

# Linear Algebra, Quantum Information Theory, and Operator Algebras

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# Simplified Quantum Mechanics via Linear Algebra

- The *states of a quantum system* are the vectors

$$\mathcal{S} = \{\vec{v} \in \mathbb{C}^n \mid \|\vec{v}\|_2 = 1\}$$

for some fixed  $n$ .

- Two states  $\vec{v}$  and  $\vec{w}$  are considered the same (modulo a phase shift) if there exists a  $z \in \mathbb{C}$  such that  $|z| = 1$  and  $\vec{v} = z\vec{w}$ .
- Quantum mechanics was derived around the idea that observables of a quantum system are *quantized* in the sense that there will exist a basis of eigenstates  $\vec{v}_1, \vec{v}_2, \dots, \vec{v}_n$  where the value of the observable for each state is a fixed number  $\lambda_1, \lambda_2, \dots, \lambda_n$  respectively, and all other states are superpositions of the eigenstates.
- Thus observables are modelled by  $A \in M_{n,n}(\mathbb{C})$  where  $\vec{v}_k$  is an eigenvector with eigenvalue  $\lambda_k$  for all  $k$ .
- Matrices do not necessarily commute  $\implies$  uncertainty!

# Valid Observables

- As

$$\langle A\vec{v}_k, \vec{v}_k \rangle = \langle \lambda_k \vec{v}_k, \vec{v}_k \rangle = \lambda_k \|\vec{v}_k\|^2 = \lambda_k,$$

it is extrapolated that the value of the observable  $A$  on the state  $\vec{v} \in \mathcal{S}$  is  $\langle A\vec{v}, \vec{v} \rangle$ .

- As observed values should be real, we require  $A$  to be self-adjoint:

$$\begin{aligned}\langle A^*\vec{v}, \vec{w} \rangle &= \frac{1}{4} \sum_{k=1}^4 i^k \langle A^*(\vec{v} + i^k \vec{w}), \vec{v} + i^k \vec{w} \rangle \\ &= \frac{1}{4} \sum_{k=1}^4 i^k \langle \vec{v} + i^k \vec{w}, A(\vec{v} + i^k \vec{w}) \rangle \\ &= \frac{1}{4} \sum_{k=1}^4 i^k \overline{\langle A(\vec{v} + i^k \vec{w}), \vec{v} + i^k \vec{w} \rangle} \\ &= \frac{1}{4} \sum_{k=1}^4 i^k \langle A(\vec{v} + i^k \vec{w}), \vec{v} + i^k \vec{w} \rangle = \langle A\vec{v}, \vec{w} \rangle.\end{aligned}$$

# Spectral Theorem for Self-Adjoint Matrices

Conversely, if  $A$  is self-adjoint, then

$$\lambda_k = \lambda_k \langle \vec{v}_k, \vec{v}_k \rangle = \langle \lambda_k \vec{v}_k, \vec{v}_k \rangle = \langle A\vec{v}_k, \vec{v}_k \rangle = \langle \vec{v}_k, A\vec{v}_k \rangle = \langle \vec{v}_k, \lambda_k \vec{v}_k \rangle = \overline{\lambda_k}$$

so  $\lambda_k \in \mathbb{R}$  for all  $k$ .

## Theorem (Spectral Theorem for Self-Adjoint Matrices)

Let  $A \in M_{n,n}(\mathbb{C})$ . Then the following are equivalent:

- 1  $A$  is self-adjoint.
- 2 There exists an orthonormal basis of eigenvectors of  $A$  corresponding to real eigenvalues.
- 3 There exists a unitary matrix  $U \in M_{n,n}(\mathbb{C})$  and a diagonal matrix  $D \in M_{n,n}(\mathbb{R})$  such that  $A = UDU^*$ .

For observables, the eigenstates are an orthonormal basis for  $\mathbb{C}^n$ .

