

The free log-normal distribution and confluent hypergeometric series

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convolution semigroups

- additive convolution
 - $(\mathcal{M}(\mathbb{R}), *)$: normal distribution
 - $(\mathcal{M}(\mathbb{R}), \boxplus)$: semicircle distribution

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Definition of the free multiplicative analog measures

Hari Bercovici and Dan Voiculescu. *Lévy-Hinčin type theorems for multiplicative and additive free convolution*, Pacific J. Math., 1992.

Voiculescu's S -transform

Let μ a compact supported probability measure on \mathbb{C} with $m_1(\mu) \neq 0$.

- $\Psi_\mu(z) := \int_{\mathbb{C}} \frac{zt}{1-zt} d\mu(t) = \sum_{n=1}^{\infty} m_n(\mu) z^n$
 $\Psi_\mu(0) = 0, \quad \Psi'_\mu(0) = m_1(\mu) \neq 0$

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- $\chi_\mu(z) := \Psi_\mu^{(-1)}(z)$

- $S_\mu(z) := \frac{z+1}{z} \chi_\mu(z)$

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Theorem (Voiculescu)

$$S_{\mu \boxtimes \nu}(z) = S_\mu(z) \cdot S_\nu(z)$$

Free multiplicative normal distribution σ_t

Definition (Bercovici, Voiculescu)

$$S_{\sigma_t}(z) := \exp\left(t\left(z + \frac{1}{2}\right)\right)$$

- $t \geq 0$: σ_t ... measure on torus \mathbb{T}
“free rolled-up normal distribution”
- $t \leq 0$: σ_t ... measure on positive, real line \mathbb{R}^+
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Properties of σ_t

- $\sigma_s \boxtimes \sigma_t = \sigma_{s+t}$ $s, t \geq 0$ or $s, t \leq 0$.
- σ_t is \boxtimes -infinite divisible.

Appearance of the measures σ_t

- Free, multiplicative central limit theorem
- Free, multiplicative Brownian Motion
- Limit distribution of some unitary random matrices

Philippe Biane. *Free Brownian motion, free stochastic calculus and random matrices*, 1997

Moments of σ_t

Theorem (Biane)

The moments of the measures σ_t are for $n \geq 1$

$$m_n(\sigma_t) = \exp\left(-\frac{nt}{2}\right) \cdot \sum_{k=0}^{n-1} \frac{(-1)^k n^{k-1}}{k!} \binom{n}{k+1} t^k$$

Proof with Lagrange's inversion theorem

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Proof with Lagrange's inversion theorem

Confluent hypergeometric function

Definition (Confluent hypergeometric series)

$${}_1F_1(a; b; z) := \sum_{n=0}^{\infty} \frac{(a)_n}{(b)_n n!} z^n$$

with Pochhammer symbol

$$(a)_n := a(a+1)\dots(a+n-1) \quad (a)_0 := 1$$

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Remark: $a \in -\mathbb{N}_0 \Rightarrow {}_1F_1(a; b; z)$ is a polynomial in z

$$\begin{aligned}m_n(\sigma_t) &= \exp\left(-\frac{nt}{2}\right) \cdot \sum_{k=0}^{n-1} \frac{(-1)^k n^{k-1}}{k!} \binom{n}{k+1} t^k \\ &= \exp\left(-\frac{nt}{2}\right) \cdot {}_1F_1(1-n; 2; nt) \\ &= \exp\left(\frac{nt}{2}\right) \cdot {}_1F_1(1+n; 2; -nt)\end{aligned}$$

Theorem

The free cumulants of the measures σ_t are

$$\kappa_n = \frac{(-nt)^{n-1}}{n!} \exp\left(-\frac{nt}{2}\right).$$

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Proof by induction on n

- moment-cumulant-formulas

$$\Rightarrow \kappa_n = \left(\frac{(-n)^{n-1}}{n!} t^{n-1} + a_{n-2}^{(n)} t^{n-2} + \dots + a_2^{(n)} t^2 \right) e^{-\frac{nt}{2}}$$

