The laplacian of a graph as a density matrix: a basic combinatorial approach to separability of mixed states

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Abstract
We study entanglement properties of mixed density matrices obtained from combinatorial laplacian matrices of graphs. We observe that some classes of graphs give rise to entangled (separable) states independently of their labelings.

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1 Introduction
Quantum information is a field which has been expanding rapidly due to the theoretical successes in fast algorithms, super-dense quantum coding, quantum error correction, teleportation and more. Most of these schemes run off entanglement in quantum states. Although entanglement in pure state systems

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is relatively “well understood”, this is much less so in the case of so-called mixed quantum states, which are statistical mixtures of pure quantum states. In this paper we aim to make some beginning steps towards improving this situation by focusing our attention to a restricted class of mixed states. The states we study here may be represented as graphs in a natural way. We hope that in this manner we may be able to make powerful statements at least about the class of states represented by what we call density matrices of graphs. We find, for example, that certain classes of graphs always represent entangled (separable) states. We also find that a number of considered states have an exactly fractional value of their concurrence — a measure of entanglement of formation in small quantum systems. The representation of a limited class of states by graphs leaves hints in the expressions we find for possibly natural ways to extend certain graph-theoretic concepts to more general objects like signed graphs and weighted graphs.

The paper is divided in six sections. In Section 2, we introduce the notion of density matrix of a graph. Theorem 2.1 characterizes the graphs with pure density matrices. Theorem 2.2 shows that the density matrix of a graph can be written as a uniform mixture of pure density matrices of graphs. In Section 3, we consider the von Neumann entropy of density matrices of graphs. Theorem 3.1 calculates the minimum and maximum von Neumann entropy that the density matrix of a graph can have, and determines the graphs for which these values are attained. Theorem 3.4 studies the von Neumann entropy of the disjoint union of cycles. In Section 4, we discuss separability. We label the $n = pq$ vertices of a graph by an ordered pair of indices, where the first index can take $p$ different values and the second index can take $q$ different values. Theorem 4.1 characterizes the graphs with separable density matrices (Lemma 4.3), and star graphs, which represent entangled states (Theorem 4.4). Theorem 4.4 shows that if a graph is a tensor product then its density matrix is separable, and the converse of it is not necessarily true. After having introduced the notion of entangled edge, we prove that if all the entangled edges of a graph on $n = 2p$ vertices form a perfect matching, then the density matrix of the graph is separable in $C^2 \otimes C^p$ (Theorem 4.10). We observe that strongly-regular graphs and transitive graphs can have entangled or separable density matrix (Corollary 4.10). We calculate the concurrence of all graph on four vertices representing entangled states. It turns out that for some of these graphs the value of the concurrence is exactly fractional. In Section 5, we describe the quantum operations that implement graph transformations like adding or deleting a vertex or an edge. In Section 6, we state open problems and conjectures. The paper is relatively self-contained. Our reference on Graph Theory and Quantum Mechanics are [2] and [8], respectively.

2 The density matrix of a graph

2.1 Definition

A graph $G = (V, E)$ is a pair defined in the following way: $V$ (or $V(G)$) is a non-empty and finite set whose elements are called vertices; $E$ (or $E(G)$) is a non-empty set of unordered pairs of vertices, which are called edges. A loop is an edge of the form $\{v_i, v_i\}$, for some vertex $v_i$. We assume that $E(G)$ does not contain only loops. A graph $G$ is said to be on $n$ vertices if $|V(G)| = n$. The adjacency matrix of a graph on $n$ vertices $G$ is an $n \times n$ matrix, denoted by $M(G)$, having rows and columns labeled by the vertices of $G$, and $ij$-th entry defined as follows:

$$[M(G)]_{i,j} = \begin{cases} 1 & \text{if } \{v_i, v_j\} \in E(G); \\ 0 & \text{if } \{v_i, v_j\} \notin E(G). \end{cases}$$

Two distinct vertices $v_i$ and $v_j$ are said to be adjacent if $\{v_i, v_j\} \in E(G)$. The degree of a vertex $v_i \in V(G)$, denoted by $d_G(v_i)$, is the number of edges adjacent to $v_i$. Two adjacent vertices are also said to be neighbours. The degree-sum of $G$ is defined and denoted by $d_G = \sum_{i=1}^{n} d_G(v_i)$. Note that $d_G = 2 |E(G)|$. The degree matrix of $G$ is an $n \times n$ matrix, denoted by $\Delta(G)$, having $ij$-th entry defined as follows:

$$[\Delta(G)]_{i,j} = \begin{cases} d_G(v_i) & \text{if } i = j; \\ 0 & \text{if } i \neq j. \end{cases}$$

The combinatorial laplacian matrix of a graph $G$ (for short, laplacian) is the matrix $L(G) \overset{\text{def}}{=} \Delta(G) - M(G)$. Notice that $L(G)$ does not change if we add or delete loops from $G$. According to our definition of graph, $L(G) \neq 0$. 

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Example 2.1 Let $I_n$ and $J_n$ be the $n \times n$ identity matrix and the $n \times n$ all-ones matrix, respectively. The complete graph on $n$ vertices, denoted by $K_n$, is defined to be the graph with adjacency matrix $(J_n - I_n)$. Then $L(K_n) = (n - 1) I_n - J_n + I_n = nI_n - J_n$.

In Standard Quantum Mechanics (that is the Hilbert space formulation of Quantum Mechanics), the state of a quantum mechanical system associated to the $n$-dimensional Hilbert space $\mathcal{H} \cong \mathbb{C}^n$ is identified with an $n \times n$ positive semidefinite, trace-one, hermitian matrix, called a density matrix. It is easy to observe that the laplacian of a graph is symmetric and positive semidefinite. The laplacian of a graph $G$, scaled by the degree-sum of $G$, has trace one and it is then a density matrix. This observation leads to the following definition.

Definition 2.2 (Density matrix of a graph) The density matrix of a graph $G$ is the matrix

$$\sigma(G) \triangleq \frac{1}{d_G} L(G).$$

Example 2.3 The density matrix of $K_n$ (cfr. Example 2.1) is $\sigma(K_n) = \frac{1}{n(n-1)} (nI_n - J_n)$.

2.2 Pure states and mixed states

Let $\text{tr}(A)$ be the trace of a matrix $A$. A density matrix $\rho$ is said to be pure if $\text{tr}(\rho^2) = 1$, and mixed, otherwise. Theorem 2.4 gives a necessary and sufficient condition on a graph $G$ for $\sigma(G)$ to be pure. We first provide some terminology and state an easy lemma. A graph $G$ is said to have $k$ components, $G_1, G_2, \ldots, G_k$, and in such a case we write $G = G_1 \uplus G_2 \uplus \cdots \uplus G_k$, if there is an ordering of $V(G)$, such that $M(G) = \bigoplus_{i=1}^k M(G_i)$. When $k = 1$, $G$ is said to be connected. From now on, we denote by $\lambda_1(A), \lambda_2(A), \ldots, \lambda_k(A)$ the $k$ different eigenvalues of an Hermitian matrix $A$ in increasing order. The set of the eigenvalues of $A$ together with their multiplicities is called spectrum of $A$.

Lemma 2.1 The density matrix of a graph $G$ has a zero eigenvalue whose multiplicity is equal to the number of components of $G$.

Proof. Given a graph $G$, it is a direct consequence of Definition 2.2 that $\lambda_1(\sigma(G)) = \frac{\lambda_1(L(G))}{d_G}$. It is well-known that $L(G)$ has a zero eigenvalue whose multiplicity is equal to the number of components of $G$ [2].

Theorem 2.4 The density matrix of a graph $G$ is pure if and only if $G = K_2$ or $G = K_2 \uplus v_1 \uplus v_2 \uplus \cdots \uplus v_l$, for some vertices $v_1, v_2, \ldots, v_l$. (These vertices are with or without loops.)

Proof. Let $G$ be a graph on $n$ vertices. Suppose that $\sigma(G)$ is pure. By the definition of pure density matrix, the different eigenvalues of $\sigma(G)$ are $\lambda_1(\sigma(G)) = 0$ and $\lambda_2(\sigma(G)) = 1$. Moreover, $\lambda_1(\sigma(G)) = 0$ has multiplicity $(n - 1)$. Then, by Lemma 2.1 the number of components of $G$ is $(n - 1)$. Since $|V(G)| = n$, it follows that $G = K_2 \uplus v_1 \uplus v_2 \uplus \cdots \uplus v_l$, where $l = n - 2$.

The next definition is based on the theorem.

Definition 2.5 (Pure density matrix of a graph) Let $G$ be a graph on $n \geq 2$ vertices. The density matrix of graph $G$ is said to be pure if $G = K_2 \uplus v_1 \uplus v_2 \uplus \cdots \uplus v_{n-2}$, for some vertices $v_1, v_2, \ldots, v_{n-2}$. (These vertices are with or without loops.)

Example 2.6 The density matrix

$$\sigma(K_2) = \begin{bmatrix} \frac{1}{2} & -\frac{1}{2} \\ -\frac{1}{2} & \frac{1}{2} \end{bmatrix}$$

is pure. In fact, $\lambda_1(\sigma(K_2)) = 0$ and $\lambda_2(\sigma(K_2)) = 1$.

A graph $H$ is said to be a factor of a graph $G$, if $V(H) = V(G)$ and there exists a graph $H'$ such that $V(H') = V(G)$ and $M(G) = M(H) + M(H')$.

Theorem 2.7 The density matrix of a graph is a uniform mixture of pure density matrices.
Proof. Let $G$ be a graph on $n$ vertices $v_1, v_2, ..., v_n$, having edges \{ $v_i, v_j$, \{ $v_i, v_j$, \{ $v_i, v_j$, \{ $v_i, v_j$, \} with $1 \leq i_1, i_2, j_1, j_2, ...$, $i_m, j_m \leq n$. Let $H_{i_k j_k}$ be the factor of $G$ such that

$$
[M(H_{i_k j_k})]_{u,w} = \begin{cases} 
1 & \text{if } u = i_k \text{ and } w = j_k \text{ or } w = i_k \text{ and } u = j_k; \\
0 & \text{otherwise}.
\end{cases}
$$

(1)

By Theorem 2.4 the density matrix $\sigma(H_{i_k j_k}) = \frac{1}{2} (\Delta(H_{i_k j_k}) - M(H_{i_k j_k}))$ is pure. Since

$$
\Delta(G) = \sum_{k=1}^m \Delta(H_{i_k j_k}) \quad \text{and} \quad M(G) = \sum_{k=1}^m M(H_{i_k j_k}),
$$

we can write

$$
\sigma(G) = \frac{1}{2m} (\Delta(G) - M(G)) = \frac{1}{m} \sum_{k=1}^m \sigma(H_{i_k j_k}),
$$

(2)

which is then a uniform mixture of pure density matrices. ■

Example 2.8 Consider a graph $G$ defined as follows: $V(G) = \{1, 2, 3\}$ and $E(G) = \{e = \{1, 2\}, f = \{2, 3\}\}$. Then

$$
M(H_{1,2_e}) = \begin{bmatrix}
0 & 1 & 0 \\
1 & 0 & 0 \\
0 & 0 & 0 
\end{bmatrix}, \quad M(H_{2,3_f}) = \begin{bmatrix}
0 & 0 & 0 \\
0 & 0 & 1 \\
0 & 0 & 0 
\end{bmatrix}
$$

and

$$
\sigma(G) = \frac{1}{2} (\sigma(H_{1,2_e}) + \sigma(H_{2,3_f})) = \begin{bmatrix}
-\frac{1}{4} & -\frac{1}{4} & 0 \\
-\frac{1}{4} & -\frac{1}{4} & -\frac{1}{4} \\
0 & -\frac{1}{4} & -\frac{1}{4} 
\end{bmatrix}.
$$

3 Von Neumann entropy

The von Neumann entropy of an $n \times n$ density matrix $\rho$ is $S(\rho) \overset{\text{def}}{=} -\sum_{i=1}^n \lambda_i(\rho) \log_2 \lambda_i(\rho)$. It is conventional to define $0 \log_2 0 = 0$. The von Neumann entropy is a quantitative measure of mixedness of the density matrix $\rho$.