# Superposition of Eigenstates

If  $\vec{v} \in \mathcal{S}$ , then there exists  $a_1, a_2, \dots, a_n \in \mathbb{C}$  such that

$$\vec{v} = a_1 \vec{v}_1 + a_2 \vec{v}_2 + \dots + a_n \vec{v}_n.$$

Note

$$\langle A\vec{v}, \vec{v} \rangle = \sum_{k=1}^n \sum_{j=1}^n a_k \bar{a}_j \langle A\vec{v}_k, \vec{v}_j \rangle = \sum_{k=1}^n \sum_{j=1}^n a_k \bar{a}_j \lambda_k \langle \vec{v}_k, \vec{v}_j \rangle = \sum_{k=1}^n |a_k|^2 \lambda_k.$$

As  $\|\vec{v}_2\| = 1$ , we have by a similar computation that

$$\sum_{k=1}^n |a_k|^2 = 1.$$

Thus we view  $\vec{v}$  to be the state which is in the eigenstate  $\vec{v}_k$  with probability  $p_k = |a_k|^2$  for all  $k$ .

## Definition

Given a matrix  $B = [b_{i,j}] \in M_{m,m}(\mathbb{C})$ , the *trace* of  $B$  is

$$\mathrm{Tr}_m(B) = \sum_{k=1}^m b_{k,k}.$$

- Note if  $B \in M_{m,n}(\mathbb{C})$  and  $C \in M_{n,m}(\mathbb{C})$ , then  $\mathrm{Tr}_m(BC) = \mathrm{Tr}_n(CB)$ .
- Thinking of  $\vec{v} \in \mathcal{S}$  as a column vector

$$\langle A\vec{v}, \vec{v} \rangle = \mathrm{Tr}_1(\vec{v}^* A\vec{v}) = \mathrm{Tr}_n(A(\vec{v}\vec{v}^*)).$$

- We call  $\vec{v}\vec{v}^* \in M_{n,n}(\mathbb{C})$  the *density matrix of the state*  $\vec{v}$ .
- Note  $\mathrm{Tr}_n(\vec{v}\vec{v}^*) = \mathrm{Tr}_1(\vec{v}^* \vec{v}) = \|\vec{v}\|_2^2 = 1$ ,  $(\vec{v}\vec{v}^*)^* = \vec{v}\vec{v}^*$ , and if  $\vec{w}$  is an eigenvector for  $\vec{v}\vec{v}^*$  with eigenvalue  $\lambda$ ,

$$\lambda \langle \vec{w}, \vec{w} \rangle = \langle \vec{v}\vec{v}^* \vec{w}, \vec{w} \rangle = \langle \langle \vec{w}, \vec{v} \rangle \vec{v}, \vec{w} \rangle = |\langle \vec{w}, \vec{v} \rangle|^2 \geq 0.$$

Hence  $\vec{v}\vec{v}^*$  is a positive matrix (i.e. positive semi-definite) of trace 1.

# Quantum Channels

A *quantum channel* is a communication channel that can transmit quantum information. Such channels are important in quantum computing where a *qubit* (a two-dimensional quantum system) is used in place of a bit.

To proceed mathematically, quantum information is encoded via the density matrices of quantum states. Then quantum channels are precisely the functions  $\Phi : M_{n,n}(\mathbb{C}) \rightarrow M_{m,m}(\mathbb{C})$  such that

- $\Phi$  preserves the trace:  $\text{Tr}_m(\Phi(A)) = \text{Tr}_n(A)$  for all  $A \in M_{n,n}(\mathbb{C})$ ,
- $\Phi$  is positive:  $\Phi(A)$  is positive for all positive  $A \in M_{n,n}(\mathbb{C})$ ,
- $\Phi$  is linear, and
- $\Phi$  is *completely positive*.

To summarize, quantum channels are completely positive, trace-preserving linear maps.

# Completely Positive Maps

Note for all  $d, n \in \mathbb{N}$  that  $M_{d,d}(M_{n,n}(\mathbb{C})) \cong M_{dn, dn}(\mathbb{C})$ .

