THE TRUNCATED TRACIAL MOMENT PROBLEM

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ABSTRACT. We present tracial analogs of the classical results of Curto and Fialkow on moment matrices. A sequence of real numbers indexed by words in non-commuting variables with values invariant under cyclic permutations of the indexes, is called a tracial sequence. We prove that such a sequence can be represented with tracial moments of matrices if its corresponding moment matrix is positive semidefinite and of finite rank. A truncated tracial sequence allows for such a representation if and only if one of its extensions admits a flat extension. Finally, we apply this theory via duality to investigate trace-positive polynomials in non-commuting variables.

1. INTRODUCTION

The moment problem is a classical question in analysis, well studied because of its importance and variety of applications. A simple example is the (univariate) Hamburger moment problem: when does a given sequence of real numbers represent the successive moments \( \int x^n \, d\mu(x) \) of a positive Borel measure \( \mu \) on \( \mathbb{R} \)? Equivalently, which linear functionals \( L \) on univariate real polynomials are integration with respect to some \( \mu \)? By Haviland’s theorem [Hav] this is the case if and only if \( L \) is nonnegative on all polynomials nonnegative on \( \mathbb{R} \). Thus Haviland’s theorem relates the moment problem to positive polynomials. It holds in several variables and also if we are interested in restricting the support of \( \mu \). For details we refer the reader to one of the many beautiful expositions of this classical branch of functional analysis, e.g. [Akh, KN, ST].

Since Schmüdgen’s celebrated solution of the moment problem on compact basic closed semialgebraic sets [Scm], the moment problem has played a prominent role in real algebra, exploiting this duality between positive polynomials and the moment problem, cf. [KM, PS, Put, PV]. The survey of Laurent [Lau2] gives a nice presentation of up-to-date results and applications; see also [Mar, PD] for more on positive polynomials.
Our main motivation are trace-positive polynomials in non-commuting variables. A polynomial is called trace-positive if all its matrix evaluations (of all sizes) have nonnegative trace. Trace-positive polynomials have been employed to investigate problems on operator algebras (Connes’ embedding conjecture [Con, KS1]) and mathematical physics (the Bessis-Moussa-Villani conjecture [BMV, KS2]), so a good understanding of this set is desired. By duality this leads us to consider the tracial moment problem introduced below. We mention that the free non-commutative moment problem has been studied and solved by McCullough [McC] and Helton [Hel]. Hadwin [Had] considered moments involving traces on von Neumann algebras.

This paper is organized as follows. The short Section 2 fixes notation and terminology involving non-commuting variables used in the sequel. Section 3 introduces tracial moment sequences, tracial moment matrices, the tracial moment problem, and their truncated counterparts. Our main results in this section relate the truncated tracial moment problem to flat extensions of tracial moment matrices and resemble the results of Curto and Fialkow [CF1, CF2] on the (classical) truncated moment problem. For example, we prove that a tracial sequence can be represented with tracial moments of matrices if its corresponding tracial moment matrix is positive semidefinite and of finite rank (Theorem 3.14). A truncated tracial sequence allows for such a representation if and only if one of its extensions admits a flat extension (Corollary 3.21). Finally, in Section 4 we explore the duality between the tracial moment problem and trace-positivity of polynomials. Throughout the paper several examples are given to illustrate the theory.

2. Basic notions

Let $\mathbb{R}\langle X \rangle$ denote the unital associative $\mathbb{R}$-algebra freely generated by $\underline{X} = (X_1, \ldots, X_n)$. The elements of $\mathbb{R}\langle X \rangle$ are polynomials in the non-commuting variables $X_1, \ldots, X_n$ with coefficients in $\mathbb{R}$. An element $w$ of the monoid $\langle X \rangle$, freely generated by $X$, is called a word. An element of the form $aw$, where $0 \neq a \in \mathbb{R}$ and $w \in \langle X \rangle$, is called a monomial and $a$ its coefficient.

We endow $\mathbb{R}\langle X \rangle$ with the involution $p \mapsto p^*$ fixing $\mathbb{R} \cup \{X\}$ pointwise. Hence for each word $w \in \langle X \rangle$, $w^*$ is its reverse. As an example, we have $(X_1X_2^2 - X_2X_1)^* = X_2^2X_1 - X_1X_2$.

For $f \in \mathbb{R}\langle X \rangle$ we will substitute symmetric matrices $A = (A_1, \ldots, A_n)$ of the same size for the variables $\underline{X}$ and obtain a matrix $f(A)$. Since $f(A)$ is not well-defined if the $A_i$ do not have the same size, we will assume this condition implicitly without further mention in the sequel.

Let $\text{Sym} \mathbb{R}\langle X \rangle$ denote the set of symmetric elements in $\mathbb{R}\langle X \rangle$, i.e.,

$$\text{Sym} \mathbb{R}\langle X \rangle = \{ f \in \mathbb{R}\langle X \rangle \mid f^* = f \}.$$  

Similarly, we use $\text{Sym} \mathbb{R}^{t \times t}$ to denote the set of all symmetric $t \times t$ matrices.

In this paper we will mostly consider the normalized trace $\text{Tr}$, i.e.,

$$\text{Tr}(A) = \frac{1}{t} \text{tr}(A) \quad \text{for } A \in \mathbb{R}^{t \times t}.$$
The invariance of the trace under cyclic permutations motivates the following definition of cyclic equivalence [KSI, p. 1817].

**Definition 2.1.** Two polynomials $f, g \in \mathbb{R} \langle X \rangle$ are cyclically equivalent if $f - g$ is a sum of commutators:

$$f - g = \sum_{i=1}^{k} (p_i q_i - q_i p_i)$$

for some $k \in \mathbb{N}$ and $p_i, q_i \in \mathbb{R} \langle X \rangle$.

**Remark 2.2.**

(a) Two words $v, w \in \langle X \rangle$ are cyclically equivalent if and only if $w$ is a cyclic permutation of $v$. Equivalently: there exist $u_1, u_2 \in \langle X \rangle$ such that $v = u_1 u_2$ and $w = u_2 u_1$.

(b) If $f \overset{cyc}{\sim} g$ then $\text{Tr}(f(A)) = \text{Tr}(g(A))$ for all tuples $A$ of symmetric matrices. Less obvious is the converse: if $\text{Tr}(f(A)) = \text{Tr}(g(A))$ for all $A$ and $f - g \in \text{Sym} \mathbb{R} \langle X \rangle$, then $f \overset{cyc}{\sim} g$ [KSI, Theorem 2.1].

(c) Although $f \overset{cyc}{\not\sim} f^*$ in general, we still have

$$\text{Tr}(f(A)) = \text{Tr}(f^*(A))$$

for all $f \in \mathbb{R} \langle X \rangle$ and all $A \in (\text{Sym} \mathbb{R}^{t \times t})^n$.

The length of the longest word in a polynomial $f \in \mathbb{R} \langle X \rangle$ is the degree of $f$ and is denoted by $\deg f$. We write $\mathbb{R} \langle X \rangle_{\leq k}$ for the set of all polynomials of degree $\leq k$.

### 3. The Truncated Tracial Moment Problem

In this section we define tracial (moment) sequences, tracial moment matrices, the tracial moment problem, and their truncated analogs. After a few motivating examples we proceed to show that the kernel of a tracial moment matrix has some real-radical-like properties (Proposition 3.8). We then prove that a tracial moment matrix of finite rank has a tracial moment representation, i.e., the tracial moment problem for the associated tracial sequence is solvable (Theorem 3.14). Finally, we give the solution of the truncated tracial moment problem: a truncated tracial sequence has a tracial representation if and only if one of its extensions has a tracial moment matrix that admits a flat extension (Corollary 3.21).

For an overview of the classical (commutative) moment problem in several variables we refer the reader to Akhiezer [Akh] (for the analytic theory) and to the survey of Laurent [Lau1] and references therein for a more algebraic approach. The standard references on the truncated moment problems are [CF1, CF2]. For the non-commutative moment problem with free (i.e., unconstrained) moments see [McC, Hel].

**Definition 3.1.** A sequence of real numbers $(y_w)$ indexed by words $w \in \langle X \rangle$ satisfying

$$y_w = y_u \text{ whenever } w \overset{cyc}{\sim} u,$$

$$y_w = y_{w^*} \text{ for all } w,$$

and $y_\emptyset = 1$, is called a (normalized) tracial sequence.
Example 3.2. Given $t \in \mathbb{N}$ and symmetric matrices $A_1, \ldots, A_n \in \text{Sym} \mathbb{R}^{t \times t}$, the sequence given by

$$y_w := \text{Tr}(w(A_1, \ldots, A_n)) = \frac{1}{t} \text{tr}(w(A_1, \ldots, A_n))$$

is a tracial sequence since by Remark 2.2 the traces of cyclically equivalent words coincide.

