

# Malliavin calculus and normal approximation

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# Malliavin Calculus

- Paul Malliavin (1925-2010) introduced in the 70's a calculus of variations with respect to the trajectories of Brownian motion.
- The purpose of this calculus was to provide a probabilistic proof of Hörmander's hypoellipticity theorem.



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- The main application of this calculus is to show the existence and smoothness of densities of functionals of Gaussian processes.
- In this talk we will present some recent applications of the Malliavin calculus, combined with Stein's method, to **normal approximations** (Nourdin-Peccati '12 : *Normal Approximations with Malliavin Calculus*).

# Multiple stochastic integrals

- $H$  is a separable Hilbert space.
- $\mathcal{H}_1 = \{X(h), h \in H\}$  is a Gaussian family of random variables in  $(\Omega, \mathcal{F}, P)$  with zero mean and covariance

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- For  $q \geq 2$  we define the  $q$ th *Wiener chaos* as

$$\mathcal{H}_q = \overline{\text{Span}\{h_q(X(h)), h \in H, \|h\|_H = 1\}},$$

where  $h_q(x)$  is the  $q$ th Hermite polynomial.

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- *Multiple stochastic integral* of order  $q$  :

$$I_q : \left( H^{\hat{\otimes} q}, \sqrt{q!} \|\cdot\|_{H^{\otimes q}} \right) \rightarrow \mathcal{H}_q$$

is a linear isometry defined by  $I_q(h^{\otimes q}) = h_q(X(h))$ , where  $H^{\hat{\otimes} q}$  is the  $q$ th *symmetric tensor product* of  $H$ .



*Example :*

Let  $B = \{B_t, t \in [0, 1]\}$  be a Brownian motion.

- Then,  $H = L^2([0, 1])$  and  $X(h) = \int_0^1 h_t dB_t$ .
- For any  $q \geq 2$ ,  $H^{\hat{\otimes} q} = L^2_{\text{sym}}([0, 1]^q)$  and  $I_q$  is the iterated Itô stochastic integral :

$$I_q(h) = q! \int_0^1 \dots \int_0^{t_2} h(t_1, \dots, t_q) dB_{t_1} \dots dB_{t_q}.$$

# Wiener chaos expansion

Assume  $\mathcal{F}$  is generated by  $\mathcal{H}_1$ . We have the orthogonal decomposition

$$L^2(\Omega) = \bigoplus_{q=0}^{\infty} \mathcal{H}_q,$$

where  $\mathcal{H}_0 = \mathbb{R}$ . Any  $F \in L^2(\Omega)$  can be written as

$$F = E(F) + \sum_{q=1}^{\infty} I_q(f_q),$$

where  $f_q \in H^{\hat{\otimes} q}$  are determined by  $F$ .

# Elements of Malliavin Calculus

- $\mathcal{S}$  is the space of random variables of the form

$$F = f(X(h_1), \dots, X(h_n)),$$

where  $h_i \in H$  and  $f \in C_b^\infty(\mathbb{R}^n)$ .

- If  $F \in \mathcal{S}$  we define its *derivative* by

$$DF = \sum_{i=1}^n \frac{\partial f}{\partial X_i}(X(h_1), \dots, X(h_n)) h_i.$$

$DF$  is a random variable with values in  $H$ .

- $\mathbb{D}^{1,2} \subset L^2(\Omega; H)$  is the closure of  $\mathcal{S}$  with respect to the norm

$$\|DF\|_{1,2} = \sqrt{E(F^2) + E(\|DF\|_H^2)}.$$

- The adjoint of  $D$  is the *divergence* operator  $\delta$  defined by the duality relationship

$$E(\langle DF, u \rangle_H) = E(F\delta(u))$$

for any  $F \in \mathbb{D}^{1,2}$  and  $u \in \text{Dom}\delta \subset L^2(\Omega; H)$ .

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- Basic formula

$$\delta(DF) = -LF,$$

where  $L$  is the generator of the *Ornstein-Uhlenbeck* semigroup defined by

$$LF = - \sum_{q=1}^{\infty} q I_q(f_q)$$

if  $F = \sum_{q=0}^{\infty} I_q(f_q)$  and  $\sum_{q=1}^{\infty} q^2 q! \|f_q\|_{H^{\otimes q}}^2 < \infty$ .