- $\kappa_n(\sigma_{s+t}) = \kappa_n(\sigma_s \boxtimes \sigma_t) = \sum_{\pi \in NC(n)} \kappa_\pi(\sigma_s) \kappa_{K(\pi)}(\sigma_t)$

$$\Rightarrow a_{n-2}^{(n)} = \dots = a_2^{(n)} = 0$$

Theorem

The free cumulants of the measures σ_t are

$$\kappa_n = \frac{(-nt)^{n-1}}{n!} \exp\left(-\frac{nt}{2}\right).$$

$${}_1F_1(1-n; 2; nt) = \sum_{\pi \in NC(n)} \prod_{\substack{V \in \pi \\ k:=|V|}} \frac{(-kt)^{k-1}}{k!}$$

$$\frac{(-n)^{n-1}}{n!} t^{n-1} = \sum_{\pi \in NC(n)} \mu(\pi, 1_n) \prod_{\substack{V \in \pi \\ k:=|V|}} {}_1F_1(1-k; 2; kt)$$

hypergeometric functions as (formal) moment sequences

$$m_n = {}_1F_1(1 - n; 2; nz) \Leftrightarrow \kappa_n = \frac{(-nz)^{n-1}}{n!}$$

$$m_n = {}_1F_1(1 + n; 2; nz) \Leftrightarrow \kappa_n = \frac{(nz)^{n-1}}{n!} \exp(nz)$$

Theorem (Addition formulae for hypergeometric functions)

$$\begin{aligned}
 {}_1F_1(1-n; 2; x+y) &= \\
 &= \sum_{\pi \in NC(n)} \left(\prod_{\substack{V \in \pi \\ k:=|V|}} \frac{\left(-\frac{k}{n}x\right)^{k-1}}{k!} \cdot \prod_{\substack{V \in K(\pi) \\ l:=|V|}} {}_1F_1\left(1-l; 2; \frac{l}{n}y\right) \right)
 \end{aligned}$$

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 \end{aligned}$$

Proof: $m_n(\sigma_{s+t}) = m_n(\sigma_s \boxtimes \sigma_t) = \sum_{\pi \in NC(n)} \kappa_\pi(\sigma_s) \cdot m_{K(\pi)}(\sigma_t)$

free cumulants of the free log-normal distribution ($t < 0$)

remark

The Bercovici-Pata bijection fails, i.e. the log-normal distribution, which is $*$ -infinite divisible, is not mapped to the free log-normal distribution.

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remark

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open question

Is the free log-normal distribution σ_t , $t < 0$, \boxplus -infinite divisible?

\Leftrightarrow Is the sequence $m_n := \frac{(n+2)^n}{(n+1)!}$ positive definite, i.e. is m_n a moment sequence?

density function of the free rolled-up normal distribution σ_t

$t > 0$: $\sigma_t \dots$ measure on the unit circle \mathbb{T}
 $\tilde{\sigma}_t \dots$ uncoiled measure on $[-\pi, \pi]$

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$t > 0$: $\sigma_t \dots$ measure on the unit circle \mathbb{T}
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Theorem

Let $t > 0$. The measure $\tilde{\sigma}_t$ is absolutely continuous. The density function is continuous, even, and monotone raising on $[-\pi, 0]$.

$$\text{supp}(\tilde{\sigma}_t) = \begin{cases} [-c, c] & 0 < t < 4 \\ [-\pi, \pi] & t \geq 4 \end{cases}$$

where

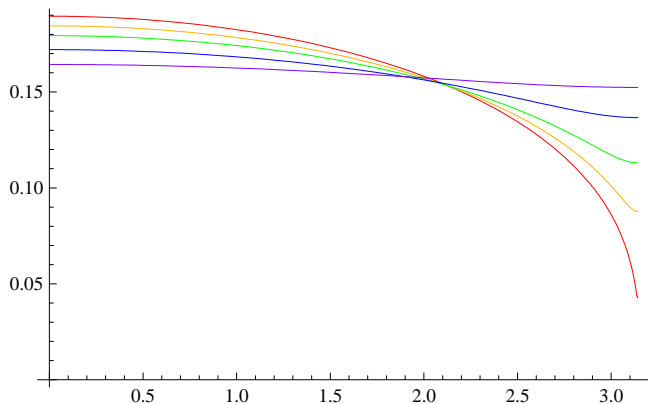
$$c = 2 \arctan \left(\sqrt{\frac{t}{4-t}} \right) + \sqrt{t \left(1 - \frac{t}{4} \right)}.$$

