Quantum Matching Theory (with new complexity theoretic, combinatorial and topological insights on the nature of the Quantum Entanglement)

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Abstract

Classical matching theory can be defined in terms of matrices with nonnegative entries. The notion of Positive operator, central in Quantum Theory, is a natural generalization of matrices with nonnegative entries. Based on this point of view, we introduce a definition of perfect Quantum (operator) matching. We show that the new notion inherits many "classical" properties, but not all of them. This new notion goes somewhere beyond matroids. For separable bipartite density matrices this new notion coincides with the full rank property of the intersection of two corresponding geometric matroids. In the classical situation, permanents are naturally associated with perfects matchings. We introduce an analog of permanents for positive operators, called Quantum Permanent and show how this generalization of the permanent is related to the Quantum Entanglement. Besides many other things, Quantum Permanents provide new rational inequalities necessary for the separability of bipartite quantum states. Using Quantum Permanents, we give deterministic poly-time algorithm to solve Hidden Matroids Intersection Problem and indicate some "classical" complexity difficulties associated with the Quantum Entanglement. Finally, we prove that the weak membership problem for the convex set of separable bipartite density matrices is NP-HARD.

1 Introduction and Main Definitions

The (classical) Matching Theory is an important, well studied but still very active part of the Graph Theory (Combinatorics). The Quantum Entanglement is one of the central topics in Quantum Information Theory. We quote from [31]: "An understanding of entanglement seems to be at the heart of theories of quantum computations and quantum cryptography, as it has been at the heart of quantum mechanics itself." We will introduce in this paper a Quantum generalization of the Matching Theory and will show that this generalization gives new and surprising insights on the nature of the Quantum Entanglement. Of course, there already exist several "bipartite" generalizations of (classical) bipartite matching theory. The most relevant to our paper is the Theory of Matroids, namely its part analyzing properties of intersections of two geometric matroids.

Definition 1.1: Intersection of two geometric matroids $MI(X,Y) = \{(x_i, y_i), 1 \leq i \leq K\}$ is a finite family of distinct 2-tuples of non-zero N-dimensional complex vectors, i.e. $x_i, y_i \in C^N$. The rank of $MI(X,Y)$ is the largest integer $m$ such that there exist $1 \leq i_1 < \ldots < i_m \leq K$ with both sets $\{x_{i_1}, \ldots, x_{i_m}\}$ and $\{y_{i_1}, \ldots, y_{i_m}\}$ being linearly independent. If $Rank(MI(X,Y))$ is equal to $N$ then $MI(X,Y)$ is called matching. The matroidal permanent $MP_{X,Y}$ is defined as follows:

$$MP_{X,Y} = \sum_{1 \leq i_1 < i_2 < \ldots < i_K \leq K} \det(\sum_{1 \leq k \leq N} x_{i_k} x_{i_k}^\dagger) \det(\sum_{1 \leq k \leq N} y_{i_k} y_{i_k}^\dagger)$$

Remark 1.2: Let us denote linear space (over complex numbers) of $N \times N$ complex matrices as $M(N)$. It is clear from this definition that $MI(X,Y)$ is matching iff $MP_{X,Y} > 0$. Moreover, $MI(X,Y)$ is matching iff the linear subspace $Lin(X,Y) \subseteq M(N)$ generated by the matrices $\{x_i y_i^\dagger, 1 \leq i \leq K\}$ contains a nonsingular matrix and, in general, $Rank(MI(X,Y))$ is equal to the maximal matrix rank achieved in $Lin(X,Y)$. The following inequality generalizes Barvinok’s [14] unbiased estimator for mixed discriminants:

$$MP_{X,Y} = E(|\det(\sum_{1 \leq i \leq K} \xi_i x_i y_i^\dagger)|^2)$$

where $\{\xi_i, 1 \leq i \leq K\}$ are zero mean independent (or even 2N-wise independent) complex valued random variables such that $E(|\xi_i|^2) = 1, 1 \leq i \leq K$. It is not clear whether the analysis from [4] can be applied to $MP_{X,Y}$.

Example 1.3: Suppose that $x_i \in \{e_1, \ldots, e_N\}, 1 \leq i \leq K$, where $\{e_1, \ldots, e_N\}$ is a standard basis in $C^N$. Define the following positive semidefinite $N \times N$ matrices:

$$Q_i = \sum_{(e_i, e_j) \in X \times Y} y_j y_j^\dagger, 1 \leq i \leq N.$$ Then it is easy to see that in this case matroidal permanents coincide with the mixed discriminant, i.e. $MP_{X,Y} = M(Q_1, \ldots, Q_N)$ where the mixed discriminant defined as follows:

$$M(Q_1, \ldots, Q_N) = \frac{\partial^n}{\partial x_1 \ldots \partial x_n} \det(x_1 Q_1 + \ldots + x_n Q_N).$$

We will also use the following equivalent definition:

$$M(Q_1, \ldots, Q_N) = \sum_{\sigma,\tau \in S_N} (-1)^{\text{sign}(\sigma)} \prod_{i=1}^N Q_i(\sigma(i), \tau(i)).$$
where $S_n$ is the symmetric group, i.e. the group of all permutations of the set $\{1, 2, \cdots, N\}$. If matrices $Q_i$, $1 \leq i \leq N$ are diagonal then their mixed discriminant is equal to the corresponding permanent.\\

Let us pose, before moving to Quantum generalizations, the following "classical" decision problem. We will call it Hidden Matroids Intersection Problem (HMIP)\\

**Problem 1.4**: Given linear subspace $L \subseteq M(N)$ and a promise that $L$ has a (hidden) basis consisting of rank one matrices. Is there exists poly-time deterministic algorithm to decide whether $L$ contains a nonsingular matrix? Or more generally, to compute maximum matrix rank achieved in $L$?\\

Below in the paper we will assume that linear subspace $L \subseteq M(N)$ in (HMIP) is given as a some rational basis in it. If this basis consists of rank one matrices then there is nothing "hidden" and one can just apply standard poly-time deterministic algorithm computing rank of intersection of two matroids. A natural (trivial) way to attack (HMIP) would be to extract a (hidden) basis consisting of rank one matrices. We are not aware about the complexity of this extraction. The following example shows that there exist linear subspaces $L \subseteq M(N)$ having a rational real basis and a "rank one" basis but without rational "rank one" basis:\n
Consider the following $2 \times 2$ matrix
\[
A = \begin{pmatrix}
0 & -2 \\
0 & 1
\end{pmatrix},
\]
and define linear subspace $IR \subseteq M(2)$ generated by $A$ and the identity $I$.

It is easy to see that $\text{Rank}(aA + bI) \leq 1$ iff $a^2 + 2b^2 = 0$.

Therefore there are no rank one rational (complex) matrices in $IR$.

From the other hand rank one matrices
\[
C = \sqrt{2}I + iB, D = \sqrt{2}I - iB
\]
form a basis in $IR$.

One of the main results of our paper is a positive answer to the nonsingularity part of (HMIP). Moreover our algorithm is rather simple and does not require to work with algebraic numbers.

And, of course, we are aware about randomized poly-time algorithms, based on Schwartz's lemma, to solve this part of (HMIP). But for general linear subspaces, i.e. without extra promise, poly-time deterministic algorithms are not known and the problem is believed to be "HARD". To move to Quantum generalization, we need to recall several, standard in Quantum Information literature, notions.

