DECOMPOSITIONS OF THE FREE ADDITIVE CONVOLUTION

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Abstract

We introduce and study a new type of convolution of probability measures called the orthogonal convolution, which is related to the monotone convolution. Using this convolution, we derive alternating decompositions of the free additive convolution $\mu \boxplus \nu$ of compactly supported probability measures in free probability. These decompositions are directly related to alternating decompositions of the associated subordination functions. In particular, they allow us to compute free additive convolutions of compactly supported measures without using free cumulants or R-transforms. In simple cases, representations of Cauchy transforms $G_{\mu \boxplus \nu}(z)$ as continued fractions are obtained in a natural way. Moreover, this approach establishes a clear connection between convolutions and products associated with the main notions of independence (free, monotone and boolean) in noncommutative probability. Finally, our result leads to natural decompositions of the free product of rooted graphs.

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1. Introduction

The addition problem for classically independent random variables leads to the classical convolution of measures. Namely, if $X_1$ and $X_2$ are independent random variables with distributions $\mu$ and $\nu$, respectively, then the classical convolution $\mu \ast \nu$ of measures $\mu$ and $\nu$ gives the distribution of $X_1 + X_2$.

In free probability there is an analogue of the addition problem which leads to the free additive convolution of probability measures. Namely, let $X_1$ and $X_2$ be random variables, i.e. elements of a noncommutative probability space $(\mathcal{A}, \varphi)$, where $\mathcal{A}$ is a unital algebra and $\varphi$ is a linear functional on $\mathcal{A}$ with $\varphi(1) = 1$, and suppose that $X_1$

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and \(X_2\) are free with respect to \(\varphi\). If \(\mu\) and \(\nu\) denote the \(\varphi\)-distributions of \(X_1\) and \(X_2\), respectively, then the \(\varphi\)-distribution of \(X_1 + X_2\), denoted \(\mu \boxplus \nu\), is called the free additive convolution of \(\mu\) and \(\nu\). This convolution was introduced by Voiculescu [24] for compactly supported probability measures on the real line. In the procedure of computing \(\mu \boxplus \nu\), a central role is played by the Cauchy transforms of probability measures. In particular, if \(X\) is a bounded self-adjoint operator on some Hilbert space \(\mathcal{H}\), then

\[
G_\mu(z) = \int_{-\infty}^{\infty} \frac{\mu(dx)}{z - x} = \sum_{n=0}^{\infty} \mu(X^n)z^{-n-1},
\]

where \(z\) lies in the open upper half plane \(\mathbb{C}^+\), is the Cauchy transform of \(\mu\) with moments \(\mu(X^n) = \varphi(X^n)\).

Voiculescu introduced the R-transform of \(\mu\) defined by \(R_\mu(z) = G_\mu^{-1}(z) - 1/z\), where \(G_\mu^{-1}(z)\) is the right inverse of \(G_\mu(z)\) with respect to the composition of formal power series (with coefficients \(r_\mu(n)\) called the free cumulants of \(\mu\)). On a suitable domain, \(R_\mu(z)\) becomes a holomorphic function. If R-transforms are used to compute \(\mu \boxplus \nu\), one has to invert the Cauchy transforms \(G_\mu(z)\) and \(G_\nu(z)\), which gives \(R_\mu(z)\) and \(R_\nu(z)\), add these up to get \(R_{\mu \boxplus \nu}(z) = R_\mu(z) + R_\nu(z)\) and then invert \(G_{\mu \boxplus \nu}^{-1}(z)\) back to obtain \(G_{\mu \boxplus \nu}(z)\). Finally, using the Stieltjes inversion formula, one can compute \(\mu \boxplus \nu\) (for details, see [24] and [25]).

The additivity of the R-transform is analogous to the additivity of the logarithm of the Fourier transform \(\mathcal{F}_\mu(it)\), or of the associated exponential moment generating function

\[
\mathcal{F}_\mu(z) = \sum_{n=0}^{\infty} \frac{\mu(X^n)}{n!}z^n
\]

for the measure \(\mu\), and the free cumulants are the analogues of the classical cumulants which appear in the power series representing \(\log \mathcal{F}_\mu(it)\). Nevertheless, in classical probability, one can express the moments of \(\mu \ast \nu\) directly in terms of the moments of \(\mu\) and \(\nu\) without using the classical cumulants since there is a ‘complete decomposition’

\[
\mathcal{F}_{\mu \ast \nu}(z) = \mathcal{F}_\mu(z) \cdot \mathcal{F}_\nu(z).
\]

In this paper we find some analogues of the above formula for the convolution \(\mu \boxplus \nu\), which allow us to compute it without using the free cumulants (or the R-transform).

For that purpose we shall use the reciprocal Cauchy transforms of probability measures. By the reciprocal Cauchy transform of \(\mu\) we understand

\[
F_\mu(z) = \frac{1}{G_\mu(z)}
\]

and the class of reciprocal Cauchy transforms of Borel probability measures on the real line \(\mathcal{M}\) we denote by \(\mathcal{RC}\). In fact, they played a central role in the approach of Maassen [18] who extended the definition of the additive free convolution to measures with finite variance. To all measures from class \(\mathcal{M}\), the definition was later extended by Bercovici and Voiculescu [5]. Another important result in the context of reciprocal Cauchy transforms is the subordination property, namely that there exist unique
functions $F_1, F_2 \in \mathcal{RC}$, called *subordination functions*, such that

$$F_{\mu \square \nu}(z) = F_{\mu}(F_1(z)) = F_{\nu}(F_2(z))$$

for $z \in \mathbb{C}^+$ (proved by Voiculescu [25] for compactly supported measures and by Biane [6] in the general case). These functions play a key role in the recent work of Belinschi [4], where complex analytic methods are used to give a detailed study of free convolutions.

The above relation was also used by Chistyakov and Goetze [11] who proved that there exist unique functions $F_1, F_2 \in \mathcal{RC}$ such that

$$F_{\mu}(F_1(z)) = F_{\nu}(F_2(z)) \quad \text{and} \quad F_{\mu}(F_1(z)) = F_1(z) + F_2(z) - z, \quad (1.6)$$

for any $\mu, \nu \in \mathcal{M}$, and therefore one can define $\mu \square \nu$ by any of the equations (1.5). It is worth mentioning here that formulas (1.6) were also used by Quenell [21] and Gutkin [12] in their study of free products of graphs and their spectra. In their approach, the functions $F_1(z)$ and $F_2(z)$ correspond to the (root) spectral distributions of the ‘branches’ of the free product of rooted graphs.

The above formulas are related to the *boolean convolution* of measures which appeared in the addition problem for boolean independent random variables [22] and the *monotone convolution* corresponding to the notion of monotone independence [19,20]. These convolutions can be defined using the reciprocal Cauchy transforms by the equations

$$F_{\mu \star \nu}(z) = F_{\mu}(F_\nu(z)) \quad \text{and} \quad F_{\mu \ast \nu}(z) = F_{\mu}(z) + F_\nu(z) - z. \quad (1.7)$$

where $\mu \ast \nu$ and $\mu \geq \nu$ denote the boolean and monotone convolutions of $\mu, \nu \in \mathcal{M}$. Using these convolutions and (1.5), equations (1.6) correspond to what might be viewed as ‘monotone’ and ‘boolean’ decompositions of the free additive convolution.

Nevertheless, functions $F_1(z)$ and $F_2(z)$ still depend on both $\mu$ and $\nu$. In this paper we find and study ‘complete’ decompositions of the corresponding measures, which express them in terms of a new ‘basic’ convolution which resembles the monotone convolution. This new convolution is denoted $\mu \rightrightarrows \nu$ and called the *orthogonal convolution* of $\mu$ and $\nu$. If we continue to use the convenient language of transforms, we can define it by its reciprocal Cauchy transform

$$F_{\mu \rightrightarrows \nu}(z) = F_{\mu}(F_\nu(z)) - F_\nu(z) + z. \quad (1.8)$$

which shows how much it differs from the monotone convolution. Equivalently, the orthogonal convolution of $\mu$ and $\nu$ can be defined as the unique measure $\mu \rightrightarrows \nu$ determined by the equation

$$K_{\mu \rightrightarrows \nu}(z) = K_{\mu}(F_\nu(z)) = K_{\mu}(z - K_\nu(z)), \quad (1.9)$$

where the $K$-transform of measure $\mu$ is given by $K_{\mu}(z) = z - F_\mu(z)$ (introduced by Speicher and Woroudi [22]). The Hilbert-space realization of the orthogonal convolution involves projections $P_1$ and $P_2$ onto $\mathcal{H} \ominus \mathbb{C}\xi_1$ and $\mathbb{C}\xi_2$, respectively, where $\xi_1$ and $\xi_2$ are cyclic vectors, which motivates our terminology. The orthogonal convolution is the main building block of the decompositions of $\mu \square \nu$ studied in this paper. In particular,
we study its combinatorics and decompose the sum $X_1 + X_2$ of free random variables on the free product of Hilbert spaces as infinite sums of replicas of $X_1$ and $X_2$ which correspond to the ‘factors’ of the ‘complete’ decompositions of $\mu \boxplus \nu$.

In these decompositions, proven here for compactly supported measures, the boolean or monotone convolutions deduced from (1.8) are followed by infinite sequences of orthogonal convolutions of alternating $\mu$ and $\nu$. Using reciprocal Cauchy transforms and $K$-transforms, the first decomposition corresponds to the ‘continued composition’ form

$$F_{\mu \boxplus \nu}(z) = F_{\mu}(z - K_{\nu}(z - K_{\mu}(z - K_{\nu}(z - \ldots)))) \quad (1.10)$$

where the right-hand side is understood as the uniform limit on compact subsets of $\mathbb{C}^+$ (a ‘twin-like’ formula is obtained by interchanging $\mu$ and $\nu$). Viewing the $K$-transform as a slight modification of the reciprocal Cauchy transform, we can treat this expression as the ‘monotone-orthogonal decomposition’ of $F_{\mu \boxplus \nu}(z)$ since it begins with the composition of $F$-transforms. It is not hard to see that this decomposition is closely related to continued fractions. Another decomposition of $F_{\mu \boxplus \nu}(z)$ is called the ‘boolean-orthogonal decomposition’ since it corresponds to the second equation of (1.6).

Let us remark that the approximants of these decompositions correspond to approximations of freeness which were studied from the point of view of product states, limit theorems and Gaussian operators. Thus, the approximants of the boolean-orthogonal decomposition correspond to the hierarchy of freeness [14,15], whereas the approximants of the monotone-orthogonal decomposition correspond to the monotone hierarchy of freeness [17]. In a different direction goes [16], where a noncommutative extension of the Fourier transform was constructed which extends both the Fourier transform and the $K$-transform. For other interpolations involving the free additive convolution, see [7-9]. In turn, the free multiplicative convolution and its decompositions will be treated in a separate paper.

The paper is organized as follows. In Section 2 we introduce basic notions. Some useful combinatorics is developed in Section 3. In Section 4 we introduce and study the concept of ‘orthogonal subalgebras’ of a noncommutative probability space and related ‘orthogonal structures’ (product of Hilbert spaces, product of $C^*$-algebras and convolution) The notion of ‘orthogonal convolution’ is studied in more detail in Section 5 (algebraic properties) and Section 6 (transforms and analytic properties). Then, in Section 7, we introduce and study ‘subordinate structures’ related to subordination functions and based on the concept of ‘$s$-free subalgebras’ of a noncommutative probability space. We show in Section 8 that they generalize the ‘branches’ of the free product of graphs. We also derive an alternating orthogonal decomposition of the $s$-free convolution. This leads to monotone-orthogonal and boolean-orthogonal decompositions of the free additive convolution given in Section 9.

2. Preliminaries

This section contains preliminaries concerning transforms of probability measures (more generally, distributions of random variables) and their convolutions associated with
notions of noncommutative independence (free, monotone and boolean).

By a non-commutative probability space we understand a pair \((A, \varphi)\), where \(A\) is a unital algebra over \(\mathbb{C}\) and \(\varphi\) is a linear functional \(\varphi : A \to \mathbb{C}\) such that \(\varphi(1) = 1\). If \(A\) is a unital \(*\)-algebra and \(\varphi\) is positive (called a state), then \((A, \varphi)\) is called a \(*\)-probability space. If, in addition, \(A\) is a \(C^*\)-algebra, then \((A, \varphi)\) is called a \(C^*\)-probability space. By the Gelfand-Naimark-Segal theorem, a \(C^*\)-probability space can always be realized as a subalgebra of bounded operators on a Hilbert space \(H\) with a distinguished unit vector \(\xi\), for which \(\varphi(a) = \langle a\xi, \xi \rangle\) for \(a \in A\).

By a random variable we will understand any element \(a\) of the considered algebra \(A\). If \(A\) is equipped with an involution, then a random variable \(a\) will be called self-adjoint if \(a = a^*\). The \(\varphi\)-distribution of a random variable \(a\) is the functional \(\mu_a : \mathbb{C}[X] \to \mathbb{C}\) given by \(\mu_a(1) = 1, \mu_a(X^n) = \varphi(a^n)\). In particular, if \((A, \varphi)\) is a \(C^*\)-probability space, then the distribution \(\mu_a\) of a self-adjoint random variable \(a \in A\) extends to a compactly supported probability measure \(\mu\) on the real line. In that case we will often use the same notation \(\mu\) for both the distribution of \(a\) and the associated compactly supported probability measure.

The additive free convolution of distributions (measures) \(\mu \boxplus \nu\) is related to the notions of freeness and free product of \(C^*\)-algebras [3,23]. Let \((A, \varphi)\) be a non-commutative probability space and let \(A_i, i \in I\), be unital subalgebras of \(A\). The family \((A_i)_{i \in I}\) is called free with respect to \(\varphi\) if

\[
\varphi(a_1 a_2 \ldots a_n) = 0 \tag{2.1}
\]

whenever \(a_j \in A_{i_j} \cap \text{Ker} \varphi\) with \(i_1 \neq i_2 \neq \ldots \neq i_n\). A family of elements \((a_i)_{i \in I}\) of \(A\) is called free if the family of unital subalgebras \((A_i)_{i \in I}\) of \(A\), each generated by \(a_i\), is free.

In turn, the monotone convolution is related to monotone independence [19] which can be defined if the set \(I\) is totally ordered. Thus, random variables \((a_i)_{i \in I}\) are monotone independent w.r.t. \(\varphi\) if

\[
\varphi(a_{i_1} \ldots a_{i_k} \ldots a_{i_n}) = \varphi(a_{i_k})\varphi(a_{i_1} \ldots a_{i_{k-1}} a_{i_{k+1}} \ldots a_{i_n}) \tag{2.2}
\]

whenever \(i_{k-1} < i_k < i_{k+1}\), with the understanding that only one of these inequalities holds if \(k \in \{1, n\}\).

In particular, we will say that the pair \((a, b)\) of elements of \(A\) is monotone independent w.r.t. \(\varphi\) if \(a = a_i\) and \(b = a_j\) with \(i < j\) and \(a_i, a_j\) are monotone independent w.r.t. \(\varphi\). In that case, if the \(\varphi\)-distributions of \(a\) and \(b\) are \(\mu\) and \(\nu\), respectively, then the \(\varphi\)-distribution of \(a + b\), denoted \(\mu \bowtie \nu\), is called the monotone convolution of \(\mu\) and \(\nu\). If \(\mu\) and \(\nu\) are probability measures on \(\mathbb{R}\), then \(\mu \bowtie \nu\) is the unique probability measure on \(\mathbb{R}\) which satisfies the equation

\[
F_{\mu \bowtie \nu}(z) = F_\mu(F_\nu(z)) \tag{2.3}
\]

where \(F_\mu(z)\) is the reciprocal Cauchy transform of \(\mu\). For details, see [20].

The third convolution, which plays an important role in our approach, is associated with the so-called boolean independence [22]. Namely, random variables \((a_i)_{i \in I}\) are called boolean independent w.r.t. \(\varphi\) if

\[
\varphi(a_{i_1} a_{i_2} \ldots a_{i_n}) = \varphi(a_{i_1}) \varphi(a_{i_2}) \varphi(a_{i_3}) \ldots \varphi(a_{i_n}) \tag{2.4}
\]
wherever \( i_1 \neq i_2 \neq \ldots \neq i_n \).