# Integration-by-parts formula

Let  $F \in \mathbb{D}^{1,2}$  with  $E(F) = 0$  and  $f \in C_b^1(\mathbb{R})$ . Using that

$$F = LL^{-1}F = -\delta(DL^{-1}F)$$

yields

$$\begin{aligned} E[f(F)F] &= -E[f(F)\delta(DL^{-1}F)] \\ &= -E[\langle D(f(F)), DL^{-1}F \rangle_H] \\ &= E[f'(F)\langle DF, -DL^{-1}F \rangle_H]. \end{aligned}$$

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- If  $F \in \mathcal{H}_q$ , with  $q \geq 1$ , then  $DL^{-1}F = -\frac{1}{q}DF$  and

$$E[f(F)F] = \frac{1}{q}E[f'(F)\|DF\|_H^2].$$

# Stein's method for normal approximation

- *Stein's lemma* :

$$Z \sim N(0, 1) \quad \Leftrightarrow \quad E[f(Z)Z - f'(Z)] = 0 \quad \forall f \in C_b^1(\mathbb{R}).$$



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- Let  $Z \sim N(0, 1)$ , and fix  $h$  such that  $E(|h(Z)|) < \infty$ . The Stein's equation associated with  $h$  is

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## Proposition

*The unique solution to Stein's equation satisfying  $\lim_{x \rightarrow \pm\infty} e^{-x^2/2} f_h(x) = 0$  is*

$$f_h(x) = e^{x^2/2} \int_{-\infty}^x (h(y) - E[h(Z)]) e^{-y^2/2} dy.$$

- Substituting  $x$  by a random variable  $F$  and taking the expectation we obtain

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- If  $\|h\|_\infty \leq 1$ , then  $\|f_h\|_\infty \leq \sqrt{\pi/2}$  and  $\|f'_h\|_\infty \leq 2$ .
- So, for any random variable  $F$ , taking  $h = \mathbf{1}_B$ ,

$$\begin{aligned} d_{TV}(F, Z) &= \sup_{B \in \mathcal{B}(\mathbb{R})} |P(F \in B) - P(Z \in B)| \\ &\leq \sup_{f \in \mathcal{C}_{TV}} |E[f'(F) - Ff(F)]|, \end{aligned}$$

where  $\mathcal{C}_{TV}$  is the class of functions with  $\|f\|_\infty \leq \sqrt{\pi/2}$  and  $\|f'\|_\infty \leq 2$ .

- If  $F \in \mathcal{H}_q$  for some  $q \geq 2$  and  $E(F^2) = 1$ , then

$$\begin{aligned}
 d_{TV}(F, Z) &\leq \sup_{f \in \mathcal{C}_{TV}} |E[f'(F) - Ff(F)]| \\
 &= \sup_{f \in \mathcal{C}_{TV}} \left| E \left[ f'(F) \left( 1 - \frac{1}{q} \|DF\|_H^2 \right) \right] \right| \\
 &\leq \frac{2}{q} \sqrt{\text{Var}(\|DF\|_H^2)},
 \end{aligned}$$

because  $E[\|DF\|_H^2] = q$ .

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because  $E[\|DF\|_H^2] = q$ .

- Moreover, using Wiener chaos expansions and product formulas for multiple stochastic integrals

$$\text{Var}(\|DF\|_H^2) \leq \frac{(q-1)q}{3} (E(F^4) - 3) \leq (q-1) \text{Var}(\|DF\|_H^2).$$

# Fourth Moment theorem

Stein's method combined with Malliavin calculus leads to a simple proof of the Fourth Moment theorem :

## Theorem (N.-Peccati '05, N.-Ortiz '07)

Fix  $q \geq 2$ . Let  $F_n = I_q(f_n) \in \mathcal{H}_q$ ,  $n \geq 1$  be such that

$$\lim_{n \rightarrow \infty} E(F_n^2) = \sigma^2.$$

The following conditions are equivalent :

- (i)  $F_n \Rightarrow N(0, \sigma^2)$ , as  $n \rightarrow \infty$ .
- (ii)  $E(F_n^4) \rightarrow 3\sigma^4$ , as  $n \rightarrow \infty$ .
- (iii) For all  $1 \leq r \leq q - 1$ ,  $\|f_n \otimes_r f_n\|_{H^{\otimes(2q-2r)}} \rightarrow 0$ , as  $n \rightarrow \infty$ .
- (iv)  $\|DF_n\|_H^2 \rightarrow q\sigma^2$  in  $L^2(\Omega)$ , as  $n \rightarrow \infty$ .