We are interested in the converse of this example (the \textit{tracial moment problem}): \textit{For which sequences $(y_w)$ do there exist $N \in \mathbb{N}$, $t \in \mathbb{N}$, $\lambda_i \in \mathbb{R} \geq 0$ with $\sum_i^N \lambda_i = 1$ and vectors $A^{(i)} = (A_1^{(i)}, \ldots, A_n^{(i)}) \in (\text{Sym} \mathbb{R}^{t \times t})^n$, such that

$$y_w = \sum_{i=1}^N \lambda_i \text{Tr}(w(A^{(i)})) \quad (3.3)$$

We then say that $(y_w)$ has a \textit{tracial moment representation} and call it a \textit{tracial moment sequence}.

The \textit{truncated tracial moment problem} is the study of (finite) tracial sequences $(y_w)_{\leq k}$ where $w$ is constrained by $\text{deg } w \leq k$ for some $k \in \mathbb{N}$, and properties (3.1) and (3.2) hold for these $w$. For instance, which sequences $(y_w)_{\leq k}$ have a tracial moment representation, i.e., when does there exist a representation of the values $y_w$ as in (3.3) for $\text{deg } w \leq k$? If this is the case, then the sequence $(y_w)_{\leq k}$ is called a \textit{truncated tracial moment sequence}.

Remark 3.3.

(a) To keep a perfect analogy with the classical moment problem, one would need to consider the existence of a positive Borel measure $\mu$ on $(\text{Sym} \mathbb{R}^{t \times t})^n$ (for some $t \in \mathbb{N}$) satisfying

$$y_w = \int w(A) \, d\mu(A). \quad (3.4)$$

As we shall mostly focus on the \textit{truncated} tracial moment problem in the sequel, the finitary representations (3.3) seem to be the proper setting. We look forward to studying the more general representations (3.4) in the future.

(b) Another natural extension of our tracial moment problem with respect to matrices would be to consider moments obtained by traces in finite \textit{von Neumann algebras} as done by Hadwin [Had]. However, our primary motivation were trace-positive polynomials defined via traces of matrices (see Definition 4.1), a theme we expand upon in Section 4. Understanding these is one of the approaches to Connes’ embedding conjecture [KS]. The notion dual to that of trace-positive polynomials is the tracial moment problem as defined above.

(c) The tracial moment problem is a natural extension of the classical quadrature problem dealing with representability via atomic positive measures in the commutative case. Taking $a^{(i)}$ consisting of $1 \times 1$ matrices
a_j^{(i)} \in \mathbb{R}$ for the $A^{(i)}$ in (3.3), we have

$$y_w = \sum_i \lambda_i w(a^{(i)}) = \int x^w d\mu(x),$$

where $x^w$ denotes the commutative collapse of $w \in \langle X \rangle$. The measure $\mu$ is the convex combination $\sum \lambda_i \delta_{a^{(i)}}$ of the atomic measures $\delta_{a^{(i)}}$.

The next example shows that there are (truncated) tracial moment sequences $(y_w)$ which cannot be written as $y_w = \text{Tr}(w(A))$.

**Example 3.4.** Let $X$ be a single free (non-commutative) variable. We take the index set $J = (1, X, X^2, X^3, X^4)$ and $y = (1, 1 - \sqrt{2}, 1, 1 - \sqrt{2}, 1)$. Then

$$y_w = \frac{\sqrt{2}}{2} w(-1) + (1 - \frac{\sqrt{2}}{2}) w(1),$$

i.e., $\lambda_1 = \frac{\sqrt{2}}{2}$, $\lambda_2 = 1 - \lambda_1$ and $A^{(1)} = -1$, $A^{(2)} = 1$. But there is no symmetric matrix $A \in \mathbb{R}^{t \times t}$ for any $t \in \mathbb{N}$ such that $y_w = \text{Tr}(w(A))$ for all $w \in J$. The proof is given in the appendix.

The (infinite) tracial moment matrix $M(y)$ of a tracial sequence $y = (y_w)$ is defined by

$$M(y) = (y_{u^*v})_{u,v}.$$

This matrix is symmetric due to the condition (3.2) in the definition of a tracial sequence. A necessary condition for $y$ to be a tracial moment sequence is positive semidefiniteness of $M(y)$ which in general is not sufficient.

The tracial moment matrix of order $k$ is the tracial moment matrix $M_k(y)$ indexed by words $u, v$ with $\deg u, \deg v \leq k$. If $y$ is a truncated tracial moment sequence, then $M_k(y)$ is positive semidefinite. Here is an easy example showing the converse is false:

**Example 3.5.** When dealing with two variables, we write $(X,Y)$ instead of $(X_1, X_2)$. Taking the index set

$$(1, X, Y, X^2, XY, Y^2, X^3, X^2Y, XY^2, Y^3, X^4, X^3Y, X^2Y^2, XYXY, XY^3, Y^4),$$

the truncated moment sequence

$$y = (1, 0, 0, 1, 1, 0, 0, 0, 4, 0, 2, 1, 0, 4)$$

yields the tracial moment matrix

$$M_2(y) = \begin{pmatrix} 1 & 0 & 0 & 1 & 1 & 1 & 1 & 0 \\ 0 & 1 & 1 & 0 & 0 & 0 & 0 & 1 \\ 0 & 1 & 1 & 0 & 0 & 0 & 0 & 1 \\ 1 & 1 & 1 & 0 & 0 & 0 & 0 & 1 \\ 1 & 0 & 0 & 0 & 0 & 0 & 0 & 1 \\ 1 & 0 & 0 & 0 & 0 & 0 & 0 & 1 \\ 1 & 0 & 0 & 0 & 0 & 0 & 0 & 1 \\ 1 & 0 & 0 & 0 & 0 & 0 & 0 & 1 \end{pmatrix},$$

with respect to the basis $(1, X, Y, X^2, XY, YX, Y^2)$. $M_2(y)$ is positive semidefinite but $y$ has no tracial representation. Again, we postpone the proof until the appendix.
For a given polynomial \( p = \sum_{w \in \mathcal{X}} p_w w \in \mathbb{R}\langle \mathcal{X} \rangle \) let \( \vec{p} \) be the (column) vector of coefficients \( p_w \) in a given fixed order. One can identify \( \mathbb{R}\langle \mathcal{X} \rangle_{\leq k} \) with \( \mathbb{R}^n \) for \( \eta = \eta(k) = \dim \mathbb{R}\langle \mathcal{X} \rangle_{\leq k} < \infty \) by sending each \( p \in \mathbb{R}\langle \mathcal{X} \rangle_{\leq k} \) to the vector \( \vec{p} \) of its entries with \( \deg w \leq k \). The tracial moment matrix \( M(y) \) induces the linear map

\[
\varphi_M : \mathbb{R}\langle \mathcal{X} \rangle \to \mathbb{R}^N, \quad p \mapsto M \vec{p}.
\]

The tracial moment matrices \( M_k(y) \), indexed by \( w \) with \( \deg w \leq k \), can be regarded as linear maps \( \varphi_{M_k} : \mathbb{R}^n \to \mathbb{R}^9 \), \( \vec{p} \mapsto M_k \vec{p} \).

**Lemma 3.6.** Let \( M = M(y) \) be a tracial moment matrix. Then the following holds:

1. \( p(y) := \sum_w p_w y_w = \mathbf{1}^* M \vec{p} \). In particular, \( \mathbf{1}^* M \vec{p} = \mathbf{1}^* M \vec{q} \) if \( p \sim q \);
2. \( \vec{p}^* M \vec{q} = \mathbf{1}^* M \vec{p} \).

**Proof.** Let \( p, q \in \mathbb{R}\langle \mathcal{X} \rangle \). For \( k := \max\{\deg p, \deg q\} \), we have

\[
\vec{p}^* M(y) \vec{q} = \mathbf{1}^* M_k(y) \vec{q}.
\]

Both statements now follow by direct calculation.

We can identify the kernel of a tracial moment matrix \( M \) with the subset of \( \mathbb{R}\langle \mathcal{X} \rangle \) given by

\[
I := \{ p \in \mathbb{R}\langle \mathcal{X} \rangle \mid M \vec{p} = 0 \}. \tag{3.6}
\]

**Proposition 3.7.** Let \( M \geq 0 \) be a tracial moment matrix. Then

\[
I = \{ p \in \mathbb{R}\langle \mathcal{X} \rangle \mid \langle M \vec{p}, \vec{p} \rangle = 0 \}. \tag{3.7}
\]

Further, \( I \) is a two-sided ideal of \( \mathbb{R}\langle \mathcal{X} \rangle \) invariant under the involution.

**Proof.** Let \( J := \{ p \in \mathbb{R}\langle \mathcal{X} \rangle \mid \langle M \vec{p}, \vec{p} \rangle = 0 \} \). The implication \( I \subseteq J \) is obvious. Let \( p \in J \) be given and \( k = \deg p \). Since \( M \) and thus \( M_k \) for each \( k \in \mathbb{N} \) is positive semidefinite, the square root \( \sqrt{M_k} \) of \( M_k \) exists. Then \( 0 = \langle M_k \vec{p}, \vec{p} \rangle = \langle \sqrt{M_k} \vec{p}, \sqrt{M_k} \vec{p} \rangle \) implies \( \sqrt{M_k} \vec{p} = 0 \). This leads to \( M_k \vec{p} = M \vec{p} = 0 \), thus \( p \in I \).