### 1.1 Positive and completely positive operators; bipartite density matrices and Quantum Entanglement

**Definition 1.5**: A positive semidefinite matrix $\rho_{A,B} : C^N \otimes C^N \rightarrow C^N \otimes C^N$ is called bipartite unnormalized density matrix (BUDM), if $\text{tr}(\rho_{A,B}) = 1$ then this $\rho_{A,B}$ is called bipartite density matrix.

It is convenient to represent bipartite $\rho_{A,B} = \rho(i_1, i_2, j_1, j_2)$ as the following block matrix:
\[
\rho_{A,B} = \begin{pmatrix}
A_{1,1} & A_{1,2} & \cdots & A_{1,N} \\
A_{2,1} & A_{2,2} & \cdots & A_{2,N} \\
\vdots & \vdots & \ddots & \vdots \\
A_{N,1} & A_{N,2} & \cdots & A_{N,N}
\end{pmatrix},
\]

where $A_{i_{1}, j_{1}} := \{ \rho(i_1, i_2, j_1, j_2) : 1 \leq i_2, j_2 \leq N \}$, $1 \leq i_1, j_1 \leq N$.

A (BUDM) $\rho$ called separable if
\[
\rho = \rho(X,Y) := \sum_{1 \leq i \leq K} x_i x_i^\dagger \otimes y_i y_i^\dagger,
\]
and entangled otherwise.

If vectors $x_i, y_i; 1 \leq i \leq K$ in (6) are real then $\rho$ is called real separable.

Quantum marginals defined as $\rho_A = \sum_{1 \leq i \leq N} A_{i,i}$ and $\rho_{A,B}(i,j) = \text{tr}(A_{i,j}); 1 \leq i, j \leq N$.

We will call (BUDM) $\rho$ weakly separable if there exists a separable $\rho_{(X,Y)}$ with the same Image as $\rho$: $\text{Im}(\rho) = \text{Im}(\rho_{(X,Y)})$. A linear operator $T : M(N) \rightarrow M(N)$ called positive if $T(X) \succeq 0$ for all $X \succeq 0$, and strictly positive if $T(X) \succeq \alpha tr(X)I$ for all $X \succeq 0$ and some $\alpha > 0$. A positive operator $T$ is called completely positive if
\[
T(X) = \sum_{1 \leq i \leq N} A_i X A_i^\dagger; A_i, X \in M(N)
\]

Choi’s representation of linear operator $T : M(N) \rightarrow M(N)$ is a block matrix $CH(T)_{i,j} := T(e_i e_j^\dagger)$. Dual to $T$ respect to the inner product $\langle X, Y \rangle := \text{tr}(X Y^\dagger)$ is denoted as $T^*$. Very useful and easy Choi’s result states that $T$ is completely positive iff $CH(T)$ is (BUDM). Using this natural (linear) correspondence between completely positive operators and (BUDM), we will freely "transfer" properties of (BUDM) to completely positive operators. For example, a linear operator $T$ is called separable iff $CH(T)$ is separable, i.e.
\[
T(Z) = T_{(X,Y)}(Z) = \sum_{1 \leq i \leq K} x_i y_i^\dagger Z y_i x_i^\dagger
\]

Notice that $CH(T_{(X,Y)}) = \rho_{(X,Y)}$ and $T_{(X,Y)} = T_{(Y,X)}$.

**Remark 1.6**: In light of definition (1.5), we will represent linear subspaces $L \subseteq M(N) \cong C^N \otimes C^N$ in (HMIP) as images of weakly separable (BUDM) $\rho$. And as the complexity measure we will use the number of bits of (rational) entries of $\rho$.

The next definition introduces the quantum permanent $QP(\rho)$, the main tool to solve (HMIP). Though it was not our original intention, it happens that $QP(\rho_{(X,Y)}) = MP(\rho_{(X,Y)})$.

**Definition 1.7**: We define quantum permanent, $QP(\rho)$, by the following equivalent formulas:
\[
QP(\rho) := \sum_{\sigma \in \delta_N} (-1)^{\text{sign}(\sigma)} M(A_{1,\sigma(1)}, \ldots, A_{N,\sigma(N)});
\]

\[
QP(\rho) = \sum_{\tau_1, \tau_2, \tau_3 \in S_N} (-1)^{\text{sign}(\tau_1 \tau_2 \tau_3)} \prod_{i=1}^N \text{trho}(i, \tau_1(i), \tau_2(i), \tau_3(i));
\]

\[
QP(\rho) = \prod_{i=1}^N \text{trho}(\tau_1(i), \tau_2(i), \tau_3(i), \tau_4(i)).
\]
Remark 1.8: The representation (6) is not unique, it follows directly from the Carathéodory Theorem that one always can choose \( K \leq N^4 \) in (6). Thus, the set of separable (BUDM), denoted by \( \text{Sep}(N,N) \), is a convex closed set. As it is known that \( \text{Sep}(N,N) \) has non-empty interior, it follows from straightforward dimensions counting that for the "most" separable (BUDM) at least \( K \geq \frac{N^4}{2} \).

In the next proposition we summarize the properties of the quantum permanents we will need later in the paper.

**Proposition 1.9:**

1. \( QP(\rho(x,y)) = MP(x,y) \) \hspace{1cm} (12)
2. \( QP(\rho) = \rho^N, Z, Z \), \hspace{1cm} (13)

where \( \rho^N \) stands for a tensor product of \( N \) copies of \( \rho \), \( <\ldots,> \) is a standard inner product and \( Z = \frac{1}{N^2} \sum (-1)^{\text{sign}(\tau_1,\tau_2)} Z_{\tau_1}^{(1)}, Z_{\tau_2}^{(2)}, \ldots, Z_{\tau_2}^{(N)}, Z_{\tau_1}^{(N)} \).

if \( j_1^i = \tau_k(i) (1 \leq i \leq N) \), \( \tau_k \in S_N(k = 1, 2) \) and zero otherwise. (The equality (13) implies that if \( \rho_1 \geq \rho_2 \geq 0 \) then \( QP(\rho_1) \geq QP(\rho_2) \geq 0 \).)
3. \( QP(A_1 \otimes A_2) = QP(A_1 \otimes A_2) QP(\rho_1) \hspace{1cm} (14)\)
4. \( QP(\rho A, B) = QP(\rho B, A) \) \hspace{1cm} (15)

**Example 1.10:** Let us present a few cases when Quantum Permanents can be computed "exactly". They will also illustrate how universal is this new notion.

1. Let \( \rho_{A,B} \) be a product state, i.e. \( \rho_{A,B} = C \otimes D \). Then \( QP(C \otimes D) = Det(C)Det(D) \).
2. Let \( \rho_{A,B} \) be a pure state, i.e., there exists a matrix \( R = R(i_1, i_2, j_1, j_2) \) such that \( \rho_{A,B}(i_1, i_2, j_1, j_2) = R(i_1, i_2)R(j_1, j_2) \).

In this case \( QP(\rho_{A,B}) = N! |Det(R)|^2 \).
3. Define blocks of \( \rho_{A,B} \) as \( A_{i,j} = R(i,j)e_ee_1 \).