In particular, if two random variables, \( a_1 = a \) and \( a_2 = b \), have \( \varphi \)-distributions \( \mu \) and \( \nu \), respectively, and are boolean independent w.r.t. \( \varphi \), then the \( \varphi \)-distribution of the sum \( a + b \) is denoted \( \mu \vartriangle \nu \) and is called the boolean convolution of \( \mu \) and \( \nu \). If \( \mu \) and \( \nu \) are probability measures on \( \mathbb{R} \), then \( \mu \vartriangle \nu \) is the unique probability measure on \( \mathbb{R} \) which satisfies the equation

\[
K_{\mu \vartriangle \nu}(z) = K_{\mu}(z) + K_{\nu}(z)
\]

(2.5)

where \( K_{\mu}(z) = z - F_{\mu}(z) \) is the so-called \( K \)-transform of \( \mu \). If \( \mu \) is a probability measure, then \( K_{\mu} : \mathbb{C}^+ \to \mathbb{C}^- \cup \mathbb{R} \) is a holomorphic function, where \( \mathbb{C}^+ \) and \( \mathbb{C}^- \) denote the open upper and lower complex half-planes, respectively. For details, see [22].

3. Combinatorics

In this section we describe the combinatorics which appears in a natural way in the context of the orthogonal convolution.

We adopt the following notations. By \( \mathcal{I}(n) \) we denote the lattice of interval partitions of the set \( \{1, 2, \ldots, n\} \). Thus, any \( \pi \in \mathcal{I}(n) \) is of the form \( \pi = \{\pi_1, \pi_2, \ldots, \pi_r\} \), where \( \pi_1 \cup \pi_2 \cup \ldots \cup \pi_r = \{1, 2, \ldots, n\} \) and \( \pi_1 < \pi_2 < \ldots < \pi_r \), where blocks are ordered in the natural way, i.e. \( \pi_k < \pi_j \) means that \( k < l \) for all \( k \in \pi_i \) and \( l \in \pi_j \). Note that there is a bijection between \( \mathcal{I}(n) \) and the set of ordered partitions of the number \( n \in \mathbb{N} \). Namely, the bijection is given by \( \pi \to (j_1, j_2, \ldots, j_r) \), where \( j_i = |\pi_i| \) for \( i = 1, \ldots, n \). Clearly, \( j_1 + j_2 + \ldots + j_r = n \). This bijection will be used in the sequel and both objects, the interval partition \( \pi \) and the corresponding tuple \((j_1, j_2, \ldots, j_r)\) will be denoted by \( \pi \).

For \( r \) odd and \( \pi \in \mathcal{I}(n) \) with blocks ordered in the natural way, we shall use the alternating decomposition of \( \pi \in \mathcal{I}(n) \) of the form \( \pi = \pi' \cup \pi'' \), where

\[
\pi' = \{\pi_1, \pi_3, \ldots, \pi_r\}, \quad \text{and} \quad \pi'' = \{\pi_2, \pi_4, \ldots, \pi_{r-1}\},
\]

(3.1)

with the associated tuples \((j_1, j_3, \ldots, j_r)\) and \((j_2, j_4, \ldots, j_{r-1})\), respectively (a similar definition can be given for \( r \) even, but we will not need it). Further, we write \( \pi \preceq \sigma \) for \( \pi, \sigma \in \mathcal{I}(n) \) if \( \pi \) is a (not necessarily proper) refinement of \( \sigma \). Finally, if \( \pi \in \mathcal{I}(n) \), then by \( \mathcal{I}(\pi) \) we denote the family of all (interval) subpartitions \( \sigma \in \mathcal{I}(n) \) of the partition \( \pi \) and by \( \mathcal{I}_{\text{odd}}(\pi) \) - its subset consisting of such (interval) subpartitions of \( \pi \) which are obtained from \( \pi \) by decomposing its every block \( \pi_k \) into an odd number of subblocks - these subpartitions will be called odd.

We will use multiplicative functions on partially ordered sets. The partially ordered set used here will be the union of lattices \( \bigcup_{n \geq 1} \mathcal{I}(n) \) with the natural partial order, again denoted \( \pi \preceq \sigma \), iff there exists \( n \in \mathbb{N} \) such that \( \pi, \sigma \in \mathcal{I}(n) \) and it holds that \( \pi \preceq \sigma \) for \( \pi, \sigma \) treated as elements of \( \mathcal{I}(n) \). In particular, for any distribution \( \mu \) we define the moment function

\[
m_\mu : \bigcup_{n \geq 1} \mathcal{I}(n) \to \mathbb{R}
\]

(3.2)

\[
m_\mu(\pi) = \mu(j_1)\mu(j_2)\ldots\mu(j_r)
\]

(3.3)
where \( j_i = |\pi_i| \), \( 1 \leq i \leq r \) and \( (\mu(n))_{n \in \mathbb{N}} \) is the collection of moments of \( \mu \). Related to the moment function is the inverse boolean cumulant function defined below (it differs from the usual boolean cumulant function \( k_\mu \) [22] with summation extending over \( \pi \leq \sigma \).

**Definition 3.1.** Let the moment function \( m_\mu \) be related to the multiplicative functions \( k^*_\mu \) on the lattice \( \bigcup_{n \geq 1} \mathcal{I}(n) \) by the formula

\[
m_\mu(\sigma) = \sum_{\pi \geq \sigma} k^*_\mu(\pi).
\]  

(3.4)

Then, functions \( k^*_\mu \), given by the Möbius inversion formula

\[
k^*_\mu(\pi) = \sum_{\sigma \geq \pi} (-1)^{|\pi| - |\sigma|} m_\mu(\sigma)
\]

(3.5)

will be called the inverse boolean cumulant function. Using the representation of \( \pi \) as \((j_1, j_2, \ldots, j_n)\), we can write \( k^*_\mu(\pi) = k^*_\mu(j_1, j_2, \ldots, j_n) \).

**Example 3.1.** Let us evaluate \( k^*_\mu(\pi) \) for the simplest partitions \( \pi \). We get

\[
\begin{align*}
k^*_\mu(n) &= \mu(n) \\
k^*_\mu(n, m) &= \mu(n)\mu(m) - \mu(n + m) \\
k^*_\mu(n, m, k) &= \mu(n)\mu(m)\mu(k) - \mu(n + m)\mu(k) - \mu(n)\mu(m + k) + \mu(n + m + k) \\
k^*_\mu(n, m, k, l) &= \mu(n)\mu(m)\mu(k)\mu(l) - \mu(n + m)\mu(k)\mu(l) - \mu(n)\mu(m + k)\mu(l) \\
&\quad - \mu(n)\mu(m)\mu(k + l) + \mu(n + m)\mu(k + l) + \mu(n + m + k)\mu(l) \\
&\quad + \mu(n)\mu(m + k + l) - \mu(n + m + k + l).
\end{align*}
\]

where \( n, m, k, l \in \mathbb{N} \).

A partition \( \pi = \{\pi_1, \pi_2, \ldots, \pi_r\} \) of the set \( \{1, 2, \ldots, n\} \) is called non-crossing if there do not exist numbers \( i < k < j < l \) such that \( i, j \in \pi_p \), \( k, l \in \pi_q \) and \( p \neq q \). By \( \mathcal{NC}(n) \) we denote the family of non-crossing partitions of the set \( \{1, 2, \ldots, n\} \). If \( \pi \in \mathcal{NC}(n) \), then its block \( \pi_p \) is inner with respect to block \( \pi_q \) if \( i < k < j \) for every \( k \in \pi_p \) and \( i, j \in \pi_q \) (then \( \pi_q \) is called outer w.r.t. \( \pi_p \)). Let \( o(\pi_p) \) be the number of blocks of \( \pi \) which are outer w.r.t. \( \pi_p \). Then the depth of \( \pi_p \) is defined as \( d(\pi_p) = o(\pi_p) + 1 \) and \( d(\pi) = \max_{1 \leq j \leq r} d(\pi_j) \) is called the depth of \( \pi \). By \( \mathcal{NC}_d(n) \) we shall denote the family of non-crossing partitions of depth smaller or equal to \( d \). In particular, \( \mathcal{NC}_1(n) = \mathcal{I}(n) \) for every \( n \in \mathbb{N} \).

Let us introduce a suitable subfamily of \( \mathcal{NC}_2(n) \).

**Definition 3.2.** A partition \( \pi \in \mathcal{NC}_2(n) \) is called decomposable if it can be decomposed as \( \pi = \pi' \cup \pi'' \), where

\[
\pi' = \{\pi'_1, \pi'_2, \ldots, \pi'_p\}, \quad \pi'' = \{\pi''_1, \pi''_2, \ldots, \pi''_q\}
\]

(3.6)

with \( p \geq 1 \) and \( q \geq 0 \), consisting of \( p \) blocks of depth 1, \( \pi'_1 \prec \pi'_2 \prec \ldots \prec \pi'_p \), and \( q \) blocks of depth 2, \( \pi''_1 \prec \pi''_2 \prec \ldots \prec \pi''_q \), such that blocks \( \pi''_i \) and \( \pi''_{i+1} \) are not neighbors.
Denote by $\mathcal{D}_2(n)$ the family of all decomposable partitions of the set $\{1, 2, \ldots, n\}$.

**Definition 3.3.** Let $\pi \in \mathcal{D}_2(n)$ be given with decomposition $\pi = \pi' \cup \pi''$ of the form (3.6). Denote by $\mathcal{P}(\pi)$ the set of refinements $\eta$ of $\pi$ of the form $\eta = \eta' \cup \eta''$, where $\eta'' = \pi''$ and

$$\eta' = \{\eta'_1, \eta'_2, \ldots, \eta'_r\},$$

is a refinement of $\pi'$ which satisfies the conditions: (i) $\eta'_1 < \eta'_2 < \ldots < \eta'_r$, (ii) if two consecutive numbers $i, i + 1$ belong to the same block of $\pi'$, they must belong to the same block of $\eta'$. By a *decomposition pair* we understand any pair $(\pi, \eta)$, where $\pi \in \mathcal{D}_2(n)$ and $\eta \in \mathcal{P}(\pi)$. Denote by $\mathcal{DP}_2(n)$ the family of decomposition pairs $(\pi, \eta)$, where $\pi \in \mathcal{D}_2(n)$ and $\eta \in \mathcal{P}(\pi)$.

![Diagram](image)

**Figure 1.** Example of a decomposition triple $\pi \geq \eta \geq \sigma$.

Finally, to given $\eta \in \mathcal{P}(\pi)$ we associate its coarsest interval subpartition $\sigma \in \mathcal{I}(n)$, i.e. $\sigma = \sigma' \cup \sigma''$, where $\sigma'' = \eta''$ and $\sigma'$ is the coarsest interval subpartition of $\eta'$. Note that $\sigma$ is also the coarsest interval subpartition of $\pi$ and every block $\pi'_j$ gives rise to an odd number of subblocks of $\sigma'$. In such a way we obtain a decomposition triple

$$\sigma \leq \eta \leq \pi$$

which can be nicely illustrated in terms of diagrams. We can think of blocks of $\pi'$ as 'bridges' lying above blocks of $\pi''$ (see partition $\pi$ in Figure 1). Now, $\eta$ is obtained from $\pi$ by erasing certain 'bridge connections' over inner blocks (the latter remain unchanged). Finally, $\sigma$ is obtained from $\eta$ by erasing the remaining 'bridge connections' over inner blocks (the latter remain unchanged).

**Example 3.2.** Consider the partition $\pi \in \mathcal{D}_2(17)$ consisting of 2 outer blocks $\pi'_1 = \{1, 2, 5, 6, 9\}, \pi'_2 = \{10, 13, 17\}$ and 4 inner blocks $\pi''_1 = \{3, 4\}, \pi''_2 = \{7, 8\}, \pi''_3 = \{11, 12\}, \pi''_4 = \{\ldots\}$. 


\[ \pi'' \equiv \{14,15,16\} \]. Let \( \eta \) be its refinement obtained by splitting the block \( \pi_1' \) into two subblocks: \( \eta_1' = \{1,2,5,6\} \) and \( \eta_2' = \{9\} \), and block \( \pi_2' \) into two subblocks: \( \eta_3' = \{10,13\} \) and \( \eta_4' = \{17\} \). Here, \( \eta' = \{\eta_1', \eta_2', \eta_3', \eta_4'\} \). Clearly, \((\pi, \eta) \in \mathcal{DP}_2(17)\) (the pair is shown in Figure 1). Finally, the coarsest interval subpartition \( \sigma \) of \( \pi \) is given by \( \sigma = \sigma' \cup \pi'' \), where \( \sigma' = \{\{1,2\}, \{5,6\}, \{9\}, \{10\}, \{13\}, \{17\}\} \).

We complete this section with two technical propositions.

**Proposition 3.1.** For every \( n \in \mathbb{N} \) there is a bijection between \( \mathcal{D}_2(n) \) and the set \( \mathcal{C}(n) \) of pairs \((\tau, \sigma)\), where \( \tau \in \mathcal{I}(n) \) and \( \sigma \in \mathcal{I}_{\text{odd}}(\tau) \).

**Proof.** Let \( \pi \in \mathcal{D}_2(n) \) be given and let \( \pi = \pi' \cup \pi'' \) be its decomposition (3.6). Let \( f : \mathcal{D}_2(n) \rightarrow \mathcal{C}(n) \), where \( f(\pi) = (\tau, \sigma) \) is defined as follows. For every \( 1 \leq j \leq r \), define the block \( \tau_j \) to be the union of the block \( \pi_j \) and all blocks of \( \pi \) which are inner w.r.t. \( \pi_j \). Then \( \sigma \) is defined to be the coarsest interval refinement of \( \pi \) (obtained by 'erasing all bridge connections' in \( \pi \)). It can be seen that \( f \) is a bijection (using diagrams, giving the pair \((\tau, \sigma)\) specifies blocks of \( \pi \) in two steps: first we give the intervals which are 'covered' by outer blocks of \( \pi \) and then we split up every such interval into an odd number of subintervals which show the positions of the inner blocks of \( \pi \)).

**Proposition 3.2.** For every \( n \in \mathbb{N} \) there is a bijection between \( \mathcal{DP}_2(n) \) and the set \( \mathcal{F}(n) \) of triples \((m, \sigma, j)\), where \( 1 \leq m \leq n \), \( \sigma \in \mathcal{I}(m) \) and \( j = (j_1, j_2, \ldots, j_{m-1}) \) is a tuple of non-negative integers whose sum is equal to \( n - m \).

**Proof.** The bijection \( g : \mathcal{DP}_2(n) \rightarrow \mathcal{F}(n) \) is given by \( g(\pi, \eta) = (m, \sigma, j) \), where the triple \((m, \sigma, j)\) is defined as follows. First, we set \( m = |\pi_1'| + |\pi_2'| + \ldots + |\pi_r'| \), i.e. \( m \) counts all numbers which belong to the outer blocks of \( \pi \). Then we define \( \sigma \in \mathcal{I}(m) \) as the unique partition of the number \( m \) which corresponds to the partition \( \eta' \) of the \( m \)-element set \( \pi_1' \cup \pi_2' \cup \ldots \cup \pi_r' \) into blocks of \( \eta' \). Finally, we set \( j = (j_1, j_2, \ldots, j_{m-1}) \), where \( j_k \) is the nonnegative integer equal to the size of the inner block of \( \pi \) which immediately follows the \( k \)-th leg of \( \pi' \) (\( \pi' \) has \( m \) legs but the last leg of \( \pi' \) ends the diagram, so it does not count). Of course, if there is no inner block following the \( k \)-th leg of \( \pi' \), then we set \( j_k = 0 \). It can be seen that the mapping \( g \) is a bijection. Using diagrams, one can say that by giving the triple \((m, \sigma, j)\), we simply draw the diagram corresponding to \((\pi, \eta)\) in the following order: first we draw all outer blocks of \( \eta' \) and then every inner block of \( \eta'' = \pi'' \) is drawn on the right side of the suitable leg of \( \eta' \).

4. **Orthogonal structures**

In this Section we introduce the notion of 'orthogonal subalgebras' of a given \((\ast, \mathcal{C}^\ast)\) algebra with respect to a pair of functionals (states) and the corresponding notions of the 'orthogonal product' of two Hilbert spaces and the 'orthogonal product' of two \((\ast, \mathcal{C}^\ast)\) algebras. We then construct 'orthogonal random variables' with prescribed
probability distributions. To some extent, these structures resemble the corresponding monotone structures and for that reason can be viewed as ‘quasi-monotone’.

The orthogonal product is neither commutative nor associative, but it turns out useful in the construction of decompositions of the free additive convolution of measures.