Remark 3.1 The $q$-entropy of an $n \times n$ density matrix $\rho$ is $(\text{tr}(\rho^q))^{1/q}$. $q$-entropies are a family of measures of mixedness for density matrices. In general, in the limit $q \rightarrow \infty$, $q$-entropies depend only on the largest eigenvalues of $\rho$, and we have $\lim_{q \rightarrow \infty} (\text{tr}(\rho^q))^{1/q} = \lambda_n(\rho)$. This eigenvalue can be considered itself as a measure of mixedness $\boxed{\lambda_n}$. If $\rho$ is the density matrix of a graph a tight upper-bound on $\lambda_n(\rho)$ is known $\boxed{\lambda_n}$.

3.1 Maximum and minimum

Theorem 3.2 Let $G$ be a graph on $n$ vertices. Then

1. $\max_G S(\sigma(G)) = \log_2 (n - 1) = S(\sigma(K_n))$;

2. $\min_G S(\sigma(G)) = 0$ and this value is attained if $\sigma(G)$ is pure.

Proof. (Proof of 1) By Lemma 2.4 $\sigma(G)$ has an eigenvalue zero with multiplicity at least one. Since $G$ is on $n$ vertices, the support of $\sigma(G)$ has dimension less or equal to $(n - 1)$. Any $n \times n$ density matrix having dimension of support less or equal to $(n - 1)$, can not have von Neumann entropy greater than $\log_2 (n - 1)$. The eigenvalues of $\sigma(K_n)$ are $\lambda_1(\sigma(K_n)) = 0$, with multiplicity 1, and $\lambda_2(\sigma(K_n)) = \frac{1}{(n-1)}$, with multiplicity $(n - 1)$. Then

$$
S(\sigma(K_n)) = -\sum_{n-1}^1 \frac{1}{n-1} \log_2 \frac{1}{n-1} = \log_2 (n - 1).
$$

(Proof of 2) Since $G$ is a graph on $n$ vertices, the maximum multiplicity of the zero eigenvalue of $\sigma(G)$ is $(n - 1)$; the other eigenvalue of $\sigma(G)$ is necessarily one. This is the case when $\sigma(G)$ is pure. When $\sigma(G)$ is pure, $S(\sigma(G)) = 0$. ■
3.2 Regular graphs

Two graphs $G$ and $H$ are said to be $L$-cospectral if $L(G)$ and $L(H)$ have the same spectrum; $\sigma$-cospectral if $\sigma(G)$ and $\sigma(H)$ have the same spectrum. Two graphs $G$ and $H$ are said to be isomorphic, and in such a case we write $G \cong H$, if there is an isomorphism between $V(G)$ and $V(H)$, that is there is a permutation matrix $P$, such that $PM(G)P^T = M(H)$. If $G \cong H$ then $G$ and $H$ are $L$-cospectral and $\sigma$-cospectral, but the converse is not necessarily true. Two graphs are $L$-cospectral and $\sigma$-cospectral if and only if they have the same degree sum. Now, a graph is said to be regular if each of its vertices has the same degree. A $d$-regular graph is a regular graph whose degree of the vertices is $d$. If $G$ is $d$-regular graph on $n$ vertices, then $\lambda_i(L(G)) = d - \lambda_i(M(G))$ and $\lambda_i(\sigma(G)) = \frac{d - \lambda_i(M(G))}{dn}$. Because we do not have any closed form of the series $\sum_{i=1}^{k} m_i (\sigma_i) \log_2 (\sigma_i)$, we have

$$S(\sigma(G)) = - \sum_{i=1}^{k} m_i (\sigma_i) \log_2 (\sigma_i) = - \frac{1}{d \cdot n} \sum_{i=1}^{k} m_i [(d - \mu_i) \log_2 (d - \mu_i)] + \frac{\log_2 (d \cdot n)}{d \cdot n} \sum_{i=1}^{k} m_i (d - \mu_i)$$

$$= - \frac{1}{d \cdot n} \sum_{i=1}^{k} m_i [(d - \mu_i) \log_2 (d - \mu_i)] + \log_2 (d \cdot n),$$

3.3 Cycles

Let $\Gamma$ be a finite group. Let $\rho_{reg}(g)$ be the (left) regular permutation representation of an element $g \in \Gamma$. The (left) Cayley graph of $\Gamma$ with respect to $S$, denoted by $X(\Gamma, S)$, is defined to be the graph with adjacency matrix $M(X(\Gamma, S)) = \sum_{s \in S} \rho_{reg}(s)$. Notice that $X(\Gamma, S)$ is connected if and only if $S$ generates $\Gamma$.

Example 3.3 Let $\Gamma = \mathbb{Z}_n$ be the group of the integers modulo $n$ and let $S = \{1, n - 1\} \subset \Gamma$. Then $M(G) = \rho_{reg}(1) + \rho_{reg}(n - 1)$. Since $S$ generates $\Gamma$, the graph $G$ is connected. The $n$-cycle, denoted by $C_n$, is a graph on $n$ vertices $v_1, v_2, ..., v_n$ and with $n$ edges $\{v_1, v_2\}, \{v_2, v_3\}, ..., \{v_{n-1}, v_n\}, \{v_n, v_1\}$. Hence, $G \cong C_n$.

Theorem 3.4 Let $G_k = X(\Gamma, S_k)$ be a Cayley graph, where $\Gamma = \mathbb{Z}_n$ and $S_k = \{k, n - k\} \subset \Gamma$. Then

1. $\max_{G_k} S(\sigma(G_k)) = S(\sigma(C_\frac{2n}{p}))$, that is when $\gcd(k, n) = 1$;
2. $\min_{G_k} S(\sigma(G_k)) = S(\sigma(C_\frac{2n}{p}))$.

Proof. (Proof of 1) We begin by observing that, given $S_k = \{k, n - k\}$, with $k = n/p$, $G_k = C_p \cup C_p \cup \cdots \cup C_p$, where $C_p$ is repeated $k$-times. This indicates that the eigenvalues of $M(G_k)$ are the eigenvalues of $M(C_p)$, each repeated $k$ times: $\lambda_j(\sigma(G_k)) = \frac{2n}{p} \lambda_j(\sigma(C_p))$, where $1 \leq j \leq p$, and each $\lambda_j(\sigma(G_k))$ has multiplicity $k$. Since, it is well-known that $\lambda_j(M(C_p)) = 2 \cos (2\pi j/p)$, where $1 \leq j \leq p$, we have $\lambda_j(\sigma(C_p)) = \frac{2 - 2\cos(2\pi j/p)}{2p} = \frac{2 \sin^2(\pi j/p)}{n}$ and $\lambda_j(\sigma(G_k)) = \frac{2 \sin^2(\pi j/p)}{n}$. By writing

$$A_p(j) = \sin^2\left(\frac{\pi j}{p}\right) \log_2 \left(\sin^2\left(\frac{\pi j}{p}\right)\right),$$

the von Neumann entropy of $\sigma(G_k)$ is given by

$$S(\sigma(G_k)) = -k \sum_{j=1}^{p} \frac{2}{n} A_p(j) = \log_2 n - 1 - \frac{2}{n} \sum_{j=1}^{p} A_p(j).$$

Because we do not have any closed form of the series $- \sum_{j=1}^{p} A_p(j)$, we use the following approximation, which is very good for large $p$:

$$- \sum_{j=1}^{p} A_p(j) \simeq - \frac{2}{p} \int_{0}^{\pi} \sin^2 x \log_2 (\sin^2 x) dx = p \cdot C, \quad \text{where} \quad C = \left(1 - \frac{\log_2 e}{2}\right) \simeq 0.2787.$$  

(3)
If \( p = 1, 2, - \sum_{j=1}^{p} A_p(j) = 0 \). So, if \( p = 1, 2 \) and \( n \) is even, \( S(\sigma(G_{n/2})) = \log_2 n - 1 \). With the use of Equation 8 we obtain \( S(\sigma(G_k)) \approx \log_2 n - 1 + 2C/k \). If \( l = n/q \) then \( S(\sigma(G_k)) - S(\sigma(G_l)) \approx 2C (\frac{1}{k} - \frac{1}{l}) \).

It follows that: \( S(\sigma(G_k)) > S(\sigma(G_l)) \) if \( l > k \); \( S(\sigma(G_k)) < S(\sigma(G_l)) \) if \( l < k \). When \( k = 1 \) then \( G_k = C_n \). Therefore, \( S(\sigma(C_n)) > S(\sigma(G_l)) \), for all \( l > 1 \). (Proof of 2) By the reasoning above, it is sufficient to observe that \( S(\sigma(G_{2^n})) = \log_2 n - 1 \).

**Example 3.5** In the table below, the values of the von Neumann entropy of the Cayley graphs \( X(Z_{12}, S) \), where \( |S| = 2 \) are given:

| \( S \) | \( \sigma(G) \) |
| --- | --- |
| \( X(Z_{12}, \{1, 11\}) \) | 3.571 |
| \( X(Z_{12}, \{2, 10\}) \) | 3.126 |
| \( X(Z_{12}, \{3, 9\}) \) | 3.084 |
| \( X(Z_{12}, \{4, 8\}) \) | 3.000 |
| \( X(Z_{12}, \{6\}) \) | 2.585 |

### 4 Separability

Let \( S_A \) and \( S_B \) be two quantum mechanical systems, associated to the \( p \)-dimensional and \( q \)-dimensional Hilbert spaces \( \mathcal{H}_A \cong \mathbb{C}^p \) and \( \mathcal{H}_B \cong \mathbb{C}^q \), respectively. The composite system \( S_A \times S_B \), which consists of the subsystems \( S_A \) and \( S_B \), is associated to the Hilbert space \( \mathcal{C}^p \otimes \mathcal{C}^q \), where “\( \otimes \)” denotes tensor product. The density matrix \( \rho_{AB} \) of \( S_{AB} \) is said to be separable if \( \rho_{AB} = \sum_{i=1}^{n} \omega_i \rho_A^{(i)} \otimes \rho_B^{(i)} \), where \( \omega_i \geq 0 \), for every \( i = 1, 2, \ldots, n \), and \( \sum_{i=1}^{n} \omega_i = 1 \); \( \rho_A^{(i)} \) and \( \rho_B^{(i)} \) are density matrices acting on \( \mathcal{H}_A \) and \( \mathcal{H}_B \), respectively. A density matrix \( \rho_{AB} \) is said to be entangled if it is not separable. In Dirac notation, a unit vector in a Hilbert space \( \mathcal{H} \cong \mathbb{C}^n \) is denoted by \( |\psi\rangle \), where \( \psi \) is a label; given the vectors \( |\varphi\rangle, |\psi\rangle \in \mathcal{H} \), the linear functional sending \( |\psi\rangle \) to the inner product \( \langle \psi | \varphi \rangle \) is denoted by \( \langle \varphi | \psi \rangle \). We write \( |\psi\rangle \langle \phi| \) for the tensor product \( |\psi\rangle \otimes |\phi\rangle \). A vector of the form \( |\psi\rangle \langle \phi| \) is called product state. For any unit vector \( |\psi\rangle \in \mathcal{H} \), the projector on \( |\psi\rangle \) is the hermitian matrix \( |\psi\rangle \langle \psi| \) which we denote by \( P \approx |\psi\rangle \langle \psi| \).