To prove that \( I \) is a two-sided ideal, it suffices to show that \( I \) is a right-ideal which is closed under \( \cdot \). To do this, consider the bilinear map

\[
\langle p, q \rangle_M := \langle M \vec{p}, \vec{q} \rangle
\]

on \( \mathbb{R}\langle \mathcal{X} \rangle \), which is a semi-scalar product. By Lemma 3.6 we get that

\[
\langle pq, pq \rangle_M = \langle (pq)^* p q \rangle(y) = \langle q q^* p p \rangle(y) = \langle pqq^*, p \rangle_M.
\]

Then by the Cauchy-Schwarz inequality it follows that for \( p \in I \), we have

\[
0 \leq \langle pq, pq \rangle_M^2 = \langle pq q^* p \rangle_M \leq \langle pq q^*, p \rangle_M \langle p, p \rangle_M = 0.
\]

Hence \( pq \in I \), i.e., \( I \) is a right-ideal.

Since \( p p^c \sim pp^c \), we obtain from Lemma 3.6 that

\[
\langle M \vec{p}, \vec{p} \rangle = \langle p, p \rangle_M = \langle p^* \rangle(y) = \langle pp^* \rangle(y) = \langle p^*, p^* \rangle_M = \langle M \vec{p}^*, \vec{p}^* \rangle.
\]

Thus if \( p \in I \) then also \( p^* \in I \).
In the commutative context, the kernel of $M$ is a real radical ideal if $M$ is positive semidefinite as observed by Scheiderer (cf. [Lau2, p. 2974]). The next proposition gives a description of the kernel of $M$ in the non-commutative setting, and could be helpful in defining a non-commutative real radical ideal.

**Proposition 3.8.** For the ideal $I$ in (3.6) we have

$$I = \{ f \in \mathbb{R} \langle X \rangle \mid (f^*)^k f \in I \text{ for some } k \in \mathbb{N} \}.$$ 

Further,

$$I = \{ f \in \mathbb{R} \langle X \rangle \mid (f^*)^{2k} + \sum g_i^* g_i \in I \text{ for some } k \in \mathbb{N}, g_i \in \mathbb{R} \langle X \rangle \}.$$ 

**Proof.** If $f \in I$ then also $f^* f \in I$ since $I$ is an ideal. If $f^* f \in I$ we have $Mf^* f = 0$ which implies by Lemma 3.6 that

$$0 = f^* Mf^* f = f^* M f = \langle Mf, f \rangle.$$ 

Thus $f \in I$. If $(f^*)^k f \in I$ then also $(f^*)^{k+1} f \in I$. So without loss of generality let $k$ be even. From $(f^*)^k f \in I$ we obtain

$$0 = f^* Mf^* f = (f^*)^{k/2} M (f^*)^{k/2},$$

implying $(f^*)^{k/2} f \in I$. This leads to $f \in I$ by induction.

To show the second statement let $(f^*)^{2k} + \sum g_i^* g_i \in I$. This leads to

$$(f^*)^k \overrightarrow{M(f^*)^k} + \sum g_i^* M g_i = 0.$$ 

Since $M(y) \geq 0$ we have $(f^*)^k M(f^*)^k \geq 0$ and $g_i^* M g_i \geq 0$. Thus $(f^*)^k \overrightarrow{M(f^*)^k} = 0$ (and $g_i^* M g_i = 0$) which implies $f \in I$ as above.  

In the commutative setting one uses the Riesz representation theorem for some set of continuous functions (vanishing at infinity or with compact support) to show the existence of a representing measure. We will use the Riesz representation theorem for positive linear functionals on a finite-dimensional Hilbert space.

**Definition 3.9.** Let $\mathcal{A}$ be an $\mathbb{R}$-algebra with involution. We call a linear map $L: \mathcal{A} \to \mathbb{R}$ a state if $L(1) = 1$, $L(a^* a) \geq 0$ and $L(a^*) = L(a)$ for all $a \in \mathcal{A}$. If all the commutators have value 0, i.e., if $L(ab) = L(ba)$ for all $a, b \in \mathcal{A}$, then $L$ is called a tracial state.

With the aid of the Artin-Wedderburn theorem we shall characterize tracial states on matrix $*$-algebras in Proposition 3.13. This will enable us to prove the existence of a tracial moment representation for tracial sequences with a finite rank tracial moment matrix; see Theorem 3.14.

**Remark 3.10.** The only central simple algebras over $\mathbb{R}$ are full matrix algebras over $\mathbb{R}$, $\mathbb{C}$ or $\mathbb{H}$ (combine the Frobenius theorem [Lam (13.12)] with the Artin-Wedderburn theorem [Lam (3.5)]). In order to understand ($\mathbb{R}$-linear) tracial states on these, we recall some basic Galois theory.
Let
\[ \text{Trd}_{\mathbb{C}/\mathbb{R}} : \mathbb{C} \to \mathbb{R}, \quad z \mapsto \frac{1}{2}(z + \bar{z}) \]
denote the field trace and
\[ \text{Trd}_{\mathbb{H}/\mathbb{R}} : \mathbb{H} \to \mathbb{R}, \quad z \mapsto \frac{1}{2}(z + \bar{z}) \]
the reduced trace [KMRT, p. 5]. Here the Hamilton quaternions \( \mathbb{H} \) are endowed with the standard involution
\[ z = a + ib + jc + kd \mapsto a - ib - jk - kd = \bar{z} \]
for \( a, b, c, d \in \mathbb{R} \). We extend the canonical involution on \( \mathbb{C} \) and \( \mathbb{H} \) to the conjugate transpose involution \( * \) on matrices over \( \mathbb{C} \) and \( \mathbb{H} \), respectively.

Composing the field trace and reduced trace, respectively, with the normalized trace yields an \( \mathbb{R} \)-linear map from \( \mathbb{C}^t \times t \) and \( \mathbb{H}^t \times t \), respectively, to \( \mathbb{R} \). We will denote it simply by \( \text{Tr} \). A word of caution: \( \text{Tr}(A) \) does not denote the (normalized) matricial trace over \( K \) if \( A \in K^t \times t \) and \( K \in \{ \mathbb{C}, \mathbb{H} \} \).

An alternative description of \( \text{Tr} \) is given by the following lemma:

**Lemma 3.11.** Let \( K \in \{ \mathbb{R}, \mathbb{C}, \mathbb{H} \} \). Then the only \( (\mathbb{R} \text{-linear}) \) tracial state on \( K^t \times t \) is \( \text{Tr} \).

**Proof.** An easy calculation shows that \( \text{Tr} \) is indeed a tracial state. Let \( L \) be a tracial state on \( \mathbb{R}^t \times t \). By the Riesz representation theorem there exists a positive semidefinite matrix \( B \) with \( \text{Tr}(B) = 1 \) such that
\[ L(A) = \text{Tr}(BA) \]
for all \( A \in \mathbb{R}^t \times t \).

Write \( B = (b_{ij})_{i,j=1}^t \). Let \( i \neq j \). Then \( A = \lambda E_{ij} \) has zero trace for every \( \lambda \in \mathbb{R} \) and is thus a sum of commutators. (Here \( E_{ij} \) denotes the \( t \times t \) matrix unit with a one in the \( (i,j) \)-position and zeros elsewhere.) Hence
\[ \lambda b_{ij} = L(A) = 0. \]
Since \( \lambda \in \mathbb{R} \) was arbitrary, \( b_{ij} = 0 \).

Now let \( A = \lambda (E_{ii} - E_{jj}) \). Clearly, \( \text{Tr}(A) = 0 \) and hence
\[ \lambda (b_{ii} - b_{jj}) = L(A) = 0. \]
As before, this gives \( b_{ii} = b_{jj} \). So \( B \) is scalar, and \( \text{Tr}(B) = 1 \). Hence it is the identity matrix. In particular, \( L = \text{Tr} \).

If \( L \) is a tracial state on \( \mathbb{C}^t \times t \), then \( L \) induces a tracial state on \( \mathbb{R}^t \times t \), so
\[ L_0 := L|_{\mathbb{R}^t \times t} = \text{Tr} \] by the above. Extend \( L_0 \) to
\[ L_1 : \mathbb{C}^t \times t \to \mathbb{R}, \quad A + iB \mapsto L_0(A) = \text{Tr}(A) \quad \text{for } A, B \in \mathbb{R}^t \times t. \]
\( L_1 \) is a tracial state on \( \mathbb{C}^t \times t \) as a straightforward computation shows. As \( \text{Tr}(A) = \text{Tr}(A + iB) \), all we need to show is that \( L_1 = L \).

