

Higher order approximation for sequences converging in the mod-Gaussian sense

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- ▶ mod-GAUSS convergence and some new results
- ▶ higher order approximation and some known results
- ▶ models with dependency and some new results

joint work with LUKAS KNICHEL, CAROLIN KLEEMANN
and MARIUS BUTZEK, Bochum

- ▶ D a domain of $\mathbb{C}$ containing 0
- ▶ ϕ infinite divisible distribution with LAPLACE transform $\exp(\eta(z))$ on D η : LÉVY exponent

Definition (KOWALSKI, NIKEGHBALI, 2010)

A sequence of real r.v. (X_n) converges mod- ϕ on D with parameter $t_n \rightarrow \infty$ and limiting function ψ if, locally uniformly on D ,

$$\exp(-t_n \eta(z)) \mathbb{E}(e^{zX_n}) \rightarrow \psi(z)$$

instead of renormalizing the variables as in CLT, we renormalize the LAPLACE transform to get access to the next term

mod- ϕ implies a CLT

Theorem

If (X_n) converges mod- ϕ on D with parameter t_n , then

$$Y_n = \frac{X_n - t_n \eta'(0)}{\sqrt{t_n \eta''(0)}} \rightarrow N(0, 1)$$

very often: classical way of proving CLTs can be adapted to prove mod- ϕ convergence

example 1

(Y_j) i.i.d., centered, $S_n = \sum_{j=1}^n Y_j$

$$\log \mathbb{E} e^{izS_n} = n \sum_{r \geq 2} \frac{\kappa^{(r)}(Y)}{r!} (iz)^r$$

$$X_n := \frac{S_n}{n^{1/\nu}}, \quad \nu \in \mathbb{N}$$

$$\log \mathbb{E} e^{izX_n} = -n^{\frac{\nu-2}{\nu}} \frac{\kappa^{(2)}(Y)}{2} z^2 + \sum_{j \geq \nu} \frac{\kappa^{(j)}(Y)}{j!} (iz)^j \frac{1}{n^{\frac{j}{\nu}-1}}$$

Hence if $n \rightarrow \infty$:

$$\frac{\mathbb{E} e^{izX_n}}{\exp\left(-n^{\frac{\nu-2}{\nu}} \sigma^2 \frac{z^2}{2}\right)} \rightarrow \exp\left(\frac{\kappa^{(\nu)}(Y)}{\nu!} (iz)^\nu\right) + \mathcal{O}\left(\frac{1}{n^{1/\nu}}\right)$$

example 1

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example 1

(X_n) converges mod-GAUSS mit parameter $t_n = \sigma^2 n^{\frac{\nu-2}{\nu}}$ and limiting function

$$\psi(z) = \exp\left(\frac{\kappa^{(\nu)}(Y)}{\nu!} (iz)^\nu\right)$$

example:

$\sigma^2 = 1$ and the law of Y is symmetric:

it could be that $\nu = 4$: $t_n = n^{1/2}$

and $\kappa^{(4)}(Y) = \mathbb{E}Y^4 - 3$

example 2

$$\log(|\det(Id - U_n)|)$$

where U_n is an unitary Haar-distributed random matrix

It converges mod-GAUSS on $\{\operatorname{Re}(z) > -1\}$ with parameter $t_n = \frac{\log n}{2}$ and limiting function

$$\psi(z) = \frac{G(1 + z/2)^2}{G(1 + z)}$$

with $G(z+1) = G(z)\Gamma(z)$, $G(1) = 1$, the BARNES G -function

KOWALSKI, NIKEGHBALI, 2010; KEATING, SNAITH, 2000

example 3

Let M_n be a GUE matrix. Then

$$\log |\det M_n| - \mathbb{E} \log |\det M_n|$$

converges mod-GAUSS on $\{|z| < 1\}$ with parameter $\frac{\log n}{2}$ and limiting function

$$\psi(z) = \frac{G(1 + z/2)^2}{G(1 + z)}$$

DÖRING, E., 2013; DAL BORGO, HOVHANNISYAN, ROUAULT, 2018

(we have studied uniform bounds on cumulants)

example 4: LAGUERRE ensemble

- ▶ Let A be a $n \times p(n)$ matrix with $p(n) \leq n$ chosen to be Gaussian over $\mathbb{C}$. Then $A^\dagger A$ is called LAGUERRE complex ensemble
- ▶ The joint pdf of $(\lambda_1, \dots, \lambda_{p(n)})$ is

$$\frac{1}{Z_{n,p(n)}} \prod_{1 \leq j < k \leq p(n)} |\lambda_j - \lambda_k|^2 \prod_{k=1}^{p(n)} \lambda_k^{n-p(n)} e^{-\frac{\lambda_k}{2}}$$