**Definition 4.1.** Let \((\mathcal{A}, \varphi, \psi)\) be a unital algebra with a pair of linear normalized functionals and let \(\mathcal{A}_1\) and \(\mathcal{A}_2\) be non-unital subalgebras of \(\mathcal{A}\). We say that \(\mathcal{A}_2\) is *orthogonal* to \(\mathcal{A}_1\) with respect to \((\varphi, \psi)\) if

\[
\begin{align*}
(i) & \quad \varphi(ba_2) = \varphi(a_1b) = 0 \\
(ii) & \quad \varphi(w_1a_1ba_2w_2) = \psi(b)(\varphi(w_1a_1a_2w_2) - \varphi(w_1a_1)\varphi(a_2w_2))
\end{align*}
\]

for any \(a_1, a_2 \in \mathcal{A}_1, b \in \mathcal{A}_2\) and any elements \(w, v\) of the algebra \(\text{alg}(\mathcal{A}_1, \mathcal{A}_2)\) generated by \(\mathcal{A}_1\) and \(\mathcal{A}_2\). We say that the pair \((a, b)\) of elements of \(\mathcal{A}\) is *orthogonal* with respect to \((\varphi, \psi)\) if the algebra generated by \(a \in \mathcal{A}\) is orthogonal to the algebra generated by \(b \in \mathcal{A}\).

**Remark 4.1.** Note that \(\varphi\) is uniquely determined on the algebra generated by \(\mathcal{A}_1\) and \(\mathcal{A}_2\) by restrictions \(\varphi|_{\mathcal{A}_1}\) and \(\psi|_{\mathcal{A}_2}\). In the case of \(\psi\), the situation is quite different. In fact, this is only for the sake of convenience that we consider two states \(\varphi, \psi\) on all of \(\mathcal{A}\) (it is natural in the Hilbert space setting, where we choose states associated with unit vectors of the ‘large’ Hilbert space). For our purposes, it would be sufficient to assume \(\psi\) to be defined only on the subalgebra \(\mathcal{A}_2\). Another observation is that ‘orthogonality’ w.r.t. \((\varphi, \psi)\), is quite different from ‘conditional freeness’ w.r.t. \((\varphi, \psi)\) studied in [7], although it also involves two states on \(\mathcal{A}\).

Let us begin with the Hilbert space setting and introduce the notion of an orthogonal product of two Hilbert spaces with distinguished unit vectors.

**Definition 4.2.** Let \((\mathcal{H}_1, \xi_1)\) and \((\mathcal{H}_2, \xi_2)\) be Hilbert spaces with distinguished unit vectors \(\xi_1\) and \(\xi_2\), respectively. The *orthogonal product* of \((\mathcal{H}_1, \xi_1)\) and \((\mathcal{H}_2, \xi_2)\) is the pair \((\mathcal{H}, \xi)\), where

\[
\mathcal{H} = \mathbb{C}\xi \oplus \mathcal{H}_1^0 \oplus (\mathcal{H}_2^0 \otimes \mathcal{H}_1^0), \tag{4.1}
\]

with \(\mathcal{H}_i^0 = \mathcal{H}_i \ominus \mathbb{C}\xi_i\) denoting the orthogonal complement of \(\mathbb{C}\xi_i, \, i = 1, 2\) and \(\xi\) being a unit vector. We denote it by \((\mathcal{H}, \xi) = (\mathcal{H}_1, \xi_1) \vdash (\mathcal{H}_2, \xi_2)\) and by \(\varphi\) - the canonical state on \(\mathcal{B}(\mathcal{H})\) associated with the vector \(\xi\).

Note that the orthogonal product of Hilbert spaces is slightly smaller than their monotone product [20]. In fact, the monotone product of \((\mathcal{H}_1, \xi_1)\) and \((\mathcal{H}_2, \xi_2)\) is equal to the direct sum of their orthogonal product and \(\mathcal{H}_2^0\). Clearly, \((\mathcal{H}_1, \xi_1) \vdash (\mathcal{H}_2, \xi_2)\) is also a truncation of the free product of Hilbert spaces \((\mathcal{H}_1, \xi_1) \ast (\mathcal{H}_2, \xi_2)\) [26]. However, in order to study representations, it is more convenient to use the tensor product \(\mathcal{H}_1 \otimes \mathcal{H}_2\) and an isometry \(U : \mathcal{H} \to \mathcal{H}_1 \otimes \mathcal{H}_2\) given by

\[
U(\xi) = \xi_1 \otimes \xi_2, \quad U(h_1) = h_1 \otimes \xi_2, \quad U(h_2 \otimes h_1) = h_1 \otimes h_2 \tag{4.2}
\]
for any $h_1 \in \mathcal{H}_1^0$ and $h_2 \in \mathcal{H}_2^0$. In particular, we have
\[UU^* = 1 - P_{c_{\xi_2} \ominus \mathcal{H}_2^0} \tag{4.3}\]
Using the isometry $U$, we shall define $\ast$-representations $\tau_i : \mathcal{B}(\mathcal{H}) \to \mathcal{B}(\mathcal{H})$ by
\[\tau_1(a) = U^*(a \otimes P_2)U, \quad \tau_2(b) = U^*(P_1^\perp \otimes b)U \tag{4.4}\]
where $P_1$, $P_2$ are the projections onto $\mathbb{C}\xi_1$ and $\mathbb{C}\xi_2$, respectively. Note that $\tau_1$ and $\tau_2$ are faithful non-unital $\ast$-homomorphisms.

In the theorem below we describe the properties of the orthogonal product of Hilbert spaces.

**Theorem 4.1.** Let $(\mathcal{H}, \xi) = (\mathcal{H}_1, \xi_1) \oplus (\mathcal{H}_2, \xi_2)$ be the orthogonal product of Hilbert spaces and let $\varphi$, $\varphi_1$ and $\varphi_2$ be the states associated with $\xi$, $\xi_1$ and $\xi_2$, respectively. Moreover, let $\psi$ be the state on $\mathcal{B}(\mathcal{H})$ associated with any unit vector $\eta \in \mathcal{H}_1^0 \subset \mathcal{H}$. Then

(i) $\mathcal{A}_2 = \tau_2(\mathcal{B}(\mathcal{H}_2))$ is orthogonal to $\mathcal{A}_1 = \tau_1(\mathcal{B}(\mathcal{H}_1))$ with respect to $(\varphi, \psi)$,

(ii) $\varphi \circ \tau_1$ agrees with the expectation $\varphi_1$ on $\mathcal{B}(\mathcal{H}_1)$

(iii) $\psi \circ \tau_2$ agrees with the expectation $\varphi_2$ on $\mathcal{B}(\mathcal{H}_2)$.

**Proof.** Let $a \in \mathcal{A}_1$, $b \in \mathcal{A}_2$ and $w_1 \in \text{alg}(\mathcal{A}_1, \mathcal{A}_2)$. First, observe that (4.3) implies that
\[\varphi(w_1 \tau_2(b)) = \varphi(w_1 U^*(P_1^\perp \otimes b)U) = 0\]
since $(P_1^\perp \otimes b)(\xi_1 \otimes \xi_2) = 0$. Similarly, $\varphi(\tau_2(b)w_1) = 0$ and thus the condition (i) of Definition 4.1 holds. We need to show the condition (ii) of that definition. We have
\[\varphi(w_1 \tau_1(a_1) \tau_2(b) \tau_1(a_2)w_2) = \varphi(w_1 U^*(a_1 \otimes P_2)UU^*(P_1^\perp \otimes b)UU^*(a_2 \otimes P_2)Uw_2)\]
\[= \varphi(w_1 U^*(a_1 \otimes P_2)UU^*(P_1^\perp a_2 \otimes bP_2)Uw_2)\]
for any $w_1, w_2 \in \text{alg}(\mathcal{A}_1, \mathcal{A}_2)$ in view of (4.3). For the same reason, we get
\[(a_1 \otimes P_2)UU^*(P_1^\perp a_2 \otimes bP_2)Uw_2 = \varphi(\tau_2(b)(a_1 \otimes P_2)UU^*(P_1^\perp a_2 \otimes P_2)Uw_2).

Below we shall demonstrate that $\varphi(\tau_2(b)) = \psi(\tau_2(b))$ for the state $\psi$ associated with any unit vector $\eta$ from $\mathcal{H}_1^0$. Therefore, we are left with computing
\[\varphi\left(w_1 U^*(a_1 \otimes P_2)UU^*(P_1^\perp a_2 \otimes P_2)Uw_2\right)\]
\[= \varphi\left(w_1 U^*(a_1 \otimes P_2)UU^*(a_2 \otimes P_2)Uw_2\right) - \varphi\left(w_1 U^*(a_1 \otimes P_2)UU^*(P_1^\perp a_2 \otimes P_2)Uw_2\right)\]
\[= \varphi\left(w_1 U^*(a_1 \otimes P_2)UU^*(a_2 \otimes P_2)Uw_2\right) - \varphi\left(w_1 U^*(a_1 \otimes P_2)UP_\xi U^*(a_2 \otimes P_2)Uw_2\right)\]
\[= \varphi(w_1 a_1 a_2 w_2) - \varphi(w_1 a_1)\varphi(a_2 w_2)\]
where we used $U^*(P_1 \otimes P_2) = P_\xi U^*(1 \otimes P_2)$, with $P_\xi$ denoting the projection onto $\mathbb{C}\xi$. Finally,
\[\varphi \circ \tau_1(a) = \langle (a \otimes P_2)\xi_1 \otimes \xi_2, \xi_1 \otimes \xi_2 \rangle = \varphi_1(a)\]
\[\psi \circ \tau_2(b) = \langle (P_1^\perp \otimes b)\eta \otimes \xi_2, \eta \otimes \xi_2 \rangle = \varphi_2(b)\]
for any $a \in \mathcal{B}(\mathcal{H}_1)$ and $b \in \mathcal{B}(\mathcal{H}_2)$, which completes the proof. ■

**Corollary 4.2.** Let $\mu, \nu$ be compactly supported probability measures on $\mathbb{R}$. Then there exist a Hilbert space $\mathcal{H}$, unit vectors $\xi, \eta \in \mathcal{H}$ and self-adjoint bounded random variables $X_1, X_2 \in \mathcal{B}(\mathcal{H})$ such that the pair $(X_1, X_2)$ is orthogonal w.r.t. $(\varphi, \psi)$, where $\varphi$ and $\psi$ are vector states associated with $\xi, \eta \in \mathcal{H}$. Moreover, the $\varphi$-distribution of $X_1$ and the $\psi$-distribution of $X_2$ coincide with $\mu$ and $\nu$, respectively. Finally, the $\varphi$-distribution of $X_1 + X_2$, denoted $\mu \vdash \nu$, is compactly supported.

**Proof.** Let $\mathcal{H}_1 = L^2(\mathbb{R}, \mu)$ and $\mathcal{H}_2 = L^2(\mathbb{R}, \nu)$ and take $\xi_1 = 1$ and $\xi_2 = 1$. Let $\hat{x}_1$ and $\hat{x}_2$ be the standard multiplication operators on these spaces, namely $\hat{x}_1 f(x) = x_1 f(x_1)$ and $\hat{x}_2 g(x_2) = x_2 g(x_2)$. They are bounded self-adjoint operators with distributions $\mu$ and $\nu$, respectively. By taking the orthogonal product $(\mathcal{H}_1, \xi_1) \perp (\mathcal{H}_2, \xi_2)$ we can construct bounded self-adjoint random variables $X_1 = \tau_1(\hat{x}_1)$ and $X_2 = \tau_2(\hat{x}_2)$ from $\mathcal{B}(\mathcal{H})$ such that the pair $(X_1, X_2)$ is orthogonal w.r.t. $(\varphi, \psi)$, where $\varphi$ is the vector state on $\mathcal{B}(\mathcal{H})$ associated with $\xi_1$, and $\psi$ is the vector state on $\mathcal{B}(\mathcal{H})$ associated with $\xi_2$, and any function $f \in L^2(\mathbb{R}, \mu)$ which satisfies $\int_{\mathbb{R}} f(x_1) \mu(dx_1) = 0$ and $\int_{\mathbb{R}} f^2(x_1) \mu(dx_1) = 1$. Finally, it is clear that the sum $X_1 + X_2$ is a bounded self-adjoint operator on $\mathcal{H}$ and thus its probability distribution extends to a compactly supported measure on the real line. ■

**Example 4.1.** Let $(G_1, e_1)$ and $(G_2, e_2)$ be two uniformly locally finite rooted graphs with adjacency matrices $A_1$ and $A_2$, respectively, which extend to bounded operators on $\mathcal{H}_1 = l_2(V_1)$ and $\mathcal{H}_2 = l_2(V_2)$, where $V_1$ and $V_2$ denote their sets of vertices. Then

$$A^{(1)} = A_1 \otimes P_{e_2} \quad \text{and} \quad A^{(2)} = P_{e_1} \otimes A_2$$

are orthogonal w.r.t. $(\varphi, \psi)$, where $\varphi$ and $\psi$ are states on $\mathcal{B}(\mathcal{H}_1) \otimes \mathcal{B}(\mathcal{H}_2)$ associated with vectors $\delta(e_1) \otimes \delta(e_2)$ and $\delta(v) \otimes \delta(e_2)$, respectively, with $v \in V_1^0 = V_1 \setminus \{e_1\}$, and $P_{e_i}$ is the projection onto $\mathbb{C} \delta(e_i)$. The sum $A = A^{(1)} + A^{(2)}$ is the adjacency matrix of a uniformly locally finite rooted graph $(G_1 \vdash G_2, e)$ obtained by attaching a replica of $G_2$ by its root to every vertex of $V_1^0$ and setting $e = e_1 \times e_2$, called the orthogonal product of $(G_1, e_1)$ and $(G_2, e_2)$. The matrix $A$ extends to a bounded operator on the orthogonal product $(\mathcal{H}_1, \delta(e_1)) \perp (\mathcal{H}_2, \delta(e_2))$. By Corollary 4.2, the spectral distribution of $A$ in the state associated with vector $\delta(e)$ is given by $\mu \vdash \nu$, where $\mu$ and $\nu$ are spectral distributions of $A_1$ and $A_2$ associated with $\delta(e_1)$ and $\delta(e_2)$, respectively. A detailed study of the orthogonal product of rooted graphs will be given in a separate paper [1].

It is now natural to define the orthogonal product in the setting of $C^*$-probability spaces. If $(A_i, \varphi_i), \ i = 1, 2$, are $C^*$-probability spaces and $(\mathcal{H}_i, \pi_i, \xi_i)$ - the corresponding GNS triples, we first construct the orthogonal product of Hilbert spaces $(\mathcal{H}, \xi) = (\mathcal{H}_1, \xi_1) \perp (\mathcal{H}_2, \xi_2)$ and then define representations

$$\iota_i : A_i \rightarrow \mathcal{B}(\mathcal{H}), \quad \iota_i = \tau_i \circ \pi_i$$

where $\tau_i, \ i = 1, 2$, are given by (4.4). Let $\mathcal{A}$ be the $C^*$-algebra generated by subalgebras $\iota_1(A_1)$ and $\iota_2(A_2)$ of $\mathcal{B}(\mathcal{H})$ and the identity $I \in \mathcal{B}(\mathcal{H})$ and let $\varphi$ denote the state on $\mathcal{A}$
associated with the vector $\xi$. Then the pair $(\mathcal{A}, \varphi)$ is called the orthogonal product of $C^*$-probability spaces $(\mathcal{A}_1, \varphi_1)$ and $(\mathcal{A}_2, \varphi_2)$ and is denoted $(\mathcal{A}_1, \varphi_1) \vdash (\mathcal{A}_2, \varphi_2)$.

**Theorem 4.3.** Let $(\mathcal{A}, \varphi) = (\mathcal{A}_1, \varphi_1) \vdash (\mathcal{A}_2, \varphi_2)$ be the orthogonal product of $C^*$-probability spaces equipped with the natural $*$-homomorphisms $\iota_i : \mathcal{A}_i \rightarrow \mathcal{A}$ and let $\psi$ be the state on $\mathcal{A}$ associated with any unit vector $\eta \in \mathcal{H}_1^0$. Then

(i) $\iota_1(\mathcal{A}_1)$ is orthogonal to $\iota_2(\mathcal{A}_2)$ w.r.t. $(\varphi, \psi)$,
(ii) $\varphi \circ \iota_1$ agrees with the expectation $\varphi_1$ on $\mathcal{A}_1$,
(iii) $\psi \circ \iota_2$ agrees with the expectation $\varphi_2$ on $\mathcal{A}_2$.

