

Integration by parts and convergence in  
distribution norms in the *CLT*

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**CLT**  $X_i \in R^d, i \in N,$  *i.i.d.* with  $E(X_i) = 0,$   $E(X_i^k X_i^p) = \delta_{k,p}$

Denote

$$S_n(X) = \frac{1}{\sqrt{n}} \sum_{i=1}^n X_i$$

Then

$$E(\phi(S_n(X))) \rightarrow E(\phi(N)) \quad \text{with} \quad N \sim N(0, I).$$

## Test functions

$$\phi \in C_b \rightarrow \text{weak convergence}$$

$$\phi \text{ measurable bounded} \rightarrow \text{convergence in total variation}$$

**THEOREM (Prohorov)** Convergence in total variation is equivalent to : there exists  $m$  such that

$$Y_1 = X_1 + \dots + X_m \sim f(x)dx + \nu(dx).$$

## Remark

$$Z = Y_1 + Y_2 \sim g(x)dx + \mu(dx) \quad g \in C(R)$$

## Consequence

$$g \geq \varepsilon \mathbf{1}_{B_r(x_0)} \geq \varphi_r(x - x_0) \quad \varphi_r \in C^\infty$$

## Splitting

$$Z \sim \xi \times V + (1 - \xi)U$$

With :  $\xi, V, U$  independent with laws

$$P(V \in dx) = \frac{1}{c_r} \varphi_r(x - x_0) dx \quad \varphi_r \simeq \mathbf{1}_{B_r(0)}$$

$$P(\xi = 1) = \varepsilon c_r, \quad P(\xi = 0) = 1 - \varepsilon c_r$$

$$P(U \in dx) = \frac{1}{1 - \varepsilon c_r} (P(Z \in dx) - \varepsilon \varphi_r(x - x_0) dx)$$

Now on we take

$$Z_i \sim \xi_i \times V_i + (1 - \xi_i)U_i \quad i \in N$$

**Strategy** : we use an "abstract Malliavin calculus" based on  $V_1, V_2, \dots, V_n, \dots$  in order to prove Prohorov's Theorem.

**Integration by parts** Suppose that

$$V \sim \varphi(x)dx$$

Then

$$\begin{aligned} E(f'(V)) &= \int_{\mathbb{R}} f'(x)\varphi(x)dx = - \int_{\mathbb{R}} f(x)\varphi'(x)dx \\ &= - \int_{\mathbb{R}} f(x) \frac{\varphi'(x)}{\varphi(x)} \times \varphi(x)dx \\ &= -E\left(f(V) \frac{\varphi'(V)}{\varphi(V)}\right) \\ &= E(f(V)H(V)) \end{aligned}$$

Conclusion

$$E(f'(V)) = E(f(V)H(V)) \quad \text{with} \quad H(V) = -\frac{\varphi'(V)}{\varphi(V)} = -\partial \ln \varphi(V).$$

**Come back on the end**

## MAIN RESULTS

**Convergence in total variation.**

$$|E(f(S_n(X))) - E(f(N))| \leq \frac{C}{\sqrt{n}} \|f\|_\infty$$

and moreover

**Convergence in (almost) distribution norms :** let  $|\alpha| = q$ .

$$|E(\partial^\alpha f(S_n(X))) - E(\partial^\alpha f(N))| \leq \frac{C}{\sqrt{n}} \|f\|_\infty + e^{-cn} \|f\|_{q,\infty}$$

with

$$\|f\|_{q,\infty} = \sum_{|\alpha| \leq q} \|\partial_\alpha f\|_\infty.$$

## Corollary

$$\gamma_\varepsilon(x) = \frac{1}{(2\pi\varepsilon)^{1/2}} e^{-\frac{|x|^2}{2\varepsilon}} \quad f_\varepsilon = f * \gamma_\varepsilon$$

