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On a surprising relation between the Marchenko–Pastur law, rectangular and square free convolutions

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Abstract. In this paper, we prove a result linking the square and the rectangular R-transforms, the consequence of which is a surprising relation between the square and rectangular versions the free additive convolutions, involving the Marchenko–Pastur law. Consequences on random matrices, on infinite divisibility and on the arithmetics of the square versions of the free additive and multiplicative convolutions are given.

Résumé. Dans cet article, on prouve un résultat reliant les versions carré et rectangulaire de la R-transformée, qui a pour conséquence une relation surprenante entre les versions carré et rectangulaire de la convolution libre additive, impliquant la loi de Marchenko–Pastur. On donne des conséquences de ce résultat portant sur les matrices aléatoires, sur l'infinie divisibilité et sur l'arithmétique des versions carré des convolutions additives et multiplicatives.

Introduction

Free convolutions are operations on probability measures on the real line which allow to compute the empirical spectral[1] or singular[2] measures of large random matrices which are expressed as sums or products of independent random matrices, the spectral measures of which are known. More specifically, the operations $\boxplus, \boxtimes$, called respectively *free additive and multiplicative convolutions* are defined in the following way [19]. Let, for each n, M_n, N_n be n by n independent random hermitian matrices, one of them having a distribution which is invariant under the action of the unitary group by conjugation, the empirical spectral measures of which converge, as n tends to infinity, to non random probability measures denoted respectively by τ_1, τ_2. Then $\tau_1 \boxplus \tau_2$ is the limit of the empirical spectral law of $M_n + N_n$ and, in the case where the matrices are positive, $\tau_1 \boxtimes \tau_2$ is the limit of the empirical spectral law of $M_n N_n$. In the same way, for any $\lambda \in [0, 1]$, the *rectangular free convolution* $\boxplus_\lambda$ is defined in [7], in the following way. Let $M_{n,p}$, $N_{n,p}$ be n by p independent random matrices, one of them having a distribution which is invariant by multiplication by any unitary matrix on any side, the symmetrized[3] empirical singular measures of which tend, as n, p tend to infinity in

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[1] The *empirical spectral measure* of a matrix is the uniform law on its eigenvalues with multiplicity.

[2] The *empirical singular measure* of a matrix M with size n by p ($n \leq p$) is the empirical spectral measure of $|M| := \sqrt{MM^*}$.

[3] The *symmetrization* of a law μ on $[0, +\infty)$ is the law ν defined by $\nu(A) = \frac{\mu(A) + \mu(-A)}{2}$ for all Borel set A. Dealing with laws on $[0, +\infty)$ or with their symmetrizations is equivalent, but for historical reasons, the rectangular free convolutions have been defined with symmetric laws. In all this paper, we shall often pass from symmetric laws to laws on $[0, +\infty)$ and vice-versa. Thus in order to avoid confusion, we shall mainly use the letter μ for laws on $[0, \infty)$ and ν for symmetric ones.

such a way that n/p tends to λ, to nonrandom probability measures ν_1, ν_2. Then the symmetrized empirical singular law of $M_{n,p} + N_{n,p}$ tends to $\nu_1 \boxplus_\lambda \nu_2$. These operations can also, equivalently, be defined in reference to free elements of a noncommutative probability space, but in this paper, we have chosen to use the random matrix point of view.

In the cases $\lambda = 0$ or $\lambda = 1$, i.e. where the rectangular random matrices considered in the previous definition are either "almost flat" or "almost square", the rectangular free convolution with ratio λ can be expressed with the additive free convolution: $\boxplus_1 = \boxplus$ and for all symmetric laws ν_1, ν_2, $\nu_1 \boxplus_0 \nu_2$ is the symmetric law the push-forward by the map $t \mapsto t^2$ of which is the free additive convolution of the push forwards of ν_1 and ν_2 by the same map. These surprising relations have no simple explanations, but they allow to hope a general relation between the operations $\boxplus_\lambda$ and $\boxplus$, which would be true for any λ. Up to now, despite many efforts, no such relation had been found, until a paper of Debbah and Ryan [13], where a relation between $\boxplus_\lambda, \boxplus$ and $\boxtimes$ is proved in a particular case. In the present paper, we give a shorter proof of a wide generalization[4] of their result: for any $\lambda \in (0, 1]$, we define μ_λ to be the law of λ times a random variable with law the Marchenko–Pastur law with mean $1/\lambda$, and we prove that for any pair μ, μ' of probability measures on $[0, +\infty)$, we have

$$\sqrt{\mu \boxtimes \mu_\lambda} \boxplus_\lambda \sqrt{\mu' \boxtimes \mu_\lambda} = \sqrt{(\mu \boxplus \mu') \boxtimes \mu_\lambda}, \tag{1}$$