Clearly, \( L_1 \) and \( L \) agree on the vector space spanned by all commutators in \( \mathbb{C}^t \times t \). This space is (over \( \mathbb{R} \)) of codimension 2. By construction, \( L_1(1) = L(1) = 1 \) and \( L_1(i) = 0 \). On the other hand,
\[ L(i) = L(i^*) = -L(i) \]
implying $L(i) = 0$. This shows $L = L_1 = \text{Tr}$.

The remaining case of tracial states over $\mathbb{H}$ is dealt with similarly and is left as an exercise for the reader. ■

**Remark 3.12.** Every complex number $z = a + ib$ can be represented as a $2 \times 2$ real matrix $z' = \begin{pmatrix} a & b \\ -b & a \end{pmatrix}$. This gives rise to an $\mathbb{R}$-linear $\ast$-map $\mathbb{C}^{t \times t} \to \mathbb{R}^{(2t) \times (2t)}$ that commutes with $\text{Tr}$. A similar property holds if quaternions $a + ib + jc + kd$ are represented by the $4 \times 4$ real matrix

$\begin{pmatrix} a & b & c & d \\ -b & a & -d & c \\ -c & d & a & -b \\ -d & -c & b & a \end{pmatrix}$.

**Proposition 3.13.** Let $A$ be a $\ast$-subalgebra of $\mathbb{R}^{t \times t}$ for some $t \in \mathbb{N}$ and $L : A \to \mathbb{R}$ a tracial state. Then there exist full matrix algebras $A^{(i)}$ over $\mathbb{R}$, $\mathbb{C}$ or $\mathbb{H}$, a $\ast$-isomorphism

$A \to \bigoplus_{i=1}^{N} A^{(i)},$  \hspace{1cm} (3.8)

and $\lambda_1, \ldots, \lambda_N \in \mathbb{R}_{\geq 0}$ with $\sum_i \lambda_i = 1$, such that for all $A \in A$,

$L(A) = \sum_i \lambda_i \text{Tr}(A^{(i)}).$

Here, $\bigoplus_i A^{(i)} = \begin{pmatrix} A^{(1)} & \cdots & A^{(N)} \\ \vdots & \ddots & \vdots \\ A^{(N)} & \cdots & A^{(1)} \end{pmatrix}$ denotes the image of $A$ under the isomorphism (3.8). The size of (the real representation of) $\bigoplus_i A^{(i)}$ is at most $t$.

**Proof.** Since $L$ is tracial, $L(U^*AU) = L(A)$ for all orthogonal $U \in \mathbb{R}^{t \times t}$. Hence we can apply orthogonal transformations to $A$ without changing the values of $L$. So $A$ can be transformed into block diagonal form as in (3.8) according to its invariant subspaces. That is, each of the blocks $A^{(i)}$ acts irreducibly on a subspace of $\mathbb{R}^t$ and is thus a central simple algebra (with involution) over $\mathbb{R}$. The involution on $A^{(i)}$ is induced by the conjugate transpose involution. (Equivalently, by the transpose on the real matrix representation in the complex of quaternion case.)

Now $L$ induces (after a possible normalization) a tracial state on the block $A^{(i)}$ and hence by Lemma 3.11 we have $L_i := L|_{A^{(i)}} = \lambda_i \text{Tr}$ for some $\lambda_i \in \mathbb{R}_{\geq 0}$. Then

$L(A) = L(\bigoplus_i A^{(i)}) = \sum_i L_i(A^{(i)}) = \sum_i \lambda_i \text{Tr}(A^{(i)})$

and $1 = L(1) = \sum_i \lambda_i$. ■

The following theorem is the tracial version of the representation theorem of Curto and Fialkow for moment matrices with finite rank [CF1].

**Theorem 3.14.** Let $y = (y_w)$ be a tracial sequence with positive semidefinite moment matrix $M(y)$ of finite rank $t$. Then $y$ is a tracial moment sequence,
i.e., there exist vectors \( A^{(i)} = (A_1^{(i)}, \ldots, A_n^{(i)}) \) of symmetric matrices \( A_j^{(i)} \) of size at most \( t \) and \( \lambda_i \in \mathbb{R}_{\geq 0} \) with \( \sum \lambda_i = 1 \) such that
\[
y_w = \sum \lambda_i \text{Tr}(w(A^{(i)})).
\]

Proof. Let \( M := M(y) \). We equip \( \mathbb{R} \langle X \rangle \) with the bilinear form given by
\[
\langle p, q \rangle_M := \langle M \overline{p}, \overline{q} \rangle = \overline{q}^* M \overline{p}.
\]

Let \( I = \{ p \in \mathbb{R} \langle X \rangle \mid \langle p, p \rangle_M = 0 \} \). Then by Proposition 3.13, \( I \) is an ideal of \( \mathbb{R} \langle X \rangle \). In particular, \( I = \ker \varphi_M \) for
\[
\varphi_M : \mathbb{R} \langle X \rangle \to \text{Ran } M, \quad p \mapsto M \overline{p}.
\]

Thus if we define \( E := \mathbb{R} \langle X \rangle / I \), the induced linear map
\[
\overline{\varphi}_M : E \to \text{Ran } M, \quad \overline{p} \mapsto M \overline{p}
\]
is an isomorphism and
\[
\dim E = \dim(\text{Ran } M) = \text{rank } M = t < \infty.
\]

Hence \( (E, (\cdot, \cdot)_E) \) is a finite-dimensional Hilbert space for \( (\overline{p}, \overline{q})_E = \overline{q}^* M \overline{p} \).

Let \( \hat{X}_i \) be the right multiplication with \( X_i \) on \( E \), i.e., \( \hat{X}_i \overline{p} := pX_i \). Since \( I \) is a right ideal of \( \mathbb{R} \langle X \rangle \), the operator \( \hat{X}_i \) is well defined. Further, \( \hat{X}_i \) is symmetric since
\[
\langle \hat{X}_i \overline{p}, \overline{q} \rangle_E = \langle M \overline{pX}_i, \overline{q} \rangle = \langle X_ip^* q \rangle(y)
\]
\[
= \langle p^* qX_i \rangle(y) = \langle M \overline{p}, \overline{qX}_i \rangle = \langle \overline{p}, \hat{X}_i \overline{q} \rangle_E.
\]

Thus each \( \hat{X}_i \), acting on a \( t \)-dimensional vector space, has a representation matrix \( A_i \in \text{Sym} \mathbb{R}^{t \times t} \).

Let \( \mathcal{B} = B(\hat{X}_1, \ldots, \hat{X}_n) = B(A_1, \ldots, A_n) \) be the algebra of operators generated by \( \hat{X}_1, \ldots, \hat{X}_n \). These operators can be written as
\[
\hat{p} = \sum_{w \in \langle X \rangle} p_w \hat{w}
\]
for some \( p_w \in \mathbb{R} \), where \( \hat{w} = \hat{X}_{w_1} \cdots \hat{X}_{w_s} \) for \( w = X_{w_1} \cdots X_{w_s} \). Observe that \( \hat{w} = w(A_1, \ldots, A_n) \). We define the linear functional
\[
L : \mathcal{B} \to \mathbb{R}, \quad \hat{p} \mapsto \overline{\overline{p}}^* M \overline{p} = p(y),
\]
which is a state on \( \mathcal{B} \). Since \( y_w = y_u \) for \( w \overset{\text{cyc}}{\sim} u \), it follows that \( L \) is tracial. Thus by Proposition 3.13 (and Remark 3.12), there exist \( \lambda_1, \ldots, \lambda_N \in \mathbb{R}_{\geq 0} \) with \( \sum \lambda_i = 1 \) and real symmetric matrices \( A_j^{(i)} (i = 1, \ldots, N) \) for each \( A_j \in \text{Sym} \mathbb{R}^{t \times t} \), such that for all \( w \in \langle X \rangle \),
\[
y_w = w(y) = L(\hat{w}) = \sum_{i} \lambda_i \text{Tr}(w(A^{(i)})),
\]
as desired.

The sufficient conditions on \( M(y) \) in Theorem 3.14 are also necessary for \( y \) to be a tracial moment sequence. Thus we get our first characterization of tracial moment sequences:
Corollary 3.15. Let \( y = (y_w) \) be a tracial sequence. Then \( y \) is a tracial moment sequence if and only if \( M(y) \) is positive semidefinite and of finite rank.

Proof. If \( y_w = \text{Tr}(w(A)) \) for some \( A = (A_1, \ldots, A_n) \in (\text{Sym} \mathbb{R}^{t \times t})^n \), then
\[
L(p) = \sum_w p_w y_w = \sum_w p_w \text{Tr}(w(A)) = \text{Tr}(p(A)).
\]
Hence
\[
\overline{p} \ast M(y) \overline{p} = L(p^* p) = \text{Tr}(p^*(A)p(A)) \geq 0.
\]
for all \( p \in \mathbb{R} \langle X \rangle \).

