

# **“Stein’s method and Malliavin calculus on the Poisson space”**

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Based on three joint works

*Stein's method and normal approximation of Poisson functionals*

(Peccati, Solé, Taqqu and Utzet, 2010)

*Multidimensional Gaussian fluctuations on the Poisson space*

(Peccati and Zheng, 2010)

*Invariance principles for homogeneous sums: universality of Wiener chaos*

(Nourdin, Peccati and Reinert, 2010)

## [Overview, I]

In a recent series of papers, Nourdin and Peccati (2009, 2010), Nourdin, Peccati and Réveillac (2010) and Nourdin, Peccati and Reinert (2009) have shown that one can combine **Malliavin calculus** with **Stein's method** on a Gaussian space, in order to obtain **explicit bounds** in the **normal** and **Gamma approximations** of **non-linear functionals** of the underlying Gaussian field. This is related to previous works by Nualart and myself (2005) and Nualart and Ortiz-Latorre (2007).

Several applications, for instance: (i) first ever Berry-Esséen bounds in the **Breuer-Major CLT** for functionals of fractional Brownian motion, (ii) exact Berry-Esséen asymptotics for **continuous-time Toeplitz quadratic forms**, (iii) generalized **second order Poincaré inequalities**, (iv) **Gaussian polymers**, (v) non-asymptotic **confidence intervals** in Hurst estimation.

## [Overview, II]

**In this talk:** we present the results proved by Peccati, Solé, Taqqu and Utzet (2010), showing how to combine **Stein's method** and **Malliavin calculus on a Poisson space** (in the form of Nualart and Vives, 1989), in order to assess the normal approximation of random variables **in the Wasserstein distance**. This refines some findings by Peccati and Taqqu (2007, 2008ab) obtained by using **decoupling techniques**.

**Motivation:** e.g., some CLTs for non-linear functionals of **Lévy moving averages**, like Ornstein-Uhlenbeck Lévy processes. These results have applications to prior specification in **Bayesian survival analysis** (Peccati and Prünster, 2008; de Blasi, Peccati and Prünster, 2009).

## [Overview, III]

In a final twist, we will show that some of our findings are indeed special instances of a very general **universality principle** for homogeneous sums, related to the notion of “**coordinate influence**” and to invariance principles by Rotar’ (1979) and Mossel, O’Donnell and Oleszkiewicz (2010).

## [General framework]

Let  $(A, \mathcal{A}, \mu)$  be a Borel measure space, with  $\mu$  a positive measure,  $\sigma$ -finite, non-atomic and s.t.  $\mu(A) = \infty$ .

We denote by  $\hat{N} = \{ \hat{N}(B) : B \in \mathcal{A} : \mu(B) < \infty \}$  a **centered Poisson random measure** controlled by  $\mu$ , that is,  $\hat{N}$  is a random signed measure such that: **(i)**  $\hat{N}(B)$  has a **centered Poisson distribution** with parameter  $\mu(B)$ , and **(ii)** for every  $B, C$  s.t.  $B \cap C = \emptyset$ ,  $\hat{N}(B)$  and  $\hat{N}(C)$  are **independent**.

We denote by  $L^2(\sigma(\hat{N}))$  the space of all real-valued square-integrable functionals of  $\hat{N}$ .

## [Chaos]

For every  $d \geq 1$ , denote by  $L_s^2(\mu^d)$  the space of (deterministic) symmetric functions on  $A^d$ , that are square-integrable w.r.t.  $\mu^d$ .

The following **Wiener-Itô chaotic decomposition** is well-known: every  $F \in L^2(\sigma(\hat{N}))$  admits a unique representation of the type

$$F = E(F) + \sum_{d=1}^{\infty} I_d(f_d) \quad (\text{convergence in } L^2),$$

where  $f_d \in L_s^2(\mu^d)$ , and  $I_d$  stands for a **multiple Wiener-Itô integral** of order  $d$ .

One can think at the r.v.  $I_d(f_d)$  as a  $d$ -fold iterated integral w.r.t.  $\hat{N}$ , over a region of  $A^d$  “not containing diagonals”.