- ▶ With SELBERG formula one obtains for the MELLIN transform

$$\mathbb{E} \left[\left(\det W_{n,p(n)}^L \right)^z \right] = 2^{p(n)z} \prod_{k=1+n-p(n)}^n \frac{\Gamma(k+z)}{\Gamma(k)}$$

example 4: LAGUERRE ensemble

- ▶ We obtain

$$\log \mathbb{E} \left[\exp(z \log(\det W_{n,p(n)}^L)) \right] = p(n)z \log 2 + L(p(n), n; z)$$

- ▶ Applying BINET formula of $\log \Gamma(z)$, BARNES G -function and ABEL-PLANA summation formula we obtain

Theorem (E., Knichel, 2018)

mod-Gauss of $\log(\det W_{n,p(n)}^L)$ with parameters

- ▶ $t_n = \log n$, if $p(n) = n$
- ▶ $t_n = \log\left(\frac{p(n)+1+c}{1+c}\right)$, if $n - p(n) = c$
- ▶ $t_n = \log\left(\frac{n}{n-p(n)}\right)$, if $n - p(n) = o(n)$

example 4: LAGUERRE ensemble

the limiting function is

$$\begin{aligned}\log \psi(z) &= \log G(z+1) - \left(z - \frac{1}{2}\right) \log \Gamma(z+1) \\ &\quad + \int_0^\infty \left(\frac{1}{2s} - \frac{1}{s^2} + \frac{1}{s(e^s - 1)} \right) \frac{e^{-sz} - 1}{e^s - 1} ds + \frac{3}{4}z^2 + \frac{z}{2}.\end{aligned}$$

$p(n) = n$: DAL BORGO, HOVHANNISYAN, ROUAULT, 2018

Theorem (E., Knichel, 2018)

If $p(n) = p$, we obtain $\text{mod-}\phi$ convergence in $i\mathbb{R}$ with $t_n = pn$ and limiting function $\psi(z) = (1+z)^{-\frac{p^2}{2}}$, ϕ is an infinitely divisible distribution

example 4: LAGUERRE ensemble, companion theorems

- ▶ **extended CLT:** for $x = o(\sqrt{\log n})$

$$P\left(\log(\det W_{n,p(n)}^L) \geq x\sqrt{\log n}\right) = P(N(0,1) \geq x) (1 + o(1))$$

- ▶ **precise deviations:** for $x > 0$:

$$P\left(\log(\det W_{n,p(n)}^L) \geq x \log n\right) = \frac{e^{-\frac{x^2}{2} \log n}}{x\sqrt{2\pi \log n}} \psi(x) (1 + o(1))$$

- ▶ **rate of convergence:**

$$d_{\text{Kol}}\left(\frac{\log(\det W_{n,p(n)}^L)}{\sqrt{t_n}}, N(0,1)\right) \leq C \frac{1}{\sqrt{t_n}}$$

stochastic geometry

Random simplices:

- ▶ $p(n) + 1$ independent random points $X_1, \dots, X_{p(n)+1}$ in $\mathbb{R}^n$, multivariate Gaussian
- ▶ consider the $p(n)$ -dimensional volume of the simplex with vertices $X_1, \dots, X_{p(n)+1}$
- ▶ dimension and number of points $p(n) + 1$ are allowed to grow simultaneously
- ▶ $\log \psi(z)$ differs by $-\frac{z^2}{2}$

see also GROTE, KABLUCHKO, THÄLE, 2018: cumulant bounds

higher order approximation

Consider a sequence of real r.v. (X_n) which converges mod-GAUSS on $D \subset i\mathbb{R}$ with parameter $t_n \rightarrow \infty$ and limiting function ψ .

Theorem (Barhoumi-Andréani, 2017)

For nice ψ , one can construct a family of random variables $H_{t_n}(\psi)$ that converges mod-GAUSS on D with parameter $t_n \rightarrow \infty$ and limiting function ψ .

The law of $H_{t_n}(\psi)$ has a LEBESGUE-density given by:

$$\frac{1}{c_n} \psi\left(\frac{x}{t_n}\right) e^{-\frac{1}{2} \left(\frac{x}{\sqrt{t_n}}\right)^2}$$

higher order approximation

Question: can we find a bound for

$$d_{\text{Kol}}(X_n, H_{t_n}(\psi)) \leq \text{ ? }$$

compare with the BERRY-ESSÉEN bound for the convergence in law of $\frac{X_n}{\sqrt{t_n}}$ to the Gaussian distribution!