## Definition

A linear map  $\Phi : M_{n,n}(\mathbb{C}) \rightarrow M_{m,m}(\mathbb{C})$  is said to be *completely positive* if for every  $d \in \mathbb{N}$  the map  $\Phi_d : M_{d,d}(M_{n,n}(\mathbb{C})) \rightarrow M_{d,d}(M_{m,m}(\mathbb{C}))$  defined by

$$\Phi_d([A_{i,j}]) = [\Phi(A_{i,j})]$$

is a positive map.



# Examples of Positive Maps

- Define  $\Phi : M_{n,n}(\mathbb{C}) \rightarrow M_{n,n}(\mathbb{C})$  by  $\Phi(A) = A^T$ . As

$$\Phi(A^*A) = (A^*A)^T = A^T(A^*)^T = A^T(A^T)^*$$

we see that  $\Phi$  is positive.

- Define  $\Phi : M_{n,n}(\mathbb{C}) \rightarrow \mathbb{C}$  by  $\Phi(A) = \text{Tr}_n(A)$ . Then  $\Phi$  is positive as  $\text{Tr}_n(A)$  will be the sum of the eigenvalues of  $A$  (counting algebraic multiplicity).
- Define  $\Phi : M_{n,n}(\mathbb{C}) \rightarrow M_{n,n}(\mathbb{C})$  by

$$\Phi([a_{i,j}]) = \text{diag}(a_1, a_2, \dots, a_n).$$

Then  $\Phi$  is positive as a positive matrix must have positive entries along the diagonal and a diagonal matrix is positive if and only if the entries along the diagonal are positive.

# Transpose and Complete Positivity

Define  $\Phi : M_{2,2}(\mathbb{C}) \rightarrow M_{2,2}(\mathbb{C})$  by  $\Phi(A) = A^T$ . Then  $\Phi$  is not completely positive. Indeed consider

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

Clearly  $A$  is self-adjoint with eigenvalues 0, 0, 0, and 2, so  $A$  is positive. However

$$\Phi_2(A) = \begin{bmatrix} \begin{bmatrix} 1 & 0 \\ 0 & 0 \end{bmatrix} & \begin{bmatrix} 0 & 0 \\ 1 & 0 \end{bmatrix} \\ \begin{bmatrix} 0 & 1 \\ 0 & 0 \end{bmatrix} & \begin{bmatrix} 0 & 1 \\ 0 & 1 \end{bmatrix} \end{bmatrix} \cong \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix}$$

which has eigenvalues 1, 1, 1, and  $-1$  and thus is not positive.

# Trace and Complete Positivity

Define  $\Phi : M_{n,n}(\mathbb{C}) \rightarrow \mathbb{C}$  by  $\Phi(A) = \text{Tr}_n(A)$ . Then  $\Phi$  is completely positive. Indeed let  $A = [A_{i,j}] \in M_{d,d}(M_{n,n}(\mathbb{C}))$  be positive. Then for all  $\vec{x} = (x_1, x_2, \dots, x_d) \in \mathbb{C}^d$  we have

$$\langle \Phi_d(A)\vec{x}, \vec{x} \rangle = \sum_{i,j=1}^d \text{Tr}_n(A_{i,j})x_j\bar{x}_i = \text{Tr}_n \left( \sum_{i,j=1}^d \bar{x}_i A_{i,j} x_j \right).$$

However, writing  $A = B^*B$  we see that

$$\sum_{i,j=1}^d \bar{x}_i A_{i,j} x_j = \begin{bmatrix} x_1 I_n \\ \vdots \\ x_d I_n \end{bmatrix}^* A \begin{bmatrix} x_1 I_n \\ \vdots \\ x_d I_n \end{bmatrix} = \left( B \begin{bmatrix} x_1 I_n \\ \vdots \\ x_d I_n \end{bmatrix} \right)^* \left( B \begin{bmatrix} x_1 I_n \\ \vdots \\ x_d I_n \end{bmatrix} \right)$$

is a positive matrix. Therefore, as  $\text{Tr}_n$  is positive,  $\langle \Phi_d(A)\vec{x}, \vec{x} \rangle \geq 0$ . Hence  $\Phi_d(A)$  is positive.