approximation of the density function

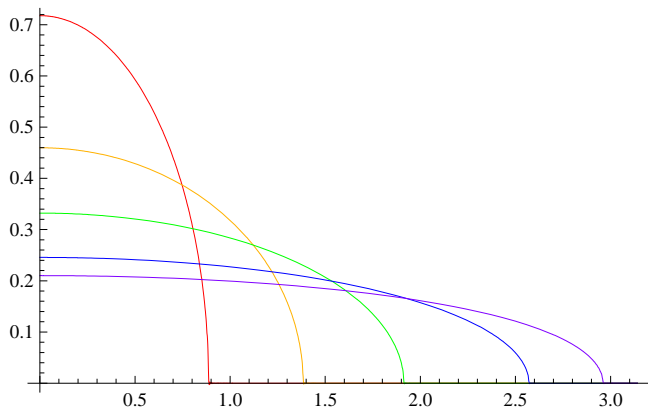
Fourier series of the density function of $\tilde{\sigma}_t$

$$f(x) = \frac{1}{2\pi} + \frac{1}{\pi} \sum_{n=1}^{\infty} \underbrace{\exp\left(-\frac{nt}{2}\right) {}_1F_1(1-n; 2; nt)}_{=m_n(\sigma_t)} \cos(nx)$$

- $0 < t \leq 4$: pointwise convergence
- $t > 4$: uniform convergence

$\tilde{\sigma}_t$ density function plots

$t = 4.1$ $t = 4.5$ $t = 5$ $t = 6$ $t = 8$

$\tilde{\sigma}_t$ density function plots

$t = 0.1$ $t = 0.5$ $t = 1$ $t = 2$ $t = 3$

Notation

Let μ a measure on the torus \mathbb{T} .

uncoiled measure $\tilde{\mu}$ on $[-\pi, \pi]$

- $\tilde{\mu}(A) := \mu(\{e^{ix} \mid x \in A\})$ for all $A \in \mathcal{B}((-\pi, \pi))$
- $\tilde{\mu}(\{-\pi\}) = \tilde{\mu}(\{\pi\}) = \frac{1}{2}\mu(\{-1\})$

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$0 < \varepsilon \leq 1$: Define measure $\mu^{(\varepsilon)}$ on \mathbb{T} by dilatation of the uncoiled measure $\tilde{\mu}$

- $\mu^{(\varepsilon)}([-\pi, -\varepsilon\pi] \cup (\varepsilon\pi, \pi]) = 0$
- $\mu^{(\varepsilon)}(A) = \tilde{\mu}(\frac{1}{\varepsilon}A)$ for all $A \in \mathcal{B}([-\varepsilon\pi, \varepsilon\pi])$

rolled-up normal distribution ν_t

$$\int_{\mathbb{T}} f(z) d\nu_t := \int_{-\infty}^{\infty} f(e^{ix}) \frac{1}{\sqrt{2\pi t}} e^{-\frac{x^2}{2t}} dx$$

Classical multiplicative CLT

Let μ a probability measure on \mathbb{T} with $m_1(\tilde{\mu}) = 0$ and let

$$\varepsilon_n = \frac{t}{\sqrt{n} \int_{-\pi}^{\pi} x^2 d\tilde{\mu}(x)}.$$

Then the sequence

$$\mu_n = \underbrace{\mu^{(\varepsilon_n)} \circledast \dots \circledast \mu^{(\varepsilon_n)}}_{n \text{ times}}$$

converge weakly to the rolled-up normal distribution ν_t .

Free multiplicative CLT

Let μ a probability measure on \mathbb{T} with *all moments real* and let

$$\varepsilon_n = \frac{t}{\sqrt{n} \int_{-\pi}^{\pi} x^2 d\tilde{\mu}(x)}.$$

Then the sequence

$$\mu_n = \underbrace{\mu^{(\varepsilon_n)} \boxtimes \dots \boxtimes \mu^{(\varepsilon_n)}}_{n \text{ times}}$$

converge weakly to the free rolled-up normal distribution σ_t .

G. P. Chistyakov and F. Götze, *Limit theorems in free probability theory II*, Cent. Eur. J. Math., 2008