#### 4.1 Tensor product of graphs

The tensor product of graphs \( G \) and \( H \) (also known in literature as strong product, cardinal product, etc.), denoted by \( G \otimes H \), is the graph whose adjacency matrix is \( M(G \otimes H) = M(G) \otimes M(H) \). Whenever we consider a graph \( G \otimes H \), where \( G \) is on \( p \) vertices and \( H \) is on \( q \) vertices, the separability of \( \sigma(G \otimes H) \) is described with respect to the Hilbert space \( \mathcal{H}_G \otimes \mathcal{H}_H \), where \( \mathcal{H}_G \) is the space spanned by the orthonormal basis \( \{|u_1\rangle, |u_2\rangle, \ldots, |u_p\rangle\} \) associated to \( V(G) \), and \( \mathcal{H}_H \) is the space spanned by the orthonormal basis \( \{|w_1\rangle, |w_2\rangle, \ldots, |w_q\rangle\} \) associated to \( V(H) \). The vertices of \( G \otimes H \) are taken as \( u_1w_1, u_1w_2, \ldots, u_1w_q, u_2w_1, u_2w_2, \ldots, u_2w_q \). We associate \( |u_1\rangle |w_1\rangle \) to \( u_1w_1 \), \( |u_1\rangle |w_2\rangle \) to \( u_1w_2 \), \ldots, \( |u_p\rangle |w_q\rangle \) to \( u pw_q \). In conjunction with this, whenever we talk about separability of any graph \( G \) on \( n \) vertices, \( v_1, v_2, \ldots, v_n \), we consider it in the space \( \mathbb{C}^n \otimes \mathbb{C}^n \), where \( n = pq \). The vectors \( |v_1\rangle, |v_2\rangle, \ldots, |v_n\rangle \) are taken as follows: \( |v_1\rangle = |u_1\rangle |w_1\rangle, |v_2\rangle = |u_1\rangle |w_2\rangle, \ldots, |v_q\rangle = |u_1\rangle |w_q\rangle, |v_{q+1}\rangle = |u_2\rangle |w_1\rangle, |v_{q+2}\rangle = |u_2\rangle |w_2\rangle, \ldots, |v_{2q}\rangle = |u_2\rangle |w_q\rangle \). We make use of the notion of partial transpose of a density matrix. Let us consider a \( pq \times pq \) density matrix \( \rho_{AB} \) acting on \( \mathcal{C}^p \otimes \mathcal{C}^q \). Let \( |v_1\rangle, |v_2\rangle, \ldots, |v_p\rangle \) and \( \{|w_1\rangle, |w_2\rangle, \ldots, |w_q\rangle\} \) be orthonormal bases of \( \mathcal{C}^p \) and \( \mathcal{C}^q \), respectively. The partial transpose of \( \rho_{AB} \) with respect to the system \( S_B \) is the \( pq \times pq \) matrix, denoted by \( \rho_{AB}^T \), and with \( (i, j; i', j') \)-th entry defined as follows: \( \rho_{AB}^T_{ij;i'j'} = \langle v_i | w_{i'} \rangle \rho_{AB} |v_{i'} \rangle \langle w_i| \), where \( 1 \leq i, i' \leq p \) and \( 1 \leq j, j' \leq q \). Regarding separability of \( \rho_{AB} \) we have the following criterion 10, 11:

**Peres-Horodecki Criterion (PH)** If \( \rho \) is a density matrix acting on \( \mathbb{C}^2 \otimes \mathbb{C}^2 \) or \( \mathbb{C}^2 \otimes \mathbb{C}^3 \), then \( \rho \) is separable if and only if \( \rho^T \) is positive semidefinite.

**Theorem 4.1** Let \( G \) and \( H \) be two graphs on \( n = p \cdot q \) vertices. If \( \sigma(G) \) is entangled in \( \mathbb{C}^p \otimes \mathbb{C}^q \) and \( G \cong H \) then \( \sigma(H) \) is not necessarily entangled in \( \mathbb{C}^p \otimes \mathbb{C}^q \).

**Proof.** Let \( G \) be a graph on the vertices 1, 2, 3 and 4, having edges \{1, 2\}, \{2, 3\} and \{3, 4\}. We associate to \( G \) the following orthonormal basis: \( \{|1\rangle = |1\rangle_A |1\rangle_B, |2\rangle = |2\rangle_A |1\rangle_B, |3\rangle = |1\rangle_A |2\rangle_B, |4\rangle = \)
where $1 \leq \sigma$. Let

$$\text{Proof.}$$

Notice that $\sigma$ and so $\sigma(H)$ is entangled. Consider the graph $H \cong G$. The edges of $H$ are $\{1, 4\}, \{4, 3\}$ and $\{3, 2\}$. We associate to $H$ the above orthonormal basis. We then have

$$\langle \sigma(H) \rangle_{ab} = \frac{1}{6} \begin{bmatrix} 1 & 0 & 0 & -1 \\ -1 & 2 & 0 & 0 \\ 0 & 1 & -1 & 0 \\ 0 & -1 & 2 & -1 \\ -1 & 0 & -1 & 2 \end{bmatrix} = \sigma(H),$$

and so $\sigma(H)$ is separable. ■

Lemma 4.1 The density matrix of the tensor product of two graphs is separable.

**Proof.** Let $G$ be a graph on $n$ vertices, $v_1, v_2, \ldots, v_n$, and $m$ edges, $\{v_{i_1}, v_{j_1}\}, \{v_{i_2}, v_{j_2}\}, \ldots, \{v_{i_m}, v_{j_m}\}$, where $1 \leq i_1, j_1, i_2, j_2, \ldots, i_m, j_m \leq n$. Let $G'$ be a graph on $p$ vertices, $v_1, v_2, \ldots, v_p$, and $q$ edges, $\{v'_{s_1}, v'_{k_1}\}, \{v'_{s_2}, v'_{k_2}\}, \ldots, \{v'_{s_q}, v'_{k_q}\}$, where $1 \leq s_1, t_1, s_2, t_2, \ldots, s_q, t_q \leq p$. By Theorem 2.7 (Equation 2), we can write

$$\sigma(G) = \frac{1}{m} \sum_{k=1}^{m} \sigma(H_{i_k,j_k})$$

and $\sigma(G') = \frac{1}{q} \sum_{l=1}^{q} \sigma(L_{s_l,t_l})$.

where $H_{i_k,j_k}$ and $L_{s_l,t_l}$ are defined according to Equation [4]. So,

$$\sigma(G \otimes G') = \frac{1}{d_{G \otimes G'}} [\Delta(G \otimes G') - M(G \otimes G')]$$

$$= \frac{1}{d_{G \otimes G'}} \sum_{k=1}^{m} \sum_{l=1}^{q} [\Delta(H_{i_k,j_k} \otimes L_{s_l,t_l}) - M(H_{i_k,j_k} \otimes L_{s_l,t_l})]$$

$$= \frac{1}{m \cdot q} \sum_{k=1}^{m} \sum_{l=1}^{q} \sigma(H_{i_k,j_k} \otimes L_{s_l,t_l})$$

$$= \frac{1}{m \cdot q} \sum_{k=1}^{m} \sum_{l=1}^{q} \frac{1}{2} \sigma(H_{i_k,j_k} \otimes \sigma(L_{s_l,t_l}) + \sigma(H_{i_k,j_k}) \otimes \sigma(L_{s_l,t_l})],$$

where

$$\sigma^+(H_{i_k,j_k}) \equiv \Delta(H_{i_k,j_k}) - \sigma(H_{i_k,j_k}) \quad \text{and} \quad \sigma^+(L_{s_l,t_l}) \equiv \Delta(L_{s_l,t_l}) - \sigma(L_{s_l,t_l}).$$

Notice that $\sigma^+(H_{i_k,j_k})$ and $\sigma^+(L_{s_l,t_l})$ are density matrices. Let

$$\sigma^+(G) \equiv \frac{1}{m} \sum_{k=1}^{m} \sigma^+(H_{i_k,j_k})$$

and

$$\sigma^+(G') \equiv \frac{1}{q} \sum_{l=1}^{q} \sigma^+(L_{s_l,t_l}).$$

Then

$$\sigma(G \otimes G') = \frac{1}{2} [\sigma(G) \otimes \sigma^+(G') + \sigma^+(G) \otimes \sigma(G')].$$

Since each of $\sigma(G), \sigma^+(G'), \sigma^+(G)$ and $\sigma(G')$ is a uniform mixture of density matrices, then $\sigma(G \otimes G')$ is separable. ■

We associate to the vertices, $v_1, v_2, \ldots, v_n$, of a graph $G$ an orthonormal basis $\{|v_1\}, |v_2\rangle, \ldots, |v_n\rangle\}$. In terms of this basis, the $uv$-th elements of the matrices $\sigma(H_{i_k,j_k})$ and $\sigma^+(H_{i_k,j_k})$ are given by $\langle u | \sigma(H_{i_k,j_k}) | v \rangle$ and $\langle u | \sigma^+(H_{i_k,j_k}) | v \rangle$, respectively. In this basis

$$\sigma(H_{i_k,j_k}) = P_2 \langle |v_{i_k}\rangle - |v_{j_k}\rangle \rangle \quad \text{and} \quad \sigma^+(H_{i_k,j_k}) = P_2 \langle |v_{i_k}\rangle + |v_{j_k}\rangle \rangle.$$
Lemma 4.2 For any \( n = p \cdot q \), the density matrix \( \sigma(K_n) \) is separable in \( \mathbb{C}^p \otimes \mathbb{C}^q \).

Proof. Let \( v_1, v_2, \ldots, v_n \) be the vertices of \( K_n \), with \( n = p \cdot q \). Let us consider the following two orthonormal bases \( \{|u_1\}, |u_2\rangle, \ldots, |u_p\rangle \) and \( \{|w_1\}, |w_2\rangle, \ldots, |w_q\rangle \) of \( \mathbb{C}^p \) and \( \mathbb{C}^q \), respectively. For all \( i = 1, 2, \ldots, n \), we then write \( |v_i\rangle = |u_{s_i}+1\rangle |w_{s_i}'\rangle \), where \( i = sq + s', 0 \leq s \leq p-1 \) and \( 1 \leq s' \leq q \). By making use of this basis, we can write \( \sigma(H_{isjk}) = P[\frac{1}{\sqrt{2}}(|u_{s_i+1}\rangle |w_{s_i}'\rangle - |u_{s_k+1}\rangle |w_{s_k}'\rangle)] \), where \( i_k = skq + s'_k, j_k = tkq + t'_k, 0 \leq s_k, t_k \leq p-1 \) and \( 0 \leq s'_k, t'_k \leq q \). By Equation 2:

\[
\sigma(K_n) = \frac{1}{m} \sum_{k=1}^{m} P[\frac{1}{\sqrt{2}}(|u_{s_k+1}\rangle |w_{s_k}'\rangle - |u_{t_k+1}\rangle |w_{t_k}'\rangle)].
\]

Since \( M(K_n) = J_n - I_n \), whenever there is a term like \( P[\frac{1}{\sqrt{2}}(|u_{s_k+1}\rangle |w_{s_k}'\rangle - |u_{t_k+1}\rangle |w_{t_k}'\rangle)] \) in the sum above, there is a term like \( P[\frac{1}{\sqrt{2}}(|u_{s_k+1}\rangle |w_{s_k}'\rangle - |u_{t_k+1}\rangle |w_{t_k}'\rangle)] \). The uniform mixture of these two terms gives arise to the separable density matrix \( \frac{1}{2} P[|u^+\rangle |w^-\rangle] + \frac{1}{2} P[|u^-\rangle |w^+\rangle] \), where \( |u^\pm\rangle = \frac{1}{\sqrt{2}}(|u_{s_k+1}\rangle \pm |u_{t_k+1}\rangle) \) and \( |w^\pm\rangle = \frac{1}{\sqrt{2}}(|w_{s_k}'\rangle \pm |w_{t_k}'\rangle) \). This shows that \( \sigma(K_n) \) is separable. ■

Remark 4.2 Separability of \( \sigma(K_n) \) does not depend upon the labeling of \( V(K_n) \). Given a graph \( G \), an isomorphism from \( V(G) \) to \( V(G) \) is called automorphism. Under composition of maps, the set of the automorphisms of \( G \) form a group, denoted by \( Aut(G) \), and called automorphism group of \( G \). Note that the separability properties of \( G \) are invariant under \( Aut(G) \). Since \( \sigma(K_n) \) is separable, and since the automorphism group of \( K_n \) is the symmetric group \( S_n \), \( G \cong \sigma(K_n) \) is also separable.