Then, with  $\varepsilon = \varepsilon_n = e^{-cn}$

$$|E(\partial_\alpha f_{\varepsilon_n}(S_n(X))) - E(\partial^\alpha f(N))| \leq \frac{C}{\sqrt{n}} \|f\|_\infty$$

## Densities : Stronger hypothesis :

**Corollary** If  $Z_i \sim p(x)dx$ ,  $p \in W^{1,1}$  ( $\nabla p \in L^1$ ) Then  $S_n \sim p_{S_n}(x)dx$  and

$$\left\| \partial^\alpha p_{S_n(X)} - \partial^\alpha \gamma \right\|_\infty \leq \frac{C}{\sqrt{n}} \quad \forall |\alpha| \leq n \quad \gamma(x) = \frac{1}{(2\pi)^{1/2}} e^{-\frac{|x|^2}{2}}$$

## Remark

$$Z_1 + \dots + Z_n \sim (p * \dots * p)(x)dx \quad (p * \dots * p) \in W^{n,1}$$

## Extensions

A.  $Z_i, i \in N$  are **not identically distributed**

B **Local theorems (Edgeworth expansions)**

$$\left| E(\partial_\alpha f(S_n)) - \int \partial_\alpha f(x) \left( \sum_{i=0}^k \frac{1}{n^{i/2}} \theta_i(x) \right) \gamma(x) dx \right| \leq \frac{C}{n^{(k+1)/2}} \|f\|_\infty + e^{-cn} \|f\|_{q,\infty}$$

## Expected number of roots for trigonometric polynomials

$$P_n(t, Y) = \frac{1}{\sqrt{n}} \sum_{k=1}^n (Y_k^1 \cos \frac{kt}{n} + Y_k^2 \sin \frac{kt}{n})$$
$$N_n(Y) = \text{card}\{t \in (0, n\pi) : P_n(t, Y) = 0\}.$$

## Invariance principle

$$\lim_n \frac{1}{n} E(N_n(Y)) = \lim_n \frac{1}{n} E(N_n(G)) = \frac{1}{\sqrt{3}}$$

## Skech of the proof

**Kac-Rice lemma** let  $f : [0, a] \rightarrow \mathbb{R}$  differentiable such that

$$\inf_{t \leq a} (|f(t)| + |f'(t)|) > 0.$$

Then

$$N_a(f) = \lim_{\delta \rightarrow 0} \int_0^a |f'(t)| \mathbf{1}_{\{|f(t)| \leq \delta\}} \frac{dt}{2\delta}.$$



We use the *CLT* for  $S_n = (S_n^1, S_n^2) \in R^2$  with

$$S_n^1(t, Y) = P_n(t, Y) = \frac{1}{\sqrt{n}} \sum_{k=1}^n (Y_k^1 \cos \frac{kt}{n} + Y_k^2 \sin \frac{kt}{n})$$

$$S_n^2(t, Y) = \partial_t P_n(t, Y) = \frac{1}{\sqrt{n}} \sum_{k=1}^n \frac{k}{n} (Y_k^2 \cos \frac{kt}{n} - Y_k^1 \sin \frac{kt}{n})$$

We take

$$\Phi_\delta(x^1, x^2) \simeq |x_2| \times \mathbf{1}_{(-\infty, 0)}^{(\delta)}(x^1) \quad \rightarrow \quad \partial_1 \Phi_\delta(x_1, x_2) \approx x_2 \frac{1}{2\delta} \mathbf{1}_{\{|x_1| \leq \delta\}}$$

K-R lemma

$$N_n(Y) = \lim_{\delta \rightarrow 0} \int_0^{n\pi} \partial_1 \Phi_\delta(S_n^1(t, Y), S_n^2(t, Y)) dt$$

with

$$\partial_1 \Phi_\delta(S_n^1(t, Y), S_n^2(t, Y)) \approx S_n^2(t, Y) \frac{1}{2\delta} \mathbf{1}_{\{|S_n^1(t, Y)| \leq \delta\}}$$