# Trace and Complete Positivity

For a fixed positive matrix  $B \in M_{n,n}(\mathbb{C})$ , define  $\Phi_B : M_{n,n}(\mathbb{C}) \rightarrow \mathbb{C}$  by  $\Phi_B(A) = \text{Tr}_n(AB)$ . Then  $\Phi_B$  is completely positive. Indeed, as  $B$  is positive, we can write  $B = C^*C$  for some  $C \in M_{n,n}(\mathbb{C})$ . Then for all positive  $A = [A_{i,j}] \in M_{d,d}(M_{n,n}(\mathbb{C}))$  we have that

$$(\Phi_B)_d(A) = [\text{Tr}_n(C^*CA_{i,j})] = [\text{Tr}_n(CA_{i,j}C^*)] = (\text{Tr}_n)_d(SAS^*)$$

where

$$S = \text{diag}(C, C, \dots, C).$$

As  $A$  is positive, we can write  $A = T^*T$  for some  $T \in M_{d,d}(M_{n,n}(\mathbb{C}))$  so

$$SAS^* = (ST)(ST)^*$$

is positive. Therefore, as  $\text{Tr}_n$  is completely positive,  $(\Phi_B)_d(A)$  is positive.

Thus 'positive' observables define completely positive maps on the density matrices.

# Diagonals and Complete Positivity

Define  $\Phi : M_{n,n}(\mathbb{C}) \rightarrow M_{n,n}(\mathbb{C})$  by

$$\Phi([a_{i,j}]) = \text{diag}(a_1, a_2, \dots, a_n).$$

Then  $\Phi$  is completely positive. Indeed let  $A = [A_{i,j}]_{i,j} \in M_{d,d}(M_{n,n}(\mathbb{C}))$  be positive. Write  $A_{i,j} = [a_{i,j,p,q}]_{p,q}$ . Then

$$\Phi_d(A) = \begin{bmatrix} \begin{bmatrix} a_{1,1,1,1} & & 0 \\ & \ddots & \\ 0 & & a_{1,1,n,n} \end{bmatrix} & \cdots & \begin{bmatrix} a_{1,d,1,1} & & 0 \\ & \ddots & \\ 0 & & a_{1,d,n,n} \end{bmatrix} \\ \vdots & & \vdots \\ \begin{bmatrix} a_{d,1,1,1} & & 0 \\ & \ddots & \\ 0 & & a_{d,1,n,n} \end{bmatrix} & \cdots & \begin{bmatrix} a_{d,d,1,1} & & 0 \\ & \ddots & \\ 0 & & a_{d,d,n,n} \end{bmatrix} \end{bmatrix}$$

# Diagonals and Complete Positivity

Reordering the basis (which is conjugating by a unitary and thus preserves positivity), we obtain that  $\Phi_d(A)$  is positive if and only if

$$C = \text{diag} \left( \begin{bmatrix} a_{1,1,1,1} & \cdots & a_{1,d,1,1} \\ \vdots & & \vdots \\ a_{d,1,1,1} & \cdots & a_{d,d,1,1} \end{bmatrix}, \dots, \begin{bmatrix} a_{1,1,n,n} & \cdots & a_{1,d,n,n} \\ \vdots & & \vdots \\ a_{d,1,n,n} & \cdots & a_{d,d,n,n} \end{bmatrix} \right)$$

is positive. By conjugating  $A$  by the same unitary matrix, we get that  $T = [[a_{i,j,p,q}]_{p,q}]_{i,j}$  is positive. As each diagonal entry in  $C$  is a diagonal minor of  $T$  and thus positive, we quickly check that  $C$  is positive.

# Quantum to Classical

For the remainder of the talk, we will use  $E$  to denote the *expectation onto the diagonal map*; that is,  $E : M_{n,n}(\mathbb{C}) \rightarrow M_{n,n}(\mathbb{C})$  is defined by

$$E([a_{i,j}]) = \text{diag}(a_1, a_2, \dots, a_n).$$

The expectation map is of interest as it takes matrices (quantum information) and produces commuting matrices (classical information).

## Question

Given a self-adjoint  $A \in M_{n,n}(\mathbb{R})$ , can we determine the possible values of  $E(A)$  based on the eigenvalues of  $A$ ?