**Proof.** This is Theorem 4.1 adapted to the $C^*$-algebra setting (for an analogous formulation in the monotone case, see [20]).

**Example 4.2.** In Example 4.1, let $\mathcal{A}_i$ be the $C^*$-algebra generated by $A_i$ and the identity $I_i$ on $\mathcal{H}_i$ and let $\varphi_i$ be the state on $\mathcal{A}_i$ associated with the vector $\delta(e_i)$, $i = 1, 2$. Then the pair $(\mathcal{A}, \varphi)$, where $\mathcal{A}$ is the $C^*$-algebra generated by $A^{(1)}$ and $A^{(2)}$ and the identity $I_1 \otimes I_2$ and $\varphi$ is the state on $\mathcal{A} \subset \mathcal{B}(\mathcal{H}_1) \otimes \mathcal{B}(\mathcal{H}_2)$ associated with the vector $\delta(e_1) \otimes \delta(e_2)$, is the orthogonal product of $(\mathcal{A}_1, \varphi_1)$ and $(\mathcal{A}_2, \varphi_2)$.

**Remark 4.2.** The notion of the orthogonal product can also be introduced in the category of noncommutative ($*$-) probability spaces. Then, conditions of Definition 4.1 can be used as defining conditions for the orthogonal product of functionals (states) on the free product $\mathcal{A}_1 \sqcup \mathcal{A}_2$ without identification of units. One can use extensions $\tilde{\mathcal{A}}_1 = \mathcal{A}_1 \ast \mathbb{C}[p_1]$ and $\tilde{\mathcal{A}}_2 = \mathcal{A}_2 \ast \mathbb{C}[p_2]$ by idempotents (projections) $p_1$ and $p_2$ to construct the unital ($*$-) homomorphism $j : \mathcal{A}_1 \sqcup \mathcal{A}_2 \rightarrow \tilde{\mathcal{A}}_1 \otimes \tilde{\mathcal{A}}_2$ as the linear and multiplicative extension of $j(a) = a \otimes p_2$ and $j(b) = p_1^+ \otimes b$ for any $a \in \mathcal{A}_1$ and $b \in \mathcal{A}_2$, where $p_1^+ = 1 - p_1$. Then $\varphi$ agrees with the functional (state) $(\tilde{\varphi}_1 \otimes \tilde{\varphi}_2) \circ j$. In particular, in the case of $*$-probability spaces, this proves positivity of $\varphi$. Therefore, the pair $(\mathcal{A}_1 \sqcup \mathcal{A}_2, \varphi)$ can be defined as the orthogonal product of noncommutative ($*$)-probability spaces $(\mathcal{A}_1, \varphi_1)$ and $(\mathcal{A}_2, \varphi_2)$.

## 5. Orthogonal Convolution

The moments of the ‘orthogonal convolution’ $\mu \vdash \nu$ of compactly supported probability measures can be computed using the Hilbert space realization of Section 4. Keeping the notations of Corollary 4.2, we obtain the following proposition.

**Proposition 5.1.** For any $\pi \in \mathcal{I}(n)$ it holds that

\[
m_{\mu \vdash \nu}(\pi) = \sum_{\sigma \in \mathcal{E}_{\text{odd}}(\pi)} (-1)^{|\sigma'|-|\pi|} k_\mu^*(\sigma') m_\nu(\sigma''),
\]

where $k_\mu^*(\sigma')$ and $m_\nu(\sigma'')$ are given by (3.3) and (3.5) and $\sigma = \sigma' \cup \sigma''$ is the decomposition given by (3.1).
Proof. First consider the case, when $\pi$ consists of one $n$-element block, which we denote $\pi = (n)$, where $n \in \mathbb{N}$. We have
\[
\langle (X_1 \otimes P_2 + P_1 \otimes X_2)^n \xi_1 \otimes \xi_2, \xi_1 \otimes \xi_2 \rangle
\]
\[
= \sum_{r=1}^{n} \sum_{j_1 + j_2 + \ldots + j_r = n \ even} \langle X_1^{j_1} P_1 \otimes X_1^{j_2} P_1 \otimes \ldots \otimes X_1^{j_r} P_1 \otimes \xi_1, \xi_1 \rangle \times \langle P_2 X_2^{j_2} P_2 \ldots X_2^{j_r-1} P_2 \xi_2, \xi_2 \rangle
\]
where we understand that the summation runs over the set of ordered partitions of the number $n$ and thus all $j_k$'s are assumed to be non-zero. Now, if we denote by $\sigma$ the interval partition associated with the tuple $(j_1, j_2, \ldots, j_n)$, and by $\sigma = \sigma' \cup \sigma''$ - the alternating decomposition (3.1), we get
\[
\langle X_1^{j_1} P_1 \otimes X_1^{j_2} P_1 \otimes \ldots \otimes X_1^{j_r} P_1 \otimes \xi_1, \xi_1 \rangle = \sum_{\eta \geq \sigma'} (-1)^{|\eta|-1} m_\mu(\eta) = (-1)^{|\eta|-1} k_\mu^*(\sigma')
\]
where we use (3.5). Moreover,
\[
\langle P_2 X_2^{j_2} P_2 \ldots X_2^{j_r-1} P_2 \xi_2, \xi_2 \rangle = \nu(j_1) \nu(j_2) \ldots \nu(j_r-1) = m_\nu(\sigma'').
\]
This gives (5.1) for $\pi = (n)$. It remains to extend this result multiplicatively to any $\pi \in \mathcal{I}(n)$. Namely, (5.1) holds for every block $\pi_j$ of $\pi$ with summation running over partitions $\sigma(j) \in \mathcal{I}_{\text{odd}}(|\pi_j|)$ with the sign factor equal to $(-1)^{|\sigma'(j)|-1}$. Thus, every block $\pi_j$ is decomposed into and odd number of subblocks. The sum over $\sigma(j)$’s gives (5.1) for any $\pi$ since $\sum_{j}(|\sigma'(j)| - 1) = |\sigma'|-|\pi|$.

One can generalize the definition of the orthogonal convolution and Proposition 5.1 to distributions of an arbitrary orthogonal pair $(a, b)$ of elements of a noncommutative probability space $\mathcal{A}$ (see Remark 4.2).

**Definition 5.1.** Let $(a, b)$ be a pair of random variables from a unital algebra $\mathcal{A}$ which is orthogonal w.r.t. to a pair of normalized linear functionals $(\varphi, \psi)$, with $\mu$ denoting the $\varphi$-distribution of $a$ and $\nu$ denoting the $\psi$-distribution of $b$. By the orthogonal convolution $\mu \triangleright \nu$ we understand the $\varphi$-distribution of $a + b$.

**Example 5.1.** Using Proposition 5.1 as well as (3.3) and (3.5), we get
\[
\begin{align*}
\mu_{a+b}(1) &= \mu_a(1) \\
\mu_{a+b}(2) &= \mu_a(2) \\
\mu_{a+b}(3) &= \mu_a(3) + (\mu_a(2) - \mu_a^2(1))\nu_b(1) \\
\mu_{a+b}(4) &= \mu_a(4) + 2\mu_a(3)\nu_b(1) + \mu_a(2)\nu_b(2) \\
&\quad -2\mu_a(2)\mu_a(1)\nu_b(1) - \mu_a^2(1)\nu_b(2)
\end{align*}
\]
It can be seen that $\mu_{a+b}$ is not symmetric with respect to $\mu_a$ and $\nu_b$. In particular, the first two moments of $a + b$ agree with the moments of $a$. 

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More generally, the moment \( \mu_{a+b}(n) \) for \( n \geq 2 \) can be expressed in terms of the moments \( \mu_n(k) \) of orders \( k \leq n \) and the moments \( \mu_n(l) \) of orders \( l \leq n - 2 \). In fact, using the language of ‘universal polynomials’, we obtain the following analogue of Proposition 4.3 of [1] or Proposition 1.2 of [24].

**Proposition 5.2.** Let \((a, b)\) be a pair of elements of a unital algebra \( A \) which is orthogonal w.r.t. \( (\varphi, \psi) \). The \( \varphi \)-distribution of \( a + b \) depends only on \( \mu_\varphi \) and \( \nu_\psi \) and there are universal polynomials with integer coefficients \( P_m(x_1, \ldots, x_m, y_1, \ldots, y_{m-2}) \) for \( m > 2 \), with \( P_1(x_1) = x_1 \) and \( P_2(x_1, x_2) = x_2 \), such that

1. \( P_m \) is homogeneous of degree \( m \) in the \( x \) and \( y \) variables taken together, where degree \( j \) is assigned to \( x_j \) and \( y_j \);
2. \( \mu_{a+b}(m) = P_n(\mu_a(1), \ldots, \mu_a(m), \nu_b(1), \ldots, \nu_b(m - 2)) \).

**Proof.** It follows directly from (5.1) for the partition \( \pi = (m) \) consisting of one block that \( \mu_{a+b} \) depends only on \( \mu_\varphi \) and \( \nu_\psi \) since \( k_\mu^n(\sigma') \) and \( m_\nu(\sigma') \) depend only on \( \mu_\varphi \) and \( \nu_\psi \). Now, each \( k_\mu^n(\sigma') \) is a polynomial in the moments of \( \mu \) with integer coefficients and \( m_\nu \) is just a product of moments of \( \nu \) as the proof of Proposition 5.1 demonstrates. Assigning the variable \( x_j \) to \( \mu(j) \) and \( y_j \) to \( \nu(j) \), we obtain the polynomial \( P_m \). Since in the expression on the right-hand side of (5.1) we have a summation over odd subpartitions of \( \pi = (m) \), moments \( \mu(j) \) of orders \( j \leq m \) and \( \nu(j) \) of orders \( j \leq m - 2 \) appear and that is why \( P_m \) depends on \( x_1, \ldots, x_m \) and \( y_1, \ldots, y_{m-2} \).

**Corollary 5.3.** Let \( a_1, a_2, \ldots, a_n \in A \) and let \( \mu_1, \mu_2, \ldots, \mu_n \) be their distributions w.r.t. normalized linear functionals \( \varphi_1, \varphi_2, \ldots, \varphi_n \) on \( A \), respectively. If the pair \((a_j, a_{j+1} + \ldots + a_n)\) is orthogonal w.r.t. \( (\varphi_j, \varphi_{j+1}) \) for every \( 1 \leq j \leq n - 1 \), then the \( m \)-th moment of \( a_1 + a_2 + \ldots + a_n \) depends only on \( \mu_r(j_r) \), where \( 1 \leq j_r \leq m - 2r + 2 \) and \( 1 \leq r \leq n \).

**Proof.** A repeated application of Proposition 5.2 gives the assertion.

**6. Reciprocal Cauchy transforms**

Our goal now is to derive a formula expressing the reciprocal Cauchy transform of distribution \( \mu \mapsto \nu \) in terms of those of \( \mu \) and \( \nu \). In the theorem below we shall do it on the level of formal power series for measures with finite moments of all orders. However, we will also show that the RHS of (6.2) is the reciprocal Cauchy transform of a probability measure if \( \mu \) and \( \nu \) are arbitrary probability measures, which gives an analytic approach to the orthogonal convolution.

Let us first prove an elementary combinatorial formula for the reciprocal Cauchy transform \( F_\mu(z) \) of distribution \( \mu \).

**Proposition 6.1.** Let \( \mu \) and \( \nu \) be probability measures with finite moments of all orders. Then the reciprocal Cauchy transform of the distribution \( \mu \) satisfies the equation
$$F_\mu(z) - z = \sum_{n=1}^{\infty} \sum_{\pi \in \mathcal{I}(n)} (-1)^{|\pi|} m_\mu(\pi) z^{-n+1}$$  \hspace{1cm} (6.1)$$

where the right-hand side is understood as a formal power series, where \(m_\mu(\pi)\) is given by (3.2)-(3.3).

**Proof.** We have

\[
F_\mu(z) = \frac{z}{1 + \sum_{n=1}^{\infty} \mu(n) z^{-n}}
\]

\[
= z(1 + \sum_{k=1}^{\infty} (- \sum_{n=1}^{\infty} \mu(n) z^{-n})^k)
\]

\[
= z(1 + \sum_{m=1}^{\infty} \sum_{j_1+j_2+\ldots+j_r=m} (-1)^r \mu(j_1) \mu(j_2) \ldots \mu(j_r) z^{-m})
\]

\[
= z + \sum_{m=1}^{\infty} \sum_{\pi \in \mathcal{I}(m)} (-1)^{|\pi|} m_\mu(\pi) z^{-m+1}
\]

which completes the proof. \(\Box\)

**Theorem 6.2.** Let \(\mu\) and \(\nu\) be probability measures with finite moments of all orders. The reciprocal Cauchy transform of \(\mu \vdash \nu\) is given by the formula

\[
F_{\mu \vdash \nu}(z) = F_\mu(F_\nu(z)) - F_\nu(z) + z
\hspace{1cm} (6.2)

where the right-hand side is understood as a formal power series.

**Proof.** Using Proposition 6.1, we obtain

\[
L : = F_{\mu \vdash \nu}(z) - z = \sum_{n=1}^{\infty} \sum_{\pi \in \mathcal{I}(m)} (-1)^{|\pi|} m_{\mu \vdash \nu}(\pi) z^{-n+1}
\]

\[
R : = F_\mu(F_\nu(z)) - F_\nu(z) = \sum_{m=1}^{\infty} \sum_{\pi \in \mathcal{I}(m)} (-1)^{|\pi|} m_\mu(\pi)(G_\nu(z))^{m-1}.
\]

In turn, the definition of \(G_\nu(z)\) gives

\[
(G_\nu(z))^{m-1} = \sum_{j_1=0}^{\infty} \ldots \sum_{j_{m-1}=0}^{\infty} \nu(j_1) \nu(j_2) \ldots \nu(j_{m-1}) z^{-j_1+j_2+\ldots+j_{m-1}-(m-1)}
\]

\[
= z^{-m+1} \sum_{k=0}^{\infty} \left( \sum_{j_1+j_2+\ldots+j_{m-1}=k} \nu(j_1) \nu(j_2) \ldots \nu(j_{m-1}) \right) z^{-k}
\]

Writing \(L\) and \(R\) in the form of formal power series

\[
L = \sum_{n=1}^{\infty} L_n z^{-n+1}, \quad R = \sum_{n=1}^{\infty} R_n z^{-n+1}
\]
we get

\[ L_n = \sum_{\tau \in \mathcal{I}(n)} (-1)^{|\tau|}m_{\mu_1}(\tau) \]
\[ R_n = \sum_{m=1}^{n} \sum_{\sigma \in \mathcal{I}(m)} (-1)^{|\tau|}m_{\mu}(\sigma) \sum_{j_1+j_2+\ldots+j_m=n-m, j_1,j_2,\ldots,j_m \geq 0} \nu(j_1)\nu(j_2)\ldots\nu(j_{m-1}) \]

and we thus need to show that \( L_n = R_n \) for every \( n \geq 1 \). Using Proposition 5.1, formula (3.5) and Proposition 3.1, we get

\[ L_n = \sum_{\tau \in \mathcal{I}(n)} (-1)^{|\tau|} \sum_{\sigma' \in \mathcal{I}_\text{odd}(\tau)} (-1)^{|\sigma'|}k^*_{\mu}(\sigma')m_{\nu}(\sigma'') \]
\[ = \sum_{\tau \in \mathcal{I}(n)} \sum_{\sigma' \in \mathcal{I}_\text{odd}(\tau), \sigma' \geq \sigma} (-1)^{|\sigma'|}m_{\mu}(\eta')m_{\nu}(\sigma''') \]
\[ = \sum_{\pi \in \mathcal{I}_2(n), \eta' \geq \pi'} m_{\mu}(\eta')m_{\nu}(\pi''') \]
\[ = \sum_{(\pi,\eta') \in \mathcal{D}_2(n)} (-1)^{|\eta'|}m_{\mu}(\eta')m_{\nu}(\pi''') \]

for every \( n \geq 1 \), where every \( \eta \) has the decomposition \( \eta = \eta' \cup \eta'' \) with every block of \( \eta' \) obtained by connecting certain blocks of \( \sigma' \) and \( \eta'' = \pi'' \). Let us finally demonstrate that \( R_n = L_n \) for every \( n \geq 1 \). In the expression for \( R_n \), there is a summation over the set \( \mathcal{F}(n) \) of triples \((m, \sigma, j)\), where \( 1 \leq m \leq n, \sigma \in \mathcal{I}(m) \) and \( j = (j_1, j_2, \ldots, j_{m-1}) \) is a tuple of non-negative integers whose sum is equal to \( n - m \). By Proposition 3.2, we get a bijection \( g : \mathcal{D}_2(n) \to \mathcal{F}(n) \) which assigns to every pair \((\pi, \eta') \in \mathcal{D}_2(n)\) the triple \((m, \sigma, j)\). Therefore, the summation over the set \( \mathcal{F}(n) \) can be replaced by a summation over the set \( \mathcal{D}_2(n) \) and since we can identify \( \sigma \) with \( \eta' \) as the proof of Proposition 3.2 shows, we have \( m_{\mu}(\sigma) = m_{\mu}(\eta') \). Moreover, \( \nu(j_1)\nu(j_2)\ldots\nu(j_{m-1}) = m_{\nu}(\pi''') \). Therefore, \( R_n = L_n \). This completes the proof. 

\[ \text{Corollary 6.3. In terms of K-transforms, the formula of Theorem 6.2 reads} \]

\[ K_{\mu_1,\nu}(z) = K_{\mu}(z - K_{\nu}(z)) \]  

\[ (6.3) \]

where \( K_{\mu}(z) \) and \( K_{\nu}(z) \) are the K-transforms of \( \mu \) and \( \nu \), respectively.

\[ \text{Proof.} \] This is an immediate consequence of Theorem 6.2 and the definition of the K-transform. 

