hermite process of order $q \geq 2$ is a $q$ iterated integral of a deterministic function with $q$ variables with respect to the Brownian motion. The Rosenblatt process is obtained in the particular case $q = 2$. It will be properly defined in Section 2. These processes have been recently studied by several authors (see [2], [4], [13], [11], [12], [19], [20]).

The Hurst parameter $H$ characterizes all the important properties of a Hermite process, as seen above. Therefore, estimating $H$ properly is of the utmost importance. Several statistics have been introduced to this end, such as wavelets, $k$-variations, variograms, maximum likelihood estimators, or spectral methods. Information on these various approaches can be found in the book of Beran [1].

One of the most popular methods to estimate the self-similarity order for stochastic process is based on the study of their variations. The $p$- variation of a process $(X_t)_{t \in [0,1]}$ is defined as the limit of the sequence (sometimes the absolute value of the increment is used in the definition)

$$V^{p,N}(X) = \frac{1}{N} \sum_{i=0}^{N-1} \left( \frac{X_{\frac{i+1}{N}} - X_{\frac{i}{N}}}{\sqrt{N}} \right)^p - 1.$$

There exists a direct connection between the behavior of the variations and the convergence of an estimator for the self-similarity order based of these variation (see [6], [20]); basically if there renormalized variation satisfies a central limit theorem then the estimator satisfies a central limit theorem and this fact is very useful for statistical aspects.

In a recent paper ([20]) the quadratic variation of the Rosenblatt process $(Z^{(H)}_t)_{t \in [0,1]}$ with self-similarity order $H \in (\frac{1}{2}, 1)$ has been studied. The following facts happen: the normalized sequence $N^{1-H} V^{2,N}(Z^{(H)})$ satisfies a non-central limit theorem, it converges in $L^2$ to the Rosenblatt random variable $Z^{(H)}$. From this, we can construct an estimator for $H$ whose behavior is still non-normal. This situation is somehow not good for statistical applications because one always prefers the estimators which are asymptotically normal. To have normal estimators we need to define some adjusted variations (as in [20]).

In the fractional Brownian motion case the well-known non-normality of the quadratic variation when $H \in (\frac{3}{4}, 1)$ can be avoided by using "longer filters" (that means, replacing the increments $X_{\frac{i+1}{N}} - X_{\frac{i}{N}}$ by $X_{\frac{i+1}{N}} - 2X_{\frac{i}{N}} + X_{\frac{i-1}{N}}$) or higher order variations (choosing a bigger $p$). In this work we will consider the second choice (the first choice has been treated in the paper [3]): we replace the quadratic variation by the cubic variation for the Rosenblatt processes to see what happens and if it is possible to find a Gaussian distribution as law of the renormalized cubic variation. In the fractional Brownian motion case, this has no sense because the third moment of a centered Gaussian random variable is zero. We use the Wiener chaos expansion for the statistics $V^{3,N}(Z^{(H)})$ and we will decompose it in several terms in
the Wiener chaos 2, 4 and 6. As in other cases ([20], [4]) the second chaos term is dominant and it has to be renormalized by $N^{1-H}$ to have a non-trivial limit. We note that the rate of convergence $N^{1-H}$ is the same as for quadratic variation, so there no gain for the speed and moreover the limit is again, modulo a constant, a Rosenblatt random variable with index $H$ (only the constant is changing). This property has been called in [4] the reproduction property of the Rosenblatt process because its variations generates again Rosenblatt random variable as limits. We conjecture that the same property holds true for the $p$ variations.

The organization of our paper is as follows. Section 2 contains the presentation of the basic tools that we will need throughout the paper: multiple Wiener-Itô integrals and their basic properties, the definition of the Rosenblatt process and its characteristics. In Section 3 we estimate the mean square of the cubic variation of the Rosenblatt process and we give its normalization and finally in Section 4 we prove a non-central limit theorem for the renormalized cubic variation.

## 2 Preliminaries

### 2.1 Multiple stochastic integrals

In this paragraph we describe the basic elements of calculus on Wiener chaos. Let $(W_t)_{t \in [0,1]}$ be a classical Wiener process on a standard Wiener space $(\Omega, \mathcal{F}, \mathbb{P})$. If $f \in L^2([0,1]^n)$ with $n \geq 1$ integer, we introduce the multiple Wiener-Itô integral of $f$ with respect to $W$. We refer to [14] for a detailed exposition of the construction and the properties of multiple Wiener-Itô integrals.

Let $f \in S_n$ be an elementary functions with $m$ variables that can be written as

$$f = \sum_{i_1, \ldots, i_m} c_{i_1, \ldots, i_m} 1_{A_{i_1} \times \ldots \times A_{i_m}},$$

where the coefficients satisfy $c_{i_1, \ldots, i_m} = 0$ if two indices $i_k$ and $i_l$ are equal and the sets $A_i \in \mathcal{B}([0,1])$ are disjoint. For a such step function $f$ we define

$$I_m(f) = \sum_{i_1, \ldots, i_m} c_{i_1, \ldots, i_m} W(A_{i_1}) \ldots W(A_{i_m})$$

where we put $W([a,b]) = W_b - W_a$. It can be seen that the application $I_n$ constructed above from $\mathcal{S}$ to $L^2(\Omega)$ is an isometry on $\mathcal{S}$, i.e.

$$\mathbb{E}[I_n(f)I_m(g)] = n!(f,g)_{L^2([0,1]^n)} \text{ if } m = n$$

and

$$\mathbb{E}[I_n(f)I_m(g)] = 0 \text{ if } m \neq n. \tag{2}$$

It also holds that

$$I_n(f) = I_n(\tilde{f})$$

where $\tilde{f}$ denotes the symmetrization of $f$ defined by $\tilde{f}(x_1, \ldots, x_n) = \frac{1}{n!} \sum_{\sigma \in S_n} f(x_{\sigma(1)}, \ldots, x_{\sigma(n)}).$

Since the set $S_n$ is dense in $L^2([0,1]^n)$ for every $n \geq 1$ the mapping $I_n$ can be extended to an isometry from $L^2([0,1]^n)$ to $L^2(\Omega)$ and the above properties hold true for this extension. Note also that $I_n$ can be viewed as an iterated stochastic integral

$$I_n(f) = n! \int_0^1 f(t_1, t_2) dW_t \ldots dW_t;$$
here the integrals are of Itô type; this formula is easy to show for elementary $f$’s, and follows for general $f \in L^2([0,1]^n)$ by a density argument.

The product for two multiple integrals can be expanded into a sum of multiple integrals (see [14]): if $f \in L^2([0,1]^n)$ and $g \in L^2([0,1]^m)$ are symmetric functions, then it holds that

$$I_n(f)I_m(g) = \sum_{l=0}^{m \wedge n} lC_m^l C_n^l I_{m+n-2l}(f \otimes_l g)$$  \hspace{1cm} (3)

where the contraction $f \otimes_l g$ belongs to $L^2([0,1]^{m+n-2l})$ for $l = 0, 1, \ldots, m \wedge n$ and it is given by

$$(f \otimes_l g)(s_1, \ldots, s_{n-l}, t_1, \ldots, t_{m-l}) = \int_{[0,1]^{l}} f(s_1, \ldots, s_{n-l}, u_1, \ldots, u_l)g(t_1, \ldots, t_{m-l}, u_1, \ldots, u_l)du_1 \ldots du_l.$$  \hspace{1cm} (4)

When $l = 0$, we will denote, throughout this paper, by $f \otimes g := f \otimes_0 g$.

