

C^∞ convergence of densities on Wiener chaoses

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The Wiener space and related operators

- dimension: n (none of our estimates depend on n)
- n -dimensional Wiener space : $(\mathbb{R}^n, \mathcal{B}(\mathbb{R}^n), \gamma_n)$
- Ornstein-Uhlenbeck operator : $L(f) = \Delta f - \sum_{i=1}^n x_i \partial_i f$
- Square field operator: $\Gamma(f, g) = \nabla f \cdot \nabla g$
- m -th Wiener chaos:

$$\mathcal{W}_m = \text{Ker}(L + m\text{Id}) = \text{Vect} \left(\prod_{i=1}^n H_{k_i}(x_i) \mid \sum_{i=1}^n k_i = m \right).$$

- Wiener orthogonal expansion:

$$L^2(\gamma_n) = \bigoplus_{m=0}^{\infty} \mathcal{W}_m.$$

- Integration by parts for $f \in \text{Dom}(\Gamma), g \in \text{Dom}(L) \subset \text{Dom}(\Gamma)$:


$$\int_{\mathbb{R}^n} \Gamma(f, g) d\gamma_n := - \int_{\mathbb{R}^n} f L(g) d\gamma_n$$

- Rephrasing using expectations:

$$\mathbb{E} \left[\Gamma \left(f(\vec{X}), g(\vec{X}) \right) \right] = -\mathbb{E} \left[f(\vec{X}) L \left(g(\vec{X}) \right) \right],$$

where $X_i : \begin{cases} (\mathbb{R}^n, \mathcal{B}(\mathbb{R}^n), \gamma_n) & \rightarrow \mathbb{R} \\ (x_1, \dots, x_n) & \rightarrow x_i, \end{cases}$ and
 $\vec{X} = (X_1, \dots, X_n) \sim \mathcal{N}(0, \mathbf{I}_n)$.

- $\mathcal{W}_m \subset \mathbb{D}^\infty := \left(\bigcap_{i \geq 1} \text{Dom}(L^i) \right) \cap \left(\bigcap_{p \geq 1} L^p(\gamma_n) \right)$

 we may skip any domain/integrability consideration on Wiener chaoses

State of the art regarding central convergence on Wiener chaoses (univariate)

- (2005) Nualart-Peccati: (F_n) sequence in \mathscr{W}_m

$$F_n \xrightarrow[n \rightarrow \infty]{\text{Law}} \mathcal{N}(0, 1) \Leftrightarrow \mathbb{E}[F_n^2] \rightarrow 1 \ \& \ \mathbb{E}[F_n^4] \rightarrow 3.$$

- (2009) Nourdin-Peccati: $F \in \mathscr{W}_m$ with $\mathbb{E}[F^2] = 1$

$$\begin{aligned} d_{\text{TV}}(F, \mathcal{N}(0, 1)) &\leq 2\sqrt{\text{Var}(\Gamma(F, F))} \\ &\leq \frac{2}{\sqrt{3}}\sqrt{\mathbb{E}[F^4 - 3]}. \end{aligned}$$

State of the art regarding central convergence on Wiener chaoses (multivariate)

- (2005) Peccati-Tudor: $\vec{F}_n := (F_{n,1}, \dots, F_{n,d})$ sequence in $\mathcal{W}_{m_1} \times \dots \times \mathcal{W}_{m_d}$:

$$\vec{F}_n \xrightarrow[n \rightarrow \infty]{\text{Law}} \mathcal{N}(0, \Sigma) \Leftrightarrow \begin{cases} \forall i & F_{n,i} \rightarrow \mathcal{N}(0, \Sigma(i, i)) \\ \forall i, j & \mathbb{E}[F_{n,i} F_{n,j}] \rightarrow \Sigma(i, j) \end{cases}$$

- (2010) Nourdin-Peccati-Reveillac:

$$d_{\text{Wass}}(\vec{F}_n, \mathcal{N}(0, \Sigma)) \leq c_{\Sigma} \sqrt{\sum_{i,j} \mathbb{E} \left[\left(\Sigma(i, j) - \frac{1}{m_j} \Gamma(F_{n,i}, F_{n,j}) \right)^2 \right]}$$

State of the art regarding central convergence on Wiener chaoses (multivariate)

- (2014) Nourdin-Peccati-Swan:
 - $\vec{F} = (F_1, \dots, F_d) \in \mathscr{W}_{m_1} \times \dots \times \mathscr{W}_{m_d}$ with $\text{Cov}(\vec{F}) = \mathbf{I}_d$
 - $\vec{N} = (N_1, \dots, N_d) \sim \mathcal{N}(0, \mathbf{I}_d)$

$$\begin{aligned} D(\vec{F} \parallel \vec{N}) &\leq C_{\vec{m}} \sqrt{\mathbb{E} \left[\|\vec{F}\|^4 - \|\vec{N}\|^4 \right]} \\ &\quad \times \left| \log \left(\mathbb{E} \left[\|\vec{F}\|^4 - \|\vec{N}\|^4 \right] \right) \right| \end{aligned}$$

Can we go beyond entropy?