## [The problem]

Given square-integrable random variables  $F, G$ , we define the **Wasserstein distance**, between the laws of  $F$  and  $G$ , as

$$d_W(F, G) = \sup_{|h'| \leq 1} |E[h(F)] - E[h(G)]|.$$

Let  $Z \sim \mathbf{N}(0, 1)$ .

**Problem.** Given  $F \in L^2(\sigma(\hat{N}))$  such that  $E(F) = 0$ , provide an **estimation** (e.g., an **upper bound**) of the quantity

$$d_W(F, Z)$$

**Idea.** Use **Stein's method** and **Malliavin calculus**.



## [Stein's method in a nutshell, I: Stein's Lemma]

(Ref. : Stein 1972, 1986)

**Stein's Lemma.** *A random variable  $F$  has a  $\mathbf{N}(0, 1)$  distribution if and only if for every smooth function  $g$*

$$E \left[ g'(F) - Fg(F) \right] = 0.$$

**Heuristically,** Stein's Lemma suggests that, if  $F$  is such that

$$E \left[ g'(F) - Fg(F) \right] \simeq 0$$

for a “**sufficiently large**” class of smooth functions  $g$ , then the law of  $F$  must be **close to Gaussian**.

## [Stein's method in a nutshell, II: Stein equations]

**Formally:** fix  $h \in \mathbf{Lip}(1)$ , take  $Z \sim \mathbf{N}(0,1)$ , and introduce the **Stein equation**

$$\boxed{g'(x) - xg(x) = h(x) - E[h(Z)], \quad x \in \mathbf{R}} \quad (\star)$$

(the unknown being  $g$ ).

Classic estimates by Stein (1986) yield that there exists a solution to  $(\star)$ , say  $g_h$ , such that

$$\boxed{|g'_h| \leq 1 \quad \text{and} \quad |g''_h| \leq 2}$$

## [Stein's method in a nutshell, III: Stein's bounds]

The previous results show that, if  $Z \sim \mathbf{N}(0, 1)$ , then

$$\begin{aligned}d_W(F, Z) &= \sup_{|h'| \leq 1} |E[h(F)] - E[h(Z)]| \\ &= \sup_{|h'| \leq 1} |E\{h(F) - E[h(Z)]\}| \\ &= \sup_{|h'| \leq 1} \left| E \left[ g'_h(F) - F g_h(F) \right] \right| \\ &\leq \sup_{\substack{|g'| \leq 1 \\ |g''| \leq 2}} \left| E \left[ g'(F) - F g(F) \right] \right|.\end{aligned}$$

which is known as the **Stein's bound** on the Wasserstein distance.

How to assess a quantity such as  $\sup_g |E[g'(F) - Fg(F)]|$  ?

## [The idea]

Our principal idea is to use **infinite-dimensional integration by parts formulae** (from Malliavin calculus) in order to write, for  $F$  centered and sufficiently smooth,

$$\begin{aligned} E [Fg(F)] &= E [g'(F) \times H(F)] + R_0(g; F) \\ |R_0(g; F)| &\leq R(F), \end{aligned}$$

where  $H$  and  $R$  **do not depend** on  $g$ . In this way, for  $Z \sim \mathbf{N}(0, 1)$ ,

$$d_W(F, Z) \leq \sup_{\substack{|g'| \leq 1 \\ |g''| \leq 2}} \left| E [g'(F) - Fg(F)] \right| \leq E |1 - H(F)| + R(F),$$

i.e.

$$d_W(F, Z) \leq E |1 - H(F)| + R(F) \leq \sqrt{E |1 - H(F)|^2} + R(F).$$

To explicitly write  $H$  and  $R$ , one must introduce **Malliavin operators**.

## [Derivatives and Integrals]

We write  $L^2(\mu \times P) = \{u(t, \omega) : E \int_A u(t)^2 \mu(dt) < \infty\}$ .