- $f_n \otimes_r f_n$  denotes the contraction of  $r$  coordinates.

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- The convergence is in **total variation**. Nourdin-Peccati '13 proved the following optimal version of the fourth moment theorem (for  $\sigma = 1$ ) :

$$c\mathbf{M}(F_n) \leq d_{TV}(F_n, Z) \leq C\mathbf{M}(F_n),$$

where  $\mathbf{M}(F_n) = \max(|E[F_n^3]|, E[F_n^4] - 3)$ .

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where  $\mathbf{M}(F_n) = \max(|E[F_n^3]|, E[F_n^4] - 3)$ .

- Peccati-Tudor '05 obtained a multidimensional extension, which can also be derived by Stein's method and Malliavin calculus.

- (i) Central limit theorem for the renormalized self-intersection local time of the  $d$ -dimensional fractional Brownian motion with Hurst parameter  $H \in [\frac{3}{2d}, \frac{3}{4})$  (Hu-N. '05).

# Applications

- (i) Central limit theorem for the renormalized self-intersection local time of the  $d$ -dimensional fractional Brownian motion with Hurst parameter  $H \in [\frac{3}{2d}, \frac{3}{4})$  (Hu-N. '05).
- (ii) Exact Berry-Esséen asymptotics for functionals of Gaussian processes (Nourdin-Peccati '10) :

$$[P(F_n \leq z) - P(Z \leq z)] \sim \varphi(n) \frac{\rho}{3q} \Phi^{(3)}(z),$$

as  $n \rightarrow \infty$ , where  $F_n \in \mathcal{H}_q$ ,  $E(F_n^2) \rightarrow 1$ ,  $\varphi(n) = \sqrt{E \left[ \left( 1 - \frac{1}{q} \|DF_n\|_H^2 \right)^2 \right]}$ ,  
 $\rho = \lim_n E(F_n \|DF_n\|_H^2)$ , and  $\Phi(z) = P(Z \leq z)$ .

- (iii) Quantitative Breuer-Major theorems for functionals of Gaussian stationary sequences (Nourdin-Peccati '12, Nourdin-Peccati-Podolskij '12, ...).

### Theorem (Brauer-Major '73)

Let  $f \in L^2(\mathbb{R}, \gamma)$ , where  $\gamma = N(0, 1)$ , with Hermite rank  $d$ , that is,

$$f(x) = \sum_{q=d}^{\infty} a_q h_q(x),$$

and  $a_d \neq 0$ . Let  $X = \{X_k, k \in \mathbb{Z}\}$  be a centered Gaussian stationary sequence with unit variance. Set  $\rho(v) = E[X_0 X_v]$  for  $v \in \mathbb{Z}$  and assume  $\sum_{v \in \mathbb{Z}} |\rho(v)|^d < \infty$ . Then,

$$V_n := \frac{1}{\sqrt{n}} \sum_{k=1}^n f(X_k) \Rightarrow N(0, \sigma^2),$$

as  $n \rightarrow \infty$ , where  $\sigma^2 = \sum_{q=d}^{\infty} q! a_q^2 \sum_{v \in \mathbb{Z}} \rho(v)^q$ .

### Sketch of the proof :

- We reduce the proof to the case  $f = a_q h_q$ ,  $q \geq d$ .
- $E[V_n^2] \rightarrow \sigma^2$ .
- Let  $H$  be the closure of  $\{(b_j, j \in \mathbb{Z})\}$  by the scalar product  $\langle b, c \rangle_H = \sum_{i, j \in \mathbb{Z}} b_i c_j \rho(i - j)$ , and assume that  $X_k = X(e_k)$ , with  $e_k = (\delta_{kj}, j \in \mathbb{Z})$ .
- It suffices to show that

$$\|DF\|_H^2 \rightarrow q q! a_q^2 \sum_{v \in \mathbb{Z}} \rho(v)^q$$

in  $L^2(\Omega)$ .