Further, the tracial moment matrix \( M(y) \) has rank at most \( t^2 \). This can be seen as follows: \( M \) induces a bilinear map
\[
\Phi: \mathbb{R} \langle X \rangle \to \mathbb{R} \langle X \rangle^*, \quad p \mapsto \left( q \mapsto \text{Tr}\left( (q^* p)(A) \right) \right);
\]
where \( \mathbb{R} \langle X \rangle^* \) is the dual space of \( \mathbb{R} \langle X \rangle \). This implies
\[
\text{rank } M = \dim(\text{Ran } \Phi) = \dim(\mathbb{R} \langle X \rangle / \ker \Phi).
\]
The kernel of the evaluation map \( \varepsilon_A: \mathbb{R} \langle X \rangle \to \mathbb{R}^{t \times t}, \quad p \mapsto p(A) \) is a subset of \( \ker \Phi \). In particular,
\[
\dim(\mathbb{R} \langle X \rangle / \ker \Phi) \leq \dim(\mathbb{R} \langle X \rangle / \ker \varepsilon_A) = \dim(\text{Ran } \varepsilon_A) \leq t^2.
\]
The same holds true for each convex combination \( y_w = \sum_i \lambda_i \text{Tr}(w(A^{(i)})) \).

The converse is Theorem 3.14.

Definition 3.16. Let \( A \in \text{Sym} \mathbb{R}^{t \times t} \) be given. A (symmetric) extension of \( A \) is a matrix \( \tilde{A} \in \text{Sym} \mathbb{R}^{(t+s) \times (t+s)} \) of the form
\[
\tilde{A} = \begin{pmatrix} A & B \\ B^* & C \end{pmatrix}
\]
for some \( B \in \mathbb{R}^{t \times s} \) and \( C \in \mathbb{R}^{s \times s} \). Such an extension is flat if \( \text{rank } A = \text{rank } \tilde{A} \), or, equivalently, if \( B = AW \) and \( C = W^* AW \) for some matrix \( W \).

The kernel of a flat extension \( M_k \) of a tracial moment matrix \( M_{k-1} \) has some (truncated) ideal-like properties as shown in the following lemma.

Lemma 3.17. Let \( f \in \mathbb{R} \langle X \rangle \) with \( \deg f \leq k - 1 \) and let \( M_k \) be a flat extension of \( M_{k-1} \). If \( f \in \ker M_k \), then \( f X_i, X_i f \in \ker M_k \).

Proof. Let \( f = \sum_w f_w w \). Then for \( v \in \langle X \rangle_{k-1} \), we have
\[
(M_k f X_i)_v = \sum_w f_w y_v^* w X_i = \sum_w f_w y_{(v X_i)}^* w = (M_k \tilde{f})_{v X_i} = 0. \tag{3.9}
\]

The matrix \( M_k \) is of the form \( M_k = \begin{pmatrix} M_{k-1} & B \\ B^* & C \end{pmatrix} \). Since \( M_k \) is a flat extension, \( \ker M_k = \ker (M_{k-1} \ B) \). Thus by (3.9), \( f X_i \in \ker (M_{k-1} \ B) = \ker M_k \). For \( X_i f \) we obtain analogously that
\[
(M_k \tilde{f} X_i)_v = \sum_w f_w y_v^* X_i w = \sum_w f_w y_{(X_i v)}^* w = (M_k \tilde{f})_{X_i v} = 0
\]
for \( v \in \langle X \rangle_{k-1} \), which implies \( X_i f \in \ker M_k \).
We are now ready to prove the tracial version of the flat extension theorem of Curto and Fialkow \cite{CF2}.

**Theorem 3.18.** Let $y = (y_w)_{w \leq 2k}$ be a truncated tracial sequence of order $2k$. If $\text{rank } M_k(y) = \text{rank } M_{k-1}(y)$, then there exists a unique tracial extension $\tilde{y} = (\tilde{y}_w)_{w \leq 2k+2}$ of $y$ such that $M_{k+1}(\tilde{y})$ is a flat extension of $M_k(y)$.

**Proof.** Let $M_k := M_k(y)$. We will construct a flat extension $M_{k+1} := \left( \begin{array}{c} M_k \\ B^* \\ C \end{array} \right)$ such that $M_{k+1}$ is a tracial moment matrix. Since $M_k$ is a flat extension of $M_{k-1}(y)$ we can find a basis $b$ of $\text{Ran } M_k$ consisting of columns of $M_k$ labeled by $w$ with $\deg w \leq k - 1$. Thus the range of $M_k$ is completely determined by the range of $M_k|\text{span } b$, i.e., for each $p \in \mathbb{R} \langle X \rangle$ with $\deg p \leq k$ there exists a unique $r \in \text{span } b$ such that $M_k \overline{r} = M_k \overline{p}$; equivalently, $p - r \in \ker M_k$.

Let $v \in \langle X \rangle$, $\deg v = k + 1$, $v = v'X_i$ for some $i \in \{1, \ldots, n\}$ and $v' \in \langle X \rangle$ with $\deg v' = k$. For $v'$ there exists an $r \in \text{span } b$ such that $v' - r \in \ker M_k$.

If there exists a flat extension $M_{k+1}$, then by Lemma \ref{Lem.C17} from $v' - r \in \ker M_k \subseteq \ker M_{k+1}$ it follows that $(v' - r)X_i \in \ker M_{k+1}$. Hence the desired flat extension has to satisfy

$$M_{k+1} \overline{v'} = M_{k+1} \overline{rX_i} = M_{k+1} \overline{wX_i}.$$  \hfill (3.10)

Therefore we define

$$B \overline{v'} := M_{k+1} \overline{rX_i}.$$  \hfill (3.11)

More precisely, let $(w_1, \ldots, w_k)$ be the basis of $M_k$, i.e., $(M_k)_{i,j} = w_i^*w_j$. Let $r_{w_i}$ be the unique element in $\text{span } b$ with $w_i - r_{w_i} \in \ker M_k$. Then $B = M_k W$ with $W = (r_{w_1X_1}, \ldots, r_{w_kX_k})$ and we define

$$C := W^*M_kW.$$  \hfill (3.12)

Since the $r_{w_i}$ are uniquely determined,

$$M_{k+1} = \left( \begin{array}{c} M_k \\ B^* \\ C \end{array} \right)$$  \hfill (3.13)

is well-defined. The constructed $M_{k+1}$ is a flat extension of $M_k$, and $M_{k+1} \succeq 0$ if and only if $M_k \succeq 0$, cf. \cite{CF2} Proposition 2.1. Moreover, once $B$ is chosen, there is only one $C$ making $M_{k+1}$ as in (3.13) a flat extension of $M_k$. This follows from general linear algebra, see e.g. \cite{CF2} p. 11. Hence $M_{k+1}$ is the only candidate for a flat extension.

Therefore we are done if $M_{k+1}$ is a tracial moment matrix, i.e.,

$$(M_{k+1})_w = (M_{k+1})_v \text{ whenever } w \preceq v.$$  \hfill (3.14)

To show this we prove that $(M_{k+1})_{X_iw} = (M_{k+1})_{wX_i}$. Then (3.14) follows recursively.

Let $w = u^*v$. If $\deg u, \deg vX_i \leq k$ there is nothing to show since $M_k$ is a tracial moment matrix. If $\deg u \leq k$ and $\deg vX_i = k + 1$ there exists an $r \in \text{span } b$ such that $r - v \in \ker M_{k-1}$, and by Lemma \ref{Lem.C17} also $vX_i - rX_i \in \ker M_{k-1}$.
ker $M_k$. Then we get
\[
(M_{k+1})_{u^*v}X_i = \overline{u^*M_{k+1}vX_i} = \overline{u^*M_{k+1}rX_i} = \overline{u^*MrX_i}
\]
\[
= (M_k)_{u^*rX_i} = (M_k)_{X_iu^*r} = (M_k)_{(uX_i)^*r}
\]
\[
\overset{*}{= \overline{uX_i}^*M_{k+1}v} = (M_k)_{(uX_i)^*v} = (M_{k+1})_{X_iw},
\]
where equality $(* \overset{\text{no}}{=} )$ holds by (3.10) which implies Lemma 3.17 by construction.

If deg $u = \text{deg } vX_i = k + 1$, write $u = X_ju'$. Further, there exist $s, r \in \text{span } b$ with $u' - s \in \ker M_{k-1}$ and $r - v \in \ker M_{k-1}$. Then
\[
(M_{k+1})_{u^*v}X_i = \overline{X_ju'X_j}M_{k+1}vX_i = \overline{X_ju'X_j}M_{k+1}vX_i
\]
\[
= (M_k)s^*X_jrX_i = (M_k)(sX_i)^*(X_jr)
\]
\[
\overset{* \overset{\text{no}}{=} }{= \overline{uX_i}^*M_{k+1}X_jv} = (M_k)(uX_i)^*X_jv = (M_{k+1})_{X_iw}.
\]
Finally, the construction of $\tilde{y}$ from $M_{k+1}$ is clear. ■

**Corollary 3.19.** Let $y = (y_w)_{w \leq 2k}$ be a truncated tracial sequence. If $M_k(y)$ is positive semidefinite and $M_k(y)$ is a flat extension of $M_{k-1}(y)$, then $y$ is a truncated tracial moment sequence.