Then \( QP(\rho_{A,B}) = M(\rho) \).

The next definition introduces Quantum Perfect Matching.

**Definition 1.11:** Let us consider a positive (linear) operator \( T : M(N) \to M(N) \) and a map \( G : C^N \to C^N \), and the following three conditions:

1. \( G(x) \in \text{Im}(T(xx^T)) \).
2. If \( \{x_1, \ldots, x_N\} \) is a basis in \( C^N \) then \( \{G(x_1), \ldots, G(x_N)\} \) is also a basis, i.e. the map \( G \) preserves linear independence.
3. If \( \{x_1, \ldots, x_N\} \) is an orthogonal basis in \( C^N \) then \( \{G(x_1), \ldots, G(x_N)\} \) is a basis.

We say that map \( G \) is Quantum Perfect Matching for \( T \) if it satisfies conditions (1,2) above; say map \( G \) is Quantum Semi-Perfect Matching for \( T \) if it satisfies conditions (1,3) above.

In the rest of the paper we will address the following topics:

1. Characterization of Quantum Perfect Matchings in spirits of Hall’s theorem.
2. Topological and algebraic properties of Quantum Perfect Matchings, i.e. properties of maps \( G \) in Definition 1.11.
3. Complexity of checking whether given positive operator is matching.
4. Quantum (or Operator) generalizations of Sinkhorn’s iterations (in the spirit of [24],[32],[30]).
5. van der Waerden Conjecture for Quantum Permanents.
6. Connections between topics above and the Quantum Entanglement.
7. Complexity to check the separability.

**2 Necessary and sufficient conditions for Quantum Perfect Matchings**

**Definition 2.1:** A positive linear operator \( T : M(N) \to M(N) \) called rank non-decreasing iff

\[ \text{Rank}(T(X)) \geq \text{Rank}(X) \text{ if } X \geq 0; \]

and called indecomposable iff

\[ \text{Rank}(T(X)) > \text{Rank}(X) \text{ if } X \geq 0 \text{ and } 1 \leq \text{Rank}(X) < N. \]

A positive linear operator \( T : M(N) \to M(N) \) called doubly stochastic iff \( T(I) = I \) and \( T^*(I) = I \); called \( \epsilon \)-doubly stochastic iff \( DS(T) := \text{tr}((T(I) - I)^2) + \text{tr}((T^*(I) - I)^2) \leq \epsilon^2 \).

The next conjectures generalize Hall’s theorem to Quantum Perfect Matchings.

**Conjecture 2.2:** Assuming that the Axiom of Choice and the Continuum Hypothesis hold, a positive linear operator \( T \) has Quantum Perfect Matching iff it is rank non-decreasing.

**Conjecture 2.3:** Assuming that the Axiom of Choice and the Continuum Hypothesis hold, a positive linear operator \( T \) has Quantum Semi-Perfect Matching iff it is rank non-decreasing.

**Remark 2.4:** We realize that the presence of the Axiom of Choice and the Continuum Hypothesis in linear finite dimensional result might look a bit weird. But we will illustrate below in this section that for some completely positive entangled operators corresponding Quantum semi-perfect matching maps \( G \) are necessary quite complicated, for instance necessary discontinuous. Moreover Conjecture 1 is plain wrong, even for doubly stochastic indecomposable completely positive operators. In separable and even weakly separable cases one does not need “exotic axioms” and one can realize Quantum perfect matching map it exists as a linear nonsingular transformation through a rather simple use of Edmonds-Rado theorem.

The next Proposition (2.5) is a slight generalization of the corresponding result in [34].

**Proposition 2.5:** Doubly stochastic operators are rank non-decreasing.

If either \( T(I) = I \) or \( T^*(I) = I \) and \( DS(T) \leq N^{-3} \) then \( T \) is rank non-decreasing. If \( DS(T) \leq (2N + 1)^{-1} \) then \( T \) is rank non-decreasing.
**Example 2.6:** Consider the following completely positive doubly stochastic operator \( SK_3 : M(3) \rightarrow M(3) \):

\[
SK_3(X) = \frac{1}{2} A_{(1,2)} X A_{(1,2)}^\dagger + A_{(1,3)} X A_{(1,3)}^\dagger + A_{(2,3)} X A_{(2,3)}^\dagger
\]

(18)

Here \( \{A_{(i,j)} : 1 \leq i < j \leq 3\} \) is a standard basis in a linear subspace of \( M(3) \) consisting of all skew-symmetric matrices , i.e. \( A_{(i,j)} = e_i e_j - e_j e_i \) and \( e_i, 1 \leq i \leq 3 \) is a standard orthonormal basis in \( C^3 \). It is easy to see that for a real normed 3-dimensional vector space \( x \) the image \( \text{Im} SK_3(xx^\dagger) \) is equal to the real orthogonal complement of \( x \), i.e. to the linear 2-dimensional subspace \( x^\perp \) consisting of all real vectors orthogonal to \( x \). Suppose that \( G \) is Quantum semi-perfect matching map , then \( G(x) \in x^\perp \) and , at least , \( G(x) \) is nonzero for nonzero vectors \( x \). By the well known topological result , impossibility to comb the unit sphere in \( R^3 \) , none of Quantum semi- perfect matchings for \( SK_3 \) is continuous. It is not difficult to show that the operator \( SK_3 \) is entangled . A direct computation shows that

\[
QP(CH(SK_3)) = 0
\]

(19)

An easy "lifting" of this construction allows to get a similar example for all \( N \geq 3 \). From the other hand , for \( N = 2 \) all rank nondecreasing positive operators have linear nonsingular Quantum perfect matchings.

**Proposition 2.7:** Assuming that the Axiom of Choice and the Continuum Hypothesis hold, \( SK_3 \) has a Quantum semi-perfect matching .

**Proof:** (Sketch) Let us well order the projective unit sphere \( PS_2 \) in \( C^3 \) : \( S_2 = (t_\alpha : \alpha \in \Gamma) \) in such way that for any \( \beta \in \Gamma \) the interval \( (t_\alpha \leq \alpha \leq t_\beta) \) is at most countable . Our goal is to build \( (g_\alpha : \alpha \in \Gamma ; g_0 \neq 0, g_\alpha \in t_\alpha^C) \) such that \( (t_\alpha_1, t_\alpha_2, t_\alpha_3) \) is orthogonal basis then \( (g_\alpha_1, g_\alpha_2, g_\alpha_3) \) is a basis.

As it usually happens in inductive constructions , we will inductively force an additional property : \( < g_\alpha, g_\beta > \neq 0 \) if \( \alpha > \beta \) and linear space \( L(g_\alpha, g_\beta) \) generated by \( (g_\alpha, g_\beta) \) is not equal to \( L(t_\alpha, t_\beta) \) if \( t_\alpha < t_\beta \) is orthogonal.

In this , orthogonal case , \( L(g_\alpha, g_\beta) = L(t_\alpha, t_\beta) \) if \( g_\alpha = t_\alpha = t_\beta \). Using countability assumption , it is easy to show that at each step of transfinite induction the set of "bad" candidates has measure zero , which allows always to choose a "good" guy \( g_\alpha \) without changing already constructed \( (g_\alpha : \alpha < \gamma) \).

The next Proposition shows that \( SK_3 \) does not have Quantum perfect matchings!

