

Convolution semigroups for operator-valued distributions

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Joint works with M. Popa and V. Vinnikov and with
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- 1 Background and context
 - Some results from scalar-valued free probability
 - Operator-valued distributions and fully matricial maps
 - Operator-valued free infinite divisibility
- 2 Operator-valued free convolution semigroups
 - A limit theorem
 - Free convolution semigroups indexed by cp maps

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Semigroups with respect to free additive convolution

Scalar-valued context.

Let $X = X^* \in (\mathcal{A}, \tau)$ be a random variable in a non-commutative probability space. We denote by μ_X its distribution with respect to τ . Some analytic tools:

- The Cauchy transform

$$G_{\mu_X}(z) = G_X(z) = \tau \left[(z - X)^{-1} \right], \quad \Im z \neq 0;$$

- The R -transform:

$$R_X(z) = G_X^{-1}(z) - \frac{1}{z}, \quad |z| \text{ small, } \left| \frac{\Re z}{\Im z} \right| \text{ bounded.}$$

Relevance: if X_1, X_2, \dots, X_t are free, then (Voiculescu '86)

$$R_{X_1}(z) + R_{X_2}(z) + \dots + R_{X_t}(z) = R_{X_1 + X_2 + \dots + X_t}(z).$$

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Notation/theorem (Voiculescu): $\mu_{X_1+X_2} = \mu_{X_1} \boxplus \mu_{X_2}$. If $\mu_{X_1} = \cdots = \mu_{X_t}$, we shall write $\mu_{X_1} \boxplus \cdots \boxplus \mu_{X_t} = \mu_{X_1}^{\boxplus t}$. Note:

$$R_{\mu_{X_1}^{\boxplus t}}(z) = tR_{\mu_{X_1}}(z). \quad (1)$$

What if $t \notin \mathbb{N}$? **Unlike** in classical probability, for any $t \in [1, +\infty)$ and probability measure μ on \mathbb{R} , there exists a probability $\mu^{\boxplus t}$ on \mathbb{R} so that (1) holds; $\{\mu^{\boxplus t} : t \geq 1\}$ forms a partial semigroup.

- Proved first by Bercovici and Voiculescu for large $t > 1$ with R -transform methods (1995);
- Then by Nica and Speicher for any $t \geq 1$ through a specific operatorial construction (1998): if $p_t = p_t^* = p_t^2$ is free from X and $\tau(p_t) = 1/t$, then $\mu_{p_t X p_t} = (1 - \frac{1}{t})\delta_0 + \frac{1}{t}D_{\frac{1}{t}}(\mu_X^{\boxplus t})$.

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Scalar-valued context.

Like in classical probability, there exists a special class of distributions μ for which $\mu^{\boxplus t}$ makes sense for any $t \geq 0$, namely the *freely infinitely divisible* ones.

Theorem

- (Bercovici-Voiculescu) A probability measure μ on \mathbb{R} is \boxplus -infinitely divisible if and only if there are $a \in \mathbb{R}$ and ρ a positive measure so that

$$R_\mu(1/z) = a - \int_{\mathbb{R}} \frac{1 + tz}{z - t} d\rho(t), \quad z \notin \mathbb{R}.$$

- (Bercovici-Pata) A probability measure μ on \mathbb{R} is \boxplus -infinitely divisible if and only if it is the distributional limit of an infinitesimal array of free random variables.

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Operator-valued distributions and fully matricial maps

Definition

(Voiculescu) Operator-valued non-commutative probability space: $(\mathcal{A}, \mathbb{E}_B, B)$, where \mathcal{A} is a unital C^* -algebra, B is a C^* -subalgebra of \mathcal{A} containing the unit of \mathcal{A} , and $\mathbb{E}_B: \mathcal{A} \rightarrow B$ is a unit-preserving *conditional expectation*.

Distribution of $X \in \mathcal{A}$ with respect to \mathbb{E}_B :

$$\mu_X = \{m_n^X: B^{n-1} \rightarrow B: m_n^X(b_1, \dots, b_{n-1}) = \mathbb{E}_B(Xb_1 \cdots Xb_{n-1}X)\}.$$

Operator valued Cauchy-Stieltjes transform of an $X = X^* \in \mathcal{A}$:

$$G_X(b) = \mathbb{E}_B \left[(b - X)^{-1} \right], \quad \Im b \geq \varepsilon$$

for some $\varepsilon > 0$ (call this $\Im b > 0$). Note:

$$\Im G_X(b) < 0, \quad \lim_{\|b^{-1}\| \rightarrow 0} bG_X(b) = \lim_{\|b^{-1}\| \rightarrow 0} G_X(b)b = 1.$$

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$G_X(b) = \mathbb{E}_B [(b - X)^{-1}] = \sum \mathbb{E}_B (b^{-1} (Xb^{-1})^n), \quad \|b^{-1}\| < \frac{1}{\|X\|},$
 so $G_X(b)$ allows us to recover the *symmetric moments* of X :
 $m_n(b^{-1}, \dots, b^{-1}) = \mathbb{E}_B (Xb^{-1}X \cdots b^{-1}X).$

Complete positivity, fully matricial maps (Voiculescu):

$\mathbb{E}_B \otimes \text{Id}_{\mathcal{M}_n(\mathbb{C})} : \mathcal{A} \otimes \mathcal{M}_n(\mathbb{C}) \rightarrow B \otimes \mathcal{M}_n(\mathbb{C})$ remains positive.