In order to apply (6.2)-(6.3) to some examples, let us recall basic facts on Jacobi continued fractions. It is well-known [2] that every probability measure \( \mu \) with finite moments of all orders is characterized by the sequences of Jacobi parameters \( \alpha = (\alpha_n) \) and \( \omega = (\omega_n) \), \( n \geq 0 \), where \( \alpha_n \in \mathbb{R} \) and \( \omega_n \geq 0 \) (we will call them Jacobi sequences). In that case we use the notation \( J(\mu) = (\alpha, \omega) \). The Cauchy transform of \( \mu \) can then
be expressed as a continued fraction of the form

\[ G_\mu(z) = \frac{1}{z - \alpha_0 - \frac{\omega_0}{z - \alpha_1 - \frac{\omega_1}{z - \alpha_2 - \frac{\omega_2}{\ldots}}}} \]  

(6.4)

and it is understood that if \( \omega_m = 0 \) for some \( m \), then the fraction terminates and, for convenience, we set \( \omega_n = \alpha_n = 0 \) for all \( n > m \). In examples, we will mainly characterize measures by giving their Jacobi sequences and refer the reader to [13] for details and explicit measures.

Let us also introduce the finite approximations of continued fractions (we shall use them in Section 7 and 8). In the case of the Cauchy transform \( G_\mu(z) \) we define them as quotients of polynomials of the form

\[ \left[ G_\mu(z) \right]_m = \frac{N_m(z)}{M_m(z)}, \quad m \geq 1, \]  

(6.5)

where the numerators and the denominators satisfy the same recurrence

\[ Y_{m+1}(z) = (z - \alpha_m)Y_k - \omega_{m-1}Y_{m-1}, \quad m \geq 1, \]

with different initial conditions: \( N_0(z) = 0, N_1(z) = 1 \) and \( M_0(z) = 1, M_1(z) = z - \alpha_0 \) (see [2]). In a similar way we define approximations of arbitrary continued fractions and expressions involving them. In particular, we have

\[ [F_\mu(z)]_m = z - [K_\mu(z)]_{m-1} = \frac{1}{\left[ G_\mu(z) \right]_m}, \quad m \geq 1, \]  

(6.6)

for the approximations of \( F_\mu(z) \) and \( K_\mu(z) \).

For any sequence \( x = (x_0, x_1, x_2, \ldots) \) of real numbers, let us also introduce the backward shift \( s(x) = (x_1, x_2, x_3, \ldots) \). When we apply this shift to Jacobi sequences, we can find a relation between the orthogonal convolution of measures and the monotone convolution.

**Corollary 6.4.** Let \( \mu \) be a probability measure with finite moments of all orders such that \( J(\mu) = (\alpha, \omega) \). Then

\[ F_{\mu_s}(z) = z - \alpha_0 - \frac{\omega_0}{F_{\mu_s}(z)} \]

where \( J(\mu_s) = (s(\alpha), s(\omega)) \), i.e. \( \mu_s \) is a measure associated with shifted Jacobi sequences.

**Proof.** In (6.2), we write \( F_\mu(w) \) as a continued fraction and then substitute \( w = F_\nu(z) \) to obtain

\[ F_{\mu_s}(z) = z - \alpha_0 - \frac{\omega_0}{F_{\mu_s}(w)} = z - \alpha_0 - \frac{\omega_0}{F_{\mu_s}(z)}. \]
which proves our assertion.

**Example 6.1** Let \( J(\mu) = (\alpha, \omega) \) and \( \nu = \delta_a \), where \( a \in \mathbb{R} \). Then \( F_\nu(z) = z - a \) and therefore, using Corollary 6.4, we get
\[
F_{\mu \vdash \nu}(z) = z - \alpha_0 - \frac{\omega_0}{F_\mu(z - a)}
\]
which shows that \( J(\mu \vdash \delta_a) = ((\alpha_0, \alpha_1 + a, \alpha_2 + a, \ldots), \omega) \). In particular, \( \mu \vdash \delta_0 = \mu \), i.e. \( \delta_0 \) is the right identity w.r.t. the operation \( \vdash \) (it is not hard to show that the left identity does not exist). In turn, if \( \mu = \delta_a \) and \( J(\nu) = (\beta, \gamma) \), where \( a \in \mathbb{R} \), then \( \alpha_0 = a \) and \( \omega_0 = 0 \) and therefore, using Corollary 6.4, we obtain \( F_{\mu \vdash \nu}(z) = z - a \), which gives \( \delta_a \vdash \nu = \delta_a \).

**Example 6.2.** Let \( \mu = p\delta_{\lambda_1} + q\delta_{\lambda_2} \), where \( p + q = 1 \) and \( J(\nu) = (\beta, \gamma) \). In that case
\[
G_\mu(z) = \frac{p}{z - \lambda_1} + \frac{q}{z - \lambda_2}
\]
and the reciprocal Cauchy transform is of the form
\[
F_\mu(z) = z - \lambda_1 p - \lambda_2 q - \frac{pq(\lambda_1 - \lambda_2)^2}{z - \lambda_1 q - \lambda_2 p}.
\]
Using (6.2), we obtain
\[
F_{\mu \vdash \nu}(z) = z - \lambda_1 p - \lambda_2 q - \frac{pq(\lambda_1 - \lambda_2)^2}{F_\nu(z) - \lambda_1 q - \lambda_2 p}.
\]
Therefore,
\[
J(\mu \vdash \nu) = \left( (\lambda_1 p + \lambda_2 q, \beta_0 + \lambda_1 q + \lambda_2 p, \beta_1, \beta_2, \ldots), (pq(\lambda_1 - \lambda_2)^2, \gamma_0, \gamma_1, \ldots) \right).
\]
In particular, if \( F_\mu(z) = z - \alpha_0 - \omega_0/z \), then
\[
F_{\mu \vdash \nu}(z) = z - \alpha_0 - \frac{\omega_0}{F_\nu(z)}
\]
and thus \( J(\mu \vdash \nu) = ((\alpha_0, \beta_0, \beta_1, \ldots), (\omega_0, \gamma_0, \gamma_1, \ldots)) \).

**Example 6.3.** A closer look at equation (6.2) shows that it is natural to consider the orthogonal convolution of measures which correspond to mixed periodic J-fractions [13] related to each other as follows:
\[
\begin{align*}
J(\mu) & = ((\alpha_0, \alpha_1, \alpha, \ldots), (\omega_0, \omega_1, \omega, \ldots)) \\
J(\nu) & = ((\beta, \beta + \alpha, \beta + \alpha, \ldots), (\gamma, \gamma + \omega, \gamma + \omega, \ldots))
\end{align*}
\]
(here, \( \alpha, \omega, \beta, \gamma \) denote numbers, not sequences). In that case we have
\[
\begin{align*}
F_\mu(z) & = z - \alpha_0 - \frac{\omega_0}{z - \alpha_1 - \omega_1 W_{(\alpha, \omega)}(z)} \\
F_\nu(z) & = z - \beta - \gamma W_{(\alpha + \beta, \omega + \gamma)}(z)
\end{align*}
\]
where $W_{(a,b)}(z)$ denotes the Cauchy transform of the Wigner measure $\sigma$ with mean $a$ and variance $b$. In this case $J(\mu) = ((\alpha_1, \alpha, \ldots), (\omega_1, \omega, \ldots))$ and thus

$$F_{\mu \triangleright \nu}(z) = F_{\nu}(z) - \alpha_1 - \omega_1 W_{(\alpha, \omega)}(F_{\nu}(z))$$

$$= z - \beta - \gamma W_{(\alpha+\beta, \omega+\gamma)}(z) - \alpha_1 - \omega_1 W_{(\alpha, \omega)}(F_{\nu}(z))$$

$$= z - \beta - \alpha_1 - (\gamma + \omega_1) W_{(\alpha+\beta, \omega+\gamma)}(z)$$

since $W_{(\alpha+\beta, \omega+\gamma)}(z) = W_{(\alpha, \omega)}(F_{\nu}(z))$. Therefore, we obtain another mixed periodic $J$-fraction $J(\mu \triangleright \nu) = ((\alpha_0, \alpha_1 + \beta, \alpha + \beta, \alpha + \beta, \ldots), (\omega_0, \omega_1 + \gamma, \omega + \gamma, \omega + \gamma, \ldots))$. For a discussion on the corresponding measures, see [13].

The notion of the orthogonal convolution can be extended to the class of all probability measures. Namely, by $\mu \triangleright \nu$ we then understand the unique probability measure defined by the reciprocal Cauchy transform of the form (6.2) - that the formula (6.2) gives in fact a function from class $\mathcal{RC}$ is proven below. Note that the binary operation $\triangleright$ is neither commutative nor associative.

**Theorem 6.5.** If $\mu$ and $\nu$ are probability measures on the real line, then the function of the form

$$F(z) = F_{\mu}(F_{\nu}(z)) - F_{\nu}(z) + z$$

(6.7)

defined on $\mathbb{C}^+$, is the reciprocal Cauchy transform of a probability measure on the real line.

**Proof.** In order to demonstrate that $F(z)$ is the reciprocal of the Cauchy transform of a probability measure, we will use the sufficiency condition of Maassen [18] and show that

$$\inf_{z \in \mathbb{C}^+} \frac{\Im(F(z))}{\Im z} = 1$$

where $\Im(u)$ denotes the imaginary part of $u \in \mathbb{C}$. Denoting $w = F_{\nu}(z)$ and using the Nevanlinna representation theorem, we can write $F(z)$ in the form

$$F(z) = z - a - \int_{\mathbb{R}} \frac{1 + x w}{w - x} \, d\tau(x)$$

$$= z - a - \int_{\mathbb{R}} \frac{(1 + x w)(\bar{w} - x)}{(w - x)(\bar{w} - x)} \, d\tau(x)$$

where $\tau$ is a positive finite measure, which gives

$$\Im F(z) = y + \Im w \int_{\mathbb{R}} \frac{1 + x^2}{|w - x|^2} \, d\tau(x) \geq y$$

for $z \in \mathbb{C}^+$ since $\Im w = -\Im(G_{\nu}(z))/|G_{\nu}(z)|^2 \geq 0$ (we have $G_{\nu} : \mathbb{C}_+ \to \mathbb{C}_-$). This implies that

$$\inf_{z \in \mathbb{C}^+} \frac{\Im F(z)}{\Im z} \geq 1.$$
Moreover, we can write
\[ \frac{F(z)}{z} = 1 - \frac{F_1(z)}{z} + \frac{F_2(z)}{z} \]
where
\[ F_1(z) = F_\nu(z) \quad \text{and} \quad F_2(z) = F_\mu(F_\nu(z)) = F_{\mu \triangleright \nu}(z) \]
and thus both \( F_1(z) \) and \( F_2(z) \) are reciprocals of Cauchy transforms of probability measures. This gives
\[ \inf_{z \in \mathbb{C}^+} \frac{\Im F_1(z)}{\Im z} = 1 \quad \text{and} \quad \inf_{z \in \mathbb{C}^+} \frac{\Im F_2(z)}{\Im z} = 1. \]
Observe now that
\[ \Im (F_1(z) - F_2(z)) = \Im \left( K_\mu \left( \frac{1}{G_\nu(z)} \right) \right) \leq 0 \]
for \( z \in \mathbb{C}^+ \) since \( 1/G_\nu : \mathbb{C}_+ \rightarrow \mathbb{C}_+ \) and \( K_\nu : \mathbb{C}^+ \rightarrow \mathbb{C}^- \cup \mathbb{R} \). Therefore
\[ \frac{\Im (F_2(z) - F_1(z))}{\Im z} \geq 0, \quad z \in \mathbb{C}^+ \]
Now, since \( F_1(z) \) and \( F_2(z) \) are holomorphic on \( \mathbb{C}^+ \), we have two real-valued functions, \( f_1(z) := \Im F_1(z)/\Im z \) and \( f_2(z) := \Im F_2(z)/\Im z \), which are continuous on \( \mathbb{C}^+ \) with \( f_2 \geq f_1 \) on \( \mathbb{C}^+ \) and \( \inf_{z \in \mathbb{C}^+} f_1(z) = \inf_{z \in \mathbb{C}^+} f_2(z) = 1 \). Thus, there exists a sequence \( (z_n) \subseteq \mathbb{C}^+ \) such that \( \lim_{n \to \infty} f_2(z_n) = 1 \) and thus \( \lim_{n \to \infty} f_1(z_n) = 1 \) which implies that \( \lim_{n \to \infty} f(z_n) = 0 \), where \( f = f_2 - f_1 \). This proves that
\[ \inf_{z \in \mathbb{C}^+} \frac{\Im F(z)}{\Im z} \leq 1. \]
which completes the proof of the sufficiency condition. \( \blacksquare \)

**Corollary 6.6.** If \( \mu \) and \( \nu \) are probability measures on the real line, then the monotone convolution of \( \mu \) and \( \nu \) can be decomposed as \( \mu \triangleright \nu = (\mu \lhd \nu) \triangleright \nu \).

**Proof.** This decomposition is a direct consequence of (2.3),(2.5) and Theorem 6.5. \( \blacksquare \)
7. Structures related to subordination functions

In analogy to Sections 4 and 5, where we studied ‘orthogonal structures’, we now define and study structures (subalgebras, products and convolutions) related to the subordination functions $F_1(z)$ and $F_2(z)$. In particular, these functions uniquely determine probability measures which can be treated as convolutions of $\mu$ and $\nu$. These convolutions resemble the free additive convolution, except that one measure can be viewed as ‘subordinate’ to the other. The same holds for the associated subalgebras and Hilbert spaces and this motivates our terminology - ‘s-free convolution’, ‘s-free subalgebras’ and ‘s-free product of Hilbert spaces’. In the case of compactly supported probability measures, these structures can also be obtained as inductive limits of ‘alternating orthogonal structures’, but it seems to be of advantage to define them directly.