2.2 The Rosenblatt process

The Rosenblatt process $(Z^{(H)}(t))_{t \in [0,1]}$ appears as a limit in the so-called Non Central Limit Theorem (see [7], [18], [10]). It is not a Gaussian process and can be defined through its representation as double iterated integral with respect to a standard Wiener process (see [19]). More precisely, the Rosenblatt process with self-similarity order $H \in (\frac{1}{2}, 1)$ is defined by

$$Z^{(H)}_t := \int_0^t \int_0^t L_t(y_1, y_2) dW_{y_1} dW_{y_2}$$  \hspace{1cm} (5)

where $(W_t, t \in [0,1])$ is a Brownian motion,

$$L^H_t(y_1, y_2) := L_t(y_1, y_2) = d(H)1_{[0,1]}(y_1)1_{[0,1]}(y_2) \int_{y_1 \vee y_2}^t \frac{\partial K^H}{\partial u}(u, y_1) \frac{\partial K^H}{\partial u}(u, y_2) du,$$  \hspace{1cm} (6)

with

$$H' := \frac{H+1}{2} \quad \text{and} \quad d(H) := \frac{1}{H+1} \left( \frac{H}{2(2H-1)} \right)^{-\frac{1}{2}}.$$  \hspace{1cm}

and with $K^H$ the standard kernel defined in (7) appearing in the Wiener integral representation of the fBm (for $t > s$ and $H > \frac{1}{2}$)

$$K^H(t,s) := c_H s^{\frac{1}{2}-H} \int_s^t (u-s)^{H-\frac{3}{2}} u^{H-\frac{3}{2}} du$$  \hspace{1cm} (7)

with $c_H = \left( \frac{H(2H-1)}{\beta(2-2H, H-\frac{1}{2})} \right)^{\frac{1}{2}}$ and $\beta(\cdot, \cdot)$ the beta function. The derivative of $K^H$ is

$$\frac{\partial K^H}{\partial t} (t,s) := \partial_1 K^H(t,s) = c_H \left( \frac{s}{t} \right)^{\frac{1}{2}-H} (t-s)^{H-\frac{3}{2}}.$$  \hspace{1cm} (8)

The two parameters function $L_t$ given by (6) will be called the kernel of the Rosenblatt process. The following key relation in our calculation and it will repeatedly used in the paper

$$\int_0^u \partial_1 K^H(u,y) \partial_1 K^H(v,y) dy = a(H)|u-v|^{2H'-2}$$  \hspace{1cm} (9)

with $a(H) = H'(2H'-1)$. Among the main properties of the Rosenblatt process, we recall
• it is \( H \)-self-similar in the sense that for any \( c > 0 \), \( (Z^{(H)}_{ct}) = (c^H Z^{(H)}_t) \), where \( " = (d) " \) means equivalence of all finite dimensional distributions;

• it has stationary increments, that is, the joint distribution of \( (Z^{(H)}_{t+h} - Z^{(H)}_t, t \in [0,1]) \) is independent of \( h > 0 \);

• \( \mathbb{E}(|Z^{(H)}_t|^p) < \infty \) for any \( p > 0 \), and \( Z^{(H)} \) has the same variance and covariance than a standard fractional Brownian motion with parameter \( H \).

• the Rosenblatt process is Hölder continuous, of order \( \delta < H \). This can easily obtained by the Kolmogorov continuity criterium.

3 Renormalization of the cubic variation

3.1 Estimation of the mean square

We will study in this paragraph the cubic variation of the Rosenblatt process obtained by putting \( p = 3 \) in (1)

\[
V^{3,N} = \frac{1}{N} \sum_{i=0}^{N-1} \left( \frac{\left( Z^{(H)}_{\frac{i+1}{N}} - Z^{(H)}_{\frac{i}{N}} \right)^3}{\mathbb{E} \left( \left( Z^{(H)}_{\frac{i+1}{N}} - Z^{(H)}_{\frac{i}{N}} \right)^3 \right)} - 1 \right)
\]

(10)

Note that this expression is immaterial is the case of the fractional Brownian motion because the third moment of a centered Gaussian random variable is zero. By denoting for \( i = 1, \ldots, N \)

\[
f_{i,N} = L^{(H)}_{\frac{i+1}{N}} - L^{(H)}_{\frac{i}{N}}
\]

we obtain \( Z^{(H)}_{\frac{i+1}{N}} - Z^{(H)}_{\frac{i}{N}} = I_2(f_{i,N}) \) where \( I_2 \) is a multiple integral of order 2 as defined in Section 2.1 and then

\[
V^{3,N} = \frac{1}{N} \sum_{i=0}^{N-1} \left( \frac{(I_2(f_{i,N}))^3}{\mathbb{E}(I_2(f_{i,N}))^3} - 1 \right).
\]

By using the product formula for multiple Wiener-Itô integrals (3), for any function \( f \in L^2([0,1]^2) \) symmetric,

\[
I_2(f)^3 = I_6 \left( (f \otimes f) \otimes (f \otimes f) \right) + 8I_4 \left( (f \otimes f) \otimes \otimes_1 f \right) + 4I_4 \left( (f \otimes_1 f) \otimes (f \otimes f) \right)
\]

\[
+ 12I_2 \left( (f \otimes f) \otimes_2 f \right) + 16I_2 \left( (f \otimes_1 f) \otimes_1 f \right) + 2(f, f)_{L^2([0,1]^2)} I_2(f) + 8((f \otimes_1 f), f)_{L^2([0,1]^2)}.
\]

Here and in the sequel \( f \otimes f \) denotes the symmetrization of the function \( f \otimes f \) which is not necessary symmetric even if \( f \) is symmetric. Applying this to \( f = f_{i,N} \) we obtain

\[
(I_2(f_{i,N}))^3 = 8(f_{i,N} \otimes_1 f_{i,N}) \otimes_2 f_{i,N} + I_2(g_{i,N}) + 4I_4(h_{i,N}) + I_6((f_{i,N} \otimes f_{i,N}) \otimes f_{i,N})
\]

(11)

Here we used the following notation

\[
g_{i,N} = 2\|f_{i,N}\|_{L^2}^2 f_{i,N} + 12(f_{i,N} \otimes_1 f_{i,N}) \otimes_2 f_{i,N} + 16(f_{i,N} \otimes_1 f_{i,N}) \otimes_1 f_{i,N}
\]

(12)
and

\[ h_{i,N} = 2(f_{i,N} \otimes f_{i,N}) \otimes f_{i,N} + f_{i,N} \otimes (f_{i,N} \otimes f_{i,N}) := h_{i,N}^{(1)} + h_{i,N}^{(2)}. \]  

(13)

Note that \( g_{i,N} \in L^2([0,1]^2) \) and \( h_{i,N} \in L^2([0,1]^4) \). On the other hand, we can simplify a little bit the above expressions since

\[
(f_{i,N} \otimes f_{i,N}) \otimes 2 f_{i,N} = \frac{1}{3} \| f_{i,N} \|^2_{L^2} f_{i,N} + \frac{2}{3} (f_{i,N} \otimes 1) f_{i,N} \otimes 1 f_{i,N}.
\]

Hence the kernel of the second chaos term can be written as

\[
g_{i,N} = 6 \| f_{i,N} \|^2_{L^2} f_{i,N} + 24 (f_{i,N} \otimes 1) f_{i,N} \otimes 1 f_{i,N}.
\]

We start with the following lemma where we compute the cubic mean of the increment of the Rosenblatt process. We already observe a significant difference from the Gaussian case: this cubic mean is not zero.

**Lemma 1** Let \( Z_t^{(H)} \in [0,1] \) be a Rosenblatt process with selfsimilarity index \( H \in (\frac{1}{2},1) \). Then, for every \( s,t \in [0,1] \)

\[
\mathbb{E} \left( Z_t^{(H)} - Z_s^{(H)} \right)^3 = C(H) |t-s|^{3H}
\]

(14)

where

\[
C(H) = 8a(H)^3 d(H)^3 \int_{[0,1]^3} (|u-v||u-u'||v-v')^{2H-2} du dv du'.
\]

**Proof.** Let us denote by

\[
f_{s,t}(x,y) = L_t(x,y) - L_s(x,y)
\]

where \( L \) is the kernel of the Rosenblatt process given by (6) and \( x,y \in [0,1] \). We will have, by using relation (9),

\[
(f_{s,t} \otimes 1 f_{s,t})(x,y) = \int_0^1 f_{s,t}(x,z) f_{s,t}(y,z) dz = d(H)^2 a(H) \left( 1_{[0,t]}(x,y) \int_x^t \int_y^{t'} \partial_1 K^{H'}(u,x) \partial_1 K^{H'}(v,y) |u-v|^{2H'-2} dudv \\
- 1_{[0,t]}(x) 1_{[0,t]}(y) \int_x^t \int_y^{t'} \partial_1 K^{H'}(u,x) \partial_1 K^{H'}(v,y) |u-v|^{2H'-2} dudv \\
- 1_{[0,t]}(x) 1_{[0,t]}(y) \int_x^t \int_y^{t'} \partial_1 K^{H'}(u,x) \partial_1 K^{H'}(v,y) |u-v|^{2H'-2} dudv \\
+ 1_{[0,t]}(x,y) \int_x^t \int_y^{t'} \partial_1 K^{H'}(u,x) \partial_1 K^{H'}(v,y) |u-v|^{2H'-2} dudv \right).
\]