- We have

$$\|DF\|_H^2 = \frac{a_q^2}{n} \sum_{i,j=1}^n h'_q(X_i)h'_q(X_j)\rho(i-j),$$

which has the same limit in  $L^2$  as the sequence

$$B_n := \frac{a_q^2}{n} \sum_{j=1}^n h'_q(X_j) \left( \sum_{m=-\infty}^{\infty} h'_q(X_{j+m})\rho(m) \right).$$

- The sequence

$$\left\{ h'_q(X_j) \left( \sum_{m=-\infty}^{\infty} h'_q(X_{j+m})\rho(m) \right), j \geq 1 \right\}$$

is strictly stationary and ergodic. By the *Ergodic Theorem*, converges in  $L^2(\Omega)$  to its expectation which is equal to

$$a_q! a_q^2 \sum_{v \in \mathbb{Z}} \rho(v)^q.$$

# Convergence in law on a finite sum of Wiener chaos

- Convergence in law is metrizable by the Fortet-Mourier distance :

$$d_{FM}(F, G) = \sup_{\varphi} |E[\varphi(F)] - E[\varphi(G)]|,$$

where the supremum is over  $\|\varphi\|_{Lip} \leq 1$  and  $\|\varphi\|_{\infty} \leq 1$ .



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## Theorem (Nourdin-Poly '12)

Let  $F_n \in \bigoplus_{k=1}^p \mathcal{H}_k$ ,  $F \Rightarrow F_{\infty}$ , and  $F_{\infty}$  is not identically zero. Then

$$d_{TV}(F_n, F_{\infty}) \leq c d_{FM}(F_n, F_{\infty})^{\frac{1}{2p+1}}.$$

A multidimensional extension :

### Theorem (Nourdin-N.-Poly '13)

Let  $F_n = (F_{1,n}, \dots, F_{d,n})$  be such that  $F_{i,n} \in \bigoplus_{k=1}^p \mathcal{H}_k$ ,  $F \Rightarrow F_\infty$ , and

$$E[\det \Gamma_n] \geq \beta > 0, \quad (1)$$

where  $\Gamma_n^{i,j} = \langle DF_{i,n}, DF_{j,n} \rangle_H$ . Then

$$d_{TV}(F_n, F_\infty) \leq cd_{FM}(F_n, F_\infty)^\gamma,$$

for any  $\gamma < [(d+1)(4d(q-1)+3)+1]^{-1}$ .

*Sketch of the proof :*

(i) Let  $\varphi : \mathbb{R}^d \rightarrow \mathbb{R}$  with  $\|\varphi\|_\infty \leq 1$ . Then,

$$\begin{aligned} |E[\varphi(F_n) - \varphi(F_m)]| &\leq |E[\varphi * \rho_\alpha(F_n) - \varphi * \rho_\alpha(F_m)]| \\ &\quad + 2 \sup_n |E[\varphi(F_n) - \varphi * \rho_\alpha(F_n)]| \\ &\leq \frac{1}{\alpha} d_{FM}(F_n, F_m) + 2R_\alpha. \end{aligned}$$

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(ii) Let  $h_\alpha = \varphi - \varphi * \rho_\alpha$ . For any  $\epsilon > 0$ ,

$$\begin{aligned} |E[h_\alpha(F_n)]| &= \left| E \left[ h_\alpha(F_n) \left( \frac{\epsilon}{\det \Gamma_n + \epsilon} + \frac{\det \Gamma_n}{\det \Gamma_n + \epsilon} \right) \right] \right| \\ &\leq 2\epsilon E[(\det \Gamma_n + \epsilon)^{-1}] + \left| E \left[ h_\alpha(F_n) \frac{\det \Gamma_n}{\det \Gamma_n + \epsilon} \right] \right|. \end{aligned}$$

(iii) For the first term we obtain

$$2\epsilon E[(\det \Gamma_n + \epsilon)^{-1}] \leq c\epsilon^{\frac{1}{2(q-1)d+1}}.$$

This follows from  $E[\det \Gamma_n] \geq \beta$  and the Carbery-Wright '01 inequality :

### Lemma

*For any polynomial  $Q$  of degree at most  $d$  we have*

$$E[Q(X)^{\frac{q}{d}}]^{\frac{1}{q}} P(|Q(X)| \leq \alpha) \leq Cq\alpha^{\frac{1}{d}},$$

*where  $X$  is a standard Gaussian vector.*

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*where  $X$  is a standard Gaussian vector.*

(iv) For the second term, if  $|F_n| \leq M$ , using Malliavin calculus we obtain

$$\left| E \left[ h_\alpha(F_n) \frac{\det \Gamma_n}{\det \Gamma_n + \epsilon} \right] \right| \leq c\epsilon^{-2} \alpha^{\frac{1}{d+1}} M^{\frac{d}{d+1}}.$$

(v) We optimize in  $\epsilon$ ,  $\alpha$  and  $M$  to get the result.