where for any probability measure ρ on $[0, +\infty)$, $\sqrt{\rho}$ denotes the symmetrization of the push-forward of ρ by the map $t \mapsto \sqrt{t}$. Our proof is based on the following relation between the R-transform[5] R_μ of a probability measure μ on $[0, +\infty)$ and the rectangular R-transform $C_{\sqrt{\mu \boxtimes \mu_\lambda}}$ with ratio λ of $\sqrt{\mu \boxtimes \mu_\lambda}$: we prove that for all z,

$$R_\mu(z) = C_{\sqrt{\mu \boxtimes \mu_\lambda}}(z).$$

This relation also allows us to prove precise relations between $\boxplus$-infinitely divisible laws and $\boxplus_\lambda$-infinitely divisible laws.

We would like to observe that formula (1) has some consequences which are far from obvious. It means that for n, p large integers such that $n/p \simeq \lambda$, for A, B, M, M' independent random matrices with respective sizes $n \times n$, $n \times n$, $n \times p$ and $n \times p$ such that A, B are invariant in law under left and right multiplication by unitary matrices and M, M' have independent Gaussian entries, then as far as the spectral measure is concerned,

$$(AM + BM')(AM + BM')^* \simeq AM(AM)^* + BM'(BM')^*. \tag{2}$$

It also means, if $1 \ll n \ll p$, that for X, Y independent $n \times p$ random matrices, as far as the spectrums are concerned,

$$(X + Y)(X + Y)^* \simeq XX^* + YY^*. \tag{3}$$

The relation (1) has also consequences on the arithmetics of free additive and multiplicative convolutions $\boxplus$ and $\boxtimes$ (Corollaries 7 and 12) which was not known yet, despite the many papers written the last years about questions related to this subject, e.g. [1,3,4,8,10–12].

1. A relation between the Marchenko–Pastur law, the square and the rectangular free convolutions

1.1. *Prerequisites on square and rectangular analytic transforms of probability measures*

1.1.1. *The square case*: *The R- and S-transforms*
These are analytic transforms of probability measures which allow to compute the operations $\boxplus$ and $\boxtimes$, like the Fourier transform for the classical convolution. The R-transform can be defined for any probability measure on the real line,

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[4]See Remark 5.

[5]Note that there are two conventions regarding the R-transform. The one we use is the one used in the combinatorial approach to freeness [17], which is not exactly the one used in the analytic approach [16]: $R_\mu^{\text{combinatorics}}(z) = zR_\mu^{\text{analysis}}(z)$.

but we shall only define it for probability measures on $[0, +\infty)$. Consider such a probability measure μ. It $\mu = \delta_0$, then $R_\mu = S_\mu = 0$. Now, let us suppose that $\mu \neq \delta_0$. Let us define the function

$$M_\mu(z) = \int_{t \in \mathbb{R}} \frac{tz}{1 - tz} \, d\mu(t).$$

Then the *R*- *and S-transforms*[6] of μ, denoted respectively by R_μ and S_μ are the analytic functions defined as follows

$$R_\mu(z) = \left[(1 + z) M_\mu^{-1}(z) \right]^{-1}, \qquad S_\mu(z) = \frac{1 + z}{z} M_\mu^{-1}(z), \tag{4}$$

where the exponent $^{-1}$ refers to the inversion of functions with respect to the operation of composition $\circ$. Note that M_μ is an analytic function defined in $\{z \in \mathbb{C}; 1/z \notin \text{support}(\mu)\}$. Hence in the case where μ is compactly supported, the functions M_μ and $(1 + z) M_\mu^{-1}(z)$ can be inverted in a neighborhood of zero as analytic functions in a neighborhood of zero vanishing at zero, with non null derivative at zero. In the case where μ is not compactly supported, these functions are inverted as functions on intervals $(-\varepsilon, 0)$ which are equivalent to (positive constant) $\times z$ at zero [9].

Note that putting together both equations of (4), one gets

$$S_\mu(z) = \frac{1}{z} R_\mu^{-1}(z) = \frac{1 + z}{z} M_\mu^{-1}(z). \tag{5}$$

The main properties of the *R*- and *S*-transforms are the fact that they characterize measures and their weak convergence and that they allow to compute free convolutions: for all μ, ν,

$$R_{\mu \boxplus \nu} = R_\mu + R_\nu \quad \text{and} \quad S_{\mu \boxtimes \nu} = S_\mu S_\nu. \tag{6}$$

