

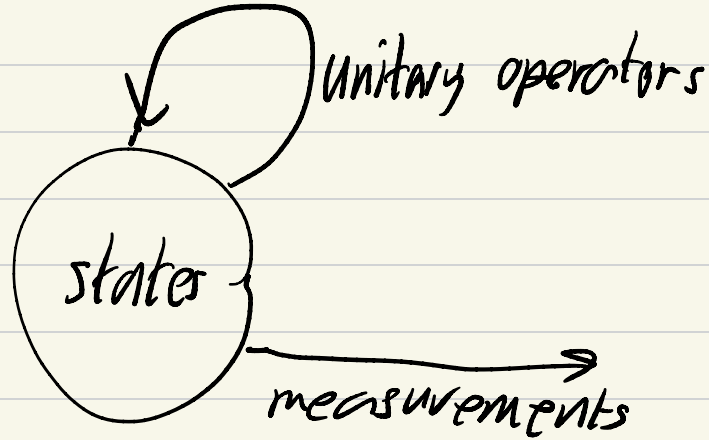


L 10/01

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# Quantum system:



# Density Matrices

A complex square matrix  $M$  is  
positive semidefinite if

$$(a) M = M^*$$

$$(b) u^* M u \geq 0, \forall u.$$

If, in addition,

$$(c) u^* M u = 0 \Rightarrow u = 0, \text{ then } M \text{ is positive definite.}$$

We write  $M \geq N$  to denote that  $M - N$  is psd.

If  $M$  is positive definite, we write  $M \succ 0$ .

## Examples

(a)  $I$

(b)  $\delta\delta^*$

(c) If  $M, N \neq 0$ , then  $M+N \neq 0$ .

(d) any projection (why?)

If  $A = [a_1, \dots, a_m]$  &  $B = [b_1, \dots, b_m]$ , both  $n \times m$ ,

then

$$AB^* = a_1 b_1^* + \dots + a_m b_m^*$$

matrix multiplication  
using outer products

It follows that

$$(e) \quad AA^* \succeq 0$$



A **density matrix** is a positive semidefinite matrix with trace one.

The simplest examples are the rank-one matrices  $zz^*$ , where  $\|z\|=1$ . Note that  $zz^*$  represents projection onto the 1-dim. subspace spanned by  $z$ . If  $a \in V(X)$ , we call  $e_a e_a^T$  a **vertex state**. If  $\|z\|=1$ , then  $zz^*$  is a **pure state**.

**Lemma** If  $M$  is positive definite, it is invertible.

**Proof** If  $Mz = 0$ , then  $z^* M z = 0$  and so  $z = 0$ .

Hence  $\ker(M) = \langle 0 \rangle$ .

□

Show that if  $M \succ 0$  and  $N \succ 0$ , then

$$M + N \succ 0.$$

**Lemma** A convex combination of density matrices is a density matrix.  $\square$

**Lemma.** Any psd matrix is a sum of positive semidefinite matrices of rank one. Any density matrix is a convex combination of pure states.  $\square$

**Theorem** For a Hermitian matrix  $M$ , the following are equivalent:

(a)  $M \succeq 0$ , i.e.  $u^* M u \geq 0$  for all  $u$ .

(b)  $M = N N^*$  for some  $n$

(c) all eigenvalues of  $M$  are non-negative.

(d) There are vectors  $z_1, \dots, z_m$  and

$$M = z_1 z_1^* + \dots + z_m z_m^* \quad \text{tr}(M^*N)$$

(e) If  $N \geq 0$ , then  $\langle M, N \rangle \geq 0$ .

We see that if  $M \geq 0$ , then  $\det(M) \geq 0$ . Also

**Lemma** Any principal submatrix of a positive semidefinite is positive semidefinite.  $\square$

$2 \times 2$  density matrices

prove this is positive  
semidefinite.

$$\rho = \begin{pmatrix} \frac{1}{2} + a & b + ic \\ b - ic & \frac{1}{2} - a \end{pmatrix}$$

$$\det(\rho) = \frac{1}{4} - a^2 - b^2 - c^2$$

If  $M \succeq 0$ , then its diagonal entries are non-negative; if the diagonal of  $M$  is zero, then  $M = 0$ .