The computation of the cubic mean of a multiple integral in the second chaos (11) implies

\[
\mathbb{E} \left( Z_t^{(H)} - Z_s^{(H)} \right)^3 = 8 \langle f_{s,t} \otimes 1 f_{s,t}, f_{s,t} \rangle_{L^2([0,1]^2)}.
\]

We compute, by (9)
\[ \langle f_{s,t} \otimes f_{s,t}, f_{s,t} \rangle_{L^2([0,1]^2)} = \int_{[0,1]^2} (f_{s,t} \otimes f_{s,t})(x,y) f_{s,t}(x,y) \, dx \, dy \]

\[ = d(H)^3 a(H)^3 \int_s^t \int_s^t \int_s^t (|u - v||u - u'||v - u'|)^{2H' - 2} \, du \, du' \, dv. \]

By the change of variables \( \bar{u} = \frac{u - s}{t - s} \) we will transform the integrals on \([s,t]\) into integrals from 0 to 1. We immediately obtain the relation (14). \( \blacksquare \)

To calculate \( \mathbb{E}(V^{3,N})^2 \) we apply the above result and we obtain

\[ \mathbb{E}[(I_2(f_{i,N}))^3] = 8(f_{i,N} \otimes f_{i,N}) \otimes f_{i,N} \]

\[ = 8d(H)^3 a(H)^3 \int_{I_i} \int_{I_i} \int_{I_i} dy_1 dy_2 dy_3 (|y_1 - y_2| \cdot |y_2 - y_3| \cdot |y_3 - y_1|)^{2H' - 2} \]

\[ = 8 \frac{d(H)^3 a(H)^3}{N^{6H' - 3}} \int_{[0,1]^3} dy_1 dy_2 dy_3 (|y_1 - y_2| \cdot |y_2 - y_3| \cdot |y_3 - y_1|)^{2H' - 2} \]

\[ = C(H) N^{-(6H' - 3)} = C(H) N^{-3H}. \]

where \( a(H) = \frac{H(H+1)}{2} \) and \( C(H) \) is defined in (15).

We can write the expression of the statistics \( V_N \) as follows

\[ V^{3,N} = \frac{1}{C(H) N^{1-3H}} \sum_{i=0}^{N-1} \left( (I_2(f_{i,N}))^3 - \mathbb{E}(I_2(f_{i,N}))^3 \right) \]

\[ = \frac{1}{C(H) N^{1-3H}} \sum_{i=0}^{N-1} \left( I_2(g_{i,N}) + 4I_4(h_{i,N}) + I_6((f_{i,N} \otimes f_{i,N}) \otimes f_{i,N}) \right). \] (16)

We prove next the following renormalization result.

**Proposition 2** Let \( V^{3,N} \) the cubic variation statistics of the Rosenblatt process. Then

\[ \mathbb{E} \left( N^{1-H} V^{3,N} \right)^2 \rightarrow_{N \to \infty} \bar{C}(H) \] (17)

where \( \bar{C}(H) := C(H)^2 C_0(H) \) with

\[ C_0(H) = \left( 9 + 36C'(H)H(2H - 1) + 144 \left[ C'(H)H(2H - 1) \right]^2 \right). \]
**Proof.** The isometry property of multiple Wiener-Itô integrals and relation (16) imply

\[
\mathbb{E}(V^{3,N})^2 = \frac{1}{N^2} \left( \mathbb{E}(I_2(f_{i,N}))^3 \right)^2 \sum_{i,j=0}^{N-1} \left[ \mathbb{E}(I_2(g_{i,N})I_2(g_{j,N})) + 16\mathbb{E}(I_4(h_{i,N})I_4(h_{j,N})) \right] \\
+ \mathbb{E} \left( I_6((f_{i,N} \otimes f_{i,N}) \otimes f_{j,N})I_6((f_{j,N} \otimes f_{j,N}) \otimes f_{j,N}) \right) \\
= \frac{1}{C(H)^2 N^{2-6H}} \sum_{i,j=0}^{N-1} \left[ 2! \left< g_{i,N} , g_{j,N} \right>_{L^2([0,1]^2)} \\
+ 4! \times 16 \left< \widetilde{h}_{i,N} , \widetilde{h}_{j,N} \right>_{L^2([0,1]^4)} \\
+ 6! \left< (f_{i,N} \otimes f_{i,N}) \otimes f_{i,N} , (f_{j,N} \otimes f_{j,N}) \otimes f_{j,N} \right>_{L^2([0,1]^6)} \right] \\
:= \frac{1}{C(H)^2 N^{2-6H}} \left[ A_N^{(2)} + B_N^{(4)} + D_N^{(6)} \right].
\]

We use the notation \( A_N^{(2)} \) to indicate that this term comes from the estimation of the second chaos summand of \( V^{3,N} \), and similarly for the terms \( B_N^{(4)} \) and \( D_N^{(6)} \). We will try to estimate the all the three terms above to see which is the dominant term of \( V_N \).

**Estimation of the term** \( A_N^{(2)} \). We start by calculating \( A_N^{(2)} \). Taking into account the expression of the second chaos kernel \( g_{i,N} \) (12)

\[
A_N^{(2)} := \sum_{i,j=0}^{N-1} 2 \left< g_{i,N} , g_{j,N} \right>_{L^2([0,1]^2)} \\
= 2 \sum_{i,j=0}^{N-1} \left[ 36 \left< f_{i,N} \right>_{L^2([0,1]^2)}^2 \left< f_{j,N} \right>_{L^2([0,1]^2)}^2 + 144 \left< f_{i,N} \right>_{L^2([0,1]^2)}^2 \left< f_{j,N} \otimes f_{j,N} \right>_{L^2([0,1]^2)} \\
+ (24)^2 \left< f_{i,N} \otimes f_{i,N} \otimes f_{j,N} \right>_{L^2([0,1]^2)} \right] \\
:= 2(36A_{1,N}^{(2)} + 144A_{2,N}^{(2)} + (24)^2 A_{3,N}^{(2)})
\]

Let us evaluate the term \( A_{1,N}^{(2)} \), we have

\[
2! \left< f_{i,N} \right>_{L^2([0,1]^2)}^2 = \mathbb{E} \left| Z_{\frac{f_{i,N}}{\frac{1}{N}}}^{(H)} - Z_{\frac{f_{i,N}}{\frac{1}{N}}}^{(H)} \right|^2 = N^{-2H}.
\]

Furthermore

\[
2 \left< f_{i,N} , f_{j,N} \right>_{L^2([0,1]^2)} = \mathbb{E} \left( Z_{\frac{f_{i,N}}{\frac{1}{N}}}^{(H)} - Z_{\frac{f_{i,N}}{\frac{1}{N}}}^{(H)} \right) \left( Z_{\frac{f_{j,N}}{\frac{1}{N}}}^{(H)} - Z_{\frac{f_{j,N}}{\frac{1}{N}}}^{(H)} \right).
\]
Hence

\[ A_{1,N}^{(2)} = \sum_{i,j=0}^{N-1} \left[ \| f_{i,N} \|_{L^2[0,1]^2}^2 \| f_{j,N} \|_{L^2[0,1]^2}^2 \langle f_{i,N}, f_{j,N} \rangle_{L^2[0,1]^2} \right] \]

\[ = \frac{1}{8} N^{-4H} \sum_{i,j=0}^{N-1} \mathbf{E} \left( \frac{Z_i^{(H)}}{N} - \frac{Z_{i,N}^{(H)}}{N} \right) \left( \frac{Z_i^{(H)}}{N} - \frac{Z_{i,N}^{(H)}}{N} \right) = \frac{1}{8} N^{-4H} \]

because \( \mathbf{E} \sum_{i,j=0}^{N-1} \left( \frac{Z_i^{(H)}}{N} - \frac{Z_{i,N}^{(H)}}{N} \right) \left( \frac{Z_i^{(H)}}{N} - \frac{Z_{i,N}^{(H)}}{N} \right) = \mathbf{E}(Z_1^{(H)})^2 = 1 \) and we have

\[ \lim_{N \to \infty} N^{4H} A_{1,N}^{(2)} = \frac{1}{8}. \quad (18) \]