1.1.2. *The rectangular case*: *The rectangular R-transform with ratio* λ

In the same way, for $\lambda \in [0, 1]$, the rectangular free convolution with ratio λ can be computed with an analytic transform of probability measures. Let ν be a symmetric probability measure on the real line. Let us define $H_\nu(z) = z(\lambda M_{\nu^2}(z) + 1)(M_{\nu^2}(z) + 1)$, where ν^2 denotes the push forward of ν by the map $t \mapsto t^2$. Then with the same conventions about inverses of functions than in the previous section, the *rectangular R-transform with ratio* λ of ν is defined to be

$$C_\nu(z) = U\left(\frac{z}{H_\nu^{-1}(z)} - 1 \right),$$

where $U(z) = \frac{-\lambda - 1 + [(\lambda + 1)^2 + 4\lambda z]^{1/2}}{2\lambda}$ for $\lambda > 0$ and $U(z) = z$ for $\lambda = 0$. By Theorems 3.8, 3.11 and 3.12 of [7], the rectangular *R*-transform characterizes measures and their weak convergence, and for all pair ν_1, ν_2 of symmetric probability measures, $\nu_1 \boxplus_\lambda \nu_2$ is characterized by the fact that

$$C_{\nu_1 \boxplus_\lambda \nu_2} = C_{\nu_1} + C_{\nu_2}. \tag{7}$$

1.2. *A relation between the square and the rectangular R-transforms*

Let us fix $\lambda \in [0, 1]$. We recall that for any probability measure ρ on $[0, +\infty)$, $\sqrt{\rho}$ denotes the symmetrization of the push-forward of ρ by the map $t \mapsto \sqrt{t}$ and that for $\lambda > 0$, we have defined μ_λ to be the law of λ times a random variable with law the Marchenko–Pastur law with mean $1/\lambda$, i.e. μ_λ is the law with support $[(1 - \sqrt{\lambda})^2, (1 + \sqrt{\lambda})^2]$ and density

$$x \mapsto \frac{\sqrt{4\lambda - (x - 1 - \lambda)^2}}{2\pi \lambda x}.$$

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[6]See footnote 5.

For $\lambda = 0$, we let μ_λ denote the Dirac mass at 1.

Theorem 1. *Let μ be a probability measure on $[0, +\infty)$. Then we have*

$$R_\mu(z) = C_{\sqrt{\mu \boxtimes \mu_\lambda}}(z).$$

Remark 2 (The cumulants point of view). *Suppose μ to be compactly supported. Let us denote the free cumulants [17] of μ by $(k_n(\mu))_{n \geq 1}$ and the rectangular free cumulants with ratio λ [7,6] of $\sqrt{\mu \boxtimes \mu_\lambda}$ by $(c_{2n}(\sqrt{\mu \boxtimes \mu_\lambda}))_{n \geq 1}$. Then the previous theorem means that for all $n \geq 1$, one has*

$$k_n(\mu) = c_{2n}\left(\sqrt{\mu \boxtimes \mu_\lambda}\right). \tag{8}$$

Proof of Theorem 1. • First of all, note that by continuity of the applications $\mu \mapsto \mu \boxtimes \mu_\lambda$, $\rho \mapsto R_\rho$ and $\nu \mapsto C_\nu$ with respect to weak convergence [7,9], it suffices to prove the result in the case where μ is compactly supported. In this case, the functions $M_\mu, R_\mu, S_\mu, M_{\mu \boxtimes \mu_\lambda}, H_{\sqrt{\mu \boxtimes \mu_\lambda}}, C_{\sqrt{\mu \boxtimes \mu_\lambda}}$ are analytic in a neighborhood of zero and the operations of inversion on these functions or related ones can be used without precaution.

• If $\lambda > 0$, the free cumulants of the Marchenko–Pastur law with mean $1/\lambda$ are all equal to $1/\lambda$, thus the ones of μ_λ are given by the formula $k_n(\mu_\lambda) = \lambda^{n-1}$ for all $n \geq 1$ and $R_{\mu_\lambda}(z) = \sum_{n \geq 1} \lambda^{n-1} z^n$. From (5), it follows that $S_{\mu_\lambda}(z) = \frac{1}{1+\lambda z}$. Hence by (6), we have $S_{\mu \boxtimes \mu_\lambda}(z) = \frac{S_\mu(z)}{1+\lambda z}$, and by (5),

$$M_{\mu \boxtimes \mu_\lambda}(z) = \left(\frac{M_\mu^{-1}(z)}{1 + \lambda z}\right)^{-1}. \tag{9}$$

Note that since $\mu_0 = \delta_1$, (9) stays true if $\lambda = 0$. Now, let us define the function $T(x) = (\lambda x + 1)(x + 1)$. Note that $T(U(x - 1)) = x$ for x in a neighborhood of zero. We have

$$H_{\sqrt{\mu \boxtimes \mu_\lambda}}(z) = z \times T \circ M_{\mu \boxtimes \mu_\lambda}(z) = z \times T \circ \left(\frac{M_\mu^{-1}(z)}{1 + \lambda z}\right)^{-1}$$

and

$$C_{\sqrt{\mu \boxtimes \mu_\lambda}}(z) = U\left(\frac{z}{(z \times T \circ (M_\mu^{-1}(z)/(1+\lambda z))^{-1})^{-1}} - 1\right). \tag{10}$$