*Sufficient conditions for  $E[\det \Gamma_n] \geq \beta > 0$  (assumption (1)) :*

- If  $F_\infty$  is normal  $N_d(0, C)$  with  $\det(C) > 0$ , then (1) holds, because  $\Gamma_n \rightarrow C$  in  $L^2(\Omega)$  (N.-Ortiz '07).
- If  $F_\infty$  has independent and non degenerate components, then (1) holds.
- If  $F_n \rightarrow F_\infty$  in  $L^2(\Omega)$ , then  $E[\det \Gamma_n] \rightarrow E[\det \Gamma_{F_\infty}]$  and (1) holds if

$$E[\det \Gamma_{F_\infty}] > 0.$$

By Kusuoka's '83 theorem, this is equivalent to say that  $F_\infty$  has an absolutely continuous law.



# Convergence of densities

- The total variation distance is equivalent to  $L^1$ -norm of the densities :

$$d_{TV}(F, Z) = \int_{\mathbb{R}} |p_F(x) - \phi(x)| dx,$$

where  $Z \sim N(0, 1)$  and  $\phi$  is the density of  $Z$ .

- Uniform convergence, however, requires stronger hypotheses :

## Theorem (Hu-Lu-N. '13)

Let  $F \in \mathcal{H}_q$ ,  $q \geq 2$ , be such that  $E(F^2) = 1$  and  $E(\|DF\|_H^{-6}) \leq M$ . Then,

$$\sup_{x \in \mathbb{R}} |p_F(x) - \phi(x)| \leq C_{M,q} \sqrt{E(F^4) - 3}.$$

- Using the notion of *Fisher information*, Nourdin and N., provided an alternative proof of this theorem under the weaker assumption  $E(\|DF\|_H^{-4-\epsilon}) \leq M$  for some  $\epsilon > 0$ .

## Sketch of the proof :

(i) Formula for the density

$$\begin{aligned} p_F(x) &= E \left[ \mathbf{1}_{\{F > x\}} \delta \left( \frac{DF}{\|DF\|_H^2} \right) \right] \\ &= E \left[ \mathbf{1}_{\{F > x\}} \frac{qF}{\|DF\|_H^2} \right] - E[\mathbf{1}_{\{F > x\}} \langle DF, D(\|DF\|_H^{-2}) \rangle_H] \\ &= E[\mathbf{1}_{\{F > x\}} F] + E[q\|DF\|_H^{-2} - 1] - E[\mathbf{1}_{\{F > x\}} \langle DF, D(\|DF\|_H^{-2}) \rangle_H]. \end{aligned}$$

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(ii) The terms  $E[|q\|DF\|_H^{-2} - 1|]$  and  $E[|\langle DF, D(\|DF\|_H^{-2}) \rangle_H|]$  can be estimated by a constant times  $\sqrt{E(F^4) - 3}$ .

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(iii) Taking into account that

$$\phi(x) = E[\mathbf{1}_{\{Z > x\}} Z],$$

where  $Z \sim N(0, 1)$ , it suffices to estimate the difference

$$E[\mathbf{1}_{\{F > x\}} F] - E[\mathbf{1}_{\{Z > x\}} Z],$$

which can be done by Stein's method and Malliavin calculus.

# Example 1

- Let  $q = 2$  and

$$F = \sum_{i=1}^{\infty} \lambda_i (X(e_i)^2 - 1),$$

where  $\{e_i, i \geq 1\}$  is a complete orthonormal system in  $H$  and  $\lambda_i$  is a decreasing sequence of positive numbers such that  $\sum_{i=1}^{\infty} \lambda_i^2 < \infty$ . Suppose  $E[F^2] = 1$ .