Example 4.3 Consider the graph \( K_4 \). The vertices of \( K_4 \) are denoted by 1, 2, 3 and 4. We associate to these vertices the orthonormal basis \( \{|1\} = |1\rangle |1\rangle, |2\rangle = |1\rangle |2\rangle, |3\rangle = |2\rangle |1\rangle, \) \( |4\rangle = |2\rangle |2\rangle \). In terms of this basis \( \sigma(K_4) \) can be written as

\[
\sigma(K_4) = \frac{1}{12} \begin{bmatrix}
3 & -1 & -1 & -1 \\
-1 & 3 & -1 & -1 \\
-1 & -1 & 3 & -1 \\
-1 & -1 & -1 & 3
\end{bmatrix} = \frac{1}{6} P[|1\rangle |1\rangle - |2\rangle |2\rangle] + \frac{1}{6} P[|1\rangle |2\rangle - |2\rangle |1\rangle]
\]

\[
+ P[|1\rangle |2\rangle - |2\rangle |1\rangle] + P[|2\rangle |1\rangle - |2\rangle |2\rangle] + \frac{1}{3} \left( \frac{1}{2} P[|1\rangle |1\rangle - |2\rangle |2\rangle] \right) + \frac{1}{2} P[|1\rangle |2\rangle - |2\rangle |1\rangle] \right). \]

Each of the first four terms in the above expression is a projector on a product state, while the last two terms give arise to the separable density matrix \( \frac{1}{2} P[|\pm\rangle |\pm\rangle] + \frac{1}{2} P[|\pm\rangle |\mp\rangle] \), where \( |\pm\rangle \coloneqq \frac{1}{\sqrt{2}} (|1\rangle \pm |2\rangle) \). Thus \( \sigma(K_4) \) is separable in \( \mathbb{C}^2 \otimes \mathbb{C}^2 \).

Lemma 4.3 The complete graph on \( n > 1 \) vertices is not a tensor product of graphs.

Proof. It is clear that, if \( n \) is prime, \( K_n \) is not a tensor product of graphs. We then assume that \( n \) is not a prime. Suppose that there exist graphs \( G \) and \( H \), respectively on \( p \) and \( s \) vertices, such that \( K_{ps} = G \otimes H \). Let \( |E(G)| = q \) and \( |E(H)| = t \). Then, by the degree-sum formula, \( 2q \leq p(p-1) \) and \( 2t \leq s(s-1) \). So, \( 2q \cdot 2t \leq p(p-1) s(s-1) = ps(p - s + 1) \). Now, observe that \( |V(G \otimes H)| = ps \) and \( |E(G \otimes H)| = 2(qt) \). Therefore, \( G \otimes H \) if and only if \( ps(p - 1) = 2 \cdot 2qt \), which is true if and only if \( p = s = 1 \). This occurs only when \( n = 1 \). ■

Theorem 4.4 Given a graph \( G \otimes H \), the density matrix \( \sigma(G \otimes H) \) is separable. However if a density matrix \( \sigma(L) \) is separable it does not necessarily mean that \( L = G \otimes H \), for some graphs \( G \) and \( H \).

Proof. The theorem follows from Lemma 14 together with Lemma 12 and Lemma 13. ■

Remark 4.5 Not always is \( \sigma(G) \otimes \sigma(G) \) the density matrix of a graph. However, we observe the following. A weighted graph is a graph with each of its edges labeled by a real number. Let \( W \) be a weighted graph defined as follows: \( V = \{ij' : i, j'=1,2\} \); the edges of \( W \) are \( \{11', 12', \ldots, 11', 21', 12', 22', 21', 22', 11', 22', 12', 21'\} \).
Consider the graph \( W := W(1, 2, 1', 2') \). Let \( G \) be a graph on the vertices \( 1, 2, ..., n \) and the edges \( \{i_1, j_1\}, \{i_2, j_2\}, ..., \{i_m, j_m\} \), where \( 1 \leq i_k, j_k \leq n \). Let \( H \) be a graph on the vertices \( 1', 2', ..., p' \) and the edges \( \{s'_1, t'_1\}, \{s'_2, t'_2\}, ..., \{s'_q, t'_q\} \), then

\[
\sigma (G) \otimes \sigma (G) = \frac{1}{mq} \sum_{k=1}^{m} \sum_{q=1}^{q} \delta (i_k, j_k; s'_q, t'_q).
\]

### 4.2 Stars

A star graph (for short, star) on \( n \) vertices \( v_1, v_2, ..., v_n \), denoted by \( K_{1,n-1} \), is the graph whose set of edges is \( \{\{v_i, v_j\} : i = 2, 3, ..., n\} \). Quantum dynamics on stars has been studied in the context of quantum chaos [6].

**Definition 4.6 (Entangled edge)** Let \( G \) be a graph on \( n = pq \) vertices, \( v_1, v_2, ..., v_n \). The \( k \)-th edge \( \{v_i, v_j\} \) of \( G \) is identified with the pure density matrix \( P \frac{1}{\sqrt{2}} (|v_i\rangle - |v_j\rangle) \), where \( |v_i\rangle = |u_{ik+1}\rangle |w_{ik}\rangle \) and \( |v_j\rangle = |u_{jk+1}\rangle |w_{jk}\rangle \), with \( i_k = s_k q + t_k \) and \( j_k = s'_k q + t'_k \), \( 0 \leq s_k, s'_k \leq p - 1 \) and \( 1 \leq t_k, t'_k \leq q \). The vectors \( |u_i\rangle \)'s and \( |u_j\rangle \)'s form orthonormal bases of \( \mathbb{C}^p \) and \( \mathbb{C}^q \), respectively. Let \( \{|w_1\rangle, |w_2\rangle, ..., |w_p\rangle\} \) and \( \{|w_1\rangle, |w_2\rangle, ..., |w_q\rangle\} \) be orthonormal bases of \( \mathbb{C}^p \) and \( \mathbb{C}^q \), respectively. So,

\[
\sigma (G) = \frac{1}{n - 1} \sum_{k=1}^{n} P \frac{1}{\sqrt{2}} (|w_1\rangle - |w_j\rangle)
\]

where \( k = s_k q + t_k \), \( 0 \leq s_k \leq p - 1 \) and \( 1 \leq t_k \leq q \). Thus

\[
\sigma (G) = \frac{1}{n - 1} \left\{ \sum_{j=2}^{q} P |w_1\rangle \frac{1}{\sqrt{2}} (|w_1\rangle - |w_j\rangle) + \sum_{i=2}^{p} P |w_1\rangle \frac{1}{\sqrt{2}} (|w_1\rangle - |w_j\rangle) \right\}.
\]

Consider now the following two dimensional projectors: \( P = |w_1\rangle \langle w_1| + |w_2\rangle \langle w_2| \) and \( Q = |w_1\rangle \langle w_1| + |w_2\rangle \langle w_2| \). Then

\[
(P \otimes Q) \sigma (G) (P \otimes Q) = \frac{1}{n - 1} \left\{ \frac{n - 1}{2} P |w_1\rangle \langle w_1| + P \frac{1}{\sqrt{2}} (|w_1\rangle - |w_2\rangle) \langle w_1| \right\}
\]

In the basis \( \{|w_1\rangle, |w_2\rangle, |w_2\rangle, |w_1\rangle, |w_2\rangle, |w_2\rangle\} \), we have

\[
[(P \otimes Q) \sigma (G) (P \otimes Q)]^\mathbb{R} = \frac{1}{n - 1}
\]

The eigenvalues of the above matrix are \( \left\{ \left\lfloor \frac{n - 1}{2} \right\rfloor, \left\lfloor \frac{n - 1}{2} \right\rfloor, 0 \right\rfloor \). As \( n \geq 4 \), \( \frac{1}{4}(1 - \sqrt{(n - 1)^2 + 8/(n - 1)}) < 0 \). Hence, by Criterion [11], the matrix \( (P \otimes Q) \sigma (G) (P \otimes Q) \) is entangled and then also \( \sigma (G) \) is entangled (Note that this matrix is not normalized.)
Example 4.8 Consider the graph $G = \{\{1, 2, 3, 4\}, \{\{1, 2\}, \{1, 3\}, \{1, 4\}, \{2, 3\}\}$. We test the separability of $\sigma(G)$ in $\mathbb{C}_A^2 \otimes \mathbb{C}_B^2$, with respect to the orthonormal basis $\{|1\rangle, |1\rangle |2\rangle, |2\rangle |1\rangle, |2\rangle |2\rangle\}$. In this basis,

$$
\sigma(G) = \frac{1}{8} \begin{pmatrix}
3 & -1 & -1 & -1 \\
-1 & 2 & -1 & 0 \\
-1 & -1 & 2 & 0 \\
-1 & 0 & 0 & 1
\end{pmatrix}.
$$

It can be easily verified that $(\sigma(G))^T_B = \sigma(G)$. As a consequence, all the eigenvalues of $(\sigma(G))^T_B$ are nonnegative, as $\sigma(G)$ is positive semidefinite. It follows from Criterion 4.1 that $\sigma(G)$ is separable in $\mathbb{C}^2 \otimes \mathbb{C}^2$. Consider now the star $K_{1,3} = \{\{1, 2, 3, 4\}, \{\{1, 2\}, \{1, 3\}, \{1, 4\}\}\}$. Observe that $K_{1,3}$ is obtained from $G$ with the removal of the edge $\{2, 3\}$. With respect to the above mentioned basis, we have

$$
\sigma(K_{1,3}) = \frac{1}{8} \begin{pmatrix}
3 & -1 & -1 & -1 \\
-1 & 1 & 0 & 0 \\
-1 & 0 & 1 & 0 \\
-1 & 0 & 0 & 1
\end{pmatrix}
$$

and $(\sigma(K_{1,3}))^T_B = \frac{1}{8} \begin{pmatrix}
3 & -1 & -1 & 0 \\
-1 & 1 & 1 & 0 \\
-1 & 1 & 1 & 0 \\
0 & 0 & 0 & 1
\end{pmatrix}$.

The eigenvalues of $(\sigma(K_{1,3}))^T_B$ are $\frac{1}{2}, \frac{1}{2} \sqrt{17} + \frac{1}{2}$ and $\frac{1}{2} \sqrt{17} - \frac{1}{2}$. It follows from the Criterion 4.1 that $\sigma(K_{1,3})$ is entangled in $\mathbb{C}^2 \otimes \mathbb{C}^2$.

Remark 4.9 A density matrix $\rho_{AB}$ acting on $\mathbb{C}_A^2 \otimes \mathbb{C}_B^2$ is said to be distillable if there exist a positive integer $k$ and two $2$-dimensional projectors $P_1 : (\mathbb{C}_A^2)^k \rightarrow \mathbb{C}^2$ and $P_2 : (\mathbb{C}_B^2)^k \rightarrow \mathbb{C}^2$ such that $(P_1 \otimes P_2) \rho_{AB} (P_1 \otimes P_2)^T_B \neq 0$. An entangled density matrix which is not distillable is called bound entangled. Theorem 4.7 actually shows that not only $\sigma(K_{1,n-1})$ is entangled but also distillable in $\mathbb{C}^p \otimes \mathbb{C}^q$ where $n = pq \geq 4$.