• Hence by (4) and (10), we have the following equivalence

$$R_\mu = C_{\sqrt{\mu \boxtimes \mu_\lambda}} \quad \Longleftrightarrow \quad ((z+1)M_\mu^{-1}(z))^{-1} = U\left(\frac{z}{(z \times T \circ (M_\mu^{-1}(z)/(1+\lambda z))^{-1})^{-1}} - 1\right)$$

$$\Longleftrightarrow \quad T \circ ((z+1)M_\mu^{-1}(z))^{-1} = \frac{z}{(z \times T \circ (M_\mu^{-1}(z)/(1+\lambda z))^{-1})^{-1}}$$

$$\Longleftrightarrow \quad \left(z \times T \circ \left(\frac{M_\mu^{-1}(z)}{1 + \lambda z}\right)^{-1}\right)^{-1} \times T \circ ((z+1)M_\mu^{-1}(z))^{-1} = z.$$

Composing both terms on the right by $(z + 1)M_\mu^{-1}(z)$, it gives

$$R_\mu = C_{\sqrt{\mu \boxtimes \mu_\lambda}} \quad \Longleftrightarrow \quad \left(z \times T \circ \left(\frac{M_\mu^{-1}(z)}{1 + \lambda z}\right)^{-1}\right)^{-1} \circ ((z+1)M_\mu^{-1}(z)) \times T(z) = (z+1)M_\mu^{-1}(z).$$

Dividing by $T(z)$, it gives

$$
\begin{aligned}
R_\mu = C_{\sqrt{\mu \boxtimes \mu_\lambda}} \quad &\iff \quad \left(z \times T \circ \left(\frac{M_\mu^{-1}(z)}{1 + \lambda z} \right)^{-1} \right)^{-1} \circ \left((z+1) M_\mu^{-1}(z) \right) = \frac{M_\mu^{-1}(z)}{1 + \lambda z} \\
&\iff \quad (z+1) M_\mu^{-1}(z) = \left(z \times T \circ \left(\frac{M_\mu^{-1}(z)}{1 + \lambda z} \right)^{-1} \right) \circ \frac{M_\mu^{-1}(z)}{1 + \lambda z} \\
&\iff \quad (z+1) M_\mu^{-1}(z) = \frac{M_\mu^{-1}(z)}{1 + \lambda z} T(z),
\end{aligned}
$$

which is obviously true by definition of $T(z)$. $\qquad\square$

1.3. *Main result of the paper*

The main theorem of this paper is the following one. $\lambda \in [0, 1]$ is still fixed.

Theorem 3. *For any pair μ, μ' of probability measures on $[0, +\infty)$, we have*

$$
\sqrt{\mu \boxtimes \mu_\lambda} \boxplus_\lambda \sqrt{\mu' \boxtimes \mu_\lambda} = \sqrt{(\mu \boxplus \mu') \boxtimes \mu_\lambda}. \tag{11}
$$

Remark 4. *Note that in the case where $\lambda = 0$, this theorem expresses what we already knew about $\boxplus_0$ (and which is explained in the second paragraph of the [Introduction]), but that the case $\lambda = 1$ is not a consequence of the already known formula $\boxplus_1 = \boxplus$.*

Remark 5. *Part of this theorem could have been deduced from Theorem 6 of [13]. However, (11) could be deduced from the theorem of Debbah and Ryan only for laws μ, μ' which can be expressed as limit singular laws of n by p (for $n/p \simeq \lambda$) corners of large $p \times p$ bi-unitarily invariant random matrices, but it follows from Theorem 14.10 of [17] that not every law has this form. Moreover, even though the idea which led us to our result was picked in the pioneer work of Debbah and Ryan, our proof is much shorter and shows the connection with the rectangular machinery in a more clear way (via Theorem 1 and Remark 2).*

Proof of Theorem 3. Define $\nu := \sqrt{\mu \boxtimes \mu_\lambda} \boxplus_\lambda \sqrt{\mu' \boxtimes \mu_\lambda}$. By (7), we have

$$
C_\nu = C_{\sqrt{\mu \boxtimes \mu_\lambda}} + C_{\sqrt{\mu' \boxtimes \mu_\lambda}}.
$$

Thus, by Theorem 1 and (6), we have

$$
C_\nu = R_\mu + R_{\mu'} = R_{\mu \boxplus \mu'} = C_{\sqrt{(\mu \boxplus \mu') \boxtimes \mu_\lambda}}.
$$

Hence by injectivity of the rectangular R-transform (Theorem 3.8 of [7]), (11) is valid. $\qquad\square$