**Proof.** By Theorem 3.18 we can extend $M_k(y)$ inductively to a positive semi-definite moment matrix $M(\tilde{y})$ with rank $M(\tilde{y}) = \text{rank } M_k(y) < \infty$. Thus $M(\tilde{y})$ has finite rank and by Theorem 3.14 there exists a tracial moment representation of $\tilde{y}$. Therefore $y$ is a truncated tracial moment sequence. ■

The following two corollaries give characterizations of tracial moment matrices coming from tracial moment sequences.

**Corollary 3.20.** Let $y = (y_w)$ be a tracial sequence. Then $y$ is a tracial moment sequence if and only if $M(y)$ is positive semidefinite and there exists some $N \in \mathbb{N}$ such that $M_{k+1}(y)$ is a flat extension of $M_k(y)$ for all $k \geq N$.

**Proof.** If $y$ is a tracial moment sequence then by Corollary 3.15, $M(y)$ is positive semidefinite and has finite rank $t$. Thus there exists an $N \in \mathbb{N}$ such that $t = \text{rank } M_N(y)$. In particular, $\text{rank } M_k(y) = \text{rank } M_{k+1}(y) = t$ for all $k \geq N$, i.e., $M_{k+1}(y)$ is a flat extension of $M_k(y)$ for all $k \geq N$.

For the converse, let $N$ be given such that $M_{k+1}(y)$ is a flat extension of $M_k(y)$ for all $k \geq N$. By Theorem 3.18 the (iterated) unique extension $\tilde{y}$ of $(y_w)_{w \leq 2k}$ for $k \geq N$ is equal to $y$. Otherwise there exists a flat extension $\tilde{y}$ of $(y_w)_{w \leq 2\ell}$ for some $\ell \geq N$ such that $M_{\ell+1}(\tilde{y}) \geq 0$ is a flat extension of $M(\tilde{y})$ and $M_{\ell+1}(\tilde{y}) \neq M_{\ell+1}(y)$ contradicting the uniqueness of the extension in Theorem 3.18.

Thus $M(y) \succeq 0$ and rank $M(y) = \text{rank } M_N(y) < \infty$. Hence by Theorem 3.14 $y$ is a tracial moment sequence. ■

**Corollary 3.21.** Let $y = (y_w)$ be a tracial sequence. Then $y$ has a tracial moment representation with matrices of size at most $t := \text{rank } M(y)$ if $M_N(y)$ is positive semidefinite and $M_N(y)$ is a flat extension of $M_N(y)$ for some $N \in \mathbb{N}$ with rank $M_N(y) = t$. 

Proof. Since rank $M(y) = \text{rank } M_N(y) = t$, each $M_{k+1}(y)$ with $k \geq N$ is a flat extension of $M_k(y)$. As $M_N(y) \succeq 0$, all $M_k(y)$ are positive semidefinite. Thus $M(y)$ is also positive semidefinite. Indeed, let $p \in \mathbb{R}\langle X \rangle$ and $\ell = \max\{\deg p, N\}$. Then $\overline{p}^* M(y) \overline{p} = \overline{p}^* M_\ell(y) \overline{p} \succeq 0$.

Thus by Corollary 3.20, $y$ is a tracial moment sequence. The representing matrices can be chosen to be of size at most rank $M(y) = t$. 

4. Positive definite moment matrices and trace-positive polynomials

In this section we explain how the representability of positive definite tracial moment matrices relates to sum of hermitian squares representations of trace-positive polynomials. We start by introducing some terminology.

An element of the form $g^* g$ for some $g \in \mathbb{R}\langle X \rangle$ is called a hermitian square and we denote the set of all sums of hermitian squares by

$$\Sigma^2 = \{ f \in \mathbb{R}\langle X \rangle \mid f = \sum g_i^* g_i \text{ for some } g_i \in \mathbb{R}\langle X \rangle \}.$$ 

A polynomial $f \in \mathbb{R}\langle X \rangle$ is matrix-positive if $f(A)$ is positive semidefinite for all tuples $A_i \in \text{Sym } \mathbb{R}^{t \times t}$, $t \in \mathbb{N}$. Helton [Hel] proved that $f \in \Sigma^2$ by solving a non-commutative moment problem; see also [McC].

We are interested in a different type of positivity induced by the trace. 

Definition 4.1. A polynomial $f \in \mathbb{R}\langle X \rangle$ is called trace-positive if

$$\text{Tr}(f(A)) \geq 0 \text{ for all } A \in (\text{Sym } \mathbb{R}^{t \times t})^n, \ t \in \mathbb{N}.$$ 

Trace-positive polynomials are intimately connected to deep open problems from e.g. operator algebras (Connes’ embedding conjecture [KS1]) and mathematical physics (the Bessis-Moussa-Villani conjecture [KS2]), so a good understanding of this set is needed. A distinguished subset is formed by sums of hermitian squares and commutators.

Definition 4.2. Let $\Theta^2$ be the set of all polynomials which are cyclically equivalent to a sum of hermitian squares, i.e.,

$$\Theta^2 = \{ f \in \mathbb{R}\langle X \rangle \mid f \overset{\text{cyc}}{=} \sum g_i^* g_i \text{ for some } g_i \in \mathbb{R}\langle X \rangle \}.$$ 

(4.1)

Obviously, all $f \in \Theta^2$ are trace-positive. However, in contrast to Helton’s sum of squares theorem mentioned above, the following non-commutative version of the well-known Motzkin polynomial [Mar, p. 5] shows that a trace-positive polynomial need not be a member of $\Theta^2$ [KS1].

Example 4.3. Let $MNC = XY^4X + YX^4Y - 3XY^2X + 1 \in \mathbb{R}\langle X,Y \rangle$. Then $MNC \notin \Theta^2$ since the commutative Motzkin polynomial is not a (commutative) sum of squares [Mar, p. 5]. The fact that $MNC(A, B)$ has nonnegative trace for all symmetric matrices $A, B$ has been shown by Schweighofer and the second author [KS1, Example 4.4] using Putinar’s Positivstellensatz [Put].
Let $\Sigma^2_k := \Sigma^2 \cap \mathbb{R}(\mathcal{X})_{\leq 2k}$ and $\Theta^2_k := \Theta^2 \cap \mathbb{R}(\mathcal{X})_{\leq 2k}$. These are convex cones in $\mathbb{R}(\mathcal{X})_{\leq 2k}$. By duality there exists a connection between $\Theta^2_k$ and positive semidefinite tracial moment matrices of order $k$. If every tracial moment matrix $M_k(y) \succeq 0$ of order $k$ has a tracial representation then every trace-positive polynomial of degree at most $2k$ lies in $\Theta^2_k$. In fact:

**Theorem 4.4.** The following statements are equivalent:

(i) all truncated tracial sequences $(y_w)_{\leq 2k}$ with positive definite tracial moment matrix $M_k(y)$ have a tracial moment representation (3.3);

(ii) all trace-positive polynomials of degree $\leq 2k$ are elements of $\Theta^2_k$.

For the proof we need some preliminary work.

**Lemma 4.5.** $\Theta^2_k$ is a closed convex cone in $\mathbb{R}(\mathcal{X})_{\leq 2k}$.

**Proof.** Endow $\mathbb{R}(\mathcal{X})_{\leq 2k}$ with a norm $\|\cdot\|$ and the quotient space $\mathbb{R}(\mathcal{X})_{\leq 2k}/\Sigma^2$ with the quotient norm

$$\|\pi(f)\| := \inf \left\{ \|f + h\| \mid h \overset{\Sigma^2}{=} 0 \right\}, \quad f \in \mathbb{R}(\mathcal{X})_{\leq 2k}. \quad (4.2)$$

Here $\pi : \mathbb{R}(\mathcal{X})_{\leq 2k} \to \mathbb{R}(\mathcal{X})_{\leq 2k}/\Sigma^2$ denotes the quotient map. (Note: due to the finite-dimensionality of $\mathbb{R}(\mathcal{X})_{\leq 2k}$, the infimum on the right-hand side of (4.2) is attained.)

Since $\Theta^2_k = \pi^{-1}(\pi(\Theta^2_k))$, it suffices to show that $\pi(\Theta^2_k)$ is closed. Let $d_k = \text{dim} \mathbb{R}(\mathcal{X})_{\leq 2k}$. Since by Carathéodory’s theorem [Bar, p. 10] each element $f \in \mathbb{R}(\mathcal{X})_{\leq 2k}$ can be written as a convex combination of $d_k + 1$ elements of $\mathbb{R}(\mathcal{X})_{\leq 2k}$, the image of

$$\varphi : (\mathbb{R}(\mathcal{X})_{\leq k})^{d_k} \to \mathbb{R}(\mathcal{X})_{2k}/\Sigma^2$$

$$(g_i)_{i=0,\ldots,d_k} \mapsto \pi\left( \sum_{i=0}^{d_k} g_i^* g_i \right)$$

equals $\pi(\Sigma^2_k) = \pi(\Theta^2_k)$. In $(\mathbb{R}(\mathcal{X})_{\leq k})^{d_k}$ we define $\mathcal{S} := \{g = (g_i) \mid \|g\| = 1\}$. Note that $\mathcal{S}$ is compact, thus $V := \varphi(\mathcal{S}) \subseteq \pi(\Theta^2_k)$ is compact as well. Since $0 \notin V$, and a sum of hermitian squares cannot be cyclically equivalent to $0$ by [KS2] Lemma 3.2 (b)], we see that $0 \notin V$.