We evaluate \( A_{3,N}^{(2)}. \) Note that

\[ d(H)^{-2} a(h)^{-1}(f_{i,N} \otimes_1 f_{i,N})(x,y) \]

\[ = 1_{\{0,1\}}^2(x,y) \int_{L_i} \int_{L_i} \partial_1 K^{H'}(u_1, x) \partial_1 K^{H'}(u_2, y) |u_1 - u_2|^{2H'-2} du_1 du_2 \]

\[ + 1_{\{0,1\}}(x) I_i(y) \int_{L_i} \int_{L_i} \partial_1 K^{H'}(u_1, x) \partial_1 K^{H'}(u_2, y) |u_1 - u_2|^{2H'-2} du_1 du_2 \]

\[ + 1_{\{0,1\}}(y) I_i(x) \int_{L_i} \int_{L_i} \partial_1 K^{H'}(u_1, x) \partial_1 K^{H'}(u_2, y) |u_1 - u_2|^{2H'-2} du_1 du_2 \]

\[ + 1_{I_i}(x) I_i(y) \int_{L_i} \int_{L_i} \partial_1 K^{H'}(u_1, x) \partial_1 K^{H'}(u_2, y) |u_1 - u_2|^{2H'-2} du_1 du_2. \]

Sometimes its useful to use to following compressed expression

\[ (f_{i,N} \otimes_1 f_{i,N})(x,y) = d(H)^2 a(H) 1_{\{0,1\}}^2(x,y) \int_{L_i} \int_{L_i} \partial_1 K^{H'}(u_1, x) \partial_1 K^{H'}(u_2, y) |u_1 - u_2|^{2H'-2} du_1 du_2 \]

and

\[ d(H)^{-1} f_{i,N}(x,z) \]

\[ = 1_{\{0,1\}}^2(x,z) \int_{L_i} \partial_1 K^{H'}(u_3, x) \partial_1 K^{H'}(u_3, z) du_3 + 1_{\{0,1\}}(x) I_i(z) \int_{L_i} \partial_1 K^{H'}(u_3, x) \partial_1 K^{H'}(u_3, z) du_3 \]

\[ + 1_{\{0,1\}}(z) I_i(x) \int_{L_i} \partial_1 K^{H'}(u_3, x) \partial_1 K^{H'}(u_3, z) du_3 \]

\[ + 1_{I_i}(x) I_i(z) \int_{L_i} \partial_1 K^{H'}(u_3, x) \partial_1 K^{H'}(u_3, z) du_3. \]

or otherwise

\[ f_{i,N}(x,z) = d(H) 1_{\{0,1\}}^2(x,z) \int_{L_i} \partial_1 K^{H'}(u_3, x) \partial_1 K^{H'}(u_3, z) du_3. \quad (20) \]
Therefore

\[
((f_i, N \otimes f_i, N) \otimes f_i, N)(y, z) = d(H)^3 a(H)^2 \left( \frac{1^{\otimes 2}}{[0, \frac{1}{N}]} \right)(y, z)
\]

\[
\times \int_{I_i} \int_{I_i} \int_{I_i} (|u_1 - u_2||u_1 - u_3|)^{2H' - 2} \partial_1 K^{H'}(u_2, y) \partial_1 K^{H'}(u_3, z) \, du_3 \, du_2 \, du_1.
\]

The norm has a nicer expression. Using the change of variables \( \bar{u} = (u - \frac{1}{N})N \) (which is now usual and it can be used systematically) we have

\[
((f_i, N \otimes f_i, N) \otimes f_i, N, (f_j, N \otimes f_j, N) \otimes f_j, N)_{L^2([0,1]^2)} =
\]

\[
d(H)^6 a(H)^6 \int_{[0,1]^3} du_1 du_2 du_3 \int_{[0,1]^3} dv_1 dv_2 dv_3
\]

\[
|u_1 - u_2|^{2H' - 2}|u_1 - u_3|^{2H' - 2}|v_1 - v_2|^{2H' - 2}|v_1 - v_3|^{2H' - 2}|u_2 - v_2|^{2H' - 2}|u_3 - v_3|^{2H' - 2}
\]

\[
= \frac{d(H)^6 a(H)^6}{N^{12H' - 6}} \int_{[0,1]^3} \int_{[0,1]^3} \int_{[0,1]^3} |v_1 - v_2|^{2H' - 2}|v_2 - v_3|^{2H' - 2}|v_3 - v_4 + i - j|^{2H' - 2}
\]

\[
\times |v_4 - v_5|^{2H' - 2}|v_5 - v_6|^{2H' - 2}|v_6 - v_1 + j - i|^{2H' - 2} dv_1 \ldots dv_6.
\]

The rate of convergence of all terms presents in this proof comes actually from how many product \( |u - v|^{2H'} \) exists with \( u \in I_i \) and \( v \in I_j \) we have. Hence

\[
A^{(2)}_{3,N} = \frac{d(H)^6 a(H)^6}{N^{12H' - 6}} \sum_{i,j=1}^{N} \int_{[0,1]^3} \int_{[0,1]^3} |v_1 - v_2|^{2H' - 2}|v_2 - v_3|^{2H' - 2}|v_3 - v_4 + i - j|^{2H' - 2}
\]

\[
\times |v_4 - v_5|^{2H' - 2}|v_5 - v_6|^{2H' - 2}|v_6 - v_1 + j - i|^{2H' - 2} dv_1 \ldots dv_6
\]

\[
= \frac{2d(H)^6 a(H)^6}{N^{12H' - 6}} \sum_{i>j=1}^{N} \int_{[0,1]^3} \int_{[0,1]^3} |v_1 - v_2|^{2H' - 2}|v_2 - v_3|^{2H' - 2}|v_3 - v_4 + i - j|^{2H' - 2}
\]

\[
\times |v_4 - v_5|^{2H' - 2}|v_5 - v_6|^{2H' - 2}|v_6 - v_1 + j - i|^{2H' - 2} dv_1 \ldots dv_6
\]

\[
= \frac{2d(H)^6 a(H)^6}{N^{12H' - 6}} \sum_{k=0}^{N-1} (N - k) \int_{[0,1]^3} \int_{[0,1]^3} |v_1 - v_2|^{2H' - 2}|v_2 - v_3|^{2H' - 2}|v_3 - v_4 + k|^{2H' - 2}
\]

\[
\times |v_4 - v_5|^{2H' - 2}|v_5 - v_6|^{2H' - 2}|v_6 - v_1 + k|^{2H' - 2} dv_1 \ldots dv_6.
\]

We put

\[
\overline{A}^{(2)}_{3,N} := \frac{1}{N^{12H' - 6}} \sum_{k=0}^{N-1} (N - k)|v_3 - v_4 + k|^{2H' - 2}|v_1 - v_6 + k|^{2H' - 2}
\]

\[
= \frac{1}{N^{8H' - 4}} \frac{1}{N} \sum_{k=0}^{N-1} (1 - \frac{k}{N}) |v_3 - v_4 + k|^{2H' - 2} \frac{v_1 - v_6}{N} + \frac{k}{N} |v_3 - v_6 + k|^{2H' - 2}
\]

and we conclude easily by a Riemann sum convergence that

\[
N^{4H} \overline{A}_{3,N} = N^{8H' - 4} \overline{A}_{3,N} \xrightarrow{N\to\infty} \int_0^1 (1 - x)x^{4H' - 4} \, dx = \frac{1}{2H - 1} - \frac{1}{2H}
\]
because the terms \(\frac{u_{22}}{N}\) are negligible with respect to \(\frac{k}{N}\) for large enough \(k\). This implies that,

\[
N^{4H} A_{3,N}^{(2)} \xrightarrow{N \to \infty} \frac{d(H)^6 a(H)^6}{H(2H - 1)} (C'(H))^2 = \frac{H^2 (2H - 1)^2}{8} (C'(H))^2 \tag{21}
\]

where

\[
C'(H) = \int_{[0,1]^3} |v_1 - v_2|^{2H'-2} |v_2 - v_3|^{2H'-2} dv_1 dv_2 dv_3.
\]

Now, we estimate the term \(A_{2,N}^{(2)}\).