**Proposition 2.8:** \( SK_3 \) does not have Quantum perfect matchings .

**Proof:** Suppose that \( G(\cdot) \) is Quantum perfect matching for \( SK_3 \). We will get a contradiction by showing that then there exists a basis \( (b_1, b_2, b_3) \) such that \( < b_1, b_2 > = 0 \) and \( G(b_1), G(b_2), G(b_3) \) are linearly dependent . For that , we need to show that there exists an orthogonal basis \( (O_1, O_2, O_3) \) such that \( O_3 \) does not belong to \( L(G(O_1), G(O_2)) \). Indeed , if non-zero \( d \in L(G(O_1), G(O_2)) \), then there is no basis \( (G(O_1), G(O_2), v) \) with \( v \in d^\perp = L(G(O_1), G(O_2)) \), but \( (O_1, O_2, d) \) is a basis since \( c < d, O_3 > = 0 \).

Take any non-zero \( x \) and an orthogonal basis \( \{y, z\} \) in \( x^\perp \) such that \( G(x) = (0, a_1, a_2) \) in \( \{x, y, z\} \) basis and \( a_1 \neq 0, a_2 \neq 0 \). Let \( G(y) = (b_1, 0, b_2) \), \( G(z) = (c_1, c_2, 0) \).

Suppose that \( z \in L(G(x), G(y)) \), and \( y \in L(G(x), G(z)) \). Then \( b_1 = 0 \) and \( c_1 = 0 \). This contradicts to \( (\{G(x), G(y), G(z)\}) \) being a basis . Thus there exists an orthogonal basis \( (O_1, O_2, O_3) \) such that \( O_3 \) does not belong to \( L(G(O_1), G(O_2)) \) and we got a final contradiction.

Next result shows that for weakly separable (and thus for separable) operators the situation is very different.

**Theorem 2.9:** Suppose that \( T : M(N) \rightarrow M(N) \) is linear positive weakly separable operator , i.e. there exists a family of rank one matrices \( \{x_1 y_1^*, ..., x_i y_i^*\} \subset M(N) \) such that for positive semidefinite matrices \( X \geq 0 \) the following identity holds:

\[
\text{Im}(T(X)) = \text{Im}(\sum_{i=1}^{l} x_i y_i x_i^*)
\]

(20)

Then the following conditions are equivalent :

1. \( T \) is rank non-decreasing .
2. The rank of intersection of two geometric matroids \( MI(X, Y) \) is equal to \( N \).
3. The exists a nonsingular matrix \( A \) such that \( \text{Im}(AXA^*) \subset \text{Im}(T(X)), X \geq 0 \).

If , additionally , \( T \) is completely positive then these conditions are equivalent to existence of nonsingular matrix \( A \) such that operator \( T(X) = T(X) - AXA^* \) is completely positive.

In this case \( QP(CH(T)) \geq N! |\text{Det}(A)|^2 > 0 \).

**Proof:** Recall Edmonds-Rado Theorem for \( MI(X, Y) \):

\[
\text{Rank of } MI(X, Y) \text{ is equal } N \text{ iff } \dim(L(x_i ; i \in A)) + \dim(L(y_j ; j \in A^c)) \geq N, \text{ where } A \subset \{1, 2, ..., l\} \text{ and } A^c \text{ is a complement of } A.
\]

Suppose that rank of \( MI(X, Y) \) is equal to \( N \). Then

\[
\text{Rank}(T(X)) = \dim(L(x_i ; i \in A)) \text{ where } A =: \{i : y_i^T X y_i \neq 0\}.
\]

As \( \dim(L(y_j ; j \in A^c) \leq \dim(Ker(X)) = N - \text{Rank}(X) \) hence , from Edmonds-Rado Theorem we get that \( \text{Rank}(T(X)) \geq N - (N - \text{Rank}(X)) = \text{Rank}(X) \).

Suppose that \( T \) is rank non-decreasing and for any \( A \subset \{1, 2, ..., l\} \) consider an orthogonal projector \( P \geq 0 \) on \( L(y_j ; j \in A^c)^\perp \). Then

\[
\dim(L(x_i ; i \in A)) \geq \text{Rank}(T) \geq \text{Rank}(P) = \text{N} - \dim(L(y_j ; j \in A^c)).
\]

It follows from Edmonds-Rado Theorem that rank of \( MI(X, Y) \) is equal to \( N \). All "equivalencies" follow now directly.

**Remark 2.10:** Let us explain why Conjectures (1,2) generalize Hall’s theorem . Consider a square weighted incidence matrix \( A_r \) of a bipartite graph \( \Gamma \), i.e. \( A_r(i, j) > 0 \) if \( i \) from the first part is adjacent to \( j \) from the second part and equal to zero otherwise. Then Hall’s theorem can be immediately reformulated as follows : A perfect matching , which is just a permutation in this bipartite case , exists iff \( |A_r x|_+ \geq |x|_+ \) for any vector \( x \) with nonnegative entries , where \( |x|_+ \) stands for a number of positive entries of a vector \( x \). One also can look at Theorem(2) as a Hall’s like reformulation of Edmonds-Rado theorem.
2.1 A preliminary summary

So far, we got necessary and sufficient conditions for the existence of Quantum Perfect Matchings and presented, based on them, a new topological insight on the nature of the Quantum Entanglement. It is not clear to us how crucial are "logical" assumptions in Prop.(2.7) . Theorem(2.9) shows that in separable (even weakly separable) case these assumptions are not needed. The next question, which we study in the next sections, is about efficient, i.e. polynomial time, deterministic algorithms to check the existence of Quantum Perfect Matchings. We will describe and analyse below in the paper a "direct" deterministic polynomial time algorithm for weakly separable case. A complexity bound for a separable case is slightly better than for just weakly separable case. Our algorithm is an operator generalization of Sinkhorn’s iterative scaling. We conjecture that without some kind of separability promise checking the existence of Quantum Perfect Matchings is "HARD" even for completely positive operators.

3 Operator Sinkhorn’s iterative scaling

Recall that for a square matrix $A = \{a_{ij} : 1 \leq i, j \leq N\}$ row scaling is defined as

$$R(A) = \left\{ \frac{a_{ij}}{\sum_j a_{ij}} \right\},$$

column scaling as $C(A) = \left\{ \frac{a_{ij}}{\sum_i a_{ij}} \right\}$ assuming that all denominators are nonzero.

The iterative process $CRCR(A)$ is called Sinkhorn’s iterative scaling (SI). There are two mainwell known properties of this iterative process, which we will generalize to positive Operators.

**Proposition 3.1:**

1. Suppose that $A = \{a_{ij} \geq 0 : 1 \leq i, j \leq N\}$. Then (SI) converges iff $A$ is matching, i.e., there exists a permutation $\pi$ such that $a_{\pi(i),i} > 0$ ($1 \leq i \leq N$).

2. If $A$ is indecomposable, i.e., $A$ has a doubly-stochastic pattern and is fully indecomposable in the usual sense, then (SI) converges exponentially fast. Also in this case there exist unique positive diagonal matrices $D_1, D_2, \det(D_2) = 1$ such that the matrix $D_1^{-1}AD_2^{-1}$ is doubly stochastic.