To be specific, if  $D = \text{diag}(x_1, x_2, \dots, x_n)$  with  $x_1, x_2, \dots, x_n \in \mathbb{R}$ , then every self-adjoint matrix with eigenvalues  $x_1, x_2, \dots, x_n$  is of the form  $UDU^*$  for some unitary matrix  $U \in M_{n,n}(\mathbb{C})$ . So what is

$$\{E(UDU^*) \mid U \in M_{n,n}(\mathbb{C}) \text{ a unitary}\}?$$

# Expectations to Doubly Stochastic Matrices

Fix  $x_1, x_2, \dots, x_n \in \mathbb{R}$  with  $x_1 \geq x_2 \geq \dots \geq x_n$  and  $D = \text{diag}(x_1, x_2, \dots, x_n)$ . Let  $U = [u_{i,j}]$  and suppose  $E(UDU^*) = \text{diag}(y_1, y_2, \dots, y_n)$ . As

$$UDU^* = \left[ \sum_{k=1}^n u_{i,k} x_k \overline{u_{j,k}} \right],$$

we see that

$$y_i = \sum_{k=1}^n |u_{i,k}|^2 x_k$$

for all  $i$ . Thus if

$$S = [|u_{i,j}|^2], \quad \vec{x} = (x_1, x_2, \dots, x_n)^T, \quad \text{and} \quad \vec{y} = (y_1, y_2, \dots, y_n)^T$$

we have  $S\vec{x} = \vec{y}$ . Moreover, the sum of any row or column of  $S$  is 1 as  $U$  is a unitary matrix. Such a matrix is an example of a *doubly stochastic matrix*.



# Doubly Stochastic Matrices to Majorization

Note

$$\sum_{i=1}^n y_i = \sum_{i=1}^n \sum_{k=1}^n |u_{i,k}|^2 x_k = \sum_{k=1}^n \sum_{i=1}^n |u_{i,k}|^2 x_k = \sum_{k=1}^n x_k$$

(i.e.  $\text{Tr}(E(UDU^*)) = \text{Tr}(D)$ ). For distinct  $i_1, i_2, \dots, i_\ell \in \{1, 2, \dots, n\}$ ,

$$\sum_{p=1}^{\ell} y_{i_p} = \sum_{p=1}^{\ell} \sum_{k=1}^n |u_{i_p,k}|^2 x_k \leq \sum_{k=1}^{\ell} x_k.$$

## Definition

Given  $\vec{x} = (x_1, x_2, \dots, x_n) \in \mathbb{R}^n$  and  $\vec{y} = (y_1, y_2, \dots, y_n) \in \mathbb{R}^n$ , it is said that  $\vec{x}$  majorizes  $\vec{y}$ , denoted  $\vec{y} \prec \vec{x}$ , if

- $\sum_{k=1}^n y_k = \sum_{k=1}^n x_k$ , and
- for all  $l$ ,

$$\max\left\{\sum_{p=1}^{\ell} y_{i_p} \mid i_1, \dots, i_\ell \in \{1, 2, \dots, n\} \text{ distinct}\right\} \leq \max\left\{\sum_{p=1}^{\ell} x_{i_p} \mid i_1, \dots, i_\ell \in \{1, 2, \dots, n\} \text{ distinct}\right\}.$$

# Majorization to Permutation Majorization

Suppose  $\vec{x} = (x_1, x_2, \dots, x_n) \in \mathbb{R}^n$  and  $\vec{y} = (y_1, y_2, \dots, y_n) \in \mathbb{R}^n$  with

$$\vec{y} \prec \vec{x}, \quad x_1 \geq x_2 \geq \dots \geq x_n \quad \text{and} \quad y_1 \geq y_2 \geq \dots \geq y_n.$$