- (2014) Nualart-Hu: quantitative \mathcal{C}^∞ -convergence of densities if $\frac{1}{\Gamma(F_n, F_n)}$ bounded in every L^p , $p \geq 1$
- (2016) Nourdin-Nualart: quantitative convergence in Fisher information if $\frac{1}{\Gamma(F_n, F_n)^{4+\delta}}$ integrable
- (2016) Nualart-Hu-Tyndel: \mathcal{C}^∞ convergence of densities for

$$\frac{1}{\sqrt{n}} \sum_{k=1}^n P(X_k),$$

with $P \in \mathbb{R}[X]$, $(X_k)_{k \geq 1}$ stationary Gaussian process with spectral density f such that $\log(f) \in L^1$.

👉 Are the above assumptions (in red) necessary or are they 'automatic'?

Our result

👉 The previous assumptions are unnecessary. Central convergence on Wiener chaos always implies bounded negative moments for $\Gamma(F_n, F_n)$ at any order.

Assumptions:

- (F_n) sequence in \mathscr{W}_m
- $F_n \xrightarrow[n \rightarrow \infty]{\text{Law}} \mathcal{N}(0, 1)$

Conclusion: $\forall p \geq 1, \limsup \mathbb{E} \left[\Gamma(F_n, F_n)^{-p} \right] < \infty$

$$\Rightarrow F_n \xrightarrow[n \rightarrow \infty]{\mathcal{C}^\infty} \mathcal{N}(0, 1)$$

$$\Leftrightarrow \forall q \geq 1, \forall \epsilon > 0, \exists \delta_{\epsilon, q, m} > 0 \text{ such that } \forall F \in \mathscr{W}_m:$$

$$\text{dist}(F, \mathcal{N}(0, 1)) \leq \delta_{\epsilon, q, m} \Rightarrow \begin{cases} F \text{ has a } \mathcal{C}^q \text{ density } f \\ \sup_{x \in \mathbb{R}} |f^{(q)}(x) - \gamma^{(q)}(x)| \leq \epsilon \end{cases}$$

Our result in higher dimensions

Assumptions:

- $\vec{F}_n = (F_{n,1}, F_{n,2}, \dots, F_{n,d})$ sequence in $\mathcal{W}_{m_1} \times \dots \times \mathcal{W}_{m_d}$
- $\vec{F}_n \xrightarrow[n \rightarrow \infty]{\text{Law}} \mathcal{N}(0, \mathbf{I}_d)$

Conclusion: $\forall p \geq 1, \limsup_n \mathbb{E} \left[\det(\Gamma(F_n, F_n))^{-p} \right] < \infty$

$$\Rightarrow \vec{F}_n \xrightarrow[n \rightarrow \infty]{\mathcal{C}^\infty} \mathcal{N}(0, \mathbf{I}_d)$$

A slightly stronger result in dimension one

Assumptions:

- (F_n) sequence in $\mathcal{W}_m \oplus \mathcal{W}_{m-1} \oplus \cdots \oplus \mathcal{W}_0$
- $\mathcal{J}_m(F_n) \xrightarrow[n \rightarrow \infty]{\text{Law}} \mathcal{N}(0, 1)$
- $\forall i \leq m - 1, \mathbb{E}(\mathcal{J}_i(F_n)^2) \leq M$

Conclusion: $\forall p \geq 1, \limsup_n \mathbb{E} \left[\Gamma(F_n, F_n)^{-p} \right] < \infty$

Example 1:

- $X_N = \left(\frac{X_{i,j}}{\sqrt{N}} \right)_{1 \leq i,j \leq N}$ with $(X_{i,j})_{i,j \geq 1}$ are i.i.d. $\mathcal{N}(0, 1)$.
- $k_1 < k_2 < \dots < k_d$
- $D = \text{Diag}(k_1, k_2, \dots, k_d)$
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$$\left(\text{Tr} \left(X_N^{k_1} \right) - \mathbb{E} \left(\text{Tr} \left(X_N^{k_1} \right) \right), \dots, \text{Tr} \left(X_N^{k_d} \right) - \mathbb{E} \left(\text{Tr} \left(X_N^{k_d} \right) \right) \right) \\ \xrightarrow[n \rightarrow \infty]{\mathcal{C}^\infty} \mathcal{N}(0, D).$$

Example 2

Let A be a Gaussian matrix of size $N \times p$.

$$\forall q \geq 1, \exists C_{q,p} > 0, \exists \delta_{q,p} > 0,$$

$$\|{}^t A \cdot A - \mathbf{I}_p\|_2 \leq \epsilon \Rightarrow \mathbb{E} \left(\det ({}^t A \cdot A)^{-q} \right) \leq C_{q,p}.$$