(Call $\mathbb{E}_B \otimes \text{Id}_{\mathcal{M}_n(\mathbb{C})} = \mathbb{E}_{\mathcal{M}_n(B)}$.) Can define

$$G_{X \otimes 1_n}(b) = \mathbb{E}_{\mathcal{M}_n(B)} \left[(b - X \otimes 1_n)^{-1} \right], \quad b \in \mathcal{M}_n(B), \Im b > 0,$$

$$G_{X \otimes 1_n}(b) = \sum_{n=0}^{\infty} \mathbb{E}_{\mathcal{M}_n(B)} (b^{-1} ((X \otimes 1_n) b^{-1})^n)$$

for any (a) $\|b^{-1}\| < \|X\|^{-1}$ or (b) “ b^{-1} nilpotent”.

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for any (a) $\|b^{-1}\| < \|X\|^{-1}$ or (b) “ b^{-1} nilpotent”.

Operator-valued distributions and fully matricial maps

Meaning: if b_1, \dots, b_n are given, and

$$b = \begin{pmatrix} 0 & b_1 & 0 & \cdots & 0 & 0 \\ 0 & 0 & b_2 & \cdots & 0 & 0 \\ \vdots & \vdots & \vdots & \vdots & \vdots & \vdots \\ 0 & 0 & 0 & \cdots & 0 & b_n \\ 0 & 0 & 0 & \cdots & 0 & 0 \end{pmatrix}, \quad \text{then}$$

$$\mathbb{E}_{\mathcal{M}_n(B)}(b((X \otimes 1_{n+1})b)^{n-1}) = \begin{pmatrix} 0 & 0 & \cdots & \mathbb{E}_B(b_1 X b_2 \cdots X b_n) \\ 0 & 0 & \cdots & 0 \\ \vdots & \vdots & \vdots & \vdots \\ 0 & 0 & \cdots & 0 \end{pmatrix},$$

and $\mathbb{E}_B \otimes \text{Id}_{\mathcal{M}_{n+1}(\mathbb{C})}(b((X \otimes 1_{n+1})b)^j) = 0$ for $j \geq n$. Set $b_1 = b_n = 1$ to get $m_{n-1}^X(b_2, \dots, b_{n-1})$

Operator-valued distributions and fully matricial maps

Free convolution: (Voiculescu '95) If $(\mathcal{A}, \mathbb{E}_B, B)$ is an op-valued noncommutative probability space and $X = X^*, Y = Y^* \in \mathcal{A}$ is free over B , then $\mu_{X+Y} = \mu_X \boxplus \mu_Y$ is determined by μ_X and μ_Y and is called the *free additive convolution* of μ_X and μ_Y .

Analytic tools for the study of $\mu_X \boxplus \mu_Y$:

1. Op-valued R -transform: $R_{X \otimes 1_n}(b) = G_{X \otimes 1_n}^{-1}(b) - b^{-1}$ satisfies

$$R_{\mu_X \boxplus \mu_Y}(b) = R_{\mu_X}(b) + R_{\mu_Y}(b), \quad \|b\| \text{ small.}$$

2. *Analytic subordination* property: there exists a unique fully matricial self-map $\omega_1^{(n)}$ of the upper half-plane $\mathcal{M}_n(B)^+$ so that

$$\mathbb{E}_{\mathcal{M}_n(B[X])} \left((b - (X + Y) \otimes 1_n)^{-1} \right) = (\omega_1^{(n)}(b) - X \otimes 1_n)^{-1}.$$

By applying $\mathbb{E}_{\mathcal{M}_n(B)}$, we get $G_{X \otimes 1_n} \circ \omega_1^{(n)} = G_{(X+Y) \otimes 1_n}$.

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Definition

(Speicher) The B -valued distribution μ_X is \boxplus -infinitely divisible if for any $n \in \mathbb{N}$ there exist B -free identically distributed selfadjoint B -valued random variables X_1, \dots, X_n so that

$$\mu_X = \mu_{X_1 + \dots + X_n} = \mu_{X_1}^{\boxplus n}.$$

Few examples: an op-valued CLT (op-valued semicircle law) of Voiculescu ('95) and an op-valued free Poisson (Speicher).

Theorem

(M. Popa, V. Vinnikov) The B -valued distribution μ is \boxplus -infinitely divisible if and only if there are $a = a^* \in B$ and $\rho: B[X] \rightarrow B$ completely positive map so that

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Assume that $\{X_{jk} : j \in \mathbb{N}, 1 \leq k \leq k_j\}$ is a triangular array of selfadjoint random variables in $(\mathcal{A}, \mathbb{E}_B, B)$ with identically distributed rows. Assume in addition that $\limsup_{j \rightarrow \infty} \|X_{j1} + \cdots + X_{jk_j}\| \leq M$ for some fixed $M \geq 0$. If $\lim_{j \rightarrow \infty} X_{j1} + \cdots + X_{jk_j}$ exists in distribution as norm-limit of moments, then the limit distribution is freely infinitely divisible.