**Definition 3.2:** [Operator scaling] Consider linear positive operator $T : M(N) \rightarrow M(N)$. Define a new positive operator, Operator scaling, $SC_{c_1,c_2}(T)$ as:

$$SC_{c_1,c_2}(T)(X) =: c_1 T(C_2^2 X C_2)C_1^2$$

(22)

Assuming that both $T(I)$ and $T^*(I)$ are nonsingular we define analogs of row and column scalings:

$$R(T) = S_{T(I)}^{-\frac{1}{2}}(T), C(T) = S_{I, T^*(I)}^{-\frac{1}{2}}(T)$$

(23)

Operator Sinkhorn’s iterative scaling (OSI) is the iterative process $...CRCR(R(T))$]

**Remark 3.3:** Using Choi’s representation of the operator $T$ in Definiton(1.5), we can define analogs of operator scaling (which are nothing but so called local transformations) and (OSI) in terms of

(24)

**BUDM:**

$$SC_{c_1,c_2}(\rho_{A,B}) = C_1 \otimes C_2(\rho_{A,B})C_1^2 \otimes C_2^2;$$

$$R(\rho_{A,B}) = \rho_A^{\frac{1}{2}} \otimes I(\rho_{A,B})\rho_A^{\frac{1}{2}} \otimes I,$$

$$C(\rho_{A,B}) = I \otimes \rho_B^{\frac{1}{2}}(\rho_{A,B})I \otimes \rho_B^{\frac{1}{2}}.$$

Let us introduce a class of locally scalable functionals (LSF) defined on a set of positive linear operators, i.e., functionals satisfying the following identity:

$$\varphi(SC_{c_1,c_2}(T)) = Det(C_1C_1^*[T] Det(C_2C_2^*[T]) \varphi(T)$$

(25)

We will call (LSF) bounded if there exists a function $f$ such that $|\varphi(T)| \leq f(tr(T(I)))$. It is clear that bounded (LSF) are natural "potentials" for analyzing (OSI). Indeed, let $T_n : T_0 = T$ be a trajectory of (OSI). $T$ is a positive linear operator. Then $T_n(I) = I$ for odd $i$ and $T_2(I) = I, i \geq 1$. Thus if $\varphi$ is (LSF) then

$$\varphi(T_{i+1}) = a(i)\varphi(T), a(i) = Det(T_i(I))^{-1}$$

if $i$ is odd,

$$a(i) = Det(T_i(I))^{-1}$$

if $i > 0$ is even.

(26)

As $tr(T(I)) = tr(T^*_I(I)) = N, i > 0$ , thus by the arithmetic/geometric means inequality we have that $|\varphi(T_{i+1})| \geq |\varphi(T)|$ and if $\varphi$ is bounded and $|\varphi(T)| \neq 0$ then $DS(T_n)$ converges to zero.

To prove a generalization of Statement 1 in Prop.(3.1) we need to "invent" a bounded (LSF) $\varphi(.)$ such that $\varphi(T) \neq 0$ iff operator $T$ is matching. We call such functionals responsible for matching. It is easy to prove that $QP(CH(T))$ is a bounded (LSF). Thus if $QP(CH(T)) \neq 0$ then $DS(T_n)$ converges to zero and, by Prop. (2.5), $T$ is rank nondecreasing. From the other hand, $QP(CH(S_{kN})) = 0$ and $S_{kN}$ is rank nondecreasing (even indecomposable). This is another "strangeness" of entangled operators, we wonder if it is possible to have "nice", say polynomial with integer coefficients, responsible for matching (LSF)? We introduce below responsible for matching bounded (LSF) and it is non-differentiable.

**Definition 3.4:** For a positive operator $T : M(N) \rightarrow M(N)$, we define its capacity as

$$Cap(T) = \inf\{Det(X) : X > 0, Det(X) = 1\}.$$

(27)

It is easy to see that $\text{Cap}(T)$ is (LSF). Since $\text{Cap}(T) \leq Det(T(I)) = (tr(T(I)))^N$, hence $\text{Cap}(T)$ is bounded (LSF).

**Lemma 3.5:** A positive operator $T : M(N) \rightarrow M(N)$ is positive rank nondecreasing if $\text{Cap}(T) > 0$.

**Proof:** Let us fix an orthonormal basis (unitary matrix) $U = \{u_1,...,u_N\}$ in $C^N$ and associate with positive operator $T$ the following positive operator:

$$T_U(X) := \sum_{1 \leq i \leq N} T(u_i u_i^*)tr(Xu_i u_i^*).$$

(28)

(In physics words, $T_U$ is a decoherence respect to the basis $U$., i.e., in this basis applying $T_U$ to matrix $X$ is the same as applying $T$ to...
the diagonal restriction of $X$. }

It is easy to see that a positive operator $T$ is rank nondecreasing iff operators $T_U$ are rank nondecreasing for all unitary $U$.

And for fixed $U$ all properties of $T_U$ are defined by the following $N$-tuple of $N \times N$ positive semidefinite matrices:

$$A_{T,U} := (T(u_1 u_1^T), \ldots, T(u_N u_N^T)).$$

(29)

Importantly for us, $T_U$ is rank nondecreasing iff the mixed discriminant $M(T(u_1 u_1^T), \ldots, T(u_N u_N^T)) > 0$.

Define capacity of $A_{T,U}$,

$$\text{Cap}(A_{T,U}) := \inf \{ \text{Det}(\sum_{1 \leq i \leq N} T(u_i u_i^T) \gamma_i) : \gamma_i > 0, \prod_{1 \leq i \leq N} \gamma_i = 1 \}.$$ 

It is clear from the definitions that $\text{Cap}(T)$ is equal to the infimum of $\text{Cap}(A_{T,U})$ over all unitary $U$.

One of the main results of [30] states that

$$M(A_{T,U}) := M(T(u_1 u_1^T), \ldots, T(u_N u_N^T)) \leq \frac{N^N}{N!} M(T(u_1 u_1^T), \ldots, T(u_N u_N^T)).$$

(30)

As the mixed discriminant is a continuous (analytic) functional and the group $SU(N)$ of unitary matrices is compact, we get the next inequality:

$$\min_{U \in SU(N)} M(A_{T,U}) \leq \text{Cap}(T) \leq \frac{N^N}{N!} \min_{U \in SU(N)} M(A_{T,U})$$

(31)

The last inequality proves that $\text{Cap}(T) > 0$ iff positive operator $T$ is rank nondecreasing.

So, the capacity is a bounded (LSF) responsible for matching, which proves the next theorem:

**Theorem 3.6:**

1. Let $T_0, T_\beta = T$ be a trajectory of (OSI), $T$ is a positive linear operator. Then $DS(T_\beta)$ converges to zero iff $T$ is rank nondecreasing.

2. Positive linear operator $T$ is rank nondecreasing iff for all $\epsilon > 0$ there exists $\epsilon$-doubly stochastic operator scaling of $T$.

The next theorem generalizes second part of Prop. (3.1) and is proved on almost the same lines as Lemmas 24,25,26,27 in [30].

**Theorem 3.7:**

1. There exist nonsingular matrices $C_1, C_2$ such that $S_{C_1, C_2}(T)$ is doubly stochastic iff the infimum in (26) is achieved.