Second proof of Theorem 4.7 Let $G$ be a graph on $n$ vertices and $m$ edges. Suppose that $G$ has $l_i$ loops at the vertex $v_i$. Then $|E(G)| = m + \sum_{i=1}^n l_i$ edges. We associate to $G$ the following density matrix

$$
\sigma_\sigma(G) \overset{\text{def}}{=} \left( 2m + \sum_{i=1}^n l_i \right)^{-1} (\Delta(G) - M(G)) + \left( 2m + \sum_{i=1}^n l_i \right)^{-1} \Delta_\sigma(G), \quad \text{where } \Delta_\sigma(G) = \bigoplus_{i=1}^n l_i P[|v_i\rangle].
$$

The matrix $\sigma_\sigma(G)$ generalizes the notion of density matrix of a graph by taking into account loops. The star $K_{1,3}$ is called claw. We denote by $K_{1,3}^{t+l}$ a claw with $l$ loops at the vertex of degree $3$. In order to prove the theorem, we first show that $K_{1,3}^{t+l}$ is entangled in $\mathbb{C}_A^2 \otimes \mathbb{C}_B^2$. The vertices of $K_{1,3}$ are denoted by $u_1 w_1, u_1 w_2, u_2 w_1, u_2 w_2$, where $d(u_1 w_1) = 3$. We then have

$$
\sigma_\sigma(K_{1,3}^{t+l}) = \frac{2}{l+6} \left\{ \frac{l}{2} P[|u_1\rangle |w_1\rangle] + P[\frac{1}{\sqrt{2}}(|u_1\rangle |w_1\rangle - |u_2\rangle |w_2\rangle)]
+ P[\frac{1}{\sqrt{2}}(|u_1\rangle - |u_2\rangle |w_1\rangle)] + P[|u_1\rangle |w_1\rangle] \right\}.
$$

One can check that $[\sigma_\sigma(K_{1,3}^{t+l})]^T_B \neq 0$. By Theorem 4.1, $\sigma_\sigma(K_{1,3}^{t+l})$ is entangled in $\mathbb{C}_A^2 \otimes \mathbb{C}_B^2$ for all $l \geq 0$. Let $A$ be the matrix in Equation 4.4. Taking $n - 4 = l$, $A = \frac{l+6}{2(l+3)} \sigma_\sigma(K_{1,3}^{t+l})$. Then $A$ is entangled in $\mathbb{C}_A^2 \otimes \mathbb{C}_B^2$. This shows that $\sigma(K_{1,n-1})$ is entangled in $\mathbb{C}_A^p \otimes \mathbb{C}_B^q$.  

Proposition 4.10 Separability of $\sigma(K_{1,n-1})$, with $n = pq \geq 4$, does not depend on the labeling of $V(K_{1,n-1})$.

Proof. In $K_{1,n-1}$, the vertex of degree $(n-1)$ is called root, the other vertices are called leaves. We define two types of isomorphisms for stars: Leaf-shuffling. An isomorphism $\iota$ acting on $V(K_{1,n-1})$ is called a leaf-shuffling if $\iota(r) = r$, where $r$ is the root of $K_{1,n-1}$; Root-swapping. An isomorphism $\iota$ acting on $V(K_{1,n-1})$ is called a root-swapping if $\iota(r) = v$, where $r$ is the root of $K_{1,n-1}$ and $v$ is a leaf. All graphs in the isomorphism class of $K_{1,n-1}$ can be obtained by combining leaf-shuffling and root-swapping. It is clear that leaf-shuffling is an automorphism and hence it does not change the
separability property of \( \sigma(K_{1,n-1}) \). We now prove that this is the case also for root-swapping. We label the vertices of a graph \( G \cong K_{1,n-1} \) as \( u_1w_1, u_1w_2, \ldots, u_1w_q, u_2w_1, u_2w_2, \ldots, u_pw_q \). Let \( u_1w_1 \) be the root of \( G \) and let \( \iota: V(G) \rightarrow V(H) \) be a root-swapping. Then, the root of \( H \) is \( \iota(u_1w_1) = u_1w_j \), where \( (1,1) \neq (i,j) \). Denote by \( P_{1+i} \) and \( Q_{1+j} \) the permutation matrices defined as follows: \( P_{1+i} | u_1 \rangle = | u_i \rangle \) and \( P_{1+i} | u_0 \rangle = | u_0 \rangle \) for \( i' \neq 1 \); \( Q_{1+j} | v_1 \rangle = | w_j \rangle \) and \( Q_{1+j} | w_0 \rangle = | w_0 \rangle \) for \( j' \neq 1 \). Then we have \( (P_{1+i} \otimes Q_{1+j}) \sigma(H) (P_{1+i} \otimes Q_{1+j})^\top = \sigma(H) \). Then \( \sigma(G) \) is entangled if and only if \( \sigma(H) \) is entangled.  

**Remark 4.11** For \( K_{1,n-1} \), with \( n = pq \geq 4 \), \( K_{1,n-1} \not\cong G \otimes H \), where \( |V(G)| = p \) and \( |V(H)| = q \).

### 4.3 Perfect matchings

A *matching* of a graph is a set of vertex-disjoint edges. A *perfect matching* of a graph \( G \) is a matching spanning \( V(G) \).

**Definition 4.12 (e-matching; pe-matching)** An e-matching is a matching having all edges entangled. Each vertex of an e-matching on \( n = pq \) vertices can be labeled by an ordered pairs \((i,j)\), where \( 1 \leq i \leq p \) and \( 1 \leq j \leq q \). A pe-matching of a graph \( G \) is an e-matching spanning \( V(G) \).

**Theorem 4.13** Let \( G \) be a graph on \( n = 2p \) vertices. If all the entangled edges of \( G \) belong to the same pe-matching then \( \sigma(G) \) is separable in \( \mathbb{C}^2 \otimes \mathbb{C}^p \).

Our proof of the theorem involves the use of the following concepts.

**Definition 4.14 (Criss-cross)** A criss-cross is a set \( \{(k,i),(l,j)\} \) of two edges belonging to an e-matching on \( n = pq \) vertices.

**Definition 4.15 (Tally-mark)** A set 

\[
\{(k,i_1),(l,i_2)\}, \{(k,i_2),(l,i_3)\}, \ldots, \{(k,i_{s+1}),(l,i_{s+2})\}, \{(k,i_{s+2}),(l,i_1)\}
\]

of \( s+2 \) edges, where \( k < l \), \( s \geq 0 \) and \( i_1 < i_2 < \cdots < i_{s+2} \), belonging to an e-matching on \( n = pq \) vertices, is called a tally-mark. (Note that a criss-cross is a tally-mark with two edges.)

**Definition 4.16 (Canonical pe-matching)** Let \( H \) be an e-matching on \( n = pq \) vertices. Then \( H \) is said to be canonical if \( H = H_1 \uplus H_2 \uplus \cdots \uplus H_k \), where every graph \( H_1, H_2, \ldots, H_k \) is a tally-mark.

**Lemma 4.4** From any pe-matching \( H \) on \( n = 2p \) vertices, labeled by \((i,j)\), where \( i = 1,2 \) and \( 1 \leq j \leq p \), we can always obtain a canonical pe-matching by applying a permutation to the second label of all the vertices of \( H \).

**Proof.** Let \( H \) be a pe-matching as in the statement of the lemma. Any pe-matching can be taken as a set of criss-crosses and e-matchings, the latter being of the forms:

- \( H_1 \) such that \( E(H_1) = \{(1,i_1),(2,j_1)\}, \{(1,i_2),(2,j_2)\}, \ldots, \{(1,i_k),(2,j_k)\} \), where \( i_1, i_2, \ldots, i_k \) = \( \{j_1,j_2,\ldots,j_k\} \);
- \( H_2 \) such that \( E(H_2) = \{(1,i'_1),(2,j'_1)\}, \{(1,i'_2),(2,j'_2)\}, \ldots, \{(1,i'_i),(2,j'_i)\} \), where \( i'_1, i'_2, \ldots, i'_i \) = \( j'_1,j'_2,\ldots,j'_i \) and \( \{i'_1,i'_2,\ldots,i'_i\} \cap \{i_1,i_2,\ldots,i_k\} = \emptyset \);
- ...

We describe an algorithm to obtain a tally-mark from any of the above e-matchings. It is sufficient to describe the algorithm for \( H_1 \). Without loss of generality we take \( i_1 < i_2 < \cdots < i_k \). We permute the \( 2^{nd} \) labels of the edges of \( H_1 \), to form one or more disjoint tally-marks. Consider the \( s^{th} \) step in the construction: if \( j_s < i_s \), we have completed a tally-mark; if \( j_s > i_s \) then we perform a permutation on the \( 2^{nd} \) label of a vertex \((-i)\) acting on indices \( i > i_s \), which maps the edge \( \{(1,i_s),(2,j_s)\} \) to the edge \( \{(1,i_s),(2,i_{s+1})\} \) (adding another downstroke to a tally-mark, yet incomplete). It is easy to see that applying this rule successively to the labels \( \{i_1, i_2, \ldots, i_k\} \), in ascending order, produces a set of one or more disjoint tally-marks.  

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**Example 4.17** In Figure 1, a pe-matching (top graph) is transformed in a canonical pe-matching by applying a permutation on the second labels of the vertices. We first apply the permutation (2 3) (central graph). We then apply the permutation (3 5) (bottom graph).

![Diagram](image)

**Figure 1**

**Lemma 4.5** Let $H$ be a tally-mark on $n = 2k + 2$ vertices. Then $\sigma(H)$ is separable in $\mathbb{C}^2 \otimes \mathbb{C}^{k+1}$.

**Proof.** Let $H$ be a tally-mark. Let us assume that $H$ is not a criss-cross. In fact, if $H$ is a criss-cross then $\sigma(H)$ is obviously separable in $\mathbb{C}^2 \otimes \mathbb{C}^2$. Let

$$E(H) = \{(1, i_0), (2, i_1), \{1, i_1\}, (2, i_2), \ldots, \{1, i_{k-1}\}, (2, i_k), (1, i_k), (2, i_0)\},$$

where $i_0 < i_1 < \cdots < i_k$. We associate the vector $|i\rangle |i_s\rangle$ to the vertex $(l, i_s) \in V(H)$, where $l = 1, 2$ and $s = 0, 1, 2, \ldots, k$. Then

$$\sigma(H) = \frac{1}{k+1} \sum_{s=0}^{k} \frac{1}{\sqrt{2}} \left(|1\rangle |i_s\rangle - |2\rangle |i_{(s+1) \text{ mod}(k+1)}\rangle\right).$$

Let us consider the permutation $g$ on the $(k + 1)$ letters $i_0, i_1, \ldots, i_k$ defined as follows: $g : i_s \mapsto i_{(s+1) \text{ mod}(k+1)}$, where $0 \leq s \leq k$. The order of $g$ is then $(k + 1)$. Let $\Gamma = \langle g \rangle \cong \mathbb{Z}_{k+1}$ and $\rho_{\text{reg}}(g) = \Pi$. One can check that

$$\sigma(H) = \frac{1}{k+1} \sum_{s=0}^{k} (I_2 \otimes \Pi^s) \frac{1}{\sqrt{2}} \left(|1\rangle |i_0\rangle - |2\rangle |i_1\rangle\right) \left(I_2 \otimes \Pi^{k+1-s}\right),$$

where $I_2$ acts on the Hilbert space spanned by the vectors $|1\rangle$ and $|2\rangle$. We are now looking for the density matrices acting on $\mathbb{C}^2 \otimes \mathbb{C}^{k+1}$, which remains invariant under the action of $\Gamma$. Let

$$|\psi_m\rangle = \frac{1}{\sqrt{k+1}} \sum_{s=0}^{k} \exp\left[\frac{2\pi im}{k+1}\right] |i_s\rangle,$$

where $m = 0, 1, 2, \ldots, k$. Observe that the vectors $|\psi_m\rangle$’s are pairwise orthonormal. Then $\Pi^l |\psi_m\rangle = \exp\left[\frac{2\pi im}{k+1}\right] |\psi_m\rangle$, for $l = 0, 1, 2, \ldots, k$ and $m = 0, 1, 2, \ldots, k$. It follows that the $|\psi_m\rangle$’s are eigenvectors of $\Pi^l$. Let

$$|\Phi\rangle = \sum_{m=0}^{k} (\alpha_m |1\rangle + \beta_m |2\rangle) |\psi_m\rangle$$

where $\sum_{m=0}^{k} (|\alpha_m|^2 + |\beta_m|^2) = 1$.