\[
\langle f_i, N \rangle \langle f_j, N \rangle L^2([0,1]^2) = \frac{d(H)^4 a(H)^4}{N^{8H'-4}} \int_{[0,1]^4} |v_2 - v_3|^{2H'-2} |v_3 - v_4|^{2H'-2}
\times \left( \sum_{i,j=1}^N |v_1 - v_2 + i - j|^{2H'-2} |v_4 - v_1 + j - i|^{2H'-2} \right) dv_1 \ldots dv_4
\]

\[
= \frac{N^{-2H}}{2} \frac{d(H)^4 a(H)^4}{N^{8H'-4}} \int_{[0,1]^4} |v_2 - v_3|^{2H'-2} |v_3 - v_4|^{2H'-2}
\times \left( 2 \sum_{k=0}^{N-1} (N - k) |v_1 - v_2 + k|^{2H'-2} |v_4 - v_1 + k|^{2H'-2} \right) dv_1 \ldots dv_4
\]

\[
= \frac{N^{-2H}}{2} \frac{d(H)^4 a(H)^4}{N^{4H'-2}} \int_{[0,1]^4} |v_2 - v_3|^{2H'-2} |v_3 - v_4|^{2H'-2}
\times \left( \frac{1}{N} \sum_{k=0}^{N-1} (1 - \frac{k}{N}) |v_1 - v_2|^{2H'-2} |v_4 - v_1|^{2H'-2} \right) dv_1 \ldots dv_4
\]

We obtain that \(N^{4H} A_{2,N}^{(2)}\) converges as \(N \to \infty\) to

\[
\left( \int_0^1 (1 - x) x^{4H'-4} dx \right) \left( \frac{d(H)^4 a(H)^4}{N^{8H'-4}} \int_{[0,1]^3} |v_2 - v_3|^{2H'-2} |v_3 - v_4|^{2H'-2} dv_2 \ldots dv_4 \right)
\]

\[
= \left( \frac{1}{2H - 1} - \frac{1}{2H} \right) \left( \frac{d(H)^4 a(H)^4}{N^{4H'-2}} \int_{[0,1]^3} |v_2 - v_3|^{2H'-2} |v_3 - v_4|^{2H'-2} dv_2 \ldots dv_4 \right)
\]
Thus
\[ N^{4H} A_{2,N}^{(2)} \xrightarrow{N \to \infty} C'(H) \frac{H(2H - 1)}{8} \] (22)

From (18), (21) and (22), we obtain that
\[ N^{4H} A_{N}^{(2)} = 2(36N^{4H} A_{1,N}^{(2)} + 144N^{4H} A_{2,N}^{(2)} + (24)^2 N^{4H} A_{3,N}^{(2)}) \] (23)
converges to \( 9 + 36C'(H)H(2H - 1) + 144 [C'(H)H(2H - 1)]^2 \) \( : = C_0(H) \) as \( N \to \infty \).

Consequently
\[ \frac{N^{4H} A_{N}^{(2)}}{C_0(H)} \xrightarrow{N \to \infty} 1. \] (24)

**Estimation of the term** \( D_{N}^{(6)} \). Now, we study the convergence of \( D_{N}^{(6)} \), using the symmetry property of every \( f_{i,N}, i = 0, \ldots, N - 1 \) on \([0, 1]^2\), there exist positive combinatorial constants \( c_1, c_2 \) and \( c_3 \) such that
\[
D_{N}^{(6)} = \sum_{i,j=0}^{N-1} 6! \langle (f_{i,N} \otimes f_{j,N}) \otimes f_{j,N}, (f_{j,N} \otimes f_{i,N}) \otimes f_{i,N} \rangle_{L^2([0,1]^6)}
\]
\[
= c_1 \sum_{i,j=0}^{N-1} \left( \langle f_{i,N}, f_{j,N} \rangle_{L^2([0,1]^2)} \right)^3
\]
\[
+ c_2 \sum_{i,j=0}^{N-1} \langle f_{i,N}, f_{j,N} \rangle_{L^2([0,1]^2)} \int_{[0,1]^4} f_{i,N}(x_1, x_2) f_{j,N}(x_2, x_3) f_{i,N}(x_3, x_4) f_{j,N}(x_4, x_1) dx_1 \ldots dx_4
\]
\[
+ c_3 \sum_{i,j=0}^{N-1} \int_{[0,1]^6} f_{i,N}(x_1, x_2) f_{j,N}(x_2, x_3) f_{i,N}(x_3, x_4) f_{j,N}(x_4, x_5) f_{i,N}(x_5, x_6) f_{j,N}(x_6, x_1) dx_1 \ldots dx_6
\]
\[
:= c_1 D_{1,N}^{(6)} + c_2 D_{2,N}^{(6)} + c_3 D_{3,N}^{(6)}.
\]

By using the same argument as above, we have
\[
D_{1,N}^{(6)} = \sum_{i,j=0}^{N-1} \left( \langle f_{i,N}, f_{j,N} \rangle_{L^2([0,1]^2)} \right)^3
\]
\[
= \frac{d(H)^6 a(H)^6}{N^{6H}} \sum_{i,j=0}^{N-1} \left( \int_{[0,1]^2} |x_1 - x_2 + i - j|^{2H - 2} \, dx_1 dx_2 \right)^3
\]
\[
= \frac{d(H)^6 a(H)^6}{N^{6H}} 2 \sum_{k=0}^{N-1} (N - k) \left( \int_{[0,1]^2} |x_1 - x_2 + k|^{2H - 2} \, dx_1 dx_2 \right)^3
\]
\[
= \frac{2d(H)^6 a(H)^6}{N^4} \sum_{k=0}^{N-1} \frac{1}{N} \left( \frac{1}{N} - \frac{k}{N} \right) \left( \int_{[0,1]^2} \left| \frac{x_1 - x_2}{N} + \frac{k}{N} \right|^{2H - 2} \, dx_1 dx_2 \right)^3.
\]
and clearly since $H < 1$ we have
\[
\lim_{N \to \infty} N^{4H} D_{1,N}^{(6)} = 0. \quad (25)
\]
By the same way, we obtain
\[
D_{2,N}^{(6)} = \sum_{i,j=0}^{N-1} \left( f_i,N \right) \left( f_j,N \right)_{L^2([0,1]^2)} \int_{[0,1]^4} f_i,N(x_1, x_2) f_j,N(x_2, x_3) f_i,N(x_3, x_4) f_j,N(x_4, x_1) \, dx_1 \ldots dx_4
\]
\[
= \sum_{i,j=0}^{N-1} N^{-4} d(H)^4 a(H)^4 \left( \left( f_i,N \right) \left( f_j,N \right)_{L^2([0,1]^2)} \right) \int_{[0,1]^4} \frac{x_1 - x_2}{N} + \frac{i - j}{N} \left| \frac{x_3 - x_2}{N} + \frac{i - j}{N} \right|^{2H'} \frac{x_3 - x_4}{N} + \frac{i - j}{N} \left| \frac{x_1 - x_4}{N} + \frac{i - j}{N} \right|^{2H'} \, dx_1 \ldots dx_4
\]
\[
= \frac{2d(H)^6 a(H)^6}{N^4} \sum_{k=0}^{N-1} \frac{1}{N} \int_{[0,1]^4} \frac{x_5 - x_6}{N} + \frac{k}{N} \left| \frac{x_5 - x_6}{N} + \frac{k}{N} \right|^{2H'} \frac{x_3 - x_4}{N} + \frac{k}{N} \left| \frac{x_3 - x_4}{N} + \frac{k}{N} \right|^{2H'} \, dx_1 \ldots dx_4.
\]
This implies that
\[
\lim_{N \to \infty} N^{4H} D_{2,N}^{(6)} = 0. \quad (26)
\]
The same manner as in previous results, we have
\[
D_{3,N}^{(6)} = \sum_{i,j=0}^{N-1} \int_{[0,1]^6} f_i,N(x_1, x_2) f_j,N(x_2, x_3) f_i,N(x_3, x_4) f_j,N(x_4, x_5) f_i,N(x_5, x_6) f_j,N(x_6, x_1) \, dx_1 \ldots dx_6
\]
\[
= \frac{d(H)^6 a(H)^6}{N^6} \sum_{i,j=0}^{N-1} \int_{[0,1]^6} \frac{x_1 - x_2}{N} + \frac{i - j}{N} \left| \frac{x_1 - x_2}{N} + \frac{i - j}{N} \right|^{2H'} \frac{x_3 - x_2}{N} + \frac{i - j}{N} \left| \frac{x_3 - x_2}{N} + \frac{i - j}{N} \right|^{2H'} \frac{x_3 - x_4}{N} + \frac{i - j}{N} \left| \frac{x_3 - x_4}{N} + \frac{i - j}{N} \right|^{2H'} \, dx_1 \ldots dx_6
\]
\[
= \frac{2d(H)^6 a(H)^6}{N^4} \sum_{k=0}^{N-1} \frac{1}{N} \int_{[0,1]^4} \frac{x_5 - x_4}{N} + \frac{k}{N} \left| \frac{x_5 - x_4}{N} + \frac{k}{N} \right|^{2H'} \frac{x_5 - x_6}{N} + \frac{k}{N} \left| \frac{x_5 - x_6}{N} + \frac{k}{N} \right|^{2H'} \frac{x_3 - x_4}{N} + \frac{k}{N} \left| \frac{x_3 - x_4}{N} + \frac{k}{N} \right|^{2H'} \, dx_1 \ldots dx_6.
\]
Hence
\[
\lim_{N \to \infty} N^{4H} D_{3,N}^{(6)} = 0. \quad (27)
\]
Thus, from (25), (26) and (27), we obtain