BARHOUMI-ANDRÉANI, 2017:

consider on $D = i\mathbb{R}$ the limiting function $\psi(z) = e^{-C \frac{z^4}{4!}}$ as being the mod-GAUSSIAN limit of $X_n := \frac{1}{n^{1/4}} \sum_{j=1}^n Y_j$, i.i.d., with $t_n = n^{1/2}$:

Assume $\sigma^2 = 1$ **and that** $C := 3 - \mathbb{E}(Y^4) > 0$

higher order approximation: STEIN'S method

develop STEIN'S method for the target distribution with density

$$g_n(x) := \frac{1}{c_n} e^{-\frac{1}{2} \frac{x^2}{\sqrt{n}} - C \frac{x^4}{4!n^2}}$$

BARHOUMI-ANDRÉANI, 2017

But: apply the **density approach** due to STEIN to obtain the STEIN equation and use the fact that the density is log-concave to apply results from PEKÖZ, RÖLLIN, ROSS, 2016

$$f'_h(x) - \frac{x}{\sqrt{n}} f_h(x) - \frac{Cx^3}{6n^2} f_h(x) = h(x) - \mathbb{E}h$$

bound solutions for certain test-functions h ...

higher order approximation: results

Theorem (BARHOUMI-ANDRÉANI, 2017)

Consider functions in

$$\mathcal{H}_1 := \{h \in C_1(\mathbb{R}) : \|h\|_\infty \leq 1, \|h'\|_\infty \leq 1\}$$

Then

$$d_{\mathcal{H}_1}(X_n, H_{\sqrt{n}}(\psi)) \leq \frac{C_1}{n^{1/4}} \|h'\|_\infty + \frac{C_2}{n^{1/2}} \|h\|_\infty$$

and the same order for the KOLMOGOROV-distance

Remark:

$$d_{\mathcal{H}_1}\left(\frac{X_n}{n^{1/4}}, \frac{H_{\sqrt{n}}(\psi)}{n^{1/4}}\right) \leq \frac{C_3}{\sqrt{n}} + \mathcal{O}\left(\frac{1}{n}\right)$$

higher order approximation: results

Theorem (BARHOUMI-ANDRÉANI, 2017)

Consider functions in

$$\mathcal{H}_2 := \{h \in C_1(\mathbb{R}) : \|h\|_\infty \leq 1, \|h'\|_\infty \leq 1, \|\mathbf{h}''\|_\infty \leq \mathbf{1}\}$$

Then

$$d_{\mathcal{H}_2}(X_n, H_{\sqrt{n}}(\psi)) \leq \frac{C_1}{\sqrt{n}} + \frac{C_2}{n^{\frac{3}{4}}}$$

Now:

$$d_{\mathcal{H}_2}\left(\frac{X_n}{n^{1/4}}, \frac{H_{\sqrt{n}}(\psi)}{n^{1/4}}\right) \leq \frac{C_3}{n} + \mathcal{O}\left(\frac{1}{n^2}\right)$$

$$d_{Kol}\left(\frac{X_n}{n^{1/4}}, \frac{H_{\sqrt{n}}(\psi)}{n^{1/4}}\right) \leq \mathcal{O}\left(\frac{1}{n^{2/3}}\right)$$

higher order approximation: results

improvement of the classical BERRY-ESSEEN bound
understood as an additive correction:

$$\left| \mathbb{E} h\left(\frac{X_n}{n^{1/4}}\right) - \mathbb{E} h\left(\frac{H_{\sqrt{n}}}{n^{1/4}}\right) \right| = \left| \mathbb{E} h\left(\frac{X_n}{n^{1/4}}\right) - \mathbb{E}(h(N)) + \text{Corr}(n, h) \right|$$

with

$$\text{Corr}(n, h) = \mathbb{E} \left(h(N) - h\left(\frac{H_{\sqrt{n}}}{n^{1/4}}\right) \right) = \mathbb{E} \left(h'\left(\frac{H_{\sqrt{n}}}{n^{1/4}} + U\Delta_n\right) \Delta_n \right)$$

where U uniform, N and $H_{\sqrt{n}}$ are independent and $\Delta_n := N - \frac{H_{\sqrt{n}}}{n^{1/4}}$

higher order approximation

Challenges:

- ▶ one has to bound the **third** derivative of STEIN-solutions !
- ▶ one has to care on the **perturbative operator**

$$\frac{C}{6n^2} \mathbb{E}(X_n^3 f(X_n))$$

(B.-A.: zero-bias techniques...)

goal: non-identically distributed and dependent variables

Questions: try to proof similar results

- ① for independent but **non-identically** distributed summands
- ② for **exchangeable pairs** (application: CURIE-WEISS model with $\beta \neq \beta_c$)
- ③ for summands with a **dependency graph** structure
- ④ other (t_n, ψ) mod-GAUSS situations...