- Then, if  $\lambda_N \neq 0$  for some  $N > 4$ , we obtain

$$\sup_{x \in \mathbb{R}} |p_F(x) - \phi(x)| \leq C_{N, \lambda_N} \sqrt{\sum_{i=1}^{\infty} \lambda_i^4}.$$

## Example 2 (Brauer-Major theorem revisited)

Fix  $q \geq 2$  and consider the sequence

$$V_n = \frac{1}{\sqrt{n}} \sum_{k=1}^n \sum_{j=d}^q a_j h_j(X_k), \quad a_d \neq 0,$$

where  $X = \{X_k, k \in \mathbb{Z}\}$  is a centered Gaussian stationary sequence with unit variance and covariance  $\rho(v)$ .

### Theorem (Hu-N.-Tindel-Xu '14)

Suppose the spectral density of  $X$ ,  $f_\rho$ , satisfies  $\log(f_\rho) \in L^1([-\pi, \pi])$ . Assume  $\sum_{v \in \mathbb{Z}} |\rho(v)|^d < \infty$ . Set  $\sigma^2 := q! a_d^2 \sum_{v \in \mathbb{Z}} \rho(v)^q \in (0, \infty)$ . Then for any  $p \geq 1$ , there exists  $n_0$  such that

$$\sup_{n \geq n_0} E[\|DV_n\|_H^{-p}] < \infty. \quad (2)$$

Therefore, if  $q = d$  and  $F_n = V_n / \sqrt{E[V_n^2]}$ , we have

$$\sup_{x \in \mathbb{R}} |p_{F_n}(x) - \phi(x)| \leq c \sqrt{E[F_n^4]} - 3.$$

### Sketch of the proof :

- From the non-causal representation  $X_k = \sum_{j=0}^{\infty} \psi_j w_{k-j}$ , where  $\{w_k, k \in \mathbb{Z}\}$  is a discrete Gaussian white noise, it follows that

$$\|DV_n\|_H^2 \geq \frac{1}{n} \sum_{m=1}^n \left( \sum_{k=m}^n \sum_{j=d}^q a_j h'_j(X_k) \psi_{k-m} \right)^2 := B_n.$$

- Fix  $N$  and consider a block decomposition  $B_n = \sum_{i=1}^N B_n^i$ , where  $B_n^i$  is the sum of  $\frac{n}{N}$  squares.
- We use the estimate

$$B_n^{-\frac{p}{2}} \leq \prod_{i=1}^N (B_n^i)^{-\frac{p}{2N}}$$

and we can apply the Carbery-Wright inequality to control the expectation of  $(B_n^i)^{-\frac{p}{2N}}$  if  $\frac{p}{2N}$  is small enough.

*Particular case :*

- Let  $B^H$  be a fractional Brownian motion with Hurst parameter  $H \in (0, 1)$  :

$$E(B_t^H B_s^H) = \frac{1}{2} \left( t^{2H} + s^{2H} - |t - s|^{2H} \right).$$

- Set  $\{X_k = B_k^H - B_{k-1}^H, k \geq 1\}$ . In this case,

$$\rho_H(v) = \frac{1}{2} (|v+1|^{2H} + |v-1|^{2H} - 2|v|^{2H}),$$

and the spectral density satisfies  $\log(f_{\rho_H}) \in L^1([-\pi, \pi])$ .

- As a consequence, we obtain the uniform convergence of densities to  $\phi$  for the sequence of Hermite variations  $F_n = V_n/E[V_n^2]$ , where

$$V_n = \frac{1}{\sqrt{n}} \sum_{k=1}^n h_q(n^H \Delta_{k/n} B^H), \quad q \geq 2,$$

for  $0 < H < 1 - \frac{1}{2q}$ , where  $\Delta_{k/n} B^H = B_{k/n}^H - B_{(k-1)/n}^H$ .