$$\lim_{N \to \infty} N^{4H} D_N^{(6)} = 0. \quad (28)$$

**Estimation of the term** $B_N^{(4)}$. Applying the same argument as in last part, there exist constants $c'_1$ and $c'_2$ such that

$$\left\langle \tilde{h}_{i,N}, \tilde{h}_{j,N} \right\rangle_{L^2([0,1]^4)}
= c'_1 \sum_{i,j=0}^{N-1} \langle f_{i,N}, f_{j,N} \rangle_{L^2([0,1]^2)} \int_{[0,1]^4} f_{i,N}(x_1, x_2) f_{j,N}(x_2, x_3) f_{i,N}(x_3, x_4) f_{j,N}(x_4, x_1) dx_i
\quad + \quad c'_2 \sum_{i,j=0}^{N-1} \int_{[0,1]^6} f_{i,N}(x_1, x_2) f_{j,N}(x_2, x_3) f_{i,N}(x_3, x_4) f_{j,N}(x_4, x_5) f_{i,N}(x_5, x_6) f_{j,N}(x_6, x_1) dx_i
\quad = \quad c'_1 D_{2,N}^{(6)} + c'_2 D_{3,N}^{(6)}.$$ 

The same terms as in the estimation of the sixth chaos kernel appear. Thus, from the convergences (26) and (27),

$$\lim_{N \to \infty} N^{4H} B_N^{(4)} = \lim_{N \to \infty} N^{4H} \sum_{i,j=0}^{N-1} \left\langle \tilde{h}_{i,N}, \tilde{h}_{j,N} \right\rangle_{L^2([0,1]^4)} = \lim_{N \to \infty} \left( c'_1 N^{4H} D_{2,N}^{(6)} + c'_2 N^{4H} D_{3,N}^{(6)} \right) = 0. \quad (29)$$

As a consequence of the convergences (24), (28) and (29), we have proved that for every $H > \frac{1}{2}$ and with the notation $\bar{C}(H) = \frac{C_0(H)^2}{C_0(H)^4}$,

$$\frac{C(H)^2}{C_0(H)} N^{2-2H} \mathbf{E}(V^{3,N})^2 = \mathbf{E} \left( \sqrt{\bar{C}(H)} N^{1-H} V^{3,N} \right)^2 \to 1. \quad (30)$$

### 3.2 Non-convergence to a Gaussian limit

We prove that the normalized variation doesn’t converge in distribution to the normal law. Of course this somehow superfluous taking into account that in the next section we show a non-central limit theorem for this statistics, but we found the calculations instructive to see why it does not converge to a Gaussian limit. Recall that by a result of [15] (Theorem 4 in this paper) a sequence $F_N = I_q(f_N)$ in the $q$ Wiener chaos with $\mathbb{E} F_N^2 \to 1$ converges to the normal law $N(0,1)$ if and only if $\|DF_N\|_{L^2[0,1]}^2$ converges to $q$ in $L^2(\Omega)$ when $N \to \infty$. Here $D$ denotes the Malliavin derivative and if $f \in L^2([0,T]^n)$ is a symmetric function, we will use the following rule to differentiate in the Malliavin sense

$$D_t I_n(f) = n I_{n-1}(f(\cdot,t)), \quad t \in [0,1].$$
We put

\[ T_N = \frac{\sqrt{C(H)} N^{1-H}}{C(H) N^{1-3H}} \sum_{i=0}^{N} I_2(g_i,N) = \frac{N^{2H}}{\sqrt{C_0(H)}} \sum_{i=0}^{N} I_2(g_i,N) \]

We derive \( T_N \) in the Malliavin sense and we obtain \( D_iT_N = \frac{2N^{2H}}{\sqrt{C_0(H)}} \sum_{i=0}^{N} I_1(g_i,N,(.,t)) \) and thus

\[
\|DT_N\|_{L^2([0,1]^2)}^2 = 4N^{4H} \frac{\int_0^1 \left( \sum_{i=0}^{N} I_1(g_i,N,(.,t)) \right)^2 dt}{C_0(H)}
\]

\[
= 4N^{4H} \frac{\int_0^1 \left( \sum_{i,j=0}^{N} I_1(g_i,N,(.,t)) I_1(g_j,N,(.,t)) \right) dt}{C_0(H)}
\]

\[
= 4N^{4H} \frac{\int_0^1 I_0(g_i,N \odot_1 g_j,N) dt}{C_0(H)} + \sum_{i,j=0}^{N} \int_0^1 I_2(g_i,N \odot_0 g_j,N) dt \right)
\]

\[
:= 4N^{4H} \frac{\left( J_{1,N} + J_{2,N} \right)}{C_0(H)}
\]

where we denoted

\[ J_{1,N} = \sum_{i,j=0}^{N} \int_0^1 I_0(g_i,N \odot_1 g_j,N) dt = \sum_{i,j=0}^{N} < g_i,N, g_j,N >_{L^2([0,1]^2)} = \frac{1}{2} A_N^{(2)} \]

From (24) we obtain

\[
\frac{4N^{4H}}{C_0(H)} J_{1,N} \xrightarrow{\text{as } N \to \infty} 2
\]

in \( L^2(\Omega) \) because the term \( A_N^{(2)} \) is deterministic. To prove that \( \|DT_N\|_{L^2([0,1]^2)}^2 \) not converges in \( L^2(\Omega) \) to 2, it is sufficient to show that

\[
\lim_{N \to \infty} \mathbb{E} \left( \frac{4N^{4H}}{C_0(H)} J_{2,N} \right)^2 > 0.
\]

where \( J_{2,N} = \sum_{i,j=0}^{N} \int_0^1 I_2(g_i,N,(.,t)g_j,N,(.,t)) dt. \)
We calculate the mean square of this term.

\[
\mathbf{E}(J_{2,N})^2 = 2 \int_{[0,1]^2} \left( \sum_{i,j=0}^{N} \int_{0}^{1} I_2(g_{i,N}(r,t)g_{j,N}(s,t))dt \right)^2 drds
\]

\[
= 2 \sum_{i,j,k,l=0}^{N} \int_{[0,1]^4} g_{i,N}(r,t)g_{j,N}(s,t)g_{k,N}(r,u)g_{l,N}(s,u)drdsdtdu
\]

\[
\geq \sum_{i,j,k,l=0}^{N} \int_{[0,1]^4} f_{i,N}(r,t)f_{j,N}(s,t)f_{k,N}(r,u)f_{l,N}(s,u)drdsdtdu
\]

\[
= 2d(H)^4 a(H)^4 N^{-8H} \sum_{i,j,k,l=0}^{N} \int_{I_i} \int_{I_j} \int_{I_k} \int_{I_l} (|r-t||s-t||r-u||s-u|)^{2H-2} drdsdtdu
\]

\[
= 2d(H)^4 a(H)^4 N^{-8H} \sum_{i,j,k,l=0}^{N} \int_{[0,1]^4} drdsdtdu
\]

\[
\frac{1}{N^4} \left( \frac{r-t+i-j}{N} - \frac{s-j+k}{N} \right) \frac{u-r+l-i}{N} \right)^{2H-2}.
\]

By using Riemann sums approximations, we obtain

\[
\lim_{N \to \infty} \mathbf{E}(N^{4H}J_{2,N})^2 \geq 2d(H)^4 a(H)^4 \int_{[0,1]^4} dx_1dx_2dx_3dx_4 (|x_1 - x_2||x_2 - x_3||x_3 - x_4||x_4 - x_1|)^{2H-2} > 0.
\]