Moreover, if $\text{Cap}(T) = \text{Det}(T(C))$ where $C > 0, \text{Det}(C) = 1$ then $T(C) \rightarrow C^{-1/2} (T)$ is doubly stochastic.

Positive operator $T$ is indecomposable iff the infimum in (27) is achieved and unique.

2. Doubly stochastic operator $T$ is indecomposable iff $\text{tr}^2(T(X)) \leq a \text{tr}(X)^2$ for some $0 \leq a < 1$ and all traceless hermitian matrices $X$.

3. If Positive operator $T$ is indecomposable then $DS(T_\beta)$ converges to zero with the exponential rate, i.e. $DS(T_\beta) \leq K \alpha^n$ for some $K$ and $0 \leq \alpha < 1$.

4 Lower and upper bounds on Quantum Permanents

The next proposition follows fairly directly from the second part of Prop.(1.9) and Cauchy-Schwarz inequality

**Proposition 4.1:** Suppose that $\rho_{A,B}$ is (BUDM). Then

$$\max_{\sigma \in S_N} |D(A_{1,\sigma(1)}, \ldots, A_{N,\sigma(N)})| = \text{Det}(A_{1,1}, \ldots, A_{1,N})$$

(32)

**Corollary 4.2:** If $\rho_{A,B}$ is (BUDM) then

$$\text{QP}(\rho_{A,B}) \leq N! \text{Det}(A_{1,1}, \ldots, A_{1,N}) \leq N! \text{Det}(\rho_A).$$

(33)

Permanent part of Example(1.10) shows that $N!$ is exact constant in both parts of (32).

The next proposition follows from the Hadamard’s inequality: if $X > 0$ is $N \times N$ matrix then $\text{Det}(X) \leq \prod_{i=1}^{N} X(i,i)$.

**Proposition 4.3:** If $X > 0$ then the following inequality holds:

$$\text{Det}(\sum_{i=1}^{K} x_i y_i^T X y_i x_i^T) \geq \text{Det}(X) \text{MP}(X,Y).$$

(34)

**Corollary 4.4:** Suppose that separable (BUDM) $\rho_{A,B}$ is Choi’s representation of completely positive operator $T$.

Then for all $X > 0$ the next inequality holds:

$$\text{Det}(T(X)) \geq \text{QP}(\rho_{A,B}) \text{Det}(X)$$

(35)

Since $\rho_A = T(I)$, hence $\text{QP}(\rho_{A,B}) \leq \text{Det}(\rho_A)$ in separable case.

Call (BUDM) $\rho_{A,B}$ doubly stochastic if it is Choi’s representation of completely positive doubly stochastic operator $T$. I.e. (BUDM) $\rho_{A,B}$ is doubly stochastic iff $\rho_A = PB = I$. As we already explained, the set of separable (BUDM) is convex and closed. Thus the set of doubly stochastic separable (BUDM), $DSEP(N,N)$, is a convex compact. Define

$$\beta(N) = \min_{\rho \in DSEP(N,N)} \text{QP}(\rho).$$

Then it follows that $\beta(N) > 0$ for all integers $N$. The next conjecture is, in a sense, a third generation of the famous van der Waerden conjecture. First generation is a permanental conjecture proved by Falikman and Egorychev ([15], [14]) in 1980 and second generation is Mixed discriminants conjecture posed by R.Bapat in 1989 and proved by the author in 1999 [19]. Mixed discriminants conjecture corresponds to block-diagonal doubly stochastic (BUDM). Any good lower bound on $\beta(N)$ will provide similarly to [20] deterministic poly-time approximations for Matroidal permanents and new sufficient conditions for the Quantum Entanglement.

**Conjecture 4.5:**

$$\beta(N) = \frac{N!}{N^N}.$$ 

(36)

It is true for $N = 2$. 

5 Polynomial time deterministic algorithm for (HMIP )

We introduced Hidden Matroids Intersection Problem (HMIP ) as a well posed computer science problem, which, seemingly, requires no "Quantum" background. Also, we explained that (HMIP ) can be formulated in terms of weakly separable (BUDM ), let us consider the following three properties of (BUDM ) $\rho_{A,B}$ . (We will view this $\rho_{A,B}$ as Choi’s representation of completely positive operator $T$, i.e. $\rho_{A,B} = CH(T)$.)

**P1** $Im(\rho_{A,B})$ contains a nonsingular matrix.

**P2** The Quantum permanent $QP(\rho_{A,B}) > 0$.

**P3** Operator $T$ is rank nondecreasing.

We proved already that $P1 \rightarrow P2 \rightarrow P3$ and illustrated that the implication $P1 \rightarrow P2$ is strict. In fact the implication $P1 \rightarrow P2$ is also strict. But, our Theorem (2.9), which is just an easy adoption of Edmonds-Rado theorem, shows that for weakly separable (BUDM ) the three properties $P1$, $P2$, $P3$ are equivalent.

Recall that to check $P1$ without the weak separability promise is the same as to check whether given linear subspace of $M(N)$ contains a nonsingular matrix and it is very unlikely that this dis-sion problem can be solved in Polynomial Deterministic time.

Next, we will describe and analyze Polynomial time deterministic algorithm to check whether $P3$ holds provided that it is promised that $\rho_{A,B}$ is weakly separable.

In terms of Operator Sinkhorn’s iterative scaling (OSI) we need to check if there exists $n$ such that $DS(T_n) \leq 1/n$. If $L = \min\{n : DS(T_n) \leq 1/n\}$ is bounded by a polynomial in $N$ and number of bits of $\rho_{A,B}$ then we have a Polynomial time Deterministic algorithm to solve (HMIP ). Algorithms of this kind for "classical" matching problem appeared independently in [24] and [25]. In the "classical" case they are just another, conceptually simple, but far from optimal, poly-time algorithms to check whether a perfect matching exists. But for (HMIP ), our Operator Sinkhorn’s iterative scaling based approach seems to be the only possibility? Assume that, without loss of generality, that all entries of $\rho_{A,B}$ are integer numbers and their maximum magnitude is $Q$. Then $Det(\rho_{A,B}) \leq (QN)^N$ by the Hadamard’s inequality. If $QP(\rho_{A,B}) > 0$ then necessary $QP(\rho_{A,B}) \geq 1$ for it is an integer number. Thus

$$QP(CH(T_n)) = \frac{QP(CH(T))}{Det(\rho_{A,B})} \geq (QN)^{-N}.$$  

Each $nth$ iteration ($n \leq L$) after the first one will multiply the Quantum permanent by $Det(X)^{-1}$, where $X > 0$, $tr(X) = N$ and $tr((X-I)^2) \geq \frac{1}{N^2}$. Using results from [23], $Det(X)^{-1} \geq (1 - \frac{1}{N})^{-1} = \frac{N}{N-1}$. Putting all this together, we get the following upper bound on $L$, the number of steps in (OSI) to reach the "boundary" $DS(T_n) \leq N$:

$$\delta^L \leq \frac{QP(CH(T))}{(QN)^{-N}}$$  

It follows from Prop.(4.2) and Cor.(4.4) that in weak separable case $QP(CH(T_n)) \leq N^N!$ and in separable case $QP(CH(T_n)) \leq 1$.