be a vector in $\mathbb{C}^2 \otimes \mathbb{C}^{k+1}$. Then $\langle I_2 \otimes \Pi^l | \Phi\rangle |\Phi\rangle = |\Phi\rangle |\Phi\rangle$ for $l = 0, 1, 2, \ldots, k$, and only if $|\Phi\rangle$ is one of the forms $(\alpha_m |1\rangle + \beta_m |2\rangle) |\psi_m\rangle$, for $m = 0, 1, 2, \ldots, k$. This shows that, for any density matrix $\rho$ acting on $\mathbb{C}^2 \otimes \mathbb{C}^{k+1}$, the following density matrix

$$\rho' = \frac{1}{k+1} \sum_{s=0}^{k} (I_2 \otimes \Pi^s) \rho \left(I_2 \otimes \Pi^{k+1-s}\right)$$
is a mixture of all the projectors $P[\frac{1}{\sqrt{2}}(\alpha_m|1\rangle + \beta_m|2\rangle)|\psi_m\rangle]$, where $m = 0, 1, 2, ..., k$. Hence $\rho$ is separable. By Equation 4.13, $\sigma(H)$ is also separable:

$$\sigma(H) = \frac{1}{k+1} \sum_{m=0}^{k} P[\frac{1}{\sqrt{2}}(1) - \exp[\frac{-2\pi im}{k+1}]|2\rangle)|\psi_m\rangle].$$

Given a graph $G$ and a factor $H$ of $G$, we denote by $G\setminus H$ the graph with adjacency matrix $M(G\setminus H) \overset{\text{def}}{=} M(G) - M(H)$.

**Proof of Theorem 4.13** Let $G$ be as in the statement of the theorem. In addition, we assume that $|E(G)| = m$. Let $H$ be the pe-matching containing all the entangled edges of $G$. Then $\sigma(G) = \overline{\sigma(G)} + \frac{m-\alpha}{m} \sigma(G\setminus H)$. The density matrix $\sigma(G\setminus H)$ is separable by assumption. Lemma 4.14 together with Lemma 4.13 shows that $\sigma(H)$ is separable in $\mathbb{C}^2 \otimes \mathbb{C}^2$. This proves the theorem.

**Theorem 4.18** The pe-matching in Figure 2 is separable in $\mathbb{C}^3 \otimes \mathbb{C}^4$.

![Figure 2](image-url)

**Proof.** Let $G$ be the pe-matching in the figure. Then $\sigma(G) = \frac{1}{6} \sum_{i=1}^{6} P[|\psi_i\rangle\rangle$, where $|\psi_i\rangle\rangle = \frac{1}{\sqrt{2}}(|1\rangle|1\rangle \pm |2\rangle|2\rangle)$, $|\psi_i\rangle\rangle = \frac{1}{\sqrt{2}}(|1\rangle|4\rangle \pm |2\rangle|3\rangle)$, $|\psi_i\rangle\rangle = \frac{1}{\sqrt{2}}(|1\rangle|2\rangle \pm |3\rangle|4\rangle)$, $|\psi_i\rangle\rangle = \frac{1}{\sqrt{2}}(|1\rangle|3\rangle \pm |3\rangle|1\rangle)$, $|\psi_i\rangle\rangle = \frac{1}{\sqrt{2}}(|2\rangle|1\rangle \pm |3\rangle|3\rangle)$ and $|\psi_i\rangle\rangle = \frac{1}{\sqrt{2}}(|2\rangle|4\rangle \pm |3\rangle|2\rangle)$). Here $(\sigma(G))^{T\alpha} = (I_3 \otimes P)\sigma(G)(I_3 \otimes P^T)$, where

$$P = \begin{bmatrix} 0 & 1 & 0 & 0 & 1 & 0 & 0 & 0 & 0 & 0 & 1 & 0 & 0 & 0 \\ 1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 & 1 & 0 & 0 & 0 \end{bmatrix}.$$ 

Then $(\sigma(G))^{T\alpha} > 0$. Since a density matrix having positive partial transpose is either separable or bound entangled [1], this holds for $\sigma(G)$. We are now going to show that $\sigma(G)$ is separable in $\mathbb{C}^3 \otimes \mathbb{C}^4$. Let $|\chi\rangle$ be in the support of $\sigma(G)$. Then $|\chi\rangle = \sum_{i=1}^{6} a_i|\psi_i\rangle\rangle = \frac{1}{\sqrt{2}}(a_1|1\rangle|1\rangle + a_2|2\rangle|2\rangle + a_3|3\rangle|3\rangle + a_4|4\rangle|4\rangle + a_5|1\rangle|3\rangle - a_6|3\rangle|1\rangle + a_7|2\rangle|4\rangle - a_8|4\rangle|2\rangle + a_9|3\rangle|5\rangle - a_{10}|5\rangle|3\rangle)$). So, $|\chi\rangle$ is separable if and only if $(a_1, a_3, a_4, a_2) = \lambda(a_5, -a_1, -a_2, a_6) = \mu(a_4, a_5, a_9, a_3)$, where $\lambda, \mu \in \mathbb{C}$. Then

$$a_1 = \lambda a_5 = \mu a_4,$$
$$a_3 = -\lambda a_1 = \mu a_6,$$
$$a_4 = -\lambda a_2 = \mu a_5,$$
$$a_2 = \lambda a_6 = \mu a_3$$

(10)

Here $\lambda \neq 0$. In fact, if $\lambda = 0$ and $\mu \neq 0$, then $a_i = 0$ for $i = 1, 2, ..., 6$, which is impossible. On the other hand, if $\lambda = \mu = 0$ then $a_1, a_2, a_3, a_4 = 0$. Then $|\chi\rangle = \frac{1}{\sqrt{2}}(|2\rangle|5\rangle|1\rangle|1\rangle + a_6|2\rangle|4\rangle - \frac{1}{\sqrt{2}}(|3\rangle|5\rangle|6\rangle|2\rangle + a_5|3\rangle|4\rangle)$, which is entangled as $|a_5|^2 + |a_6|^2 \neq 0$. Similarly, it can be shown that $\mu \neq 0$. Therefore, from Equation (10) $\lambda^2 = 1$ and $\mu^2 = \lambda$, and we can distinguish the following cases.

**Case 1.** ($\lambda = \mu = 1$) We have $a_2 = a_3 = a_6 = -a_1$ and $a_4 = a_5 = a_1$. So $|\chi\rangle = \frac{a_1}{\sqrt{2}}(|1\rangle|2\rangle - |3\rangle|1\rangle) - (|2\rangle|3\rangle - |4\rangle).$

**Case 2.** ($\lambda = 1, \mu = -1$) We have $a_2 = a_5 = a_6 = a_1$ and $a_3 = a_4 = -a_1$. So $|\chi\rangle = \frac{a_1}{\sqrt{2}}(|1\rangle|2\rangle + |3\rangle|1\rangle - (|2\rangle|3\rangle - |4\rangle).$

**Case 3.** ($\lambda = \omega = e^{2\pi i/3}, \mu = -\omega^2$) We have $a_2 = a_3 = a_4 = -a_5, a_5 = a_6 = \omega^2 a_1$. So $|\chi\rangle = \frac{a_1}{\sqrt{2}}(|1\rangle|2\rangle + \omega^2|3\rangle(|1\rangle - |2\rangle - \omega|3\rangle + |4\rangle).$

**Case 4.** ($\lambda = \omega, \mu = \omega^2$) We have $a_2 = a_3 = a_4 = -a_5, a_5 = a_6 = \omega^2 a_1$. So $|\chi\rangle = \frac{a_1}{\sqrt{2}}(|1\rangle + \omega^2|2\rangle - \omega|3\rangle)(|1\rangle - |2\rangle + \omega|3\rangle - |4\rangle).$

**Case 5.** ($\lambda = \omega^2, \mu = \omega$) We have $a_2 = a_1, a_3 = a_4 = -\omega a_1, a_5 = a_6 = \omega a_1$. So $|\chi\rangle = \frac{a_1}{\sqrt{2}}(|1\rangle + \omega|2\rangle - \omega^2|3\rangle)(|1\rangle - \omega^2|2\rangle + \omega^2|3\rangle - |4\rangle).$
Case 6. \((\lambda = \omega^2, \mu = -\omega)\) We have \(a_2 = a_1, a_3 = a_4 = -\omega^2a_1, a_5 = a_6 = \omega a_1\). So \(|\chi\rangle = \frac{3}{\sqrt{6}}(|1\rangle + \omega|2\rangle + \omega^2|3\rangle)(1) - \omega^2|2\rangle - \omega^2|3\rangle + |4\rangle).

Thus we can observe that the range of the rank six density matrix \(\sigma(G)\) contains only the following six separable states: \(|\chi_1\rangle = \frac{1}{\sqrt{3}}(|1\rangle + |2\rangle - |3\rangle), |\chi_2\rangle = \frac{1}{\sqrt{3}}(|1\rangle - |2\rangle + |3\rangle - |4\rangle), |\chi_3\rangle = \frac{1}{\sqrt{3}}(|1\rangle + \omega^2|2\rangle + \omega|3\rangle), |\chi_4\rangle = \frac{1}{\sqrt{3}}(|1\rangle - \omega^2|2\rangle - \omega|3\rangle), |\chi_5\rangle = \frac{1}{\sqrt{3}}(|1\rangle + \omega|2\rangle + \omega^2|3\rangle - |4\rangle), |\chi_6\rangle = \frac{1}{\sqrt{3}}(|1\rangle + \omega|2\rangle + \omega^2|3\rangle)\). These states are pairwise orthogonal. As \(\sigma(G)\) is proportional to a six dimensional projector, we can write \(\sigma(G) = \frac{1}{3} \sum_{i=1}^{6} |\chi_i\rangle\langle \chi_i|\), and hence \(\sigma(G)\) is separable. 

Remark 4.19 The pe-matching \(G\) in Figure 3 is entangled in \(\mathbb{C}^3 \otimes \mathbb{C}^4\). In fact, it can be shown that \((\sigma(G))^{\top B} \neq 0\).

4.4 The Petersen graph

Let \(v, k\) and \(i\) be fixed positive integers, with \(v \geq k \geq i\). Let \(S\) be an \(n\)-elements set. The Johnson graph \(J(v, k, i)\) is defined as follows: the vertices of \(J(v, k, i)\) are the \(k\)-elements subsets of \(S\); two vertices are adjacent if their intersection has size \(i\). The graph \(J(5, 2, 0)\) is called Petersen graph and it has a number of important properties. For example, it is strongly-regular and transitive. A graph \(G\) that is not complete is said to be strongly-regular if it is regular, every pair of adjacent vertices has a the same number of common neighbours, and every pair of nonadjacent vertices has the same number of common neighbours. A graph \(G\) is said to be transitive if \(\text{Aut}(G)\) acts transitively on \(V(G)\). A permutation group \(\Gamma\) acts transitively on a set \(S\) if, for any \(s, t \in S\), there exists \(g \in \Gamma\), such that \(g(s) = t\).

Theorem 4.20 Let \(G\) be a Petersen graph. Then \(\sigma(G)\) is either separable or entangled in \(\mathbb{C}^2 \otimes \mathbb{C}^5\), depending on the labelling of \(G\).