Let $(\mathcal{H}_i, \xi_i)$, $i = 1, 2$, be Hilbert spaces with distinguished unit vectors. Then their Hilbert space free product $(\mathcal{H}_1, \xi_1) * (\mathcal{H}_2, \xi_2)$ is $(\mathcal{H}, \xi)$ where

$$\mathcal{H} = \mathbb{C} \xi \oplus \bigoplus_{n=1}^{\infty} \bigoplus_{j \neq i_1, \ldots, i_n} \mathcal{H}_{i_1}^0 \otimes \mathcal{H}_{i_2}^0 \otimes \ldots \otimes \mathcal{H}_{i_n}^0$$

(7.1)

with $\mathcal{H}_i^0 = \mathcal{H}_i \ominus \mathbb{C} \xi_i$ and $\xi$ denoting a unit vector (canonical scalar product is used). For any $h \in \mathcal{H}_i$, denote by $\bar{h}_i$ the orthogonal projection of $h$ onto $\mathcal{H}_i^0$. Moreover, let

$$\mathcal{H}^{(n)}(j) = \bigoplus_{i_1 \neq i_2 \neq \ldots \neq i_n} \mathcal{H}_{i_1}^0 \otimes \mathcal{H}_{i_2}^0 \otimes \ldots \otimes \mathcal{H}_{i_n}^0$$

(7.2)

$$\mathcal{K}^{(n)}(j) = \bigoplus_{i_1 \neq i_2 \neq \ldots \neq i_n} \mathcal{H}_{i_1}^0 \otimes \mathcal{H}_{i_2}^0 \otimes \ldots \otimes \mathcal{H}_{i_n}^0$$

(7.3)

for any $j = 1, 2$, and $m \in \mathbb{N}$. For convenience, also set $\mathcal{H}^{(0)}(j) = \mathcal{K}^{(0)}(j) = \mathbb{C} \xi$ with the canonical projection $P_0 : \mathcal{H} \rightarrow \mathbb{C} \xi$. We will also use

$$\mathcal{H}(j) = \bigoplus_{n=1}^{\infty} \mathcal{H}^{(n)}(j), \quad \mathcal{K}(j) = \bigoplus_{n=1}^{\infty} \mathcal{K}^{(n)}(j)$$

(7.4)

for $j = 1, 2$, and $\mathcal{H}^{(n)} = \mathcal{H}^{(n)}(1) \oplus \mathcal{H}^{(n)}(2)$ for $n \in \mathbb{N}$. Thus, we have

$$\mathcal{H} = \mathbb{C} \xi \oplus \mathcal{H}(1) \oplus \mathcal{H}(2) = \mathbb{C} \xi \oplus \mathcal{K}(1) \oplus \mathcal{K}(2)$$

hence $\mathcal{H}$ can be decomposed as the union of $\mathbb{C} \xi$ and two ‘branches’ (originating or ending with $\mathcal{H}_1^0$ or $\mathcal{H}_2^0$).

Using these notations, we can also decompose the free product of Hilbert spaces (7.1) as the orthogonal direct sums

$$\mathcal{H} = \bigoplus_{n \geq 1} \mathcal{H}^{(n-1)}(\bar{j}) \oplus \mathcal{H}^{(n)}(\bar{j})$$

(7.5)

where we adopt the notation $\bar{1} = 2$ and $\bar{2} = 1$. Moreover, let $P_j(n)$ denote the orthogonal projection onto $\mathcal{H}^{(n-1)}(j) \oplus \mathcal{H}^{(n)}(j)$. 

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The above notations will always be used in the following context. Let \((A_i, \varphi_i), i = 1, 2,\) be \(C^*\)-noncommutative probability spaces. We denote by \((\mathcal{H}_i, \pi_i, \xi_i)\) the GNS triple of \((A_i, \varphi_i),\) i.e. \(\mathcal{H}_i\) is a Hilbert space, \(\xi_i\) is a cyclic (unit) vector in \(\mathcal{H}_i\) and \(\pi_i : A_i \to \mathcal{B}(\mathcal{H}_i)\) is a \(*\)-homomorphism, such that \(\varphi_i(a) = \langle \pi_i(a) \xi_i, \xi_i \rangle\) for all \(a \in A_i,\) where \(\langle \cdot, \cdot \rangle\) denotes the scalar product in \(\mathcal{H}_i\) (for simplicity, we use the same notation for all scalar products).

Recall the definition of the free product representation. On \(\mathcal{H}\) we define a \(*\)-representation \(\lambda_i : A_i \to \mathcal{B}(\mathcal{H})\) of each algebra \(A_i, i = 1, 2,\) as follows:

\[
\lambda_i(a)(h_1 \otimes h) = (\pi_i(a) h_1)^0 \otimes h + \langle \pi_i(a) h_i, \xi_i \rangle h \\
\lambda_i(a)(\xi) = (\pi_i(a) \xi_i)^0 + \langle \pi_i(a) \xi_i, \xi_i \rangle \xi
\]

for any \(h \in \mathcal{H}(i)\) and \(h_1 \in \mathcal{H}_i,\) where identifications \(\xi_i \otimes h \equiv h\) are made for any \(h_1 \in \mathcal{H}_i^0\) and \(h \in \mathcal{H}(i).\) The free product of \((\lambda_i)_{i \in \mathcal{I}}\) is the representation \(\lambda = \ast_{i \in \mathcal{I}} \lambda_i : \ast_{i \in \mathcal{I}} A_i \to \mathcal{B}(\mathcal{H})\) given by the linear extension of

\[
(\ast_{i \in \mathcal{I}} \lambda_i)(a_1 a_2 \ldots a_n) = \lambda_i(a_1) \lambda_i(a_2) \ldots \lambda_i(a_n)
\]

for \(a_j \in A_{i_j},\) where \(i_1, i_2, \ldots, i_n\) is a sequence of alternating 1’s and 2’s. Finally, on \(\mathcal{B}(\mathcal{H})\) we define the so-called vacuum state \(\varphi(\cdot) = \langle \Omega, \Omega \rangle\) and the free product of states \((\varphi_i)_{i \in \mathcal{I}}\) as the functional \(\ast_{i \in \mathcal{I}} \varphi_i : \ast_{i \in \mathcal{I}} A_i \to \mathbb{C}\) given by the composition \(\ast_{i \in \mathcal{I}} \varphi_i = \varphi \circ \lambda.\)

Let \(a_1 \in \mathcal{B}(\mathcal{H}_1)\) and \(a_2 \in \mathcal{B}(\mathcal{H}_2)\) be fixed random variables with distributions \(\mu\) and \(\nu,\) respectively. The corresponding free random variables \(\lambda(a_1)\) and \(\lambda(a_2)\) can be decomposed according to the Hilbert space decompositions (7.5) and can be interpreted as consisting of sums of replicas of \(a_1\) and \(a_2,\) respectively (see Proposition 7.1).

**Proposition 7.1.** According to the decomposition (7.5), the free random variable \(\lambda(a_j)\) is the strongly convergent series

\[
\lambda(a_j) = \sum_{n=1}^{\alpha} a_j(n)
\]

where \(a_j(n) = P_j(n) a_j P_j(n)\) are replicas of \(a_j,\) where \(j = 1, 2.\)

**Proof.** Note that the subspace \(\mathcal{H}^{(n-1)}(j) \oplus \mathcal{H}^{(n)}(j)\) is left invariant by \(\lambda(a_j)\) for every \(n \in \mathbb{N}.\) Using the direct sum decomposition (7.5), we get the assertion. \(\blacksquare\)

**Remark 7.1.** Decompositions of free random variables of type given by (7.6) were studied in the algebraic framework of \(*\)-algebras [15], where it was shown that they can be viewed as ‘closed operators’ w.r.t. a suitable topology implemented by a sequence of projections.

From now on, when speaking of free random variables \(a_1, a_2\) as elements of \(\mathcal{B}(\mathcal{H}),\) we will understand that \(a_j = \lambda(a_j) \in \mathcal{B}(\mathcal{H}),\) where \(j = 1, 2.\) The Hilbert space setting will be used below to study the ‘free convolution product’ of \(a_1\) and \(a_2\) as well as the ‘subordinate convolution’ related to the subordination functions. We begin, however,
with a general formulation.

**Definition 7.1.** Let \((A, \varphi, \psi)\) be a unital algebra with a pair of linear normalized functionals. Let \(A_1\) be a unital subalgebra of \(A\) and let \(A_2\) be a non-unital subalgebra with an ‘internal’ unit 1\(_2\), i.e. 1\(_2b = b = b1\(_2\) for every \(b \in A_2\). We say that the pair \((A_1, A_2)\) is free with subordination, or simply \(s\)-free, with respect to \((\varphi, \psi)\) if \(\psi(1\(_2\)) = 1\) and it holds that

\[
\begin{align*}
(i) \quad &\varphi(a_1a_2 \ldots a_n) = 0 \text{ whenever } a_j \in A_{i_j}^0 \text{ and } i_1 \neq i_2 \neq \ldots \neq i_n \\
(ii) \quad &\varphi(w_11\(_2\)w_2) = \varphi(w_1w_2) - \varphi(w_1)\varphi(w_2) \text{ for any } w_1, w_2 \in \text{alg}(A_1, A_2),
\end{align*}
\]

where \(A_i^0 = A_i \cap \ker \varphi \) and \(A_0^2 = A_2 \cap \ker \psi\). We say that the pair \((a, b)\) of random variables from \(A\) is \(s\)-free with respect to \((\varphi, \psi)\) if the unital algebra generated by \(a\) and the (non-unital) algebra generated by \(b\) have this property.

The notion of ‘\(s\)-freeness’ reminds freeness except that the ‘internal’ unit 1\(_2\) is mapped by the GNS representation onto \(P_{\xi}^1 = 1 - P_\xi\), where \(\xi\) is the distinguished unit vector of the Hilbert space, instead of the unit 1 (see below). Note also that condition (ii) resembles condition (ii) of Definition 4.1, but it is weaker since it has 1\(_2\) ‘in the middle’ and not an arbitrary \(b \in A_2\). In particular, since \(A_1\) is unital, it also follows from (ii) that \(\varphi\) vanishes on \(A_2\) (cf. (ii) of Definition 4.1). Let us also point out that by conditions (i)-(ii) of Definition 7.1, \(\varphi\) is uniquely determined on \(\text{alg}(A_1, A_2)\) by restrictions \(\varphi|A_1\) and \(\psi|A_2\) and it vanishes on \(A_2\) (as in the orthogonal case). Finally, note that ‘\(s\)-freeness’ w.r.t. \((\varphi, \psi)\) differs from ‘conditional freeness’ w.r.t. \((\varphi, \psi)\), as in the orthogonal case (in particular, it is not symmetric w.r.t. \(A_1\) and \(A_2\)).

The corresponding Hilbert space setting can be given as follows.

**Definition 7.2.** Let \((H_1, \xi_1)\) and \((H_2, \xi_2)\) be Hilbert spaces with distinguished unit vectors \(\xi_1\) and \(\xi_2\), respectively. The \(s\)-free product of \((H_1, \xi_1)\) and \((H_2, \xi_2)\) is the pair \((K, \xi)\), where \(K = \mathbb{C}\xi \oplus K(2)\). We denote it by \((K, \xi) = (H_1, \xi_1) \oplus (H_2, \xi_2)\) and by \(\varphi\) - the canonical state on \(B(H)\) associated with \(\xi\).

Let us define \(*\)-representations \(\rho_i : B(H_i) \to B(K)\) by strongly convergent series

\[
\rho_1(a_1) = \sum_{r=0}^{\infty} a_1(2r + 1), \quad \rho_2(a_2) = \sum_{r=1}^{\infty} a_2(2r) \quad (7.7)
\]

where \(a_1 \in B(H_1), a_2 \in B(H_2)\). Note that \(\rho_1\) (\(\rho_2\)) is a faithful unital (non-unital) \(*\)-homomorphism. Using these representations, we can describe the properties of the \(s\)-free product of Hilbert spaces. Of course, if we consider \((H_2, \xi_2) \oplus (H_1, \xi_1)\), we need to define a different pair of \(*\)-representations (with the roles of \(B(H_1)\) and \(B(H_2)\) interchanged).

**Theorem 7.2.** Let \((K, \xi) = (H_1, \xi_1) \oplus (H_2, \xi_2)\) be the \(s\)-free product of Hilbert spaces and let \(\varphi, \varphi_1\) and \(\varphi_2\) be the states associated with unit vectors \(\xi, \xi_1\) and \(\xi_2\), respectively. Moreover, let \(\psi\) be the state on \(B(K)\) associated with any unit vector \(\eta \in H_1^0 \subset H\). Finally, let \(A_i = \rho_i(B(H_i))\), where \(i = 1, 2\). Then
(i) the pair \((A_1, A_2)\) is s-free w.r.t. \((\varphi, \psi)\),
(ii) \(\varphi \circ \rho_1\) agrees with the expectation \(\varphi_1\) on \(B(H_1)\),
(iii) \(\psi \circ \rho_2\) agrees with the expectation \(\varphi_2\) on \(B(H_2)\).

Proof. Let \(a_j \in A_{i_j}^0, 1 \leq j \leq n \) with \(i_1 \neq i_2 \neq \ldots \neq i_n\). In order to see that condition (i) of Definition 7.1 holds, first observe that it holds whenever \(a_n \in A_2^0\), since \(\rho_2(b)\xi = 0\) for any \(b \in B(H_2)\). Therefore, assume that \(a_n \in A_1^0\). In that case \(a_n\xi = h_n \in H_1^0, a_{n-1}h_n = h_{n-1} \otimes h_n \in H_1^0 \otimes H_1^0, a_{n-2}h_{n-1} \otimes h_n = h_{n-2} \otimes h_{n-1} \otimes h_n \in H_1^0 \otimes H_1^0 \otimes H_1^0\), etc. Continuing this process, we get condition (i) of Definition 7.1. Finally, let \(w_1, w_2 \in \text{alg}(A_1, A_2)\) and observe that \(1_2 = \rho_2(1_{H_2}) = 1 - \rho_1\). This proves condition (ii) of Definition 7.1 and thus completes the proof of (i). Verification of (ii) and (iii) is straightforward.

COROLLARY 7.3. Let \(\mu, \nu\) be compactly supported probability measures on \(\mathbb{R}\). Then there exist a Hilbert space \(K\), unit vectors \(\xi, \eta \in K\) and self-adjoint bounded random variables \(X_1, X_2 \in B(K)\) such that the pair \((X_1, X_2)\) is s-free w.r.t. \((\varphi, \psi)\), where \(\varphi\) and \(\psi\) are vector states associated with \(\xi, \eta \in K\). Moreover, the \(\varphi\)-distribution of \(X_1\) and the \(\psi\)-distribution of \(X_2\) coincide with \(\mu\) and \(\nu\), respectively. Finally, the \(\varphi\)-distribution of \(X_1 + X_2\), denoted \(\mu \boxplus \nu\), is compactly supported.

Proof. The proof is similar to that of Corollary 4.2 - replace \(H\) by \(K = \mathbb{C}\xi \oplus \mathbb{K}(2)\) and \(*\)-homomorphisms \(\tau_1\) and \(\tau_2\) by \(\rho_1\) and \(\rho_2\), respectively.

The s-free product of \(C^*\)-probability spaces can be defined along the lines of Section 4 (together with a \(C^*\)-version of Theorem 7.2). Computations of convolutions \(\mu \boxplus \nu\) are postponed till Section 8, where the transforms are studied, which provide the natural tools. However, we give here an example which shows a natural connection between the s-free product and branches of the free product of graphs (convolutions \(\mu \boxplus \nu\) will then give their spectral distributions).

EXAMPLE 7.1. Consider two rooted graphs as in Example 4.1. Let \((B_1, e)\) be the (uniformly locally finite) rooted graph called ‘branch subordinate to \(G_1\)’ (or, simply, ‘branch’) obtained from the free product of rooted graphs \((G_1, e_1) \ast (G_2, e_2)\) by restricting the set of vertices \(V_1 \ast V_2\) to the set \(V\) consisting of the empty word \(e\) and words ending with a vertex (letter) from \(V_1^0\). Then the adjacency matrix of \((B_1, e)\) can be decomposed as \(A(B_1) = A^{(1)} + A^{(2)}\), where the summands

\[
A^{(1)} = \sum_{n \text{ odd}} A_1(n), \quad A^{(2)} = \sum_{n \text{ even}} A_2(n),
\]

(with \(n\) positive) are bounded operators on \(l_2(V)\) which are s-free w.r.t. \((\varphi, \psi)\), where \(\varphi(.) = \langle \delta(e), \delta(e) \rangle\) and \(\psi(.) = \langle \delta(v), \delta(v) \rangle\) with \(v \in V_1^0\). Moreover, the pair \((A, \varphi)\), where \(A\) is the \(C^*\)-algebra generated by \(A^{(1)}\), \(A^{(2)}\) and the identity \(I\) on \(l_2(V)\), is the s-free product of the \(C^*\)-probability spaces \((A_1, \varphi_1)\) and \((A_2, \varphi_2)\), where \(A_i\) is generated by \(A_i\) and the unit \(I_i\) on \(l_2(V_i)\), and \(\varphi_i\) is the state defined by the vector \(\delta(e_i), i = 1, 2\). Thus the branch \(B_1\) can be viewed as the s-free product of two subgraphs obtained
from the coverings of $B_1$ built from replicas of $G_1$ and $G_2$, respectively. The 'branch subordinate to $G_2$' can be decomposed in a similar way. A detailed study of 'branches' of the free product of rooted graphs will be given in [1].