1. non-identically distributed

(Y_j) independent, centered, use exchangeable pair-approach

$$X_n = \frac{1}{n^{\frac{1}{4}}} \sum_{j=1}^n Y_j, \quad \mathbb{E}(X'_n | X_n) = \left(1 - \frac{1}{n}\right) X_n$$

$$\left| \mathbb{E} \left(f'(X_n) - \frac{X_n}{\sqrt{n}} f(X_n) \right) \right| \leq \frac{C_1}{n^{3/2}} \sum_{j=1}^n \mathbb{E}(Y_j^4) + \frac{C_2}{n} \left(\sum_{j=1}^n \text{Var}(Y_j^2) \right)^{1/2}$$

1. non-identically distributed

with $\tilde{f}(x) := x^2 f(x)$, $\lambda = \frac{1}{n}$, $\mathbb{E}(X'_n | X_n) = (1 - \lambda)X_n$ and

$$\hat{K}(t) := (X_n - X'_n)(1_{\{-(X_n - X'_n) \leq t \leq 0\}} - 1_{\{0 \leq t \leq -(X_n - X'_n)\}})$$

we obtain

$$\frac{C}{6n^2} \mathbb{E}(X_n^3 f(X_n)) = \frac{C}{6n^2} \mathbb{E}(X_n \tilde{f}(X_n)) = \frac{C}{6n^2} \frac{1}{\lambda} \mathbb{E} \int_{\mathbb{R}} \tilde{f}'(X_n + t) \hat{K}(t) dt$$

second equality well known...

1. non-identically distributed

notice that $\tilde{f}'(X_n + t) = (X_n + t)^2 f'(X_n + t) + 2(X_n + t)f(X_n + t)$

after some calculations we obtain:

$$\begin{aligned} \left| \frac{C}{6n} \mathbb{E} \int_{\mathbb{R}} \tilde{f}'(X_n + t) \hat{K}(t) dt \right| &\leq \frac{C_1}{n^3} \sum_{j=1}^n \mathbb{E} Y_j^4 + \frac{C_2}{n} \\ &\quad + \frac{C_3}{n^{5/2}} \sum_{j=1}^n \mathbb{E} |Y_j|^3 + \frac{C_4}{n^{3/2}} \sum_{j=1}^n \mathbb{E} |Y_j| \end{aligned}$$

hence we have obtained the aimed general bound for $d_{\mathcal{H}_2}$ (and $d_{\mathcal{H}_1}$)

2. CURIE-WEISS model

n magnetic particles $Y_1, \dots, Y_n$: each with spin $+1, -1$
density:

$$\frac{1}{Z_n} \exp\left(\frac{\beta}{2n} \sum_{i,j=1}^n y_i y_j\right) d\varrho^{\otimes n}(y)$$

with $\varrho = \frac{1}{2}\delta_{-1} + \frac{1}{2}\delta_{+1}$

the particles try to align themselves together with the same spin, β inverse temperature

Theorem (ELLIS, NEWMAN, 1978)

for $0 < \beta < 1$ the magnetization $\frac{1}{n} \sum Y_j$ fulfils:

$$W_n := \frac{1}{\sqrt{n}} \sum_{j=1}^n Y_j \Rightarrow N\left(0, \frac{1}{1-\beta}\right)$$

$\beta = 1$: critical temperature (phase transition)

2. STEIN-pair for CURIE-WEISS

$$W'_n := W_n + \frac{Y'_I - Y_I}{\sqrt{n}}$$

$$\mathbb{E}(W'_n | W_n) = \left(1 - \frac{1 - \beta}{n}\right) W_n + R = (1 - \lambda) W_n + R$$

known (BARBOUR 1980, CHATTERJEE, SHAO 2010; E., LÖWE 2010):
for $0 < \beta < 1$:

$$\sup_{z \in \mathbb{R}} |\mathbb{P}_{\text{CW}}(W_n \leq z) - \Phi_\beta(z)| \leq \frac{C}{\sqrt{n}}$$

2. CURIE-WEISS model

Theorem (BUTZEK, E., 2019)

Consider the CURIE-WEISS model and let $X_n := n^{1/4}W_n$.