Proof. Let \(G\) (left) and \(H\) (right) be the graphs in Figure 4:

Both, \(G\) and \(H\) are isomorphic to the Petersen graph. The density matrix \(\sigma(G)\) is separable, since every edge of \(G\) is separable. The density matrix of \(H\) is entangled in \(\mathbb{C}^2_A \otimes \mathbb{C}^5_B\), since it can be shown that \((\sigma(H))^{\top B} \neq 0\).

Corollary 4.1 The density matrices of strongly-regular graphs and transitive graphs can be separable or entangled.

Proof. By Theorem 4.20, since the Petersen graph is strongly-regular and transitive.
4.5 Concurrence

Let $|\psi\rangle_{AB} \in \mathbb{C}^2_A \otimes \mathbb{C}^2_B$. The notion of concurrence was introduced by Wootters [12]. The concurrence of $|\psi\rangle_{AB}$ is denoted and defined as follows: $C(\psi) = \sqrt{2(1 - \text{tr}(\rho_{AB}^2))}$, where $\rho_{A} = \text{tr}_B(|\psi\rangle_{AB}\langle\psi|)$. Let $\rho_{AB}$ be a density matrix acting on $\mathbb{C}^2_A \otimes \mathbb{C}^2_B$. The concurrence of $\rho_{AB}$ is denoted and defined as follows: $C(\rho_{AB}) = \inf \{ \sum_i \omega_i |\psi_i\rangle \langle \psi_i| : \rho_{AB} = \sum_i \omega_i |\psi_i\rangle \langle \psi_i|, 0 \leq \omega_i \leq 1, \sum \omega_i = 1 \}$. Let $\sigma_y = -i|1\rangle\langle 2| + i|2\rangle\langle 1|$, where $|1\rangle$ and $|2\rangle$ are the eigenvectors of the matrix

$$\sigma_z = \begin{bmatrix} 1 & 0 \\ 0 & -1 \end{bmatrix},$$

corresponding to the eigenvalues 1 and $-1$, respectively. Let $M^*$ be the conjugate of a complex matrix $M$. An analytical formula for $C(\rho_{AB})$, is given by $C(\rho_{AB}) = \max\{0, \lambda_1 - \lambda_2 - \lambda_3 - \lambda_4\}$, where $\lambda_1, \lambda_2, \lambda_3$ and $\lambda_4$ are the square roots of the eigenvalues of $\rho_{AB}\rho_{AB}^*$. There are 12 nonisomorphic graphs on 4 vertices. Seven of these graphs have entangled density matrix, independently of the labeling. In the table below, the graphs and the respective concurrence are given. Note that in three cases the value of the concurrence is exactly fractional.

| Concurrence | 0.33326668 | 1/3 | 1/5 | 1 | 0.25005352 | 0.500131893 | 0.333236542 |
|-------------|-----------|-----|-----|---|------------|-------------|-------------|
| Graph       | 🟢 | 🟢 | 🟢 | 🟢 | 🟢 | 🟢 | 🟢 |

5 Graph operations

A graph operation is a map that takes a graph to another one. In Graph Theory, the study of graph operations consists of a vast literature [10]. The following are two examples of graph operations.

**Example 5.1** Deleting an edge $\{v_i, v_j\}$ from a graph $G$ means to transform $G$ into the graph $G - \{v_i, v_j\} \overset{def}{=} (V(G), E(G)\setminus \{v_i, v_j\})$. Adding an edge $\{v_i, v_j\}$ to a graph $G$, where $\{v_i, v_j\} \notin E(G)$, means to transform $G$ into the graph $G + \{v_i, v_j\} \overset{def}{=} (V(G), E(G) \cup \{v_i, v_j\})$. Deleting a vertex $v_i$ from a graph $G$ means to transform $G$ into the graph $G - v_i \overset{def}{=} (V(G) \setminus \{v_i\}, E(G) \setminus E_i)$, where $E_i$ is the set of all edges incident to $v_i$. Adding a vertex $v_i$ to a graph $G$ means to transform $G$ into the graph $G + v_i + T_i \overset{def}{=} (V(G) \cup \{v_i\}, E(G) \cup T_i)$, where $T_i$ is an arbitrary set of edges incident to $v_i$.

Let $\mathcal{B}(\mathcal{H}^n)$ be the space of all bounded linear operators on $\mathcal{H}^n$. A linear map $\Lambda : \mathcal{B}(\mathcal{H}^n) \rightarrow \mathcal{B}(\mathcal{H}^m)$ is said to be hermiticity preserving if for every hermitian operator $O \in \mathcal{B}(\mathcal{H}^n)$, $\Lambda(O)$ is an hermitian operator in $\mathcal{B}(\mathcal{H}^m)$. A hermiticity preserving map $\Lambda : \mathcal{B}(\mathcal{H}^n) \rightarrow \mathcal{B}(\mathcal{H}^m)$ is said to be positive if for any positive operator $O \in \mathcal{B}(\mathcal{H}^n)$, $\Lambda(O)$ is a positive operator in $\mathcal{B}(\mathcal{H}^m)$. A positive map $\Lambda : \mathcal{B}(\mathcal{H}^n) \rightarrow \mathcal{B}(\mathcal{H}^m)$ is said to be completely positive if for each positive integer $k$, $(\Lambda \otimes I_k) : \mathcal{B}(\mathcal{H}^n \otimes \mathcal{H}^k) \rightarrow \mathcal{B}(\mathcal{H}^m \otimes \mathcal{H}^k)$ is again a positive map. A completely positive map $\Lambda : \mathcal{B}(\mathcal{H}^n) \rightarrow \mathcal{B}(\mathcal{H}^m)$ is said to be trace preserving if $\text{tr}(\Lambda(O)) = \text{tr}(O)$, for all $O \in \mathcal{B}(\mathcal{H}^n)$. A quantum operation is a trace preserving completely positive map (for short, TPCP). In Standard Quantum Mechanics, any physical transformation of a quantum mechanical system is described by a quantum operation. We are going to use the following result:

**(Kraus representation theorem)** Given a quantum operation $\Lambda : \mathcal{B}(\mathcal{H}^n) \rightarrow \mathcal{B}(\mathcal{H}^m)$, there exist $n \times m$ matrices $A_i$, such that $\Lambda(A) = \sum_i A_i \rho A_i^\dagger$, where $\rho$ is any density matrix acting on $\mathcal{H}^n$ and $\sum_i A_i^\dagger A_i = I_m$. (The converse is also true.) The matrices $A_i$’s are called Kraus operators.

A projective measurement $\mathcal{M} = \{P_i : i = 1, 2, ..., n\}$, on a quantum mechanical system $S$ whose state is $\rho$, consists of pairwise orthogonal projectors $P_i : \mathcal{H}_S \rightarrow \mathcal{H}_S$, such that $\sum_{i=1}^n P_i = I_{\dim(\mathcal{H}_S)}$. The $i$-th outcome of the measurement occurs with probability $\text{tr}(P_i \rho)$ and the post-measurement state of $S$ is $\frac{P_i \rho P_i}{\text{tr}(P_i \rho)}$. Whenever the $i$-th outcome of the measurement occurs, we say that $P_i \; \text{clicks}$. 


5.1 Deleting and adding an edge

Here we describe how to delete or add an edge by means of TPCP. Let $G$ be a graph on $n$ vertices, $v_1, v_2, \ldots, v_n$, and $m$ edges, $\{v_{i_1}, v_{j_1}\}, \{v_{i_2}, v_{j_2}\}, \ldots, \{v_{i_m}, v_{j_m}\}$, where $1 \leq i_1, j_1, i_2, j_2, \ldots, i_m, j_m \leq n$. Our purpose is to delete the edge $\{v_{i_k}, v_{j_k}\}$. Then we have

$$\sigma(G) = \frac{1}{m} \sum_{l=1}^{m} \sigma(H_{i_l,j_l}) = \frac{1}{m} \sum_{l=1}^{m} P\left[\frac{1}{\sqrt{2}}(|v_{i_l}| - |v_{j_l}|)\right]$$

and

$$\sigma(G - \{v_{i_k}, v_{j_k}\}) = \frac{1}{m-1} \sum_{l=1,l\neq k}^{m} \sigma(H_{i_l,j_l}) = \frac{1}{m-1} \sum_{l=1,l\neq k}^{m} P\left[\frac{1}{\sqrt{2}}(|v_{i_l}| - |v_{j_l}|)\right].$$

A measurement in the following basis, $\mathcal{M} = \{\frac{1}{\sqrt{2}}(|v_{i_l}| \pm |v_{j_l}|), |v_i| : i \neq i_k, j_k \text{ and } i = 1, 2, \ldots, n\}$ is performed on the system prepared in the state $\sigma(G)$. The probability that $P\left[\frac{1}{\sqrt{2}}(|v_{i_l}| + |v_{j_l}|)\right]$ clicks is

$$\frac{m-1}{2m} \left(\langle v_{i_l} | + \langle v_{j_l} | \right) \sigma(G - \{v_{i_k}, v_{j_k}\}) \left( |v_{i_l}| + |v_{j_l}| \right) = \frac{1}{4m} \sum_{l=1,l\neq k}^{m} (\delta_{i_l,i_k} - \delta_{i_l,j_k} + \delta_{j_l,i_k} - \delta_{j_l,j_k})^2. \quad (11)$$

The state after the measurement is $P\left[\frac{1}{\sqrt{2}}(|v_{i_l}| - |v_{j_l}|)\right]$. Let $U^{+}_{kl}$ be an $n \times n$ unitary matrix, such that

$$U^{+}_{kl} \frac{1}{\sqrt{2}}(|v_{i_l}| + |v_{j_l}|) = \frac{1}{\sqrt{2}}(|v_{i_l}| - |v_{j_l}|), \text{ for } l = 1, 2, \ldots, k-1, k+1, \ldots, m.$$  

With probability $1/(m-1)$ we apply $U^{+}_{kl}$ on $P\left[\frac{1}{\sqrt{2}}(|v_{i_l}| + |v_{j_l}|)\right]$, for each $l = 1, 2, \ldots, k-1, k+1, \ldots, m$. Finally we obtain $\sigma(G - \{v_{i_k}, v_{j_k}\})$ with probability given by Equation 11. The probability that $P\left[\frac{1}{\sqrt{2}}(|v_{i_l}| - |v_{j_l}|)\right]$ clicks is

$$\frac{1}{2m} \sum_{l=1}^{m} (\delta_{i_l,i_k} - \delta_{i_l,j_k} - \delta_{j_l,i_k} + \delta_{j_l,j_k})^2. \quad (12)$$

The state after the measurement is $P\left[|v_{i_l}|\right]$. Let $U_{il}$ be an $n \times n$ unitary matrix, such that

$$U_{il} \frac{1}{\sqrt{2}}(|v_{i_l}| - |v_{j_l}|) = \frac{1}{\sqrt{2}}(|v_{i_l}| - |v_{j_l}|), \text{ for } l = 1, 2, \ldots, k-1, k+1, \ldots, m.$$  

With probability $1/(m-1)$ we apply $U_{il}$ on $P\left[|v_{i_l}|\right]$, for each $l = 1, 2, \ldots, k-1, k+1, \ldots, m$. We obtain $\sigma(G - \{v_{i_k}, v_{j_k}\})$ with probability given by Equation 12. This completes the process. The set of Kraus operators that realize the TPCP for deleting the edge $\{v_{i_k}, v_{j_k}\}$ is then

$$\left\{ \frac{1}{\sqrt{m-1}} U^{+}_{kl} P\left[\frac{1}{\sqrt{2}}(|v_{i_l}| + |v_{j_l}|)\right] : l = 1, 2, \ldots, k-1, k+1, \ldots, m \right\}$$

$$\cup \left\{ \frac{1}{\sqrt{m-1}} U^{-}_{kl} P\left[\frac{1}{\sqrt{2}}(|v_{i_l}| - |v_{j_l}|)\right] : l = 1, 2, \ldots, k-1, k+1, \ldots, m \right\},$$

and the state after the measurement is $P\left[|v_{i_l}|\right]$. Let $U_{il}$ be an $n \times n$ unitary matrix, such that

$$U_{il} \frac{1}{\sqrt{2}}(|v_{i_l}| - |v_{j_l}|) = \frac{1}{\sqrt{2}}(|v_{i_l}| - |v_{j_l}|), \text{ for } l = 1, 2, \ldots, k-1, k+1, \ldots, m.$$  