The **derivative operator**  $D$  is such that  $\text{dom}(D) \subset L^2(\sigma(\hat{N}))$ . For  $F = E(F) + \sum_{d \geq 1} I_d(f_d) \in \text{dom}(D)$

$$D_t F = \sum_{d=1}^{\infty} d \times I_{d-1}(f_d(t, \cdot)), \quad t \in A.$$

Alternatively,  $D_t F(\omega) = F(\omega + \varepsilon_t) - F(\omega)$  ( $\varepsilon_t = \text{Dirac mass at } t$ ).

We denote by  $\delta$  the **adjoint** of  $D$ , called the **Skorohod integral**. One has that  $\text{dom}(\delta) \subset L^2(\mu \times P)$  and  $\delta : \text{dom}(\delta) \rightarrow L^2(\sigma(\hat{N}))$ . For every  $u \in \text{dom}(\delta)$  and  $F \in \text{dom}(D)$

$$E[\delta(u) F] = E \int_A [u(t) D_t F] \mu(dt)$$

## [Ornstein-Uhlenbeck operators, I]

The **generator of the Ornstein-Uhlenbeck semigroup**, noted  $L$ , is such that  $\text{dom}(L) \subset L^2(\sigma(\hat{N}))$ , for  $F = E(F) + \sum_{d \geq 1} I_d(f_d) \in \text{dom}(L)$

$$LF = - \sum_{d=1}^{\infty} d \times I_d(f_d).$$

The **pseudo-inverse** of  $L$ , noted  $L^{-1}$ , is such that, for every  $F = E(F) + \sum_{d \geq 1} I_d(f_d) \in L^2(\sigma(\hat{N}))$

$$L^{-1}F = - \sum_{d=1}^{\infty} \frac{1}{d} \times I_d(f_d).$$

Plainly,  $L^{-1}F = L^{-1}(F - E(F))$ , and

$$LL^{-1}F = F - E(F).$$

## [Ornstein-Uhlenbeck operators, II]

One fundamental relation is the following

$$\delta D = -L,$$

yielding that, for every centered  $F \in L^2(\sigma(\hat{N}))$

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

## [The crucial computation]

Observe first that, if  $|g'| \leq 1$  and  $|g''| \leq 2$  and  $F \in \mathbf{dom}(D)$ , then (by using the **difference representation** of  $D$ )

$$D_t g(F) = g'(F) D_t F + R(D_t F), \quad |R(y)| \leq y^2.$$

It follows that, if  $F \in \mathbf{dom}(D)$  is also centered,

$$\begin{aligned} E[Fg(F)] &= E\left(\delta\left(-DL^{-1}F\right)g(F)\right) \\ &= E\int_A \left[-D_t L^{-1}F \times D_t g(F)\right] \mu(dt) \\ &= E\left[g'(F) \int_A \left[-D_t L^{-1}F \times D_t F\right] \mu(dt)\right] \\ &\quad + E\left[\int_A \left[-D_t L^{-1}F \times R(D_t F)\right] \mu(dt)\right] \end{aligned}$$



## [A general bound]

If  $Z \sim \mathbf{N}(0, 1)$ , then

$$\begin{aligned} d_W(F, Z) &\leq \sup_{|g'| \leq 1; |g''| \leq 2} \left| E \left[ g'(F) - Fg(F) \right] \right| \\ &\leq E \left| \mathbf{1} - \left\langle DF, -DL^{-1}F \right\rangle_\mu \right| + E \int_A [D_t F]^2 |D_t L^{-1}F| \mu(dt). \end{aligned}$$

If  $F = I_d(f)$ , then  $L^{-1}F = -\frac{1}{d}F$ , and

$$\begin{aligned} \left\langle DF, -DL^{-1}F \right\rangle_\mu &= \frac{1}{d} \|DF\|_\mu^2 \\ \int_A [D_t F]^2 |D_t L^{-1}F| \mu(dt) &= \frac{1}{d} \int_A |D_t F|^3 \mu(dt). \end{aligned}$$