Let $(f_\ell)_{\ell}$ be a sequence in $\pi(\Theta^2_k)$ which converges to $\pi(f)$ for some $f \in \mathbb{R}(\mathcal{X})_{\leq 2k}$. Write $f_\ell = \lambda_\ell v_\ell$ for $\lambda_\ell \in \mathbb{R}_{\geq 0}$ and $v_\ell \in V$. Since $V$ is compact there exists a subsequence $(v_{\ell_j})_j$ of $v_\ell$ converging to $v \in V$. Then

$$\lambda_{\ell_j} = \frac{\|f_{\ell_j}\|}{\|v_{\ell_j}\|} \to \frac{\|f\|}{\|v\|}.$$ 

Thus $f_\ell \to f = \frac{\|f\|}{\|v\|} v \in \pi(\Theta^2_k)$.

**Definition 4.6.** To a truncated tracial sequence $(y_w)_{\leq k}$ we associate the (tracial) Riesz functional $L_y : \mathbb{R}(\mathcal{X})_{\leq k} \to \mathbb{R}$ defined by

$$L_y(p) := \sum_w p_w y_w \quad \text{for} \quad p = \sum_w p_w w \in \mathbb{R}(\mathcal{X})_{\leq k}.$$
We say that \( L_y \) is strictly positive (\( L_y > 0 \)), if
\[
L_y(p) > 0 \quad \text{for all trace-positive} \quad p \in \mathbb{R}\langle X \rangle_{\leq k}, \ p^\circ \neq 0.
\]
If \( L_y(p) \geq 0 \) for all trace-positive \( p \in \mathbb{R}\langle X \rangle_{\leq k} \), then \( L_y \) is positive (\( L_y \geq 0 \)).

Equivalently, a tracial Riesz functional \( L_y \) is positive (resp., strictly positive) if and only if the map \( \bar{L}_y \) it induces on \( \mathbb{R}\langle X \rangle_{\leq 2k}/\mathcal{C}_\mathbb{C} \) is nonnegative (resp., positive) on the nonzero images of trace-positive polynomials in \( \mathbb{R}\langle X \rangle_{\leq 2k}/\mathcal{C}_\mathbb{C} \).

We shall prove that strictly positive Riesz functionals lie in the interior of the cone of positive Riesz functionals, and that truncated tracial sequences \( y \) with strictly positive \( L_y \) are truncated tracial moment sequences (Theorem 4.8 below). These results are motivated by and resemble the results of Fialkow and Nie [FN, Section 2] in the commutative context.

**Lemma 4.7.** If \( L_y > 0 \) then there exists an \( \varepsilon > 0 \) such that \( L_{\tilde{y}} > 0 \) for all \( \tilde{y} \) with \( \|y - \tilde{y}\|_1 < \varepsilon \).

**Proof.** We equip \( \mathbb{R}\langle X \rangle_{\leq 2k}/\mathcal{C}_\mathbb{C} \) with a quotient norm as in (1.2). Then
\[
\mathcal{S} := \{ \pi(p) \in \mathbb{R}\langle X \rangle_{\leq 2k}/\mathcal{C}_\mathbb{C} \mid p \in \mathcal{C}_k, \|\pi(p)\| = 1 \}
\]
is compact. By a scaling argument, it suffices to show that \( \bar{L}_y > 0 \) on \( \mathcal{S} \) for \( \tilde{y} \) close to \( y \). The map \( y \mapsto \bar{L}_y \) is linear between finite-dimensional vector spaces. Thus
\[
|\bar{L}_y'(\pi(p)) - \bar{L}_{y''}(\pi(p))| \leq C\|y' - y''\|_1
\]
for all \( \pi(p) \in \mathcal{S} \), truncated tracial moment sequences \( y', y'' \), and some \( C \in \mathbb{R}_{>0} \).

Since \( \bar{L}_y \) is continuous and strictly positive on \( \mathcal{S} \), there exists an \( \varepsilon > 0 \) such that \( \bar{L}_y(\pi(p)) \geq 2\varepsilon \) for all \( \pi(p) \in \mathcal{S} \). Let \( \tilde{y} \) satisfy \( \|y - \tilde{y}\|_1 < \frac{\varepsilon}{2} \). Then
\[
\bar{L}_y(\pi(p)) \geq 2\varepsilon \quad \Rightarrow \quad \bar{L}_y(\pi(p)) - C\|y - \tilde{y}\|_1 \geq \varepsilon > 0.
\]

**Theorem 4.8.** Let \( y = (y_w)_{\leq k} \) be a truncated tracial sequence of order \( k \).
If \( L_y > 0 \), then \( y \) is a truncated tracial moment sequence.

**Proof.** We show first that \( y \in \overline{T} \), where \( \overline{T} \) is the closure of
\[
T = \{(y_w)_{\leq k} \mid \exists \exists \exists \lambda_i \in \mathbb{R}_{\geq 0} : y_w = \sum \lambda_i \operatorname{Tr}(A^{(i)})\}.
\]
Assume \( L_y > 0 \) but \( y \notin \overline{T} \). Since \( \overline{T} \) is a closed convex cone in \( \mathbb{R}^\eta \) (for some \( \eta \in \mathbb{N} \)), by the Minkowski separation theorem there exists a vector \( \overline{p} \in \mathbb{R}^\eta \) such that \( \overline{p}^* y < 0 \) and \( \overline{p}^* w \geq 0 \) for all \( w \in T \). The non-commutative polynomial corresponding to \( \overline{p} \) is trace positive since \( \overline{p}^* z \geq 0 \) for all \( z \in T \). Thus \( 0 < L_y(p) = \overline{p}^* y < 0 \), a contradiction.

By Lemma 4.7 \( y \in \text{int}(\overline{T}) \). Thus \( y \in \text{int}(\overline{T}) \subseteq T \) [Ber, Theorem 25.20].

We remark that assuming only non-strict positivity of \( L_y \) in Theorem 4.8 would not suffice for the existence of a tracial moment representation (3.3) for \( y \). This is a consequence of Example 3.5.
Proof (of Theorem 4.4). To show (i) \(\Rightarrow\) (ii), assume \(f = \sum_w f_w w \in \mathbb{R}\langle X\rangle_{\leq 2k}\) is trace-positive but \(f \notin \Theta^2_k\). By Lemma 4.5, \(\Theta^2_k\) is a closed convex cone in \(\mathbb{R}\langle X\rangle_{\leq 2k}\), thus by the Minkowski separation theorem we find a hyperplane which separates \(f\) and \(\Theta^2_k\). That is, there is a linear form \(L : \mathbb{R}\langle X\rangle_{\leq 2k} \to \mathbb{R}\) such that \(L(f) < 0\) and \(L(p) \geq 0\) for \(p \in \Theta^2_k\). In particular, \(L(f) = 0\) for all \(f \in \mathbb{R}\langle X\rangle_{\leq 2k}\), i.e., without loss of generality, \(L\) is tracial. Since there are tracial states strictly positive on \(\Sigma^2_k \setminus \{0\}\), we may assume \(L(p) > 0\) for all \(p \in \Theta^2_k\), \(p \in \mathbb{R}\langle X\rangle_{\leq 2k}\). Hence the bilinear form given by

\[(p, q) \mapsto L(pq)\]

can be written as \(L(pq) = \varphi^* M \varphi\) for some truncated tracial moment matrix \(M > 0\). By assumption, the corresponding truncated tracial sequence \(y\) has a tracial moment representation

\[y_w = \sum \lambda_i \text{Tr}(w(A^{(i)}))\]

for some tuples \(A^{(i)}\) of symmetric matrices \(A^{(i)}_j\) and \(\lambda_i \in \mathbb{R}_{\geq 0}\) which implies the contradiction

\[0 > L(f) = \sum \lambda_i \text{Tr}(f(A^{(i)})) \geq 0.\]

Conversely, if (ii) holds, then \(L_g > 0\) if and only if \(M(y) > 0\). Thus a positive definite moment matrix \(M(y)\) defines a strictly positive functional \(L_g\) which by Theorem 4.8 has a tracial representation. \(\blacksquare\)

As mentioned above, the Motzkin polynomial \(M_{nc}\) is trace-positive but \(M_{nc} \notin \Theta^2_k\). Thus by Theorem 4.4 there exists at least one truncated tracial moment matrix which is positive definite but has no tracial representation.