The formula (11) gives us a new insight on rectangular free convolutions: it allows to express it, in certain cases, in terms of the free convolutions "of square type" $\boxplus$ and $\boxtimes$. However, only laws which can be expressed under the form

$$
\sqrt{\mu \boxtimes \mu_\lambda} \quad \left(\mu \text{ probability measure on } [0, +\infty) \right) \tag{12}
$$

can have their rectangular convolution computed via formula (11). Thus it seems natural to ask whether all symmetric laws can be expressed like in (12). Note that it is equivalent to the fact that any law on $[0, +\infty)$ can be expressed under the form $\mu \boxtimes \mu_\lambda$, which is equivalent to the fact that the Dirac mass at one δ_1 can be expressed under the form $\mu \boxtimes \mu_\lambda$. Indeed, if $\delta_1 = \mu \boxtimes \mu_\lambda$, then any law τ on $[0, +\infty)$ satisfies $\tau = \tau \boxtimes \delta_1 = (\tau \boxtimes \mu) \boxtimes \mu_\lambda$. The following proposition shows that it is not the case. However, Theorem 11 will show that many symmetric laws can be expressed like in (12).

Proposition 6. *Unless $\lambda = 0$, the law $\frac{\delta_1 + \delta_{-1}}{2}$ cannot be expressed under the form $\sqrt{\mu \boxtimes \mu_\lambda}$ for μ probability measure on $[0, +\infty)$.*

Proof. Suppose that $\lambda > 0$ and that there is a probability measure μ on $[0, +\infty)$ such that $\frac{\delta_1 + \delta_{-1}}{2} = \sqrt{\mu \boxtimes \mu_\lambda}$. Then $\delta_1 = \mu \boxtimes \mu_\lambda$. This is impossible, by Corollary 3.4 of [2], which states that the free multiplicative convolution of two laws which are not Dirac masses has always a nonnull absolutely continuous part (there is another, more direct way to see that it is impossible: by (6), such a law μ has to satisfy $S_\mu(z) = 1 + \lambda z$, which implies that for z small enough, $M_\mu(z) = \frac{z - 1 + [(1-z)^2 + 4\lambda z]^{1/2}}{2\lambda}$: such a function does not admit any analytic continuation to $\mathbb{C} \setminus [0, +\infty)$, thus no such probability measure μ exists). $\qquad\square$

Theorem 3 has a consequence on the free convolutions "of square type" which was not known yet, despite the many papers written the last years about questions related to the arithmetics of these convolutions, e.g. [1,3,4,8,10–12].

Corollary 7. *For any pair μ, μ' of probability measures on $[0, +\infty)$ we have*

$$\sqrt{\mu \boxtimes \mu_1} \boxplus \sqrt{\mu' \boxtimes \mu_1} = \sqrt{(\mu \boxplus \mu') \boxtimes \mu_1}. \tag{13}$$

Proof. It is an obvious consequence of Theorem 3 and of the fact that $\boxplus_1 = \boxplus$. $\qquad\square$

Remark 8. *The referee of the paper communicated to us a proof of (13) which is not, as ours, based on computations on the R- and S-transforms, but on the direct proof of (8) in the special case $\lambda = 1$. Let us briefly outline this proof. When $\lambda = 1$, by [5], Eq. (4.1), (8) reduces to*

$$k_n(\mu) = k_{2n}\big(\sqrt{\mu \boxtimes \mu_1}\big). \tag{14}$$

Let a, s are free elements in a tracial non commutative probability space with respective distributions μ and the standard semicircle law. By [17], Proposition 12.13, s^2 has distribution μ_1, hence sas has distribution $\mu \boxtimes \mu_1$. It follows, by [17], Proposition 12.18, that for all n, the nth moment of μ is equal to $k_n(\mu \boxtimes \mu_1)$. But by [17], Proposition 11.25, for all n, we have

$$k_n(\mu \boxtimes \mu_1) = \sum_{\pi \in \mathrm{NC}(n)} \prod_{V \in \pi} k_{2|V|}\big(\sqrt{\mu \boxtimes \mu_1}\big).$$

It follows, using the expression of the nth moment of μ in terms of its free cumulants, that for all n,

$$\sum_{\pi \in \mathrm{NC}(n)} \prod_{V \in \pi} k_{|V|}(\mu) = \sum_{\pi \in \mathrm{NC}(n)} \prod_{V \in \pi} k_{2|V|}\big(\sqrt{\mu \boxtimes \mu_1}\big),$$

and that for all n, $k_n(\mu) = k_{2n}(\sqrt{\mu \boxtimes \mu_1})$.

2. Consequences on square and rectangular infinite divisibility

2.1. *Prerequisites on infinite divisibility and Lévy–Kinchine formulas*

Infinite divisibility is a fundamental probabilistic notion, at the base of Lévy processes, and which allows to explain deep relations between limit theorems for sums of either independent random variables, square or rectangular random matrices. Let us briefly recall basics of this theory [5,8,9,15,18].