## [First example]

Consider a function  $h \in L^2(\mu)$ , as well as the stochastic integral

$$F = I_1(h) = \int_A h(x) \hat{N}(dx).$$

Observe:  $EF^2 = \|h\|_\mu^2$ . Then, for  $Z \sim \mathbf{N}(0, 1)$ ,

$$d_W(F, Z) \leq \left| 1 - \|h\|_\mu^2 \right| + \int_A |h(x)|^3 \mu(dx).$$

For instance, if  $\{\hat{N}_t : t \geq 0\}$  is a centered Poisson process with intensity 1,

$$d_W(\hat{N}_t/\sqrt{t}, Z) \leq \frac{1}{\sqrt{t}}$$

(not surprising, but consistent with usual Berry-Esséen estimates).

## [The example of double integrals]

Consider a function  $h \in L_s^2(\mu^2)$ , and define the **double stochastic integral**

$$F = I_2(h) = \int_A \int_A h(x, y) \hat{N}(dx) \hat{N}(dy) \quad (\text{no diagonals}).$$

One has  $EF^2 = 2 \|h\|_{\mu^2}^2$ . For  $Z \sim \mathbf{N}(0, 1)$ ,

$$\begin{aligned} d_W(F, Z) \leq & \left| 1 - 2 \|h\|_{\mu^2}^2 \right| + C_1 \sqrt{\int_A \int_A [h *_{\frac{1}{2}} h(x, y)]^2 \mu(dx) \mu(dy)} \\ & + C_2 \sqrt{\int_A [h *_{\frac{1}{2}} h(x)]^2 \mu(dx)} \\ & + C_3 \sqrt{\int_A \int_A h(x, y)^4 \mu(dx) \mu(dy)}, \end{aligned}$$

where  $C_1, C_2, C_3$  are explicit universal constants.

## [Contractions]

Here,

$$h *_{\mathbf{1}}^1 h(x, y) = \int_A h(t, x) h(t, y) \mu(dt),$$

is the usual **contraction** of  $h$  with itself, whereas

$$h *_{\mathbf{2}}^1 h(x) = \int_A h(t, x) h(t, x) \mu(dt)$$

is a **degenerate** contraction, obtained as the restriction of  $h *_{\mathbf{1}}^1 h$  on  $\{x = y\}$ .

From the previous bound, one deduces a **CLT**: if  $E I_2(h_k)^2 \rightarrow 1$ ,  $\|h_k *_{\mathbf{1}}^1 h_k\|_{\mu^2} \rightarrow 0$ ,  $\|h_k *_{\mathbf{2}}^1 h_k\|_{\mu} \rightarrow 0$  and  $\int h_k^4 d\mu^2 \rightarrow 0$ , then

$$I_2(h_k) \xrightarrow{\text{LAW}} Z \sim \mathbf{N}(0, 1).$$

## [Decoupling]

The **same CLT** was obtained (without explicit bounds) by Peccati and Taqqu (2007, 2008ab) by a **decoupling technique**, named the **principle of conditioning (POC** – Jakubowski, 1987). The POC implies that to establish the weak convergence of random variables of the type

$$\sum_{ij} h(i, j) \xi_i \xi_j, \quad \{\xi_i\} \text{ independent}$$

it is sufficient to consider

$$\sum_{ij} h(i, j) \xi_i \xi'_j, \quad \{\xi'_i\} \text{ independent copy of } \{\xi_i\}.$$

Therefore, one can study  $I_2(h_k)$  by first considering  $\int_{A^2} h_k d\hat{N} d\hat{N}'$ , where:  $\hat{N}'$  independent copy of  $\hat{N}$ .

## [Higher order integrals]

If

$$F = I_d(h) = \int_{A^d} h(x_1, \dots, x_d) \hat{N}(dx_1) \cdots \hat{N}(dx_d) \quad (\text{no diagonals}),$$

then a similar bounds holds: for  $Z \sim \mathbf{N}(0, 1)$

$$\begin{aligned} d_W(F, Z) \leq & \left| \mathbf{1} - EF^2 \right| + C \times \sum_{r=1}^{d-1} \|h *_{r}^r h\|_{\mu^{2d-2r}} + \\ & + \text{norms of degenerate contractions } (r = 1, \dots, d-1) \\ & + \sqrt{\int_{A^d} h^4 d\mu^d}. \end{aligned}$$

These bounds yield **explicit sufficient conditions for CLTs**. It seems unfeasible to obtain the same CLTs by decoupling.