Taking logarithm we get that in weak separable case

$$L \leq \approx 3N(N \ln(N) + N(N + \ln(Q)))$$  

and in separable case

$$L \leq \approx 3N(N \ln(N) + \ln(Q)).$$  

In any case, $L$ is polynomial in the dimension $N$ and the number of bits log($Q$).

To finish our analysis, we need to evaluate a complexity of each step of (OSI).

Recall that $T_n(X) = L_n(T(R_n^1X R_n^2)) L_n^*$

and $T_n(I) = L_n(T(R_n^1 R_n^2)) L_n^*$.

To evaluate $DS(T_n)$ we need to compute $tr((T_n^*(I) - I)^2)$ for odd $n$, and $tr((T_n(I) - I)^2)$ for even $n$.

Define $P_n = L_n^* L_n$, $Q_n = R_n^* R_n$. It is easy to see that the matrix $T_n(I)$ is similar to $P_n T(Q_n)$, and $T_n^*(I)$ is similar to $Q_n T^*(P_n)$. As traces of similar matrices are equal, therefore to evaluate $DS(T_n)$ it is sufficient to compute matrices $P_n$, $Q_n$.

And this leads to standard, rational, matrix operations with $O(N^3)$ per one iteration in (OSI).

Notice that our original definition of (OSI) requires computation of an operator square root. It can be replaced by the Cholesky factorization, which still requires computing scalar square roots. But our final algorithm is rational!

6 Weak Membership Problem for a convex compact set of normalized bipartite separable density matrices is NP-HARD

One of the main research activities in Quantum Information Theory is a search for "operational" criterium for the separability. We will show in this section that, in a sense defined below, the problem is NP-HARD even for bipartite normalized density matrices provided that each part is large (each "particle" has large number of levels).

First, we need to recall some basic notions from computational convex geometry.

6.1 Algorithmic aspects of convex sets

We will follow [18].

**Definition 6.1**: A proper (i.e. with nonempty interior) convex set $K \subset R^n$ called well-bounded is centered if there exist rational vector $a \in K$ and positive (rational) numbers $r$, $R$ such that $B(a, r) \subset K$ and $K \subset B(a, R)$ (here $B(a, r) = \{x : \|x-a\| \leq r\}$ and $\|\cdot\|$ is a standard euclidean norm in $R^n$). Encoding length of such convex set $K$ is

$$\langle K \rangle = n + \langle r \rangle + \langle R \rangle + \langle a \rangle,$$

where $\langle r \rangle$, $\langle R \rangle$, $\langle a \rangle$ are the number of bits of corresponding rational numbers and rational vector.

Following [18] we define $S(K, \delta)$ as a union of all $\delta$-balls with centers belonging to $K$; and $S(K, -\delta) = \{x \in K : B(x, \delta) \subset K\}$.

**Definition 6.2**: The Weak Membership Problem (WMEM($K$, $y$, $\delta$)) is defined as follows:

Given a rational vector $y \in R^n$ and a rational number $\delta > 0$ either

(i) assert that $y \in S(K, \delta)$ or

(ii) assert that $y \notin S(K, -\delta)$.

The Weak Validity Problem (WVAL($K$, $c$, $\gamma$, $\delta$)) is defined as follows:

Given a rational vector $y \in R^n$, rational number $\gamma$ and a rational number $\delta > 0$ either

(i) assert that $\langle c, x \rangle = \gamma + \delta$ for all $x \in S(K, -\delta)$, or

(ii) assert that $\langle c, x \rangle \geq \gamma - \delta$ for some $x \in S(K, \delta)$.


Remark 6.3: Define $M(K,c) =: \max_{x \in K} < c, x >$. It is easy to see that

\[ M(K,c) \geq M(S(K,-\delta),c) \geq M(K,c) - ||c||\delta; \]
\[ M(K,c) \leq M(S(K,\delta),c) \geq M(K,c) + ||c||\delta. \]

Recall that seminal Yudin - Nemirovskii theorem (\cite{1}, \cite{12}) implies that if there exists a deterministic algorithm solving \textit{WSEP}(K,\delta) in \textit{Poly}(< K > + < y > + < \delta > + < \gamma > + < \gamma >) steps then there exists a deterministic algorithm solving \textit{WVL}(K,\gamma,\delta) in \textit{Poly}(< K > + < c > + < \delta > + < \gamma >) steps.

Let us denote as $\text{NSEP}(M,N)$ a compact convex set of separable density matrices $\rho_{A,B} : C^M \otimes C^N \rightarrow C^M \otimes C^N$, \textit{tr}(\rho_{A,B}) = 1 , $M \geq N$. Recall that

$\text{NSEP}(M,N) = CO(\{ x^\dagger \otimes y^\dagger : x \in C^M, y \in C^N; ||x|| = ||y|| = 1 \})$

where $CO(X)$ stands for a convex hull generated by a set $X$. Our goal is to prove that Weak Membership Problem for $\text{NSEP}(M,N)$ is NP-HARD. As we are going to use Yudin - Nemirovskii theorem, it is sufficient to prove that \textit{WVL}(\text{NSEP}(M,N),\gamma,\delta) is NP-HARD respect to the complexity measure ($M + < c > + < \delta > + < \gamma >$) and to show that $\text{NSEP}(M,N)$ is polynomial in $M$.

### 6.2 Geometry of $\text{NSEP}(M,N)$

First, $\text{NSEP}(M,N)$ can be viewed as a proper convex subset of the hyperplane in $R^{N^2M^2}$. The standard euclidean norm in $R^{N^2M^2}$ corresponds to the Frobenius norm for density matrices, i.e. $||\rho||_F = \text{tr}(\rho^\dagger \rho)$. The matrix $\frac{1}{\sqrt{M}} I \in \text{NSEP}(N,N)$ and $\frac{1}{\sqrt{M}} I - xx^\dagger \otimes yy^\dagger ||_F \leq 1$ for all norm one vectors $x, y$. Thus $\text{NSEP}(N,N)$ is covered by the ball $B(\frac{1}{\sqrt{M}} I, 1)$. Next we will show that $B(\frac{1}{\sqrt{M}} I, \frac{1}{\sqrt{N^2}}) \subset \text{NSEP}(N,N)$. Recall that $\rho \in \text{SEP}(N,N)$ iff $\text{tr}(CH(T)\rho) \geq 0$ for all positive operators $T : M \rightarrow M(N)$. This rather straightforward result was first proved in \cite{3}. Let $\rho = \{ A_{i,j} : 1 \leq i,j \leq N \}$ be a block matrix as in (5). For a linear operator $\Psi : M(N) \rightarrow M(N)$ define $\rho^\Psi = \{ \Psi(A_{i,j}) : 1 \leq i,j \leq N \}$.

The following proposition is an easy consequence of the above Woronowicz’s criterium.

**Proposition 6.4:** $\rho \in \text{SEP}(N,N)$ if and only if $\rho^\Psi \succeq 0$ for all linear positive operators $\Psi$ such that $\Psi(I) = I$.

**Lemma 6.5:** Suppose that $\Psi : M(N) \rightarrow M(N)$ is linear positive operator and $\Psi(I) = I$.

Then $||\Psi(\rho)||_F \leq \sqrt{N} ||\rho||_F$.