8. Subordinate branches

In this Section we study ‘complete’ decompositions of the sum $X_1 + X_2$ of free bounded random variables with distributions $\mu$ and $\nu$, respectively. This leads to the concept of ‘free branches subordinate to $X_1$ or $X_2$’ (terminology is motivated by Example 7.1) which give a Hilbert space realization of the decomposition of the s-free convolution of $\mu$ and $\nu$ into a sequence of orthogonal convolutions of alternating $\mu$ and $\nu$.

**Definition 8.1.** Let $(X_1, X_2)$ be a pair of self-adjoint random variables from $B(H)$ which are free w.r.t. $\varphi$. The self-adjoint random variable from $B(H)$ given by the strongly convergent series

$$B_j(k) = \sum_{r=0}^{\infty} X_j(2r + k) + \sum_{r=0}^{\infty} X_j(2r + k + 1)$$

(8.1)

where $j = 1, 2$ and $k \in \mathbb{N}$, is called the $k$-th free branch subordinate to $X_j$.

In particular, for $k = 1$, the 1-st free branch subordinate to $X_1$ takes the form

$$B_1(1) = \rho_1(X_1) + \rho_2(X_2)$$

which, for simplicity, we shall denote $B_1$. Free branches of higher orders can then be obtained from the recursions

$$B_j(k) = X_j(k) + B_j(k + 1)$$

which shows that we get two disjoint sequences of ‘free branches’ with alternating subordination, beginning either with $B_1 \equiv B_1(1)$ or $B_2 \equiv B_2(1)$.

Viewing the ‘free convolution product’ of $X_1$ and $X_2$ as a tree-like structure with two types of ‘leaves’ (which it really is when $X_1$ and $X_2$ are taken to be adjacency matrices of graphs, cf. Example 7.1), one can interpret free branches as follows. The branch $B_1$ is the ‘half’ of the ‘tree’ which originates with ‘leaf’ $X_1(1)$ (the first replica of $X_1$), followed by ‘leaves’ produced by $X_2(2)$ (the second replica of $X_2$), then ‘leaves’ produced by $X_1(3)$ (the third replica of $X_3$), etc. In the case of branches of higher order, $B_1(k)$ is the part of the ‘tree’ which originates from ‘leaves’ $X_1(k)$ at height $k$, followed by ‘leaves’ $X_2(k + 1)$, then $X_1(k + 2)$, etc (it is similar for branches which are ‘subordinate’ to $X_2$).

**Proposition 8.1.** Let $(X_1, X_2)$ be a pair of free self-adjoint bounded random variables from $B(H)$ with distributions $\mu$ and $\nu$, respectively. For $k \in \mathbb{N} \cup \{0\}$, let $\varphi_k$
and $\psi_k$ denote states on $\mathcal{B}(\mathcal{H})$ associated with arbitrary unit vectors $\zeta_k \in \mathcal{H}(k)(2)$ and $\eta_k \in \mathcal{H}(k)(1)$, respectively. Then the $\psi_{k-1}$-distribution of $B_1(k)$ and the $\varphi_{k-1}$-distribution of $B_2(k)$ are given by $\mu \boxplus \nu$ and $\nu \boxplus \mu$, respectively, for every $k \in \mathbb{N}$.

Proof. For $k = 1$ the assertion is a direct consequence of Definition 8.1 and Corollary 7.3. For $k > 1$, without loss of generality, consider the $\psi_{k-1}$-distribution of $B_1(k)$. We need to show that $\psi_{k-1}(B_1^n(k)) = \varphi(B_1^n)$ for every $k \geq 2$ and $n \in \mathbb{N}$. Therefore we can write

$$B_1^n(k)\eta_{k-1} = (B_1^n \xi_1) \otimes \eta_{k-1}$$

if we identify $\xi_1 \otimes \eta_{k-1}$ with $\eta_{k-1}$. In fact, in the definition of $B_1(k)$, we only have $X_1(j)$’s with $j \geq k$ and $X_2(j)$’s with $j \geq k + 1$, thus $X_1(k - 1)$ is missing, which is the only summand among all $X_1(r)$’s and $X_2(r)$’s in (7.6) which can map $\eta_{k-1}$ onto a vector which is not in $\mathbb{C}\eta_{k-1}$. Therefore

$$B_1^n(k)\eta_{k-1} = \varphi(B_1^n)\eta_{k-1} \mod (\mathcal{H} \ominus \mathcal{H}(k-1)(1))$$

which gives $\psi_{k-1}(B_1^n(k)) = \varphi(B_1^n)$ and that completes the proof. $\blacksquare$

Lemma 8.2. For every $k \in \mathbb{N}$, the pair of random variables $(X_1(k), B_2(k+1))$ is orthogonal with respect to $(\psi_{k-1}, \varphi_k)$ and the pair $(X_2(k), B_1(k+1))$ is orthogonal with respect to $(\varphi_{k-1}, \psi_k)$.

Proof. Without loss of generality, consider the pair $(X_2(k), B_1(k+1))$. For simplicity, denote $b = B_1(k+1)$, $y = Y(k)$ and identify each element $w$ of the algebra generated by $b$ and $y$ with $\lambda(w)$. We need to show two orthogonality conditions of Definition 4.1:

$$\varphi_{k-1}(w_1 b) = \varphi_{k-1}(b w_1) = 0$$

$$\varphi_{k-1}(w_1 b^n y w_2) = \psi_k(b^n)(\varphi_{k-1}(w_1 y^2 w_2) - \varphi_{k-1}(w_1 y)\varphi_{k-1}(yw_2))$$

for any $w_1, w_2 \in \text{alg}(b, y)$. We have

$$\varphi_{k-1}(w_1 b) = \langle w_1 b \zeta_{k-1}, \zeta_{k-1} \rangle$$

where $\zeta_{k-1} \in \mathcal{H}(k-1)(2)$ is a unit vector (it is convenient to think of a simple tensor of length $k - 1$ which begins with a vector from $\mathcal{H}_1^0$). By Definition 8.1, we have $b \zeta_{k-1} = 0$. This proves that $\varphi_{k-1}(w_1 b) = 0$ (the equation $\varphi_{k-1}(b w_2) = 0$ can be obtained from this by taking the adjoints). This gives the first orthogonality condition.

To prove the second orthogonality condition, consider now $y w_2 \zeta_{k-1}$. Using the definition of $b$ again, we obtain

$$y w_2 \zeta_{k-1} = \alpha \zeta_{k-1} + h_2 \otimes \zeta_{k-1}$$

for some $\alpha \in \mathbb{C}$ and $h_2 \in \mathcal{H}_2^0$. Now, we know that $b^n \zeta_{k-1} = 0$ for every $n \in \mathbb{N}$ since $b \zeta_{k-1} = 0$, and, moreover,

$$b^n h_2 \otimes \zeta_{k-1} = \psi_k(b^n)h_2 \otimes \zeta_{k-1} \mod (\mathcal{H}^{k+1} \oplus \ldots \oplus \mathcal{H}^{k+n}).$$
This gives

\[
\varphi_{k-1}(w_1 y^n y w_2) = \langle w_1 y^n y w_2 \zeta_{k-1}, \zeta_{k-1} \rangle
\]

\[
= \langle w_1 y^n (\alpha \zeta_{k-1} + h_2 \otimes \zeta_{k-1}), \zeta_{k-1} \rangle
\]

\[
= \psi_k(b^n)\langle w_1 y h_2 \otimes \zeta_{k-1}, \zeta_{k-1} \rangle
\]

\[
= \psi_k(b^n)\langle w_1 y (y w_2 \zeta_{k-1} - \alpha \zeta_{k-1}), \zeta_{k-1} \rangle
\]

\[
= \psi_k(b^n)\langle w_1 y^2 w_2 \zeta_{k-1}, \zeta_{k-1} \rangle - \alpha \langle w_1 y \zeta_{k-1}, \zeta_{k-1} \rangle
\]

\[
= \psi_k(b^n)(\varphi_{k-1}(w_1 y^2 w_2) - \varphi_{k-1}(y w_2)\varphi_{k-1}(w y))
\]

since \( \alpha = \langle y w_2 \zeta_{k-1}, \zeta_{k-1} \rangle = \varphi_{k-1}(y w_2) \). Thus, we proved the second orthogonality condition for \((X_2(k), B_1(k+1))\), which completes the proof of the lemma.

**Lemma 8.3.** Let \( \mu \) and \( \nu \) be the \( \varphi \)-distributions of \( X_1 \) and \( X_2 \), respectively, and let \((\mu \vdash_n \nu)_{n \geq 1}\) be a sequence of distributions defined recursively by

\[
\mu \vdash_1 \nu = \mu \vdash \nu, \quad \mu \vdash_n \nu = \mu \vdash (\nu \vdash_{n-1} \mu), \quad n \geq 2.
\]

Then the moments of \( B_1 \) and \( B_2 \) of orders \( \leq 2m \) in the state \( \varphi \) agree with the corresponding moments of \( \mu \vdash_{m-1} \nu \) and \( \nu \vdash_{m-1} \mu \), respectively.

**Proof.** Let us consider only the \( \varphi \)-distribution of \( b = B_1 \) since the proof for \( B_2 \) is identical. We have

\[
b = Z_1 + Z_2 + \ldots + Z_m
\]

where \( Z_1 = X_1(1), Z_2 = X_2(2), Z_3 = X_1(3), \ldots, Z_{m-1} = X_1(m-1), Z_m = B_2(m) \). In view of Lemma 8.2, the pairs

\[
(Z_1, Z_2 + \ldots + Z_m), (Z_2, Z_3 + \ldots + Z_m), \ldots, (Z_{m-1}, Z_m)
\]

are orthogonal w.r.t. \((\psi_0, \varphi_1), (\varphi_1, \psi_2), \ldots, (\psi_{m-2}, \varphi_{m-1})\), respectively. Thus,

\[
\varphi(b^r) = \mu_1 \vdash (\mu_2 \vdash (\ldots (\mu_{m-1} \vdash \mu_m) \ldots))(r)
\]

for every natural \( r \), where \( \mu_1, \mu_2, \ldots, \mu_m \) are the \( \psi_0, \varphi_1, \psi_2, \ldots, \varphi_{m-1} \)-distributions of \( Z_1, Z_2, \ldots, Z_m \), respectively. We know that \( \mu_1 = \mu, \mu_2 = \nu, \ldots, \mu_{m-1} = \mu \) and \( \mu_m \) is the distribution of \( B_2(m) \). Thus, in view of Corollary 5.3, \( \varphi(b^r) \) depends on moments of \( \mu_i \) of orders \( 1 \leq j_i \leq r - 2i + 2 \) for \( 1 \leq i \leq \lfloor 1/2(r+1) \rfloor \). In particular, it depends on the moments of \( \mu_m \) if and only if \( r \geq 2m - 1 \). However, if \( r < 2m \), then only the moments of \( \mu_m \) of orders \( \leq 2 \) may come into play. The latter, however, agree with the moments of \( \nu \). This proves our assertion.

**Theorem 8.4.** For compactly supported \( \mu \) and \( \nu \), it holds that

\[
K_{\mu \oplus \nu}(z) = \lim_{m \to \infty} K_{\mu \vdash_m \nu}(z)
\]

where the convergence is uniform on compact subsets of \( \mathbb{C}^+ \).
Proof. In view of Lemma 8.3, we have convergence of moments and thus (since all measures involved have compact support) weak convergence

\[ w - \lim_{m \to \infty} (\mu \uplus_m \nu) = \mu \boxplus \nu \]

which implies that \( K_{\mu \uplus_m \nu}(z) \) converges uniformly to \( K_{\mu \boxplus \nu}(z) \) on compact subsets of \( \mathbb{C}^+ \). Using Corollary 6.3, we obtain

\[
K_{\mu \uplus_m \nu}(z) = K_{\mu_1}(z - K_{\mu_2}(z - K_{\mu_3}(\ldots (K_{\mu_m}(z))))))
\]

where \( \mu_1 = \mu, \mu_2 = \nu, \mu_3 = \mu, \ldots, \mu_m = \mu \) (if \( m \) even) or \( \mu_m = \nu \) (if \( m \) odd). This completes the proof. ■

Remark 8.1. An informal way of writing (8.3) is to use the ‘continued composition’ form

\[ K_{\mu \boxplus \nu}(z) = K_{\mu}(z - K_{\nu}(z - K_{\mu}(z - K_{\nu}(\ldots)))) \tag{8.4} \]

which is particularly appealing when the \( K \)-transforms involved are simple and can be easily associated with Jacobi continued fractions.

Remark 8.2. It follows from Lemma 8.3 that the moments of \( \mu \boxplus \nu \) depend only on the moments of \( \mu \) and \( \nu \). Moreover, from Corollary 7.3 we know that \( \boxplus \) is a binary operation on \( \mathcal{M}_c \) (neither commutative nor associative). Finally, using Theorem 8.4 (see also Example 8.1 below), we can see that \( \delta_0 \) is the right identity w.r.t. \( \boxplus \) (the left identity does not exist).

Remark 8.3. More generally, if \( \mu \) and \( \nu \) are arbitrary probability measures, one can define \( \mu \boxplus \nu \) and \( \nu \boxplus \mu \) as the unique probability measures associated with the reciprocal Cauchy transforms \( F_1(z) \) and \( F_2(z) \), respectively, and thus \( \boxplus \) extends to a binary operation on \( \mathcal{M} \). Moreover, \( \delta_0 \) is the right identity w.r.t. the operation \( \boxplus \) on all of \( \mathcal{M} \) since \( F_1(z) = z, F_2(z) = F_{\mu}(z) \) are the unique functions from class \( \mathcal{RC} \) which satisfy (1.8).

Example 8.1. If \( \mu = \delta_a \) and \( J(\nu) = (\beta, \gamma) \), then we can use properties of the orthogonal convolution (see Example 6.1) to obtain \( \delta_a \boxplus \nu = \delta_a \uplus (\nu \boxplus \delta_a) = \delta_a \) since \( \delta_a \uplus \sigma = \delta_a \) for any \( \sigma \). Similarly, if \( \nu = \delta_a \) and \( J(\mu) = (\alpha, \omega) \), we get \( \mu \boxplus \delta_a = \mu \uplus (\delta_a \boxplus (\mu \boxplus \delta_a)) = \mu \uplus \delta_a \) and we already know that \( J(\mu \uplus \delta_a) = ((\alpha_0, \alpha_1 + a, \alpha_2 + a, \ldots), \omega) \), which gives \( J(\mu \boxplus \nu) \).

Example 8.2. Let \( K_{\mu}(z) = \alpha + \omega/z \) and \( K_{\nu}(z) = \beta + \gamma/z \), where \( \omega \neq 0 \neq \gamma \). Using Theorem 8.4 and the definition of the \( K \)-transform, we obtain

\[
K_{\mu \boxplus \nu}(z) = \alpha + \frac{\omega}{z - \beta - \frac{\gamma}{z - \alpha - \frac{\omega}{z - \beta - \gamma}}}\ldots
\]
and therefore, \( G_{\mu \boxdot \nu}(z) \) is a 2-periodic J-fraction. Algebraic calculations give
\[
G_{\mu \boxdot \nu}(z) = \frac{P(z) - 2\gamma - \sqrt{P^2(z) - 4A^2}}{2\Delta(z)}
\]
where \( P(z) = -\gamma - \omega + (z - \alpha)(z - \beta), \Delta(z) = \gamma(z - \alpha) \) and \( A^2 = \gamma\omega \). The absolutely continuous part of this measure is of the form
\[
f(x) = \frac{\sqrt{4A^2 - P^2(x)}}{2\pi(x - \alpha)}
\]
on the ‘stable band’ with end-points \( \alpha + \beta \pm \sqrt{(\alpha - \beta)^2 + 4\gamma + 4\omega + 8A} \), with a possible atom at \( x = \alpha \) if \( \gamma \neq \omega \). For details, see [13].