## [Applications in Bayesian survival analysis]

These results can be naturally applied in order to establish CLTs for objects of the type

$$\int_0^T M_t^p dt, \quad T \rightarrow \infty,$$

where  $M$  is a **Lévy moving average**, that is

$$M_t = \int_0^t k(t-s) dL_s, \quad L = \text{Lévy process.}$$

For instance  $k(t-s) = \sqrt{2\lambda} \exp[-\lambda(t-s)]$  gives the **Ornstein-Uhlenbeck Lévy processes**. **CLTs** of this type are useful in **prior specification in Bayesian non-parametric survival analysis**, where moving averages model **random hazard rates**. **See:** Peccati and Prünster (2008; prior), de Blasi, Peccati and Prünster (2008; posterior)

## [Multidimensional case]

In Peccati and Zheng (2010), one can find multidimensional extensions of these results. These results partially generalize findings known in the Gaussian world, namely that **inside Wiener chaos componentwise convergence to Gaussian implies joint convergence.**

By replacing Stein's method with the so-called "smart path method" (popular in spin glasses), we can deal with covariance matrices that are not necessarily positive definite.

The problem of lifting these findings from a finite-dimensional framework to a functional level is still open.



## [Comparing Gaussian and Poisson chaos, I]

See: Nualart and Peccati (2005; i–iii), Nualart and Ortiz (2007; iv) and Nourdin and Peccati (2009, estimate);  $W =$  Gaussian field.

**Theorem.** Suppose  $F_k = I_d^W(h_k)$ , is such that  $EF_k^2 = 1$ . Then, the following are equivalent:

(i)  $F_k \xrightarrow{\text{LAW}} Z \sim \mathbf{N}(0, 1)$ ;      (ii)  $EF_k^4 \rightarrow 3$ ;

(iii)  $\|h_k *_r^r h_k\| \rightarrow 0$ ,  $r = 1, \dots, d - 1$ ;      (iv)  $\frac{1}{d} \|DF_k\|^2 \rightarrow 1$  in  $L^2$ .

Moreover,

$$d_W(F_k, Z) \leq E \left| 1 - \frac{1}{d} \|DF_k\|^2 \right| \leq \sqrt{\frac{|EF_k^4 - 3|}{3}}.$$

## [Comparing Gaussian and Poisson chaos, II]

Here is what we can deduce from our results on Poisson functionals.

**Theorem.** Suppose  $F_k = I_d^{\hat{N}}(h_k)$ , is such that  $EF_k^2 = 1$ . If **(a)**  $\|h *_{r}^r h\| \rightarrow 0$ ,  $r = 1, \dots, d - 1$ , **(b)** all degenerate contractions converge to zero, and **(c)**  $\int h_k^4 d\mu^d \rightarrow 0$ , then

$$F_k \xrightarrow{\text{LAW}} Z \sim \mathbf{N}(0, 1).$$

Moreover,

$$d_W(F_k, Z) \leq E \left| 1 - \frac{1}{d} \|DF_k\|_{\mu}^2 \right| + \frac{1}{d} \int |D_t F_k|^3 \mu(dt).$$

## [Homogeneous sums]

Let  $\hat{N} = \{\hat{N}(i) : i \geq 1\}$  be a sequence of i.i.d. of centered Poisson r.v.'s with parameter 1 (for simplicity). For  $d \geq 2$ , consider the sequence  **$d$ -homogeneous sum**

$$Q_d(K, \hat{N}) = \sum_{1 \leq i_1, \dots, i_d \leq K} f_K(i_1, \dots, i_d) \hat{N}(i_1) \cdots \hat{N}(i_d), \quad K \geq 1,$$

where  $EQ_d(K, f_K, \hat{N})^2 = 1$  and each  $f_K$  “vanishes on diagonals”.