As  $\vec{y} \prec \vec{x}$ ,  $x_1 \geq y_1$  and if  $x_k = y_k$  for all  $1 \leq k < \ell$ , then  $x_\ell \geq y_\ell$ . Let  $p$  be the first index where  $x_p > y_p$ . As  $\sum_{k=1}^n x_k = \sum_{k=1}^n y_k$ , there exists a first index  $q$  where  $x_q < y_q$ . So  $x_p > y_p \geq y_q > x_q$ . Choose  $\theta \in [0, 2\pi]$  such that if

$$x'_p = \cos^2(\theta)x_p + \sin^2(\theta)x_q \quad \text{and} \quad x'_q = \sin^2(\theta)x_p + \cos^2(\theta)x_q$$

then either  $x'_p = y_p$  and  $x'_q < y_q$ , or  $x'_p > y_p$  and  $x'_q = y_q$ . As

$$x'_p + x'_q = x_p + x_q$$

and  $x_k \geq y_k$  for all  $p < k < q$ , if  $x'_k = x_k$  for all  $k \neq p, q$ , then

$$(y_1, y_2, \dots, y_n) \prec (x'_1, x'_2, \dots, x'_n) \prec (x_1, x_2, \dots, x_n).$$

# Permutation Majorization to Expectations

Repeating, we get a chain

$$\vec{y} = \vec{x}_r \prec \vec{x}_{r-1} \prec \cdots \prec \vec{x}_1 \prec \vec{x}_0 = \vec{x}$$

where  $\vec{x}_k$  and  $\vec{x}_{k+1}$  differ in only two entries in the above way and the pairs of entries differ from all other pairs of  $(\vec{x}_\ell, \vec{x}_{\ell+1})$ . With the notation above, if

$$U_1 = \cos(\theta)E_{p,p} - \sin(\theta)E_{p,q} + \sin(\theta)E_{q,p} + \cos(\theta)E_{q,q} + \sum_{\substack{k=1 \\ k \neq p,q}}^n E_{k,k},$$

then the diagonal entries of  $U_1 \text{diag}(x_1, x_2, \dots, x_n) U_1^*$  are  $\vec{x}_1$ . This can be repeated with each progressive unitary  $U_k$  not disturbing the previously 'corrected' diagonal entries in lieu of new off-diagonal entries so that when we take the product of the  $U_k$  we get a unitary  $U$  such that

$$E(UDU^*) = \text{diag}(y_1, y_2, \dots, y_n).$$

# Schur-Horn Theorem

## Theorem (Schur-Horn Theorem; 1923, 1954, etc.)

Let  $\vec{x} = (x_1, x_2, \dots, x_n) \in \mathbb{R}^n$  and  $\vec{y} = (y_1, y_2, \dots, y_n) \in \mathbb{R}^n$ . The following are equivalent:

- If  $D = \text{diag}(x_1, x_2, \dots, x_n)$ , there exists a unitary  $U \in M_{n,n}(\mathbb{R})$  such that

$$E(UDU^*) = \text{diag}(y_1, y_2, \dots, y_n).$$

- There exists a doubly stochastic matrix  $A \in M_{n,n}(\mathbb{R})$  such that  $A\vec{x} = \vec{y}$ .
- $\vec{y} \prec \vec{x}$ .
- There exists a chain

$$\vec{y} = \vec{x}_r \prec \vec{x}_{r-1} \prec \dots \prec \vec{x}_1 \prec \vec{x}_0 = \vec{x}$$

where  $\vec{x}_k$  and  $\vec{x}_{k+1}$  differ in only two entries in the above way and the pairs of entries differ from all other pairs of  $(\vec{x}_\ell, \vec{x}_{\ell+1})$ .

# Convex Hull of Unitary Orbit

Convexity is important in functional analysis. Given  $A \in M_{n,n}(\mathbb{C})$ , let

$$\text{conv}(\mathcal{U}(A)) = \left\{ \sum_{k=1}^m t_k U_k A U_k^* \mid \begin{array}{l} U_1, U_2, \dots, U_m \in M_{n,n}(\mathbb{C}) \text{ unitaries,} \\ t_1, t_2, \dots, t_m \in (0,1), \text{ and} \\ t_1 + t_2 + \dots + t_m = 1 \end{array} \right\}.$$