- For  $q = 2$  we have

$$V_n = \frac{1}{\sqrt{n}} \sum_{k=1}^n [(n^H \Delta_{k/n} B^H)^2 - 1].$$

- If  $H \in (0, \frac{3}{4})$  and  $F_n = V_n / E[V_n^2]$  we have (Biermé-Bonami-León '11)

$$\sup_{x \in \mathbb{R}} |p_{F_n}(x) - \phi(x)| \leq c \sqrt{E(F_n^4) - 3} \leq c_H \begin{cases} n^{-\frac{1}{2}} & \text{if } H \in (0, \frac{5}{8}) \\ n^{-\frac{1}{2}} (\log n)^{\frac{3}{2}} & \text{if } H = \frac{5}{8} \\ n^{4H-3} & \text{if } H \in (\frac{5}{8}, \frac{3}{4}) \end{cases}$$

# Generalizations

- (i) One can show the uniform approximation of the  $m$ th derivative of  $p_F$  by the corresponding  $m$ th derivative of the Gaussian density  $\phi^{(m)}$  under the stronger assumption  $E(\|DF\|_H^{-\beta}) < \infty$  for some  $\beta > 6m + 6 (\lfloor \frac{m}{2} \rfloor \vee 1)$ .

# Generalizations

- (i) One can show the uniform approximation of the  $m$ th derivative of  $p_F$  by the corresponding  $m$ th derivative of the Gaussian density  $\phi^{(m)}$  under the stronger assumption  $E(\|DF\|_H^{-\beta}) < \infty$  for some  $\beta > 6m + 6$  ( $\lfloor \frac{m}{2} \rfloor \vee 1$ ).
- (ii) Consider a  $d$ -dimensional vector  $F$ , whose components are in a fixed chaos, and such that  $E[(\det \Gamma_F)^{-p}] < \infty$  for all  $p$ , where  $\Gamma_F$  denotes the Malliavin matrix of  $F$ . In this case for any multi-index  $\beta = (\beta_1, \dots, \beta_k)$ ,  $1 \leq \beta_i \leq d$ , one can show

$$\sup_{x \in \mathbb{R}^d} |\partial_\beta p_F(x) - \partial_\beta \phi_d(x)| \leq c \left( \|C - I\|^{\frac{1}{2}} + \sum_{j=1}^d \sqrt{E[F_j^4] - 3(E[F_j^2])^2} \right)$$

where  $C$  is the covariance matrix of  $F$ ,  $\phi_d$  is the standard  $d$ -dimensional normal density, and  $\partial_\beta = \frac{\partial^k}{\partial x_{\beta_1} \cdots \partial x_{\beta_k}}$ .

# Further developments

- Rate of convergence in stable limit theorems when the limit is a mixture of Gaussian distributions ?

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- *Examples* :
  - (i) Fluctuations of the error in approximation schemes for SDE.
  - (ii) Weighted Hermite variations of stationary Gaussian processes.
  - (iv) Central limit theorems for the second and third moments in space of Brownian local time increments.

# Further developments

- Rate of convergence in stable limit theorems when the limit is a mixture of Gaussian distributions ?
- *Examples* :
  - (i) Fluctuations of the error in approximation schemes for SDE.
  - (ii) Weighted Hermite variations of stationary Gaussian processes.
  - (iv) Central limit theorems for the second and third moments in space of Brownian local time increments.
- General results on the rate of convergence have been obtained by Nourdin-N.-Peccati '14 using an interpolation method and Malliavin calculus.

*Example (Weighted quadratic variation of the fBm) :*

$$F_n = \frac{1}{\sqrt{n}} \sum_{k=1}^n f(B_{(k-1)/n}^H) \left[ (n^H \Delta_{k/n} B^H)^2 - 1 \right].$$

### Theorem (Nourdin-N.-Peccati '14)

*Let  $H \in (\frac{1}{4}, \frac{3}{4})$  and  $f \in C^4(\mathbb{R})$  such that  $|f^{(i)}(x)| \leq c_1 \exp(c_2|x|^\beta)$  for some  $\beta < 2$  and for  $i = 0, \dots, 4$ . Let*

$$S = \sqrt{\sigma_H \int_0^1 f^2(B_s^H) ds},$$

*with  $\sigma_H^2 = \sum_{k=-\infty}^{\infty} \rho_H(k)^2$ . Suppose  $E[S^{-\alpha}] < \infty$  for some  $\alpha > 2$ . Then,*

$$|E[\varphi(F_n)] - E[\varphi(S\eta)]| \leq C_{f,H} \max_{1 \leq i \leq 5} \|\varphi^{(i)}\|_\infty n^{-(|2H-\frac{1}{2}| \wedge |2H-\frac{3}{2}|)},$$

*where  $\eta$  is a  $N(0, 1)$  random variable independent of  $B^H$ .*