**Example 4.9.** Taking the index set

\[(1, X, Y, X^2, XY, YX, Y^2, X^2Y, YX^2, Y^2X, X^3, Y^3, XYX, YXY),\]

the matrix

\[
M_3(y) :=
\begin{pmatrix}
1 & 0 & 0 & \frac{7}{5} & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0
\\
0 & \frac{7}{5} & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0
\\
0 & 0 & \frac{7}{5} & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0
\\
0 & 0 & 0 & \frac{24}{5} & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0
\\
0 & 0 & 0 & 0 & \frac{10}{3} & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0
\\
0 & 0 & 0 & 0 & 0 & \frac{10}{3} & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0
\\
0 & 0 & 0 & 0 & 0 & 0 & \frac{10}{3} & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0
\\
0 & 0 & 0 & 0 & 0 & 0 & 0 & \frac{24}{5} & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0
\\
0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & \frac{24}{5} & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0
\\
0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & \frac{24}{5} & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0
\\
0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & \frac{24}{5} & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0
\\
0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & \frac{24}{5} & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0
\\
0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & \frac{24}{5} & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0
\\
0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & \frac{24}{5} & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0
\\
0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & \frac{24}{5} & 0 & 0 & 0 & 0 & 0 & 0 & 0
\\
0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & \frac{24}{5} & 0 & 0 & 0 & 0 & 0 & 0
\\
0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & \frac{24}{5} & 0 & 0 & 0 & 0 & 0
\\
0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & \frac{24}{5} & 0 & 0 & 0 & 0
\\
0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & \frac{24}{5} & 0 & 0 & 0
\\
0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & \frac{24}{5} & 0 & 0
\\
0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & \frac{24}{5}
\end{pmatrix}
\]
is a tracial moment matrix of degree 3 in 2 variables and is positive definite. But
\[ L_y(M_{nc}) = M_{nc}(y) = -\frac{5}{16} < 0. \]
Thus \( y \) is not a truncated tracial moment sequence, since otherwise \( L_y(p) \geq 0 \) for all trace-positive polynomials \( p \in \mathbb{R}\langle X, Y \rangle_{\leq 6} \).

On the other hand, the (free) non-commutative moment problem is always solvable for positive definite moment matrices \([McC\text{, Theorem 2.1}]\). In our example this means there are symmetric matrices \( A, B \in \mathbb{R}^{15 \times 15} \) and a vector \( v \in \mathbb{R}^{15} \) such that
\[ y_w = \langle w(A) v, v \rangle \]
for all \( w \in \langle X, Y \rangle_{\leq 3} \).

**Remark 4.10.** A trace-positive polynomial \( f \in \mathbb{R}\langle X \rangle \) of degree \( 2k \) lies in \( \Theta_k^2 \) if and only if \( L_y(f) \geq 0 \) for all truncated tracial sequences \( (y_w)_{\leq 2k} \) with \( M_k(y) \succeq 0 \). This condition is obviously satisfied if all truncated tracial sequences \( (y_w)_{\leq 2k} \) with \( M_k(y) \succeq 0 \) have a tracial representation.

Using this we can prove that trace-positive binary quartics, i.e., homogeneous polynomials of degree 4 in \( \mathbb{R}\langle X, Y \rangle \), lie in \( \Theta_2^2 \). Equivalently, truncated tracial sequences \( (y_w) \) indexed by words of degree 4 with a positive definite tracial moment matrix have a tracial moment representation.

Furthermore, trace-positive binary biquadratic polynomials, i.e., polynomials \( f \in \mathbb{R}\langle X, Y \rangle \) with \( \deg_X f, \deg_Y f \leq 2 \), are cyclically equivalent to a sum of hermitian squares. Example 3.5 then shows that a polynomial \( f \) can satisfy \( L_y(f) \geq 0 \) although there are truncated tracial sequences \( (y_w)_{\leq 2k} \) with \( M_k(y) \succeq 0 \) and no tracial representation.

Studying extremal points of the convex cone
\[ \{ (y_w)_{\leq 2k} \mid M_k(y) \succeq 0 \} \]
of truncated tracial sequences with positive semidefinite tracial moment matrices, we are able to impose a concrete block structure on the matrices needed in a tracial moment representation.

These statements and concrete sum of hermitian squares and commutators representations of trace-positive polynomials of low degree will be published elsewhere \([Bur]\).

**Appendix A. Proofs of the claims made in Examples 3.4 and 3.5**

**Example 3.4 revisited.** We take the index set \( J = (1, X, X^2, X^3, X^4) \) and \( y = (1, 1 - \sqrt{2}, 1, 1 - \sqrt{2}, 1) \). Then there is no symmetric matrix \( A \in \mathbb{R}^{t \times t} \) for any \( t \in \mathbb{N} \) such that
\[ y_w = \text{Tr}(w(A)) \quad \text{for all } w \in J. \quad \text{(A.1)} \]
Without loss of generality we can choose \( A \) to be diagonal with diagonal elements \( a_1, \ldots, a_t \). Then \( y_w = \text{Tr}(w(A)) \) if and only if the following
equations hold:

\[
\sum_{i=1}^{t} a_i = \sum_{i=1}^{t} a_i^3 = (1 - \sqrt{2})t, \tag{A.2}
\]

\[
\sum_{i=1}^{t} a_i^2 = \sum_{i=1}^{t} a_i^4 = t. \tag{A.3}
\]

In the general means inequality

\[
\frac{\sum_{i=1}^{t} x_i}{t} \geq \sqrt{\frac{\sum_{i=1}^{t} x_i^2}{t}}
\]

for the arithmetic and the quadratic mean of \( x = (x_1, \ldots, x_t) \in \mathbb{R}_{\geq 0}^t \), equality holds if and only if all the \( x_i \) are the same. Hence (A.3) rewritten as

\[
\frac{\sum_{i=1}^{t} a_i^2}{t} = 1 = \sqrt{\frac{\sum_{i=1}^{t} a_i^4}{t}},
\]

gives \( a_1^2 = \cdots = a_t^2 = 1 \). Therefore,

\[
\sum_{i=1}^{t} a_i = \sum_{i=1}^{t} a_i^3 \in \mathbb{Z}.
\]

Since \((1 - \sqrt{2})t \notin \mathbb{Z}\), this contradicts (A.3) and there is no representation (A.1) of \( y \).

**Example 3.5 revisited.** The truncated tracial moment matrix

\[
M_2(y) = \begin{pmatrix}
1 & 0 & 1 & 0 & 1 & 1 & 1 \\
0 & 1 & 1 & 0 & 0 & 0 & 0 \\
1 & 1 & 1 & 0 & 0 & 0 & 0 \\
1 & 0 & 0 & 4 & 0 & 0 & 2 \\
1 & 0 & 0 & 0 & 2 & 1 & 0 \\
1 & 0 & 0 & 1 & 2 & 0 & 0 \\
1 & 0 & 0 & 2 & 0 & 0 & 4
\end{pmatrix}
\]

is positive semidefinite but with respect to the index set

\( (1, X, Y, X^2, XY, YX, Y^2) \),

\( y \) has no tracial moment representation (3.3).

Assume \( y_w = \sum_{i=1}^{N} \lambda_i \text{Tr}(w(A_1^{(i)}, A_2^{(i)})) \) for some symmetric matrices \( A_j^{(i)} \) and \( \lambda_i \in \mathbb{R}_{\geq 0} \) with \( \sum_i \lambda_i = 1 \). Setting

\[
T^{(i)} := (\text{Tr}(u^*v(A_1^{(i)}, A_2^{(i)})))_{u,v}
\]

we have \( M_2(y) = \sum_{i=1}^{N} \lambda_i T^{(i)} \). Each \( T^{(i)} \) is positive semidefinite, thus in particular \( T_{22}^{(i)} = T_{33}^{(i)} = T_{23}^{(i)} =: t_i \) holds for all \( i = 1, \ldots, N \). Let \( d_i \) be the size of the symmetric matrices \( A_j^{(i)} \), \( j = 1, 2 \). From

\[
\frac{1}{d_i^2} \langle A_1^{(i)}, A_1^{(i)} \rangle \langle A_2^{(i)}, A_2^{(i)} \rangle = \text{Tr}(A_1^{(i)})^2 \text{Tr}(A_2^{(i)})^2 = t_i^2
\]

\[
= (\text{Tr}(A_1^{(i)}A_2^{(i)}))^2 = \frac{1}{d_i^2} \langle A_1^{(i)}, A_2^{(i)} \rangle^2
\]
we obtain by the Cauchy-Schwarz inequality that $A_1(i) = \alpha_i A_2(i)$ for some $\alpha_i \in \mathbb{R}$, $i = 1, \ldots, N$. But then we derive the contradiction

$$2 = M_2(y)_{55} = \sum_{i=1}^N \lambda_i T_{55}^{(i)} = \sum \lambda_i \text{Tr}(A_1^{(i)} A_2^{(i)}) = \sum \lambda_i \alpha_i^2 \text{Tr}(A_2^{(i)})$$

$$= \sum \lambda_i \text{Tr}(A_1^{(i)} A_1^{(i)} A_2^{(i)} A_2^{(i)}) = M_2(y)_{45} = 1.$$

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