If  $A = \text{diag}(a_1, a_2, \dots, a_n) \in M_{n,n}(\mathbb{R})$ , if  $B = \text{diag}(b_1, b_2, \dots, b_n)$  is the above matrix, and  $U_k = [u_{i,j,k}]$  then

$$\vec{b} = \left( \sum_{k=1}^m t_k [ |u_{i,j,k}|^2 ] \right) \vec{a},$$

so  $\vec{b} \prec \vec{a}$ . Conversely, by using the chain of majorizations, and

$$\begin{bmatrix} x'_p & 0 \\ 0 & x'_q \end{bmatrix} = \cos^2(\theta) \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} \begin{bmatrix} x_p & 0 \\ 0 & x_q \end{bmatrix} \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} + \sin^2(\theta) \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix} \begin{bmatrix} x_p & 0 \\ 0 & x_q \end{bmatrix} \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}$$

if  $\vec{b} \prec \vec{a}$  then  $\text{diag}(b_1, b_2, \dots, b_n) \in \text{conv}(\mathcal{U}(\text{diag}(a_1, a_2, \dots, a_n)))$ .

## Theorem

Let  $A, B \in M_{n,n}(\mathbb{C})$  be self-adjoint. Then the following are equivalent:

- The eigenvalues of  $B$  are majorized by the eigenvalues of  $A$ .
- $B \in \text{conv}(\mathcal{U}(A))$ .

Let  $P = \text{diag}(1, 0, 0, \dots, 0) \in M_{n,n}(\mathbb{R})$ . Note if  $B \in M_{n,n}(\mathbb{C})$ , then  $B \prec P$  if and only if  $\text{Tr}(B) = 1$  and  $B$  is positive. Thus for any positive matrix  $B$  of trace 1 we can write

$$B = \sum_{k=1}^m t_k U_k P U_k^*.$$

If  $\vec{u}_1, \dots, \vec{u}_m$  are the first rows of  $U_1, \dots, U_m$  respectively, then

$$B = \sum_{k=1}^m t_k \vec{u}_k \vec{u}_k^*$$

Thus  $B$  represents the density matrix of a superposition of states!

# To Infinity and Beyond

- A *Hilbert space* is an inner product space that is complete (i.e. every Cauchy sequence converges) with respect to the norm

$$\|\vec{x}\| = \sqrt{\langle \vec{x}, \vec{x} \rangle}.$$

- Given a Hilbert space  $\mathcal{H}$ , an *orthonormal basis* is a maximal orthonormal set.
- Every Hilbert space with an infinitely countable orthonormal basis is isomorphic to

$$\ell_2 = \left\{ (x_n)_{n \geq 1} \mid \sum_{n=1}^{\infty} |x_n|^2 < \infty \right\}.$$

- If  $L_2[0, 1]$  is the square-integrable 'functions' on  $[0, 1]$ , then  $L_2[0, 1] \cong \ell_2$ .

# Bounded Linear Maps

- Given two Hilbert spaces  $\mathcal{H}$  and  $\mathcal{K}$ , the *bounded linear maps from  $\mathcal{H}$  to  $\mathcal{K}$* , denoted  $\mathcal{B}(\mathcal{H}, \mathcal{K})$ , are all linear maps  $T : \mathcal{H} \rightarrow \mathcal{K}$  such that

$$\|T\| = \sup \{ \|T(\vec{x})\|_{\mathcal{K}} \mid \vec{x} \in \mathcal{H}, \|\vec{x}\|_{\mathcal{H}} \leq 1 \} < \infty.$$

- If  $T \in \mathcal{B}(\mathcal{H}, \mathcal{K})$ , then there exists a  $T^* \in \mathcal{B}(\mathcal{K}, \mathcal{H})$  such that

$$\langle T^*(\vec{x}), \vec{y} \rangle_{\mathcal{H}} = \langle \vec{x}, T(\vec{y}) \rangle_{\mathcal{K}}$$

for all  $\vec{x} \in \mathcal{K}$  and  $\vec{y} \in \mathcal{H}$ .