4 The non-central limit theorem for the cubic variation of the Rosenblatt process

Denote by \( L_t \) the kernel of the Rosenblatt process

\[
L_t(x,y) = d(H)1_{[0,1]}(x,y) \int_{0}^{t} \int_{0}^{t} \left( \int_{x<s<y} \partial_1 K^{H'}(s,x)\partial_1 K^{H'}(s,y)ds \right)
\]

and recall the notation

\[
f_{i,N}(x,y) = L_{i+1}^{(H)}(x,y) - L_{i}^{(H)}(x,y).
\]

We proved in the previous section that the dominant term of the statistics \( V_{3,N} \) which gives its normalization is

\[
C(H)^{-1} N^{3H-1} \sum_{i=0}^{N-1} I_2(g_{i,N})
\]

where

\[
g_{i,N} = 6\|f_{i,N}\|_{L^2([0,1]^2)}^2 f_{i,N} + 24 (f_{i,N} \otimes f_{i,N}) \otimes f_{i,N} = 3N^{-2H} f_{i,N} + 24 (f_{i,N} \otimes f_{i,N}) \otimes f_{i,N} := 3g_{i,N}^{(1)} + 24g_{i,N}^{(2)}.
\]

More precisely, it follows from the proof of Proposition 2 that

\[
\mathbf{E} \left[ N^{1-H} \left( N^{3H-1} \sum_{i=0}^{N-1} I_2(g_{i,N}^{(1)}) \right) \right]^2 = N^{4H} A_{1,N}^{(2)} \to_{N \to \infty} 1/8
\]
and
\[
E \left[ N^{1-H} \left( N^{3H-1} \sum_{i=0}^{N-1} I_2(g_{i,N}^{(2)}) \right) \right] = \left. N^{4H} A_{3,N}^{(2)} \to_{N \to \infty} \frac{H^2(2H-1)^2}{8} (C'(H))^2. \right]
\]

Consequently, the limit of the sequence $V_{3,N}$ is the same as the limit of the sequence
\[
C(H)^{-1} N^{1-H} N^{3H-1} \left( 3 \sum_{i=0}^{N-1} I_2(g_{i,N}^{(1)}) + 24 \sum_{i=0}^{N-1} I_2(g_{i,N}^{(2)}) \right).
\]

We prove here our main result.

**Theorem 3** The renormalized cubic variation statistics based on the Rosenblatt process $N^{1-H} V_{3,N}$ with $V_{3,N}$ given by (10) converges in $L^2(\Omega)$ as $N \to \infty$ to the Rosenblatt random variable $D(H) Z_1^{(H)}$ where $D(H) = C(H)^{-1}(3 + 24d(H)^2 a(H)^2 C'(H))$.

**Proof.** To see the limit of $N^{1-H} V_{3,N}$ we need therefore to study the convergence of $N^{1-H} \left( N^{3H-1} \sum_{i=0}^{N-1} I_2(g_{i,N}^{(1)}) \right)$ and of $N^{1-H} \left( N^{3H-1} \sum_{i=0}^{N-1} I_2(g_{i,N}^{(2)}) \right)$.

Is easy to treat the first part. In fact we have
\[
N^{1-H} N^{3H-1} \sum_{i=0}^{N-1} I_2(g_{i,N}^{(1)}) = N^{2H} \sum_{i=0}^{N-1} N^{-2H} I_2(f_{i,N}) = \sum_{i=0}^{N-1} I_2(f_{i,N}) = Z_1^{(H)} \quad (32)
\]
where $Z_1^{(H)}$ is a Rosenblatt random variable with self-similarity order $H$.

We find then the limit of the second part of the dominant term. We have
\[
N^{1-H} N^{3H-1} \sum_{i=0}^{N-1} I_2(g_{i,N}^{(2)}) = N^{2H} \sum_{i=0}^{N-1} I_2((f_{i,N} \otimes_1 f_{i,N}) \otimes_1 f_{i,N}).
\]

Let us denote by
\[
l^{H'}(x,y,z,t) := \partial_1 K^{H'}(x,y)\partial_1 K^{H'}(z,t)
\]
and by
\[
l_0^{H'}(x,y,z) := \partial_1 K^{H'}(x,y)\partial_1 K^{H'}(x,z) = l^{H'}(x,y,x,t)
l_1^{H'}(x,y,z) := \partial_1 K^{H'}(x,z)\partial_1 K^{H'}(y,z) = l^{H'}(x,z,y,z).
\]

Using the relations (19) and (20) we get
\[
((f_{i,N} \otimes_1 f_{i,N}) \otimes_1 f_{i,N})(y_1,y_2) = d(H)^3 a(H)^2 \int_{I_{[y_1,y_2]}}^2 du_1 du_2 du_3 l^{H'}(u_1,y_1,u_3,y_2) ||u_1 - u_2||u_2 - u_3||^{2H'-2}
\]
\[
:= b_{i,N}^{(1)}(y_1,y_2) + b_{i,N}^{(2)}(y_1,y_2)
\]

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with
\[ b^{(1)}_{i,N}(y_1, y_2) = d(H)^3 a(H)^2 \int_{[y_1, y_2]} d\nu_1 d\nu_2 d\nu_3 (u_1, y_1, y_2) \int_{[1]} \int_{[1]} d\nu_1 d\nu_2 d\nu_3 \left( u_1 - u_2 \right) \left| u_1 - u_2 \right| dH - 2 \]

and
\[ b^{(2)}_{i,N}(y_1, y_2) = d(H)^3 a(H)^2 \times \left[ 1 \int_{[y_1, y_2]} d\nu_1 d\nu_2 d\nu_3 (u_1, y_1, y_2) \int_{[1]} \int_{[1]} d\nu_1 d\nu_2 d\nu_3 \left( u_1 - u_2 \right) \left| u_1 - u_2 \right| dH - 2 \]

We show that the \( I_2(\sum_{i=0}^{N-1} b^{(2)}_{i,N}) \) converges to zero in \( L^2(\Omega) \) and it has no contribution to the limit. Indeed,
\[
\begin{align*}
E \left( I_2(\sum_{i=0}^{N-1} b^{(2)}_{i,N}) \right)^2 &\leq 2d(H)^6 a(H)^4 N^{4H} \sum_{i=0}^{N-1} \int_{[y_1, y_2]} d\nu_1 d\nu_2 d\nu_3 \left( u_1 - u_2 \right) \left| u_1 - u_2 \right| dH - 2 \\
&\times \left[ \int_{[y_1, y_2]} d\nu_1 d\nu_2 d\nu_3 (u_1, y_1, y_2) \int_{[1]} \int_{[1]} d\nu_1 d\nu_2 d\nu_3 \left( u_1 - u_2 \right) \left| u_1 - u_2 \right| dH - 2 \right] \\
&\leq 2d(H)^6 a(H)^4 N^{4H} \sum_{i=0}^{N-1} \int_{[y_1, y_2]} d\nu_1 d\nu_2 d\nu_3 d\nu_4 \left( u_1 - u_2 \right) \left| u_1 - u_2 \right| dH - 2 \\
&\times \left[ \int_{[y_1, y_2]} d\nu_1 d\nu_2 d\nu_3 \left( u_1 - u_2 \right) \left| u_1 - u_2 \right| dH - 2 \right] \\
&\leq 2d(H)^6 a(H)^4 N^{4H} \sum_{i=0}^{N-1} \int_{[y_1, y_2]} d\nu_1 d\nu_2 d\nu_3 d\nu_4 \left( u_1 - u_2 \right) \left| u_1 - u_2 \right| dH - 2 \\
&\times \left[ \int_{[y_1, y_2]} d\nu_1 d\nu_2 d\nu_3 \left( u_1 - u_2 \right) \left| u_1 - u_2 \right| dH - 2 \right] \\
&\leq 2d(H)^6 a(H)^4 N^{4H} N N^{-6} N^{12-12H} \left( \int_{[0,1]} d\nu_1 d\nu_2 d\nu_3 d\nu_4 \left( u_1 - u_2 \right) \left| u_1 - u_2 \right| dH - 2 \right) \\
&\leq 2d(H)^6 a(H)^4 N^{1-2H}.
\end{align*}
\]
Combining with the fact that $H > \frac{1}{2}$, we conclude that
\[ I_2(N^{2H} \sum_{i=0}^{N-1} b^{(2)}_{i,N}) \xrightarrow{N \to \infty} 0 \quad \text{in } L^2(\Omega), \] (33)

and then we need to find the limit of
\[
N^{2H} \sum_{i=0}^{N-1} b^{(1)}_{i,N} = d(H)^3 a(H)^2 N^{2H} \sum_{i=0}^{N-1} 1_{[0,\frac{1}{N}]}(y_1, y_2)
\times \int_{R^3} du_1 du_2 du_3 h'(u_1, y_1, u_3, y_2) \left[ |u_1 - u_2| |u_2 - u_3| \right]^{2H'-2}
= d(H)^3 (a(H))^2 \sum_{i=0}^{N-1} 1_{[0,\frac{1}{N}]}(y_1, y_2) N^{-1}
\times \int_{[0,1]^3} dv_1 dv_2 dv_3 h' \left( \frac{v_1 + i}{N}, y_1, \frac{v_3 + i}{N}, y_2 \right) \left[ |v_1 - v_2| |v_2 - v_3| \right]^{2H'-2}.
\]