CLT (invariance principle)

$$\lim_n E(\partial_1 \Phi_\delta(S_n^1(t, Y), S_n^2(t, Y))) = \lim_n E(\partial_1 \Phi_\delta(S_n^1(t, G), S_n^2(t, G)))$$

## Variance

$$\lim_n \frac{1}{n} \text{Var}(N_n(G)) = c(G) = 0.56.$$

We prove (non-universality) :

$$\lim_n \frac{1}{n} \text{Var}(N_n(Y)) = c(G) + 30(E(Y_1^4) - 3).$$

Basic quantity

$$\frac{1}{n} \int_0^{\pi n} dt \int_0^{\pi n} ds \quad \partial_1 \Phi_\delta(S_n(t, Y)) \partial_1 \Phi(S_n(s, Y))$$

We use *CLT* for

$$\bar{S}_n(t, s, Y) = (S_n(t, Y), S_n(s, Y)) \in R^4$$

with the Edgorth developpement of order 4.

## Difficulties

1 Degeneracy on the diagonal :

$$s \rightarrow t \quad \Rightarrow \quad S_n(s, Y) \rightarrow S_n(t, Y)$$

2 Magnitude order

$$\frac{1}{n} \int_0^{\pi n} dt \int_0^{\pi n} ds \quad \sim \quad \frac{1}{n} \times n^2 = n$$

So we need

$$n \times \left| \bar{S}_n(t, s, Y) - \bar{S}_n(t, s, G) \right| \rightarrow 0.$$

Estimates of order  $n^{-3/2}$  in the *CLT*. Use Edgworth at order 3.

## Integration by parts (abstract Malliavin)

$$F = f(\xi_i V_i + (1 - \xi_i) U_i, i = 1, \dots, n)$$
$$P(V \in dx) = \frac{1}{c_r} \varphi_r(x - x_0) dx \quad \varphi_r \simeq \mathbf{1}_{B_r(0)}$$

Derivatives

$$D_k F = \xi_k \frac{\partial}{\partial V_k} f(\xi_i V_i + (1 - \xi_i) U_i, i = 1, \dots, n)$$

Duality

$$E(\langle DF, DG \rangle) = \sum_{k=1}^n E(D_k F D_k G) = E(FLG)$$

with

$$LG = - \sum_{k=1}^N D_k D_k G + D_k G \partial_v \ln \varphi_r(V_k - x_0)$$

## Proof Standard Integration by parts :

$$\begin{aligned}
 & E(D_k F D_k G) \\
 &= E\left(\xi_k \frac{\partial}{V_k} f(\xi_k V_k + (1 - \xi_k)U_k, \dots) \frac{\partial}{V_k} g(\xi_k V_k + (1 - \xi_k)U_k, \dots)\right) \\
 &= -E\left(\xi_k \int \partial_k f(\xi_k v_k, \dots) \partial_k g(\xi_k v_k, \dots) \varphi_r(v_k - x_0) dv_k\right) \\
 &= -E\left(\xi_k \int f \times (\partial_k^2 g \times \varphi_k + \partial_k g \varphi_k') dv_k\right) \\
 &= -E\left(\xi_k \int f \times \left(\partial_k^2 g + \partial_k g \frac{\varphi_k'}{\varphi_k}\right) \times \varphi_r(v_k - x_0) dv_k\right) \\
 &= -E\left(F(D_k D_k G + D_k G \partial_v \ln \varphi_r(V_k - x_0))\right)
 \end{aligned}$$

**Covariance matrix** :  $F = (F_1, \dots, F_d)$

$$\sigma_F^{i,j} = \langle DF_i, DF_j \rangle, \quad i, j = 1, \dots, d.$$

Difficulty

$$P(\det \sigma_F = 0) \geq P(\xi_1 = \dots = \xi_n = 0) > 0.$$

Hoeffding's inequality

$$P\left(\frac{1}{\sqrt{n}} \sum_{i=1}^n \xi_i < \frac{c_r}{2}\right) \leq e^{-an}.$$