**Lemma** If  $M \succeq 0$ , there is a unique psd matrix  $N$  s.t.  $M = N^2$ .  $\square$

A **quantum state** is a density matrix.

Physicists will say that the state space of a quantum system is a separable Hilbert space (our state spaces will be finite) and a state is a unit vector. In fact the state is a 1-dim subspace of  $\mathbb{C}^n$ . The projection onto this subspace is a rank-one density matrix, that is, a pure state.



Operations on quantum states are given by unitary operators. If  $D$  is a state of a system &  $U$  is unitary, then applying  $U$  brings the system to a state  $UDU^\dagger$ .

If  $D = |z\rangle\langle z|$  (with  $\|z\|=1$ ), then

$$UDU^\dagger = U|z\rangle\langle z|U^\dagger = U|z\rangle\langle z|U^\dagger$$

$|z\rangle \rightarrow U|z\rangle$

Bounds on cocliques

Let  $A$  be the adjacency matrix of a graph.

The  $A$  is real & symmetric, hence has a spectral decomp

$$A = \theta_0 E_0 + \dots + \theta_d E_d$$

Our default assumption is that  $\theta_0 \geq \dots \geq \theta_d$

Since  $\text{tr}(A) = 0$ , we have  $\sum_i \theta_i = 0$ . So either  $\theta_0 > 0$  and  $\theta_1 < 0$ , or  $\theta_0 = \dots = \theta_j = 0$  and  $A = 0$ .  $\square$

**Lemma** If  $k$  is the maximum valency of a vertex of  $X$ , then  $\theta_0 \leq k$ .

**Proof** Assume  $z \neq 0$  and  $Az = \theta_0 z$ . Then

$$\theta_0 z_r = \sum_{i=1}^r z_i$$



and, taking absolute values, we have

$$\theta_0 |z_r| \leq \sum |z_i|.$$

There are at most  $k$  terms in this sum. If  $r$  is

chosen so  $|z_r|$  is maximal;  $\theta_0 \leq \sum_i \frac{|z_i|}{|z_r|} \leq k$ .  $\square$

**Lemma** A graph  $X$  is regular if and only if  $\underline{1}$  is an eigenvector for  $A$ . □

**Lemma** If  $X$  is  $k$ -regular, then  $k$  is an eigenvalue of  $A$  with multiplicity equal to the number of components. □

**Lemma** Let  $X$  be a graph on  $n$  vertices. The following are equivalent:

(a)  $X$  is connected and regular

(b)  $J$  is a polynomial in  $A$ ,

$$(c) E_0 = \frac{1}{n} J$$

A *co clique* in a graph  $X$  is a set  $S$  of vertices such that no two vertices in  $S$  are adjacent.

The maximum size of (a.k.a. independent, stable)

a *co clique* in  $X$  is denoted by  $\alpha(X)$ .



Hoffman, Debarde

**Theorem** If  $X$  is  $k$ -regular with least eigenvalue  $\tau$ ,

then

$$\alpha(X) \leq \frac{v}{1 - \frac{k}{\tau}}$$

If  $x$  is the characteristic vector of a coclique of

size  $v/(1 - \frac{k}{\tau})$ , then  $x_{\mathcal{C}} - \frac{|\mathcal{C}|}{v} \mathbf{1}$  is an eigenvector

for  $X$  with eigenvalue  $\tau$ .

**Proof:** Assume  $S \subseteq V(X)$  with characteristic vector  $x$ . Then  $x^T A x$  is the number of pairs  $(u, v)$  in  $S \times S$  with  $u \sim v$ . Hence  $S$  is a coclique if & only if  $x^T A x = 0$ .

Next, set

$$B = A - \frac{k-\tau}{v} I$$

Then  $B \neq 0$  and so  $x^T B x \geq 0$ . Now

$$\begin{aligned} 0 \leq x^T B x &= x^T A x - \frac{k-\tau}{\nu} x^T J x - \tau x^T x \\ &= 0 - \frac{(k-\tau)|S|^2}{\nu} - \tau |S| \end{aligned}$$

and therefore  $|S| \leq \frac{\nu}{1-k\tau}$ .

The rest is left as an exercise.  $\square$