Take $t_n = \sqrt{n}$, $\psi(z) = e^{-Cz^4}$ and any $0 < \beta < 1$:

$$d_{\mathcal{H}_2}(X_n, H_{\sqrt{n}}(\psi)) \leq \frac{C_1}{\sqrt{n}} + \frac{C_2}{n^{3/4}}$$

and

$$d_{\mathcal{H}_2}\left(W_n, \frac{H_{\sqrt{n}}(\psi)}{n^{1/4}}\right) \leq \frac{C}{n} + \mathcal{O}\left(\frac{1}{n^2}\right) \quad d_{Kol}\left(W_n, \frac{H_{\sqrt{n}}(\psi)}{n^{1/4}}\right) \leq \frac{C}{n^{2/3}}$$

2. sketch of the proof

we apply a recent bound due to SHAO, ZHANG, 2018:

$$\mathbb{E} \left| \frac{1}{2\lambda} \mathbb{E}((W_n - W'_n) | W_n - W'_n | | W_n) \right| \leq C \frac{1}{\sqrt{n}}$$

to obtain

$$|\mathbb{E}(f'(X_n) - \frac{X_n}{\sqrt{n}} f(X_n))| = \mathcal{O}(\frac{1}{\sqrt{n}})$$

again:

$$\begin{aligned} \frac{C}{6n^2} \mathbb{E}(X_n^3 f(X_n)) &= \frac{C}{6n^2} \mathbb{E}(X_n \tilde{f}(X_n)) \\ &= \frac{C}{6n^2} \frac{1}{2\lambda} \mathbb{E} \int_{\mathbb{R}} \tilde{f}'(X_n + t) \hat{K}(t) dt + \frac{C}{6n^2} \frac{1}{\lambda} \mathbb{E}(\tilde{f}(X_n) R(X_n)) \\ &= \mathcal{O}(\frac{1}{\sqrt{n}}) \end{aligned}$$

3. dependency graphs

a graph L with vertex set A is a **dependency graph** for $(Y_\alpha)_{\alpha \in A}$ if

- ▶ if A_1 and A_2 are disconnected subsets in L , then $(Y_\alpha)_{\alpha \in A_1}$ and $(Y_\alpha)_{\alpha \in A_2}$ are independent
- ▶ roughly: there are edges between pairs of *dependent* random variables
- ▶ example: $G(n, p)$ -model, counting Δ s; edge between two, if they share an edge

3. dependency graph

setting:

- ▶ $(Y_{n,i})_{1 \leq i \leq N_n}$ family of bounded random variables, $|Y_{n,i}| < A$.
- ▶ there is a dependency graph L_n with maximal degree $D_n - 1$
- ▶ consider $S_n = \sum_{i=1}^{N_n} Y_{n,i}$ and $\sigma_n^2 = \mathbb{V}(S_n)$

JANSON, 1988: $\kappa_r(S_n) \leq C_r N_n D_n^{r-1} A^r$, C_r a constant

assume that $\frac{\sigma_n^2}{D_n N_n} \rightarrow \sigma^2 > 0$: error term in the CLT is $\mathcal{O}\left(\sqrt{\frac{D_n}{N_n}}\right)$

3. dependency graph

Theorem (E., KLEEMANN, 2019)

Assume that $\frac{\sigma_n^2}{N_n D_n} \rightarrow \sigma^2 > 0$ and $\frac{\kappa_4(S_n)}{N_n D_n^3} \rightarrow K > 0$. Moreover assume that $\mathbb{E} Y_{n,i}^3 = 0$ for all n, i . Then

$$X_n := \frac{S_n - \mathbb{E}(S_n)}{D_n^{3/4} N_n^{1/4}}$$

converges mod-Gauß with $t_n = \left(\frac{N_n}{D_n}\right)^{1/2}$ and limiting function $\psi(z) = \exp\left(-\frac{K}{4} z^4\right)$. Moreover

$$d_{\mathcal{H}_1}(X_n, H_{t_n}(\psi)) \leq \mathcal{O}\left(\left(\frac{D_n}{N_n}\right)^{1/4}\right).$$

needs good bounds on cumulants!

thank you