**IbyP**

$$E(\partial^\alpha f(F)G) = E(f(F)H_\alpha(F, G))$$

with

$$\|H_\alpha(F, G)\|_p \leq E\left(\frac{1}{(\det \sigma_F)^{p'}}\right) \times (\|F\|_{q,p} + \|G\|_{q,p}).$$

**Regularization lemma**  $|\alpha| = m$

$$\begin{aligned} |E(\partial^\alpha f(F)) - E(\partial^\alpha(f * \gamma_\varepsilon)(F))| &\leq C \|f\|_{m,\infty} P^{1/2}(\det \sigma_F \leq \eta) + \frac{\varepsilon}{\eta^{2m}} \|f\|_\infty \\ &\leq C \|f\|_{m,\infty} e^{-an} + \frac{\varepsilon}{\eta^{2m}} \|f\|_\infty \end{aligned}$$

Here  $\gamma_\varepsilon$  is a **super kernel**

In particular

$$\begin{aligned} |E(f(F)) - E(f * \gamma_\varepsilon)(F)| &\leq (P^{1/2}(\det \sigma_F \leq \eta) + \frac{\varepsilon}{\eta^{2m}}) \|f\|_\infty \\ &\leq C(e^{-an} + \frac{\varepsilon}{\eta^{2m}}) \|f\|_\infty \end{aligned}$$

**Proof**  $\gamma_\varepsilon \geq 0$  is a super kernel if  $\int \gamma_\varepsilon(x) = 1$  and

$$\int x^p \gamma_\varepsilon(x) dx = 0 \quad \forall p \geq 1.$$

It follows that, for every  $m$

$$\begin{aligned}(f * \gamma_\varepsilon)(F) - f(F) &= \int \gamma_\varepsilon(y) f(F + y) dy - f(F) \\ &= \int \gamma_\varepsilon(y) (f(F + y) - f(F)) dy \\ &= \sum_{i=1}^m \int \gamma_\varepsilon(y) \times \frac{y^i}{i!} dy \times f^{(i)}(F) \\ &\quad + \int \gamma_\varepsilon(y) R_m(f)(y) dy\end{aligned}$$

with

$$R_m(f)(y) = \frac{y^{m+1}}{(m+1)!} \int_0^1 d\lambda f^{(m+1)}(\lambda F + (1-\lambda)y)$$

And we use *IbP* in order to handel

$$E\left(\int \gamma_\varepsilon(y) R_m(f)(y) dy\right)$$



## Estimates

$$\begin{aligned} & |E(f(F)) - E(f(G))| \\ & \leq |E(f(F)) - E((f * \gamma_\varepsilon)(F))| + |E(f(G)) - E((f * \gamma_\varepsilon)(G))| \\ & \quad + |E(f * \gamma_\varepsilon(F)) - E((f * \gamma_\varepsilon)(G))| \\ & \leq C(e^{-an} + \frac{\varepsilon}{\eta^{2m}}) \|f\|_\infty + \frac{1}{\varepsilon} E|F - G| \times \|f\|_\infty \end{aligned}$$

1. Bally, V., Caramellino, L., Poly, G. (2018) Convergence in distribution norms in the CLT for non identical distributed random variables. *Electronic Journal of Probability*, 23, no. 45, 1-51.
2. Bally, V., Caramellino, L., Poly, G. (2017) On the non universality for the variance of the number of real roots of random trigonometric polynomials. Preprint *arXiv* :1711.03364 43 pages. November 2017 *Probability Theory and Related Fields*
3. Bally, V., Caramellino, L., Poly, G. (2019) : Regularization lemmas and convergence in total variation. *ArXiv* :1907.12328, 2019. *Electronic Journal of Probability*