Let $*$ denote the classical convolution of probability measures on the real line. Firstly, recall that a probability measure μ is said to be $*$-infinitely divisible (resp. $\boxplus$-, $\boxplus_\lambda$-infinitely divisible) if for all integer n, there exists a probability measure ν_n such that $\nu_n^{*n} = \mu$ (resp. $\nu_n^{\boxplus n} = \mu$, $\nu_n^{\boxplus_\lambda n} = \mu$). In this case, there exists a $*$- (resp. $\boxplus$-, $\boxplus_\lambda$-) semigroup $(\mu_t)_{t \geq 0}$ such that $\mu_0 = \delta_0$ and $\mu_1 = \mu$. For all t, μ_t is denoted by μ^{*t} (resp. $\mu^{\boxplus t}$, $\mu^{\boxplus_\lambda t}$). Infinitely divisible distributions have been classified: μ is $*$- (resp. $\boxplus$-) infinitely divisible if and only if there exists a real number γ

and a positive finite measure on the real line σ such that the Fourier transform is $\hat{\mu}(t) = \exp[i\gamma t + \int_{\mathbb{R}}(e^{itx} - 1 - \frac{itx}{x^2+1})\frac{x^2+1}{x^2}\,d\sigma(x)]$ (resp. $R_\mu(z) = \gamma z + z\int_{\mathbb{R}}\frac{z+t}{1-tz}\,d\sigma(t)$). Moreover, in this case, such a pair (γ, σ) is unique, it is called the *Lévy pair* of μ and we denote μ by $\nu_*^{\gamma,\sigma}$ (resp. $\nu_{\boxplus}^{\gamma,\sigma}$). For all $t \geq 0$, μ_t has Lévy pair $(t\gamma, t\sigma)$. In the same way, a symmetric probability measure ν is $\boxplus_\lambda$-infinitely divisible if and only if there exists a positive finite symmetric measure on the real line G such that $C_\nu(z) = z\int_{\mathbb{R}}\frac{1+t^2}{1-zt^2}\,dG(t)$. In this case, the measure G is unique, and ν will be denoted by $\nu_{\boxplus_\lambda}^G$. The correspondences $\nu_*^{\gamma,\sigma} \longleftrightarrow \nu_{\boxplus}^{\gamma,\sigma}$ (for any pair (γ, σ) as above) and $\nu_*^{0,G} \longleftrightarrow \nu_{\boxplus_\lambda}^G$ (for any G as above) are called *Bercovici–Pata bijections*. These bijections have many deep properties [5,8], some of which will be mentioned in the proof of the following lemma.

Lemma 9. *Let γ be a real number and σ be a positive finite measure on the real line. Then we have equivalence between*:

   (i) *For all $t \geq 0$, $\nu_*^{t\gamma, t\sigma}$ is supported on $[0, +\infty)$.*
   (ii) *For all $t \geq 0$, $\nu_{\boxplus}^{t\gamma, t\sigma}$ is supported on $[0, +\infty)$.*
   (iii) *We have $\sigma((-\infty, 0]) = 0$ and the integral $\int \frac{1}{x}\,d\sigma(x)$ is finite and $\leq \gamma$.*

Proof. The equivalence between (i) and (iii) follows from Theorem 24.7 and Corollary 24.8 of [18]. Let us prove the equivalence between (i) and (ii). In order to do that, let us recall a fact proved in [8]: for any Lévy pair (γ, σ) and any sequence (ν_n) of probability measures, one has

$$\nu_n^{*n} \text{ converges weakly to } \nu_*^{\gamma,\sigma} \quad \Longleftrightarrow \quad \nu_n^{\boxplus n} \text{ converges weakly to } \nu_{\boxplus}^{\gamma,\sigma}. \tag{15}$$

Let us suppose (i) (resp. (ii)) to be true. Let us fix $t \geq 0$. For all n, we have

$$\left(\nu_*^{t\gamma/n, t\sigma/n}\right)^{*n} = \nu_*^{t\gamma, t\sigma} \quad \left(\text{resp. } \left(\nu_{\boxplus}^{t\gamma/n, t\sigma/n}\right)^{\boxplus n} = \nu_{\boxplus}^{t\gamma, t\sigma}\right).$$

Thus by (15),

$$\left(\nu_*^{t\gamma/n, t\sigma/n}\right)^{\boxplus n} \text{ converges weakly to } \nu_{\boxplus}^{t\gamma, t\sigma} \quad \left(\text{resp. } \left(\nu_{\boxplus}^{t\gamma/n, t\sigma/n}\right)^{*n} \text{ converges weakly to } \nu_*^{t\gamma, t\sigma}\right).$$