Idea of proof: note that, as in the scalar case, $G_{X_{j1} + \cdots + X_{jk_j}}(b) = G_{\boxplus k_j}(b) = G_{\mu_{X_{j1}}}(\omega_j(b))$ (Voiculescu's subordination) and $b = k_j \omega_j(b) + (k_j - 1)G_{\mu_{X_{j1}}}(\omega_j(b))^{-1}$ (R -transform). So (i) $\omega_j(b) = G_{\nu_j}(b)^{-1}$, where $\nu_j = (\mu_{X_{j1}}^{\boxplus k_j})^{\uplus 1 - \frac{1}{k_j}}$, and (ii) the R -transform of ν_j is $R_{\nu_j}(b^{-1}) = (k_j - 1)(b - G_{\mu_{X_{j1}}}(b))^{-1}$.

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Free convolution semigroups indexed by cp maps

Curran: Generalization of the Nica-Speicher semigroup to op-valued distributions. In an op-valued noncommutative probability space $(\mathcal{A}, \mathbb{E}_B, B, \tau)$ where τ is a tracial state on \mathcal{A} so that $\tau = \tau \circ \mathbb{E}_B$, a projection p of trace t which is free over B from $X = X^* \in \mathcal{A}$ and classically independent from B wrt τ , shows a subordination property in Voiculescu's sense for G_{pX} .

Aside: One can show, using Dykema's version of the S -transform, that under the hypotheses that $\mathbb{E}_B(p)$ is invertible and belongs to the centre of B , that

$$\mathbb{E}_B(p)^{-1} \mathbf{R}_{D_{\mathbb{E}_B(p)}(\mu_X)}(\mathbb{E}_B(p)^{-1} b) = \mathbb{E}_B(p)^{-1} \Psi_{Xp} \left(\mathbb{E}_B(p)^{-1} b [1 + \mathbb{E}_B(p)^{-1} \mathbf{R}_{D_{\mathbb{E}_B(p)}(\mu_X)}(\mathbb{E}_B(p)^{-1} b)]^{-1} \right),$$

which is essentially the proof of Nica and Speicher from the scalar-valued context.

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Free convolution semigroups indexed by cp maps

This (among others) suggests “generalizing” $t > 1$ to a cp map:

Theorem

For any cp map $\alpha: B \rightarrow B$ so that $\alpha - 1$ is still cp and invertible, and for any B -valued selfadjoint random variable X , there exists a B -valued selfadjoint random variable X_α so that

$$\alpha(R_{\mu_X}(b)) = R_{\mu_{X_\alpha}}(b).$$

Denote μ_{X_α} by $\mu_X^{\boxplus \alpha}$.

One (rough) idea of proof: (i) show that $\mu_X^{\boxplus \alpha}$ exists for \boxplus -infinitely divisible distributions (easier, also thanks to the Popa-Vinnikov characterization); (ii) show that $\mu_X^{\uplus \alpha}$ exists for any cp map α ; (iii) Show that $\mu_X^{\boxplus \alpha} = (\mathbb{B}(\mu^{\uplus \alpha}))^{\uplus(\alpha-1)^{-1}\alpha}$, where \mathbb{B} is the *Boolean-to-free Bercovici-Pata bijection*.

Free convolution semigroups indexed by cp maps

As in the scalar-valued context, the Boolean-to-free Bercovici-Pata bijection from the operator-valued context embeds in a semigroups of maps on the space of distributions:

$$\{\mathbb{B}_\alpha(\mu) = \left(\mu^{\boxplus(1+\alpha)}\right)^{\boxplus(1+\alpha)^{-1}} : \alpha: \mathcal{B} \rightarrow \mathcal{B} \text{ cp map}\}.$$

$\mathbb{B}_\alpha(\mu)$ is \boxplus -infinitely divisible for any μ when $\alpha = 1$; we get then the Boolean-to-free Bercovici-Pata bijection. Moreover, again as in the scalar-valued context, the map $h(\alpha, b) = G_{\mathbb{B}_\alpha(\mu)}(b)^{-1} - b$ satisfied a “complex Burgers equation:”

$$\frac{\partial h(\alpha, b)}{\partial \alpha}(\rho) - \frac{\partial h(\alpha, b)}{\partial b}(\rho(h(\alpha, b))) = 0,$$

where ρ, α are cp, and $\Im b > 0$.

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Concluding remarks

- One can generalize many results related to convolution semigroups to the operator-valued context;
- Analytic tools (the fully matricial maps) work in the operator-valued case **almost** as well as in the scalar valued case;
- The **full strength** of Voiculescu's subordination result has not been used in the proofs. (So there must be more...)

Thank you!