**Proof:** For $A \in M(N)$ denote $||A||$ the operator norm induced by a standard euclidean norm in $C^N$ (i.e. $||A||$ is the largest singular value of $A$). Recall that $||A||^2 \leq ||A||_F^2 \leq N ||A||^2$. Let $B$ be a hermitian $N \times N$ complex matrix, then $||B||_F \geq ||B|| \geq ||B||_I$. Thus using positivity and linearity we get that $||B||_I \succeq ||B|| \succeq \sqrt{N} ||B||_F$. We conclude that $||\Psi(\rho)||_F \leq ||\rho||_F$ for hermitian $B$ (40)

(The last inequality is in fact true for all matrices $B$).

Let us consider an arbitrary $A \in M(N)$ and decompose it uniquely as $A = H_1 + iH_2$ where matrices $H_1, H_2$ are hermitian: $2H_1 = A + A^\dagger, 2H_2 = -i(A - A^\dagger)$. It is easy to check that

$\frac{1}{2} ||A||_F^2 = ||H_1||_F^2 + ||H_2||_F^2$.

Therefore

$||\Psi(A)||_F^2 = ||\Psi(H_1)||_F^2 + ||\Psi(H_2)||_F^2 \leq N(||\Psi(H_1)||^2 + ||\Psi(H_2)||^2)$.

By (40) we get that

$||\Psi(H_1)||^2 + ||\Psi(H_2)||^2 \leq ||H_1||^2 + ||H_2||^2 \leq ||H_1||_F^2 + ||H_2||_F^2 = ||A||_F^2$.

Putting all together, we finally get that

$||\Psi(A)||_F \leq \sqrt{N} ||A||_F$ (41)

**Theorem 6.6:** Let $\Delta$ be a block hermitian matrix as in (5). If $||\Delta||_F \leq \sqrt{N}$ then the the block matrix $I + \Delta$ is separable.

**Proof:** Let us consider positive linear operator $\Psi : M(N) \rightarrow M(N)$ satisfying $\Psi(I) = I$.

Then $(I + \Delta)^\Psi = I + \Delta^\Psi$. Applying inequality (41) to each block of $\Delta$ and summing all of them we get that $||\Delta^\Psi||_F \leq 1$. As the matrix $\Delta^\Psi$ is hermitian, we conclude that $(I + \Delta)^\Psi \succeq 0$. It follows from Proposition(6.4) that $I + \Delta$ is separable.

**Summarizing,** we get that

$B(\frac{1}{\sqrt{N^2}} I, \frac{1}{\sqrt{N^2}}) \subset \text{NSEP}(N,N) \subset B(\frac{1}{\sqrt{N^2}} I, 1)$

and conclude that $\text{NSEP}(N,N) \succeq \text{Poly}(N)$. It is easy to get from the last inequality that $\text{NSEP}(N,N) \succeq \text{Poly}(\text{max}(M,N)$.

It is left to prove that $\text{WVL}(\text{NSEP}(M,N),\gamma,\delta) \text{ is } \text{NP-HARD}$ respect to the complexity measure ($M + < c > + < \delta > + < \gamma >$).

### 6.3 Proof of Hardness

Let us consider the following hermitian block matrix:

$C = \begin{pmatrix} 0 & A_1 & \ldots & A_{M-1} \\
A_1 & 0 & \ldots & 0 \\
\vdots & \vdots & \ddots & \vdots \\
A_{M-1} & 0 & \ldots & 0 \end{pmatrix}$

i.e. $(i,j)$ blocks are zero if either $i \neq 1$ or $j \neq 1$ and (1,1) block is also zero; $A_1, \ldots, A_{M-1}$ are real symmetric $\times N$ matrices.

**Proposition 6.7:**

$\max_{\rho \in \text{NSEP}(M,N)} \text{tr}(C\rho) = \max_{\rho \in \text{R},||\rho||=1} \sum_{1 \leq i \leq N-1} (y^tA_iy)^2$.

**Proof:** First, by linearity and the fact that the set of extreme points

$\text{Ext}(\text{NSEP}(M,N)) = \{ xx^\dagger \otimes yy^\dagger : x \in C^M, y \in C^N; ||x|| = ||y|| = 1 \}$

we get that

$\max_{\rho \in \text{NSEP}(M,N)} \text{tr}(C\rho) = \max_{xx^\dagger \otimes yy^\dagger : x \in C^M, y \in C^N, ||x|| = ||y|| = 1} \text{tr}(C(xx^\dagger \otimes yy^\dagger))$.

But $\text{tr}(C(yy^\dagger \otimes xx^\dagger)) = \text{tr}(A(y)xx^\dagger)$, where real symmetric $M \times M$ matrix $A(y)$ is defined as follows:

$A(y) = \begin{pmatrix} 0 & a_1 & \ldots & a_{M-1} \\
a_1 & 0 & \ldots & 0 \\
\vdots & \vdots & \ddots & \vdots \\
a_{M-1} & 0 & \ldots & 0 \end{pmatrix}$; $a_i = \text{tr}(A(yy^\dagger))$, $1 \leq i \leq M-1$. 

Thus
\[
\lambda_{\max} A(y) = \max_{\rho \in \text{NSPH}(M,N)} tr(C \rho) = \\
\max_{y \in \mathbb{R}^N, \|y\|=1} tr(C y) = \\
\max_{y \in \mathbb{R}^N, \|y\|=1} (y^T A y)^2.
\]

(Above \(\lambda_{\max} A(y)\) is a maximum eigenvalue of \(A(y)\))

It is easy to see \(A(y)\) has only two non-zero eigenvalues \((d, -d)\), where \(d = \frac{1}{N} \sum_{i=1}^{N-1} (tr(A_i y^2))^2\).

As \(A_i, 1 \leq i \leq N - 1\) are real symmetric matrices we finally get that
\[
\lambda_{\max} A(y) = \max_{\rho \in \text{NSPH}(M,N)} tr(C \rho) = \\
\max_{y \in \mathbb{R}^N, \|y\|=1} \sum_{i=1}^{N-1} (y^T A_i y)^2.
\]

**Proposition (6.7):** The Weak Membership Problem for \(NSPH(M,N)\) is NP-HARD if \(N \leq M \leq \frac{N(N-1)}{2} + 2\).

**Remark 6.10:** It is easy exercise to prove that \(BUDM_{A,B}\) written in block form \((5)\) is real separable iff it is separable and all the blocks in \((5)\) are real symmetric matrices. It follows that this is NP-HARDness to check the positivity of a given operator \([1]\).

**7 Concluding Remarks**

Many ideas of this paper were suggested by [1]. The world of mathematical interconnections is very unpredictable (and thus is so exciting). The main technical result is a very recent breakthrough in Communication Complexity [33] is a rediscovery of particular rank one, case of a general matrix tuples scaling, result proved in [30] with much simpler proof than in [33]. Perhaps this new paper will produce something new in Quantum Communication Complexity.

We still don't know whether there is deterministic poly-time algorithm to check whether a given completely positive operator is rank nondecreasing. And this question is related to lower bounds on \(C_{an}(T')\) provided that Choi's representation \(CH(T')\) is an integer semidefinite matrix.

Theorem (6.9) together with other results from our paper gives a new classical complexity based, insight on the nature of the Quantum Entanglement and, in a sense, closes a long line of research in Quantum Information Theory. Still many open questions remain.

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