With probability $1/(m-1)$ we apply $U_{il}$ on $P\left[|v_{i_l}|\right]$, for each $l = 1, 2, \ldots, k-1, k+1, \ldots, m$. We obtain $\sigma(G - \{v_{i_k}, v_{j_k}\})$ with probability given by Equation 13. This completes the process. The set of Kraus operators that realize the TPCP for adding back the edge $\{v_{i_k}, v_{j_k}\}$ to $G - \{v_{i_k}, v_{j_k}\}$ is

$$\left\{ \frac{1}{\sqrt{m}} V^{+}_{kl} P\left[\frac{1}{\sqrt{2}}(|v_{i_l}| + |v_{j_l}|)\right] : l = 1, 2, \ldots, m \right\}$$

$$\cup \left\{ \frac{1}{\sqrt{m}} V^{-}_{kl} P\left[\frac{1}{\sqrt{2}}(|v_{i_l}| - |v_{j_l}|)\right] : l = 1, 2, \ldots, m \right\},$$

where $V_{kl}^{+}$, $V_{kl}^{-}$, and $U_{il}$ are $n \times n$ unitary matrices defined as follows:

$$V_{kl}^{+} \frac{1}{\sqrt{2}}(|v_{i_l}| + |v_{j_l}|) = \frac{1}{\sqrt{2}}(|v_{i_l}| - |v_{j_l}|), \quad \text{for } l = 1, 2, \ldots, m;$$

$$V_{kl}^{-} \frac{1}{\sqrt{2}}(|v_{i_l}| - |v_{j_l}|) = \frac{1}{\sqrt{2}}(|v_{i_l}| - |v_{j_l}|), \quad \text{for } l = 1, 2, \ldots, m;$$

$$V_{il} |v_{i_l}| = \frac{1}{\sqrt{2}}(|v_{i_l}| - |v_{j_l}|), \quad \text{for } i = 1, 2, \ldots, n; i \neq i_k, j_k; l = 1, 2, \ldots, m.$$
5.2 Deleting and adding a vertex

Here we describe how to delete or add a vertex by means of TPCP. Let $G$ be a graph on $n$ vertices, $v_1, v_2, ..., v_n$, and $m$ edges, $\{v_{i_1}, v_{j_1}\}, \{v_{i_2}, v_{j_2}\}, ..., \{v_{i_m}, v_{j_m}\}$, where $1 \leq i_1, j_1, i_2, j_2, ..., i_m, j_m \leq n$. Our purpose is to delete a vertex $v_i$. We first delete all the edges incident to $v_i$ (cfr. Section 5.1).

In this way, we obtain a new graph, say $H$. We then perform the following projective measurement on $\sigma(H)$: $M = \{I_n - P([v_i]), P([v_i])\}$. Given that, possible loops in $H$ do not appear on $\sigma(H)$, when $M$ is performed on $\sigma(H)$, $I_n - P([v_i])$ clicks with probability one. The state after the measurement is $\sigma(G - v_i)$, which is the state of the desired graph. Let $G$ be a graph on $n$ vertices, $v_1, v_2, ..., v_n$, and $m$ edges, $\{v_{i_1}, v_{j_1}\}, \{v_{i_2}, v_{j_2}\}, ..., \{v_{i_m}, v_{j_m}\}$, where $1 \leq i_1, j_1, i_2, j_2, ..., i_m, j_m \leq n$. Our purpose is to obtain the graph $G + v_i = G \uplus \{x\}$. Consider the following density matrix $\rho = (\frac{1}{2} \sum_{i=1,2} P([u_i])) \otimes \sigma(G)$, where $\{\{u_1\}, \{u_2\}\}$ forms an orthonormal basis of $C^2$. We associate the vertex $u_i$ to the state $|u_i\rangle$ for $i = 1, 2$. Consider the graph $H = \{\{u_1, u_2\}, \{\{u_1, u_1\}, \{u_2, u_2\}\}\}$. It is easy to check (cfr. Equation 3) that $\sigma_G(H) = \frac{1}{2} \sum_{i=1,2} P([u_i])$. Also observe that $\rho = \sigma(G \otimes G)$. Thus $H \otimes G$ is a graph on $2n$ vertices labeled by $u_1v_1, u_1v_2, ..., u_1v_n, u_2v_1, u_2v_2, ..., u_2v_n$ and with $2m$ edges

$$\{u_1v_{i_1}, u_1v_{j_1}\}, ..., \{u_1v_{i_m}, u_1v_{j_m}\}, \{u_2v_{i_1}, u_2v_{j_1}\}, ..., \{u_2v_{i_m}, u_2v_{j_m}\}.$$

So, $H \otimes G = H_1 \uplus H_2$, where $H_1 = \{\{u_1v_1, ..., u_1v_n\}, \{\{u_1v_{i_1}, u_1v_{j_1}\}, ..., \{u_1v_{i_m}, u_1v_{j_m}\}\}\}$ and $H_2 = \{\{u_2v_1, ..., u_2v_n\}, \{\{u_2v_{i_1}, u_2v_{j_1}\}, ..., \{u_2v_{i_m}, u_2v_{j_m}\}\}\}$. We first delete all the edges of $H \otimes G$ which are incident to the vertex $u_2v_1 \in V(H_2)$. Now, we perform the following projective measurement on $\sigma(G \otimes H)$:

$$M = \{I_{2n} - \frac{n}{i=2} P([u_2]\langle v_i|), \frac{n}{i=2} P([u_2]\langle v_i|)\}.$$

The probability that $I_{2n} - \frac{n}{i=2} P([u_2]\langle v_i|)$ clicks one and the state after the measurement is $\sigma(H_1 + u_2v_1)$, where $H_1 \cong G$.

5.3 LOCC

A local operation and classical communication (or short, LOCC) is a TPCP $\Lambda : B(\mathcal{H}_A \otimes \mathcal{H}_B) \rightarrow B(\mathcal{K}_A \otimes \mathcal{K}_B)$ defined in the following way. The TPCP $\Lambda$ is an LOCC if, for some $n > 0$, there exist sequences of Hilbert spaces $\mathcal{H}_A^{n+1}$ and $\mathcal{H}_B^{n+1}$ with $\mathcal{H}_A^n = \mathcal{H}_A$, $\mathcal{H}_B^n = \mathcal{H}_B$, $\mathcal{H}_A^{n+1} = \mathcal{K}_A$ and $\mathcal{H}_B^{n+1} = \mathcal{K}_B$, such that $\Lambda$ can be written in the following form $\Lambda(\sigma) = \sum_{i_1, ..., i_n} V_{i_1, ..., i_n}^{AB} \otimes \sigma(V_{i_1, ..., i_n}^{AB})$ for all $\sigma \in B(\mathcal{H}_A \otimes \mathcal{H}_B)$. Let $I_A^n : \mathcal{H}_A^n \rightarrow \mathcal{H}_A^n$ and $I_B^n : \mathcal{H}_B^n \rightarrow \mathcal{H}_B^n$ be identity operators. In $A(\sigma)$, $V_{i_1, ..., i_2n}^{AB} : \mathcal{H}_A \otimes \mathcal{H}_B \rightarrow \mathcal{K}_A \otimes \mathcal{K}_B$ is given by

$$V_{i_1, ..., i_2n}^{AB} = (I_A^{i_1} \otimes W_{2n-1-i_1}^{i_1}) (V_{2n-1-i_1}^{i_1}) (I_B^{i_1} \otimes W_{2n-2-i_1}^{i_1}) (I_A^{i_1} \otimes W_{2n-3-i_1}^{i_1}) ... (I_A^{i_1} \otimes W_{i_1}^{i_1}) (I_B^{i_1} \otimes W_{i_1}^{i_1}).$$

The sequences $(V_{2k-1-i_1}^{i_1} : \mathcal{H}_A^n \rightarrow \mathcal{H}_A^{n+1})_{k=1}^n$ and $(W_{2k-1-i_1}^{i_1} : \mathcal{H}_B^n \rightarrow \mathcal{H}_B^{n+1})_{k=1}^n$ are families of operators. For each sequence of indices $(i_2, ..., i_1)$ and for $k = 0, 1, ..., n-1$, $\sum_{i_{2k+1}^{i_1}, ..., i_n} V_{2k-1-i_1}^{i_1} = I_A^{i_1}$, for each sequence of indices $(i_2, ..., i_1)$ and for $k = 0, 1, ..., n-1$, $\sum_{i_{2k+2}^{i_1}, ..., i_n} W_{2k+1-i_1}^{i_1} = I_B^{i_1}$.

Thermodynamic Principle. One can not obtain an entangled state from a separable state by using LOCC.

A consequence of this principle is that, given two (possibly isomorphic) graphs $G$ and $H$ on $n = pq$ vertices, we can always obtain $\sigma(H)$ from $\sigma(G)$ by using LOCC only, if $\sigma(G)$ is separable or entangled and $\sigma(H)$ is separable, in $C^p \otimes C^q$.

Example 5.2 Let $G \cong K_2$ and let $\{11, 22\}, \{12, 21\} \in E(G)$. Then

$$\sigma(G) = \frac{1}{2} P(\frac{1}{\sqrt{2}}(|1\rangle\langle 1| - |2\rangle\langle 2|) + \frac{1}{2} P(\frac{1}{\sqrt{2}}(|1\rangle\langle 2| - |2\rangle\langle 1|)).$$

This density matrix is separable. Can we delete an edge of $G$ by LOCC? The answer is no. If we can delete $\{12, 21\}$ (or, equivalently, $\{11, 22\}$) by LOCC, we obtain $\sigma(G - \{12, 21\}) = P(\frac{1}{\sqrt{2}}(|1\rangle\langle 1| - |2\rangle\langle 2|)$, which is entangled. This fact violates the thermodynamic principle.
Problem 6.3 Let $G \cong K_4 - e$, for some edge $e$. Let $f$ be the edge of $G$ incident with the vertices of degree 3. Then $\sigma(G - f)$ is separable independent of the labeling. From $G$ we can always obtain $G - f$ by LOCC.

Example 5.3 Let $G$ be a graph with $|V| = pq$. If $G$ has only one entangled edge then $\sigma(G)$ is entangled; if all the entangled edges of $G$ are incident to the same vertex then $\sigma(G)$ is entangled.

Conjecture 6.5 Let $G$ be a graph $(|V| = pq)$. If $G$ has only one entangled edge then $\sigma(G)$ is entangled; if all the entangled edges of $G$ are incident to the same vertex then $\sigma(G)$ is entangled.

Conjecture 6.6 Let $G$ be a graph $(|V| = pq)$ with $m$ edges. If $\sigma(G)$ is entangled in $C_A^q \otimes C_B^q$ then $E_F(\rho) \approx \frac{1}{m} \sum_{k=1}^m E_F(\sigma(k))$, where $\sigma(k)$ is the pure density matrix associated to the $k$-th edge of $G$.

Conjecture 6.7 Let $G_n$ be the set of all connected graphs on $n$ vertices. Let $G \in G_n$ $(|V| = pq)$. Then $\max_{G_n} E_F(\sigma(G)) = E_F(\sigma(K_{1,n-1}))$.

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