Thus since any free (resp. classical) additive convolution and any weak limit of measures with supports on $[0, +\infty)$ has support on $[0, +\infty)$, (ii) (resp. (i)) holds. $\square$

Remark 10. *Note that* (i) *is equivalent to the fact that there exists $t > 0$ such that $\nu_*^{t\gamma, t\sigma}$ is supported on $[0, +\infty)$* [18], *Corollary* 24.8. *However, the same is not true for the free infinitely divisible laws. Indeed, let, for each $t \geq 0$, MP_t denote the Marchenko–Pastur law with mean t* [16], *Example* 3.3.5, *and let us define, for each t, $\mu_t = \mathrm{MP}_t * \delta_{-t/4}$. Then since free and classical convolutions with Dirac masses are the same, $(\mu_t)_{t\geq 0}$ is a convolution semi-group with respect to $\boxplus$. But μ_4 is supported on $[0, +\infty)$, whereas for each $t \in (0, 1]$, the support of μ_t contains a negative number* (*namely $-t/4$*).

2.2. *Main result of the section*

The following theorem allows us to claim that even though not every symmetric law can be expressed under the form $\sqrt{\mu \boxtimes \mu_\lambda}$ for μ law on $[0, +\infty)$ (see Proposition 6), many of them have this form. $\lambda \in [0, 1]$ is still fixed.

For G measure on the real line, we let G^2 denote the push-forward of G by the function $t \mapsto t^2$.

Theorem 11. (i) *Let μ be a $\boxplus$-infinitely divisible law such that for all $t \geq 0$, $\mu^{\boxplus t}$ is supported on $[0, +\infty)$. Then the law $\sqrt{\mu \boxtimes \mu_\lambda}$ is $\boxplus_\lambda$-infinitely divisible, with Lévy measure the only symmetric measure G such that*

$$G^2 = \left(\gamma - \int \frac{1}{x}\,d\sigma(x)\right)\delta_0 + \frac{1+x^2}{x(1+x)}\,d\sigma(x), \tag{16}$$

where (γ, σ) denotes the Lévy pair of μ.

(ii) *Reciprocally, any $\boxplus_\lambda$-infinitely divisible law ν has the form $\sqrt{\mu \boxtimes \mu_\lambda}$ for some $\boxplus$-infinitely divisible law μ such that for all $t \geq 0$, $\mu^{\boxplus t}$ is supported on $[0, +\infty)$. Moreover, the Lévy pair (γ, σ) of μ is defined by*

$$\gamma = \int_{[0,+\infty)} \frac{1+x}{1+x^2} \, dG^2(x) \quad and \quad \sigma = \frac{x(1+x)}{1+x^2} \, dG^2(x), \tag{17}$$

where G denotes the Lévy measure of ν.

Proof. (i) Note that by Theorem 3, the map $\mu \mapsto \sqrt{\mu \boxtimes \mu_\lambda}$ is a morphism from the set of laws on $[0, +\infty)$ to the set on symmetric laws on the real line endowed respectively with the operations $\boxplus$ and $\boxplus_\lambda$. Thus if μ is $\boxplus$-infinitely divisible, then $\sqrt{\mu \boxtimes \mu_\lambda}$ is $\boxplus_\lambda$-infinitely divisible. Moreover, if the Lévy pair of μ is (γ, σ), then its R-transform is $R_\mu(z) = \gamma z + z \int_{t \in \mathbb{R}} \frac{z+t}{1-zt} \, d\sigma(t)$. By Theorem 1, it implies that $C_{\sqrt{\mu \boxtimes \mu_\lambda}}(z) = \gamma z + z \int_{t \in \mathbb{R}} \frac{z+t}{1-zt} \, d\sigma(t)$. But by uniqueness, the Lévy measure G of $\sqrt{\mu \boxtimes \mu_\lambda}$ is characterized by the fact that $C_{\sqrt{\mu \boxtimes \mu_\lambda}}(z) = z \int_{\mathbb{R}} \frac{1+t^2}{1-zt^2} \, dG(t)$. Thus to prove (16), it suffices to prove that for G given by (16), for all z, one has

$$\gamma z + z \int_{t \in \mathbb{R}} \frac{z+t}{1-zt} \, d\sigma(t) = z \int_{\mathbb{R}} \frac{1+t^2}{1-zt^2} \, dG(t),$$

which can easily be verified.