Ideas of proof: (1) a decoupling procedure

$$F = \phi(X_1, \dots, X_n)$$

Sharp operator:

$$\sharp F = \sum_{k=1}^n \frac{\partial F}{\partial X_k} \hat{X}_k, \quad (X_k, \hat{X}_k)_{k \geq 1} \text{ i.i.d. } \sim \mathcal{N}(0, 1).$$

$$\mathbb{E} \left(e^{it \sharp F} \right) = \mathbb{E} \left(\exp \left(-\frac{t^2}{2} \Gamma(F, F) \right) \right).$$

$\Gamma(F, F)$ has negative moments

$\Leftrightarrow \sharp F$ has a smooth law

Ideas of proof: (2) decoupling twice

$\#F$ has a smooth law $\Leftrightarrow \Gamma[\#F, \#F]$ has negative moments
 $\Leftrightarrow \#[\#F]$ has a smooth law

👉 One is left to study the smoothness of the law of

$$\sum_{i,j \leq n} \frac{\partial^2 F}{\partial X_i \partial X_j} \hat{X}_i \hat{X}_j.$$

A random quadratic form

- Set $M = \left(\frac{\partial^2 F}{\partial X_i \partial X_j} \right)_{1 \leq i, j \leq n}$
- if $F \in \mathcal{W}_m$ then $M \in \mathcal{M}_n(\mathcal{W}_{m-2})$.

$$\begin{aligned} & \left| \mathbb{E} \left(e^{it \sum_{i,j \leq n} \frac{\partial^2 F}{\partial X_i \partial X_j} \hat{X}_i \hat{X}_j} \right) \right| = \left| \mathbb{E} \left(e^{it \sum_{k=1}^n \lambda_k \hat{X}_k^2} \right) \right| \\ &= \left| \mathbb{E} \left(\prod_{k=1}^n \frac{1}{\sqrt{1 + 2it\lambda_k}} \right) \right| \leq \mathbb{E} \left(\prod_{k=1}^n \frac{1}{(1 + 4t^2\lambda_k^2)^{\frac{1}{4}}} \right) \end{aligned}$$

☞ One is left to prove negative moments for

$$\mathcal{R}_q(M) = \sum_{k_1 \neq k_2 \cdots \neq k_q} \lambda_{k_1}^2 \cdots \lambda_{k_q}^2$$

☞ It is equivalent to get negative moments for

$$\mathcal{D}_q(M) = \inf_{\text{rank}(B) \leq q-1} \|M - B\|_2$$

A kind of compress sensing argument

For any matrix A we have $\|A\|_2^2 = q \mathbb{E}(\|AX\|^2)$ for X random matrix with standard Gaussian independent entries of dimensions $n \times q$

👉 For a typical realization of X one should have

$$\begin{aligned} \mathbb{P}(D_q(M) \leq \epsilon) &\approx_q \mathbb{P}(D_q(MX) \leq \epsilon) \\ &\approx_q \mathbb{P}\left(\inf_{a \in S^{q-1}} \|MX^t a\| \leq \epsilon\right) \end{aligned}$$

Dealing with a fixed $a \in \mathcal{S}^{q-1}$:

We consider $Y \sim \mathcal{N}(0, \mathbf{I}_n)$ then

$$F \approx \mathcal{N}(0, 1) \text{ \& } F \in \mathcal{W}_m \Rightarrow \frac{1}{\sqrt{m(m-1)}} {}^t Y M X \approx \mathcal{N}(0, \mathbf{I}_q)$$
$$\underbrace{\Rightarrow}_{\text{conditionning}} \gamma_{n,q} \left(\chi \in \mathcal{M}_{n,q}(\mathbb{R}) \mid \frac{1}{\sqrt{m(m-1)}} {}^t Y M \chi \approx \mathcal{N}(0, \mathbf{I}_q) \right) \approx 1$$

Then for χ typical and $a \in \mathcal{S}^{q-1}$ one should have

$$\frac{1}{\sqrt{m(m-1)}} {}^t Y M \chi {}^t a \approx \mathcal{N}(0, 1).$$

Reasoning by induction on m

Set $M_{\chi^t a} = (z_1, \dots, z_n)$,

$$\begin{aligned}\mathbb{E} \left(e^{it \frac{1}{\sqrt{m(m-1)}} {}^t Y M_{\chi^t a}} \right) &\approx \mathbb{E} \left(\exp \left(-\frac{t^2}{2} \sum_i z_i^2 \right) \right) \\ &= O \left(\frac{1}{t^p} \right), \quad p \gg 1.\end{aligned}$$

Where we use that ${}^t Y M_{\chi^t a}$ may be seen as an element of \mathcal{W}_{m-1} so we can proceed by induction.

☞ $\mathbb{P}(\|M_{\chi^t a}\| \leq \epsilon) \leq \epsilon^p$ for large p .

☞ We manage the inversion of inf and \mathbb{P} using a discretization of the sphere, the union bound and optimizing the constants.