The last sequence has the same limit pointwise (for every $y_1, y_2$) as
\[
d(H)^3 a(H)^2 \int_{[0,1]^3} dv_1 dv_2 dv_3 |v_1 - v_2| |v_2 - v_3|^{2H'-2}
\times \sum_{i=0}^{N-1} 1_{[0,\frac{1}{N}]}(y_1, y_2) N^{-1} \int_{\{0,1\}^3} dv_1 dv_2 dv_3 h' \left( \frac{v_1 + i}{N}, y_1, \frac{v_3 + i}{N}, y_2 \right).
\]

This last term is a Riemann sum that converges to
\[
d(H)^3 a(H)^2 \int_{[0,1]^3} dv_1 dv_2 dv_3 |v_1 - v_2| |v_2 - v_3|^{2H'-2} \int_{y_1 \vee y_2} dx h'(x, y_1, x, y_2)
= d(H)^3 a(H)^2 C'(H) \int_{y_1 \vee y_2} dx \partial_1 K^{H'}(x, y_1) \partial_1 K^{H'}(x, y_2) = d(H)^2 a(H)^2 C'(H) L^{(1)}_1(x, y)
\]
where $L^{(1)}_1$ is the standard kernel of the Rosenblatt process (6).

We need a Cauchy sequence argument as in [20] to conclude the proof. That is, we will show that the sequence $N^{2H} \sum_{i=0}^{N-1} b_{i,N}^{(1)}$ (or equivalently $N^{2H} \sum_{i=0}^{N-1} g^{(2)}_{i,N}$) is Cauchy in the Hilbert space $L^2([0,1]^2)$. This will imply that the sequence of random variable $I_2 \left( N^{2H} \sum_{i=0}^{N-1} g^{(2)}_{i,N} \right)$ is Cauchy, so convergent, in the space $L^2(\Omega)$ and it is easy to deduce that its limit coincides
with the multiple integral of the pointwise limit of the kernel. We compute, for \( M, N \geq 1 \)

\[
\left\| N^{2H} \sum_{i=0}^{N-1} b_i^{(1)} - M^{2H} \sum_{i=0}^{M-1} b_i^{(1)} \right\|_{L^2([0,1]^2)}^2
\]

\[
= \big(H\big)^6 a(H)^4 \left[ N^{4H} \sum_{i,j=0}^{N-1} \int_{I_i^2} \int_{I_j^2} du_1 du_2 du_3 du_4 du_5 du_6 \left[ |u_1 - u_2| |u_2 - u_3| |u_1' - u_2' - u_3' - u_4'| \right]^{2H' - 2}
\]

\[
\times \int_0^{u_1 \wedge u_1'} \int_0^{u_2 \wedge u_2'} \int_0^{u_3 \wedge u_3'} dy_1 dy_2 H'(u_1, y_1, u_3, y_2) t^{H'}(u_1', y_1, u_3', y_2)
\]

\[
+ M^{4H} \sum_{i,j=0}^{M-1} \int_{I_i^2} \int_{I_j^2} du_1 du_2 du_3 du_4 du_5 du_6 \left[ |u_1 - u_2| |u_2 - u_3| |u_1' - u_2' - u_3' - u_4'| \right]^{2H' - 2}
\]

\[
\times \int_0^{u_1 \wedge u_1'} \int_0^{u_2 \wedge u_2'} \int_0^{u_3 \wedge u_3'} dy_1 dy_2 H'(u_1, y_1, u_3, y_2) t^{H'}(u_1', y_1, u_3', y_2)
\]

\[
- 2N^{2H} M^{2H} \sum_{i=0}^{N-1} \sum_{j=0}^{M-1} \int_{I_i^2} \int_{I_j^2} du_1 du_2 du_3 du_4 du_5 du_6 \left[ |u_1 - u_2| |u_2 - u_3| |u_1' - u_2' - u_3' - u_4'| \right]^{2H' - 2}
\]

\[
\times \int_0^{u_1 \wedge u_1'} \int_0^{u_2 \wedge u_2'} \int_0^{u_3 \wedge u_3'} dy_1 dy_2 H'(u_1, y_1, u_3, y_2) t^{H'}(u_1', y_1, u_3', y_2)
\]

\[
= d(H)^6 a(H)^4 \left[ N^{4H} \sum_{i=0}^{N-1} \int_{I_i^2} \int_{I_i^2} du_1 du_2 du_3 du_4 du_5 du_6 \left[ |u_1 - u_2| |u_2 - u_3| |u_1' - u_2' - u_3' - u_4'| \right]^{2H' - 2}
\]

\[
\times \left[ |u_1 - u_2| |u_2 - u_3| |u_1' - u_2' - u_3' - u_4'| \right]^{2H' - 2}
\]

\[
+ M^{4H} \sum_{i,j=0}^{M-1} \int_{I_i^2} \int_{I_j^2} du_1 du_2 du_3 du_4 du_5 du_6 \left[ |u_1 - u_2| |u_2 - u_3| |u_1' - u_2' - u_3' - u_4'| \right]^{2H' - 2}
\]

\[
\times \left[ |u_1 - u_2| |u_2 - u_3| |u_1' - u_2' - u_3' - u_4'| \right]^{2H' - 2}
\]

\[
- 2N^{2H} M^{2H} \sum_{i=0}^{N-1} \sum_{j=0}^{M-1} \int_{I_i^2} \int_{I_j^2} du_1 du_2 du_3 du_4 du_5 du_6 \left[ |u_1 - u_2| |u_2 - u_3| |u_1' - u_2' - u_3' - u_4'| \right]^{2H' - 2}
\]

\[
\times \left[ |u_1 - u_2| |u_2 - u_3| |u_1' - u_2' - u_3' - u_4'| \right]^{2H' - 2}
\]

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and this equal to
\[
d(H)^6 a(H)^4 \left[ N^{-2H} \sum_{i,j=0}^{N-1} \int_{[0,1]^6} du_1 du_2 du_3 \sum_{i,j=0}^{N-1} \int_{[0,1]^6} du_1' du_2' du_3' \right. \\
\times \left. \left[ |v_1 - v_2| |v_2 - v_3| |v_1' - v_2'| |v_2' - v_3'| v_1 - v_1' + i - j |v_3 - v_3' + i - j | \right]^{2H'-2} \right. \\
\times \left. \left[ |v_1 - v_2| |v_2 - v_3| |v_1' - v_2'| |v_2' - v_3'| v_1 - v_1' + i - j |v_3 - v_3' + i - j | \right]^{2H'-2} \right. \\
- 2N^{-1} M^{-1} \sum_{i=0}^{N-1} \sum_{j=0}^{M-1} \int_{[0,1]^6} du_1' du_2' du_3' \\
\times \left[ |v_1 - v_2| |v_2 - v_3| |v_1' - v_2'| |v_2' - v_3'| v_1 - v_1' + i - j |v_3 - v_3' + i - j | \right]^{2H'-2} \right]
\]

The same way as in above this las term when \( N \to \infty \) and \( N \to \infty \) converges to
\[
d(H)^6 a(H)^4 \left[ 2 \int_0^1 (1-x) x^{2H-2} dx + 2 \int_0^1 (1-x) x^{2H-2} dy + 2 \int_0^1 \int_0^1 |x-y|^{2H-2} dxdy \right]
= \frac{1}{H(2H-1)} + \frac{1}{H(2H-1)} - \frac{2}{H(2H-1)} = 0.
\]

We obtained that \( \{ N^{2H} \sum_{i=0}^{N-1} b^{(1)}_{i,N}, N \geq 0 \} \) is a Cauchy sequence and this completes the proof. ■

References