Let  $G = \{G(i) : i \geq 1\}$  be an i.i.d. centered standard Gaussian sequence, and define analogously

$$Q_d(K, G) = \sum_{1 \leq i_1, \dots, i_d \leq K} f_K(i_1, \dots, i_d) G(i_1) \cdots G(i_d), \quad K \geq 1.$$

## [Bounds for homogeneous sums]

In this framework, our results allow to deduce the (simpler) bounds: for  $Z \sim \mathbf{N}(0, 1)$ ,

$$d_W(Q_d(K, \hat{N}), Z) \leq C_1 \sum_{r=1}^{d-1} \|f_K *_{r}^r f_K\|$$
$$d_W(Q_d(K, G), Z) \leq C_2 \sum_{r=1}^{d-1} \|f_K *_{r}^r f_K\|$$

In view of of the criterion by Nualart and Peccati (2005), one infers that the following implication holds, as  $K \rightarrow \infty$ :

$$\text{if } Q_d(K, G) \xrightarrow{\text{LAW}} Z, \text{ then } Q_d(K, \hat{N}) \xrightarrow{\text{LAW}} Z.$$

Is this an instance of a more general phenomenon?

## [Answer: Universality of the Gaussian Wiener chaos]

The answer is: **Yes**.

**Theorem (Nourdin Peccati Reinert).** *With the above notation and for  $d \geq 2$ , suppose that*

$$Q_d(K, G) \xrightarrow{\text{LAW}} Z \sim \mathbf{N}(0, 1).$$

*Then, for every sequence of independent r.v.'s  $X = \{X(i) : i \geq 1\}$  such that*  
**(a)**  $EX(i) = 0$ , **(b)**  $EX(i)^2 = 1$  and **(c)**  $\sup_i E|X(i)|^{2+\varepsilon}$  (some  $\varepsilon > 0$ )

$$Q_d(K, X) \xrightarrow{\text{LAW}} Z \quad (\text{in the Kolmogorov distance}).$$

*If the  $\{X(i)\}$  are also identically distributed, then the requirement **(c)** can be dropped (at the cost of the Kolmogorov distance).*

**NB:** the conclusion is **false** for  $d = 1$  ! **Also:** multidimensional versions and “Gamma versions”.

## [Influence functions and Lindeberg method]

**Idea of the proof.** Due to a remarkable invariance principle by Rotar' (1979) and Mossel, O'Donnell and Oleszkiewicz (2009), one knows that the “**maximum over influence functions**”

$$\tau_K = \max_i \sum_{a_1, \dots, a_{d-1}} f_K(i, a_1, \dots, a_{d-1})^2$$

**measures the distance** between the laws of  $Q_d(K, G)$  and  $Q_d(K, X)$ . This fact can be established by using the “Lindeberg method” for probabilistic approximations. One also knows that

$$\|f_K *_{d-1}^{d-1} f_K\|^2 = \sum_{i,j} \left[ \sum_{a_1, \dots, a_{d-1}} f_K(i, a_1, \dots, a_{d-1}) f_K(j, a_1, \dots, a_{d-1}) \right]^2 \geq \tau_K.$$

The conclusion follows once again from the previous criterion for CLT on a Wiener chaos.

[1] Peccati, Solé, Taqqu and Utzet (2010). Stein's method and normal approximation of Poisson functionals. *Ann. Probab.*

[2] Peccati and Zheng (2010). Multi-dimensional Gaussian fluctuations on the Poisson space. *Electronic J. Probab.*

[2] Nourdin, Peccati and Reinert (2010). Invariance principles for homogeneous sums: universality of Wiener chaos. *Ann. Probab.*

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[4] Nourdin and Peccati (2010). Stein's method and exact Berry-Esséen asymptotics for functionals of Gaussian fields. *Ann. Probab.*

[6] de Blasi, Peccati and Prünster (2009). Posteriors for random hazards. *Ann. Stat.*

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# Wiener Chaos: Moments, Cumulants and Diagrams

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