- A *C\*-algebra* is a norm-closed subalgebra  $\mathfrak{A}$  of  $\mathcal{B}(\mathcal{H})$  such that if  $A \in \mathfrak{A}$  then  $A^* \in \mathfrak{A}$ .
- An  $A \in \mathfrak{A}$  is *positive* if  $A = B^*B$  for some  $B \in \mathfrak{A}$ ; equivalently  $\langle A\vec{x}, \vec{x} \rangle_{\mathcal{H}} \geq 0$  for all  $\vec{x} \in \mathcal{H}$ .
- A  *$II_1$  factor* is a unital C\*-algebra with a faithful tracial state, trivial centre, and is closed in the weak operator topology.



# Schur-Horn Theorem in $II_1$ Factors

## Theorem (Ravichandran; preprint 2012)

Let  $\mathfrak{M}$  be a type  $II_1$  factor, let  $\mathcal{A}$  be a MASA in  $\mathfrak{M}$ , and let  $E_{\mathcal{A}} : \mathfrak{M} \rightarrow \mathcal{A}$  denote the normal conditional expectation. Given self-adjoint elements  $T \in \mathfrak{M}$  and  $S \in \mathcal{A}$ , the following are equivalent:

- 1  $S \prec T$ ,
- 2 there exists an element  $R \in \overline{\mathcal{U}}(T)$  such that  $E_{\mathcal{A}}(R) = S$ .

## Theorem (Kennedy, Skoufranis; 2015)

Let  $\mathfrak{M}$  be a type  $II_1$  factor, let  $\mathcal{A}$  be a MASA in  $\mathfrak{M}$ , and let  $E_{\mathcal{A}} : \mathfrak{M} \rightarrow \mathcal{A}$  denote the normal conditional expectation. Given elements  $T \in \mathfrak{M}$  and  $S \in \mathcal{A}$ , the following are equivalent:

- 1  $S \prec_w T$ ,
- 2 there exists an element  $R \in \mathfrak{M}$  with  $\sigma_R = \sigma_T$  such that  $E_{\mathcal{A}}(R) = S$ .

In addition, the above is logically equivalent to Ravichandran's result.

# Closed Convex Hull of Unitary Orbits in $C^*$ -Algebras

Let  $\mathfrak{A}$  be a unital  $C^*$ -algebra and let  $\mathcal{T}(\mathfrak{A})$  denote all 'unbounded traces'; that is, all maps  $\tau : \mathfrak{A}_+ \rightarrow [0, \infty]$  such that

- $\tau(T + S) = \tau(T) + \tau(S)$  for all  $T, S \in \mathfrak{A}_+$ ,
- $\tau(\alpha T) = \alpha\tau(T)$  for all  $T \in \mathfrak{A}_+$  and  $\alpha \in \mathbb{R}_+$  ( $0 \cdot \infty = 0$ ),
- $\tau(X^*X) = \tau(XX^*)$  for all  $X \in \mathfrak{A}$ , and
- $\tau$  is lower semicontinuous.

## Theorem (Ng, Robert, Skoufranis; 2018)

Let  $\mathfrak{A}$  be a unital  $C^*$ -algebra and let  $T, S \in \mathfrak{A}$  be self-adjoint. The following are equivalent:

- $S \in \overline{\text{conv}}(\mathcal{U}(T))$ .
- $\tau((S - \alpha)_+) \leq \tau((T - \alpha)_+)$  and  $\tau((-S - \alpha)_+) \leq \tau((-T - \alpha)_+)$  for all  $\tau \in \mathcal{T}(\mathfrak{A})$  and  $\alpha \in \mathbb{R}$ .

# Courses to Take

## Undergraduate:

- MATH 2022: Linear Algebra II
- MATH 3001: Analysis II
- MATH 4011: Metric Spaces
- MATH 4012: Lebesgue Measure Theory
- PHYS 4010: Quantum Mechanics
- EECS 4141: Introduction to Quantum Computing?

## Graduate:

- MATH 6280: Measure Theory
- MATH 6450: Topology
- MATH 6461: Functional Analysis I
- MATH 6462: Functional Analysis II
- PHYS 5000: Quantum Mechanics I

Thanks for Listening!