Example 8.3. Consider \( \mu \) to be the Wigner measure with mean \( \alpha \) and variance \( \omega \). Then it holds that \( J(\mu \boxdot \mu) = ((\alpha, 2\alpha, 2\alpha, \ldots), (\omega, 2\omega, 2\omega, \ldots)) \). To show this, it is enough to derive a suitable formula for \( J_m(z) := [K_{\mu \boxdot \mu}(z)]_m \). Using the continued fraction
\[
K_{\mu}(z) = \alpha + \frac{\omega}{z - \alpha - \frac{\omega}{z - \alpha - \frac{\omega}{\ldots}}}
\]
and its finite approximations (6.6), we obtain by the induction argument
\[
J_m(z) = [K_{\mu + m-1 \mu}(z)]_m
\]
\[
= [K_{\mu}(z - J_{m-1}(z))]_m
\]
\[
= \alpha + \frac{\omega}{z - \alpha - J_{m-1}(z) - \frac{\omega}{z - \alpha - J_{m-2}(z) - \frac{\omega}{\ldots}}}
\]
and thus \( K_{\mu \boxdot \mu}(z) = \omega W_{2\alpha, 2\omega}(z) \), which proves our claim that we get a mixed periodic J-fraction. The analytic form of the Cauchy transform is
\[
G_{\mu \boxdot \mu}(z) = \frac{\Lambda_1(z) - \sqrt{P^2(z) - A^2}}{2\Gamma_1(z)}
\]
where \( P(z) = (z - 2\alpha)^2, A^2 = 2\omega^2, \Lambda_1(z) = 3z - 2\alpha \) and \( \Gamma_1(z) = z^2 - z\alpha + \omega^2 \), using the notation of [Theorem 3.4, Ref.13] (with \( N = 1, M = 1 \)). Thus \( \mu \boxdot \mu \) is the measure with the absolutely continuous part given by the density
\[
f(x) = \frac{\sqrt{A^2 - P^2(x)}}{2\pi\Gamma_1(x)}
\]
on the ‘stable band’ with endpoints at $\alpha \pm 2\sqrt{2}\omega$, with possible atoms at $1/2(\alpha \pm \sqrt{\alpha^2 - 4\omega^2})$ if $\alpha^2 - 4\omega^2 \geq 0$.

It follows from Theorem 8.4 that every compactly supported probability measure whose Cauchy transform can be represented in the form of a 2-periodic continued fraction is the weak limit of a sequence $\mu \vdash_m \nu$ for suitably chosen $\mu$ and $\nu$. This corollary can be easily generalized to make every compactly supported probability measure the weak limit of a sequence of $m$-fold orthogonal convolutions. It can be shown that this setting corresponds to the model of freeness with infinitely many states [10], but we will not discuss this connection here.

**Corollary 8.5.** Let $\mu$ be a compactly supported probability measure on the real line such that $J(\mu) = (\alpha, \omega)$ and let $(\mu_n)_{n \in \mathbb{N}}$ be discrete probability measures with $K$-transforms

$$K_{\mu_1}(z) = \alpha_0 + \frac{\omega_0}{z - \alpha_1}, \quad K_{\mu_n}(z) = \frac{\omega_{n-1}}{z - \alpha_n}, \quad n \geq 2.$$ 

Then $\mu = w - \lim_{m \to \infty} \mu_1 \vdash (\mu_2 \vdash (\ldots \vdash \mu_m))$.

**Proof.** One can observe that the sequence of truncations of the $K$-transform of $\mu$ has the form

$$[K_{\mu}(z)]_m = K_{\mu_1}(z - K_{\mu_2}(z - \ldots (z - K_{\mu_m}(z)))) = K_{\mu_1 \vdash (\mu_2 \vdash (\ldots \vdash \mu_m))}(z)$$

and therefore it converges uniformly to $K_{\mu}(z)$ on the compact subsets of $\mathbb{C}^+$, from which the assertion follows. $\blacksquare$

**Remark 8.4.** We end this section with a comment on graphs and their products (cf. Examples 4.1, 4.2, 7.1) which actually motivated some of our terminology. In the context of rooted graphs, Theorem 8.4 corresponds to the ‘complete’ decomposition of branches (briefly discussed in Example 7.1) as orthogonal products of (replicas of) alternating $\mathcal{G}_1$ and $\mathcal{G}_2$. Namely, taking two rooted graphs as in Example 4.1, we obtain

$$\mathcal{B}_1 = \mathcal{G}_1 \vdash (\mathcal{G}_2 \vdash (\mathcal{G}_1 \vdash (\mathcal{G}_2 \vdash (\ldots))))$$

in the case of $\mathcal{B}_1$. The spectral distributions of this branch associated with the vector $\delta(e)$ is expressed in terms of $\mu$ and $\nu$ (with the notation of Example 4.1) by their s-free convolution $\mu \boxplus \nu = w - \lim_{m \to \infty} (\mu \vdash_m \nu)$. Similar formulas hold for $\mathcal{B}_2$. For more details, see [1].

9. **Decomposition of the additive free convolution**

Using decompositions of free branches of Section 8, we can show now that the distribution of the sum $X_1 + X_2$ of self-adjoint free random variables $X_1, X_2 \in \mathcal{B}(\mathcal{H})$ with $\varphi$-distributions $\mu$ and $\nu$, respectively, can be ‘completely’ decomposed.
Lemma 9.1. In the decomposition $X_1 + X_2 = B_1 + B_2$, the branches $B_1$ and $B_2$ are boolean independent with respect to $\varphi$. Therefore, $\mu \boxdot \nu = (\mu \boxdot \nu) \varphi (\nu \boxdot \mu)$ is the corresponding decomposition of the free additive convolution.

Proof. Let $w$ be any element of the $*$-algebra generated by $B_1$ and $B_2$ and, for simplicity, denote $x = X_1(1)$. In order to show boolean independence of $B_1$ and $B_2$, we compute the moment with $B_1^k$ at the ‘end’ of the moment (the other case is similar):

\[
\varphi (w B_2 B_1^k) = \langle w B_2 B_1^k \xi, \xi \rangle \\
= \langle w B_2 x^k, \xi \rangle \\
= \langle x^k, \xi \rangle \langle w B_2, \xi \rangle + \langle w B_2 (x^k)^0, \xi \rangle \\
= \varphi (x^k) \varphi (w B_2) \\
= \varphi (B_1^k) \varphi (w B_2)
\]

since $B_2 h_1 = 0$ for every $h_1 \in \mathcal{H}^0_1$. This completes the proof. ■

Lemma 9.2. In the decomposition $X_1 + X_2 = X_j(1) + Z_j$, where $j = 1, 2$, the pair of random variables $(X_j(1), Z_j)$ is monotone independent w.r.t. $\varphi$. Therefore,

\[
\mu \boxplus \nu = \mu \Rightarrow (\nu \boxplus \mu), \quad \nu \boxplus \mu = \nu \Rightarrow (\mu \boxplus \nu) \quad (9.1)
\]

are the corresponding decompositions of the free additive convolution for $j = 1, 2$, respectively.

Proof. It is enough to consider the decomposition for $j = 1$ (the proof for $j = 2$ is similar). For simplicity, denote $z = Z_1$, $x = X_1(1)$, $b_1 = B_1 = B_2(1)$, $b_2 = B_2(2)$ and $p = P_1(1)$. In order to show that (2.2) holds for the pair $(x, z)$, consider two cases.

Case 1. Suppose that $z^n, n \in \mathbb{N}$, is at the ‘end’ of the moment. We need to show that

\[
\varphi (w x z^n) = \varphi (w x) \varphi (z^n) \quad (9.2)
\]

for any $w \in \text{alg}(x, z)$. Write

\[
\varphi (w x z^n) = \langle w x z^n, \xi \rangle
\]

and examine $z^n \xi$. Since $z = b_1 + b_2$ does not contain $X_1(1)$, it holds that

\[
z^n \xi = b_1^n \xi \in \mathbb{C} \xi \oplus \mathcal{K}^{(1)}(1) \oplus \ldots \oplus \mathcal{K}^{(n)}(1)
\]

which implies that

\[
z^n \xi = \varphi (z^n) \xi \mod (\mathcal{K}^{(1)}(1) \oplus \ldots \oplus \mathcal{K}^{(n)}(1)) \quad (9.3)
\]

for any $n \in \mathbb{N}$. In particular, this gives $\varphi (z^n) = \varphi (b_1^n)$, i.e. that $z$ and $b_1$ are $\varphi$-identically distributed. Now, since $x = pXp$ and $p : \mathcal{H} \to \mathbb{C} \xi \oplus \mathcal{H}^0_1$, we have

\[
x z^n \xi = \varphi (z^n) x \xi
\]

from which our claim follows.
Case 2. Suppose now that $z^n$ is in the ‘middle’ of the moment. We need to show that
\[ \varphi(w_1 xz^n x w_2) = \varphi(z^n) \varphi(w_1 x^2 w_2) \] (9.4)
for any $w_1, w_2 \in \text{alg}(x, z)$. Let $h \in \mathcal{H}_1^0$ be a vector of norm $\| h \| = 1$. Using decomposition (8.1) for $b_1$ and $b_2$, we get
\[ z^n h = \sum_{(i_1, j_1), \ldots, (i_n, j_n) \in I_2} X_{i_1}(j_1) X_{i_2}(j_2) \ldots X_{i_n}(j_n) h = b_2^n h \]
where $I_2$ consists of sequences $((i_1, j_1), \ldots, (i_n, j_n))$ ending with $(i_n, j_n) = (2, 2)$ and such that for any $2 \leq r \leq n$, either $(i_{r-1}, j_{r-1}) = (i_r, j_r)$, or $|i_{r-1} - i_r| = 1$ and $j_{r-1} = j_r + 1$ (this is because $X_{i}(j) X_{i'}(j') = 0$ unless $(i, j) = (i', j')$ or $|i - i'| = 1$ and $j = j' + 1$). Therefore,
\[ z^n h = \langle b_2^n h, h \rangle h \mod (\mathcal{K}^{(2)}(2) \oplus \cdots \oplus \mathcal{K}^{(n+1)}(2)) \]
\[ = \varphi(b_2^n) h \mod (\mathcal{K}^{(2)}(2) \oplus \cdots \oplus \mathcal{K}^{(n+1)}(2)) \]
where we used Proposition 8.1. This, together with (9.3) and the fact that $z$ and $b_1$ are $\varphi$-identically distributed, implies that
\[ z^n(h + \alpha \xi) - \varphi(z^n)(h + \alpha \xi) \perp \mathbb{C} \xi \oplus \mathcal{H}_1^0 \]
for any $\alpha \in \mathbb{C}$. Since we have $xz^nx = pxpz^npzp$ and $p$ is the canonical projection onto $\mathbb{C} \xi \oplus \mathcal{H}_1^0$, we get (9.4). This proves that the pair $(x, z)$ is monotone independent w.r.t. $\varphi$. Since $z$ and $b_1$ are $\varphi$-identically distributed, we get the desired decompositions of the additive free convolution. \qed

It remains to connect the ‘orthogonal’ decomposition of $\mu \boxplus \nu$ given by Theorem 8.4 with the ‘boolean’ and ‘monotone’ decompositions of $\mu \boxplus \nu$ given by Lemmas 9.1-9.2. We formulate this result, using the associated transforms: the $K$-transform in the ‘symmetric’ boolean case and the reciprocal Cauchy transform in the ‘non-symmetric’ monotone case.

**Theorem 9.3.** If $\mu$ and $\nu$ are the distributions of $X_1$ and $X_2$, then
\[ F_{\mu \boxplus \nu}(z) = \lim_{m \to \infty} F_{\mu}(F_{\nu \uparrow m \mu}(z)) \] (9.5)
\[ K_{\mu \boxplus \nu}(z) = \lim_{m \to \infty} (K_{\mu \uparrow m \nu}(z) + K_{\nu \uparrow m \mu}(z)) \] (9.6)
where the convergence is uniform on compact subsets of $\mathbb{C}^+$. 

**Proof.** For any $m \in \mathbb{N}$, the moments $(\mu_1 \gg_2 \mu_2)(m)$ and $(\mu_1 \ll_2 \mu_2)(m)$ of compactly supported measures depend only on the moments of $\mu_1$ and $\mu_2$ of orders $\leq m$. Therefore, in view of (2.3), Lemma 8.3 and Lemma 9.2, the moments of $X_1 + X_2$ of orders $\leq m$ agree with the corresponding moments of $\mu \gg (\nu \uparrow m \mu)$ (the same holds for the
moments of $\nu \triangleright (\mu \triangleright_{m-1} \nu)$). Similarly, in view of (2.5), Lemma 8.3 and Lemma 9.1, they agree with the corresponding moments of $(\mu \upharpoonright_{m-1} \nu) \uplus (\nu \upharpoonright_{m-1} \mu)$. Therefore,

$$
\mu \uplus \nu = w - \lim_{n \to \infty} \mu \triangleright (\nu \upharpoonright_n \mu)
$$

$$
\mu \uplus \nu = w - \lim_{n \to \infty} ((\mu \upharpoonright_n \nu) \uplus (\nu \upharpoonright_n \mu))
$$

which, by Theorem 7.4, gives the assertion.

**Remark 9.1.** One can also write (9.5)-(9.6) in the ‘continued composition’ form, similar to (8.4). In particular, (9.5) gives

$$
F_{\mu \uplus \nu}(z) = F_{\mu}(z - K_{\nu}(z - K_{\mu}(z - \ldots)))
$$

(9.7)

which is particularly useful when $\mu$ and $\nu$ have simple $K$-transforms and computations on continued fractions can be carried out (roughly speaking, the difficulty in computing the free additive convolution is related to the difficulties arising in the addition of continued fractions).

**Example 9.1.** We have $\mu \uplus \delta_a = \mu \triangleright (\delta_a \uplus \mu) = \mu \triangleright \delta_a \sim \mu_a$, where by $\mu_a$ we denote the measure associated with the Jacobi sequences $((\alpha_0 + a, \alpha_1 + a, \ldots), \omega)$). On the other hand, $\delta_a \uplus \mu = \delta_a \triangleright (\mu \uplus \delta_a) = \mu_a$ since $J(\mu \uplus \delta_a) = ((\alpha_0, \alpha_1 + a, \alpha_2 + a, \ldots), \omega)$ and $F_{\delta_a \uplus \omega}(z) = F_{\nu}(z) - a$ for any $\nu$.

**Example 9.2.** Let $F_{\mu}(z) = z - \alpha - \omega/z$ and $F_{\nu}(z) = z - \beta - \gamma/z$. Using Example 8.2 and Theorem 9.3, we obtain

$$
F_{\mu \uplus \nu}(z) = F_{\mu}(F_{\nu \uplus \mu}(z))
$$

$$
= F_{\nu \uplus \mu}(z) - \alpha - \frac{\omega}{F_{\nu \uplus \mu}(z)}.
$$

This is the algebraic relation which can be used to compute the explicit form of the (reciprocal) Cauchy transform of $\mu \uplus \nu$. Since the solution has a complicated form and it is known (see, for instance [12]), we shall not give it here.

**Example 9.3.** Let $\mu$ be the Wigner measure with mean $\alpha$ and variance $\omega$. Then it holds that $J(\mu \uplus \mu) = ((2\alpha, 2\alpha, \ldots), (2\omega, 2\omega, \ldots))$, which can be shown by using Theorem 9.3 and the result of Example 8.3. Namely,

$$
[F_{\mu \uplus \mu}(z)]_m = [F_{\mu}(F_{\mu \uplus \mu}(z)))_m
$$

$$
= z - J_m(z) - \alpha - \frac{\omega}{z - J_{m-1}(z) - \alpha - \frac{\omega}{z - J_{m-2}(z) - \alpha - \ldots}}
$$

$$
= z - 2J_m(z)
$$

which proves our assertion.
Remark 9.2. The decompositions of the free additive convolution of Theorem 9.3 correspond to natural decompositions of the free product of graphs. If we consider two rooted graphs as in Examples 4.1 and 7.1, Theorem 9.3 gives decompositions

$$G_1 \ast G_2 = G_1 \triangleright (G_2 \triangleright (G_1 \triangleright (\ldots)))$$
$$G_1 \ast G_2 = (G_1 \triangleright (G_2 \triangleright (G_1 \triangleright (\ldots)))) \ast (G_2 \triangleright (G_1 \triangleright (G_2 \triangleright (\ldots))))$$

where $\triangleright$ and $\ast$ denote the so-called comb- and star products of rooted graphs, respectively. One can show that both decompositions are related to natural inductive definitions of the free product of rooted graphs and to the so-called $m$-free [14] and $m$-monotone hierarchies [17] of product states. Recall that the comb product $(G_1 \triangleright G_2, e)$ is obtained by attaching a replica of $G_2$ by its root to every vertex of $V_1$, whereas the star product $(G_1 \ast G_2, e)$ is obtained by gluing $G_1$ and $G_2$ together at their roots. The associated convolutions are the monotone and boolean convolutions, respectively. Details will be discussed in [1].

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