(ii) Let ν be a $\boxplus_\lambda$-infinitely divisible law with Lévy measure denoted by G. Let (γ, σ) be the Lévy pair defined by (17). Note that (γ, σ) satisfies (iii) of Lemma 9, thus, for $\mu := \nu_{\boxplus}^{\gamma, \sigma}$, for all $t \geq 0$, the law $\mu^{\boxplus t}$ is actually supported by $[0, +\infty)$. Thus by (i), $\sqrt{\mu \boxtimes \mu_\lambda}$ is $\boxplus_\lambda$-infinitely divisible with Lévy measure the only symmetric measure H satisfying

$$H^2 = \left(\gamma - \int \frac{1}{x} \, d\sigma(x) \right) \delta_0 + \frac{1+x^2}{x(1+x)} \, d\sigma(x).$$

To prove that $\sqrt{\mu \boxtimes \mu_\lambda} = \nu$, it suffices to prove that $H = G$, which can easily be verified. $\square$

One of the consequences of this theorem is that it gives us a description of the free multiplicative convolution of two Marchenko–Pastur laws (i.e. free Poisson laws), one of them having a mean ≥ 1. For all $t > 0$, the Marchenko–Pastur law MP_t with mean t has been introduced at Remark 10.

Corollary 12. *Consider $a, c > 0$ such that $a > 1$. Then $MP_c \boxtimes MP_a$ is the push forward, by the map $x \mapsto ax^2$, of the $\boxplus_\lambda$-infinitely divisible law with Lévy measure $\frac{c}{4}(\delta_1 + \delta_{-1})$ for $\lambda = 1/a$.*

Proof. It suffices to notice that for $\lambda = 1/a$, MP_a is the push-forward, by the map $x \mapsto ax$, of the law μ_λ, that MP_c is the $\boxplus$-infinitely divisible law with Lévy pair $(c/2, c/2\delta_1)$, and then to apply (i) of Theorem 11. $\square$

This corollary can be interpreted as the coincidence of the limit laws of two different matrix models. Indeed, the $\boxplus_\lambda$-infinitely divisible law with Lévy measure $\frac{c}{4}(\delta_1 + \delta_{-1})$ was already known [5], Proposition 6.1, to be the limit symmetrized singular law of the random matrix $M := \sum_{k=1}^{p} u_k v_k^*$, for n, p, q tending to infinity in such a way that $p/n \to c$ and $n/q \to \lambda$ and $(u_k)_{k \geq 1}, (v_k)_{k \geq 1}$ two independent families of independent random vectors such that for all k, u_k, v_k are uniformly distributed on the unit spheres of respectively $\mathbb{C}^n, \mathbb{C}^q$. Thus, if, for large n, p, q's such that $p/n \simeq c$ and $n/q \simeq \lambda$, one considers such a random matrix M and also two independent random matrices T, Q with respective dimensions $n \times p, n \times q$, the entries of which are independent real standard Gaussian random variables, then the empirical spectral measures of the random matrices MM^* and $\frac{1}{nq} TT^* QQ^*$ are close to each other, as illustrated by Fig. 1.
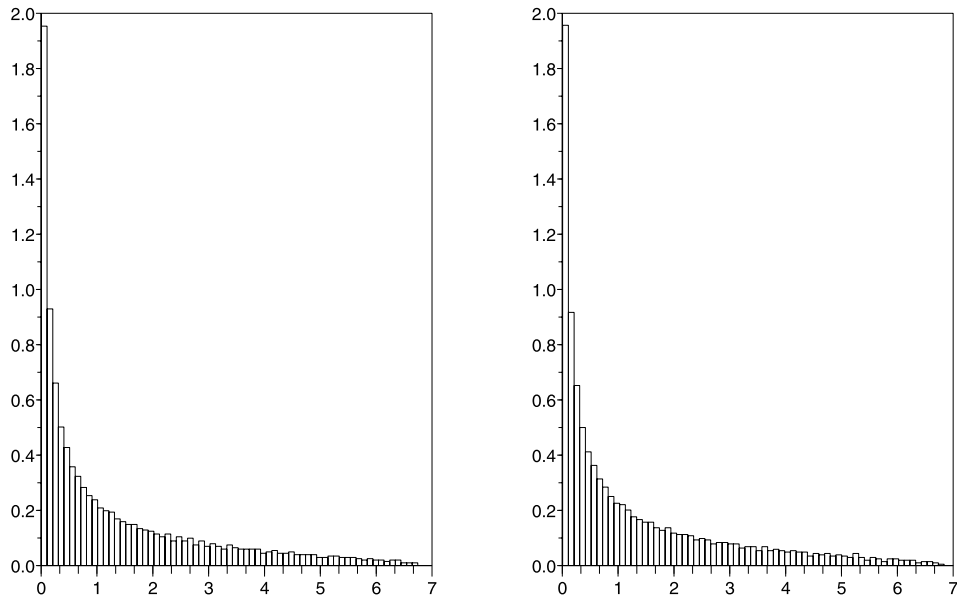
Acknowledgment

Fig. 1. Histograms of the spectrums of MM^* (left) and $\frac{1}{nq}TT^*QQ^*$ (right) for $n = 2000$, $\lambda = 0.6$, $c = 1.3$.

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