Ungleichungen und ähnlich verwirrende Konzepte

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Outline

- Introduction
 - ksjladüys
- Local and quantum correlation matrices
 - Local correlation matrices
 - Quantum correlation matrices
 - The relations between quantum correlation and local correlation matrices
- Grothendieck Inequality

Let's get on the same page

- We should know what a state is
- We should know what the tensor product does
- We should be familiar with the Dirac notation

Quantum systems

- A quantum system is a portion of the whole universe. For example a set electrons.
- A quantum system X is associated with a copy of \mathbb{C}^k
- It may consist of subsystems X_1, \ldots, X_N each of which is associated with a copy of \mathbb{C}^{n_i} . In this case $k = n_1 \ldots n_N$

Measurements

- ullet A measurement can be performed on a system X that is in state ho
- ullet Let $\mathcal A$ be a finite set of outcomes of the measurement
- The measurement itself is defined by a set of psd matrices $\{F^a\}_{a\in\mathcal{A}}\subseteq\mathbb{C}^{n\times n}$ that sum up to the identity matrix, i.e. $\sum_{a\in\mathcal{A}}F^a=I$

Measurements

- A projective measurement is defined by psd matrices that satisfy $F^aF^b=\delta_{ab}F^a\ \forall a,b\in\mathcal{A}$
- The outcome of a measurement is a random variable χ with probability distribution: $\mathbb{P}[\chi=a]=\mathrm{Tr}(\rho F^a)$
- ullet To define an expected value we define outcomes in ${\mathcal A}$ as real numbers



Measurements

- $\mathbb{E}[\chi] = \sum_{a \in \mathcal{A}} a \mathrm{Tr}(\rho F^a) = \mathrm{Tr}(\rho(\sum_{a \in \mathcal{A}} a F^a))$
- $\sum_{a \in \mathcal{I}} aF^a$ is called observable
- A simple case we will use later are $\{-1,1\}$ -valued observables
- if we consider projective measurements we have

$$(F^{+} - F^{-})^{2} = \underbrace{F^{+^{2}}}_{=F^{+}} - \underbrace{F^{+}F^{-}}_{\delta_{+-}=0} + \underbrace{F^{-^{2}}}_{F^{-}} = F^{+} + F^{-} = I$$

ullet i.e. a $\{-1,1\}$ -valued observable is both unitary an Hermitian



Doling out subsystems

- Consider a system X consisting of subsystems $X_1, \ldots X_N$ which we distribute among N parties, which may be located anywhere in the universe
- The parties *share* the state X is in
- Every party may perform a measurement on their subsystem X_i , i.e. there are N sets of psd matrices $\{F^{a_1}\}_{a_1\in\mathcal{A}_1}\in\mathbb{C}^{n_1\times n_1},\ldots,\{F^{a_N}\}_{a_N\in\mathcal{A}_N}\in\mathbb{C}^{n_N\times n_N}$

The joint probability distribution of the N measurement outcomes χ_1,\ldots,χ_N is

$$\mathbb{P}\left[\chi_1=\mathsf{a}_1,\chi_2=\mathsf{a}_2,\ldots,\chi_N=\mathsf{a}_N\right]=\mathsf{Tr}\big(\rho \mathsf{F}_1^{\mathsf{a}_1}\otimes\cdots\otimes \mathsf{F}_N^{\mathsf{a}_N}\big)$$

Entanglement

- We will only consider pure states meaning states that they have rank 1 and therefore can be written as $\rho=|\psi\rangle\langle\psi|$
- A state is called product state if it can be written as $|\psi\rangle = |\psi_1\rangle |\psi_2\rangle \dots |\psi_N\rangle$
- When a vector $|\psi\rangle$ is referred to as a state we mean the matrix $|\psi\rangle\langle\psi|$
- A state that is not a product state is called entangled

Example

- Let $|\psi\rangle = |\psi_A\rangle |\psi_B\rangle$ be a system and give $|\psi_A\rangle$ to Alice and $|\psi_B\rangle$ to Bob
- Let them perform measurements $\{G^b\}_{b\in\mathcal{B}}$ and $\{F^a\}_{a\in\mathcal{A}}$ on their respective quantum systems
- What is the probability of Alice getting measurement outcome $\chi_A = a$ and Bob getting $\chi_B = b$?

Example

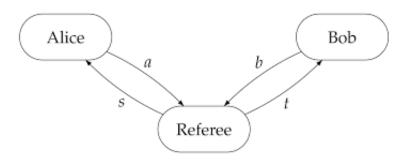
$$\begin{aligned} \operatorname{Tr}(|\psi\rangle\langle\psi|F^{a}\otimes G^{b}) &= \langle\psi|F^{a}\otimes G^{b}|\psi\rangle \\ &= (\langle\psi_{A}|\otimes\langle\psi_{B}|)(F^{a}\otimes G^{b})(|\psi_{A}\rangle\otimes|\psi_{B}\rangle) \\ &= ((\langle\psi_{A}|F^{a})\otimes(\langle\psi_{B}|G^{b}))(|\psi_{A}\rangle\otimes|\psi_{B}\rangle) \\ &= \langle\psi_{A}|F^{a}|\psi_{A}\rangle\otimes\langle\psi_{B}|G^{b}|\psi_{B}\rangle \\ &= \langle\psi_{A}|F^{a}|\psi_{A}\rangle\langle\psi_{B}|G^{b}|\psi_{B}\rangle \end{aligned}$$

This is equal to the product of the probabilities of Alice measuring a and Bob measuring b, i.e. the outcome do not correlate.

Nonlocal games

- Three participants: Alice, Bob and a referee
- Referee doles out a question s to Alice and a question t to Bob
- Alice and Bob are assumed to be located anywhere in the universe respectively
- Alice and Bob must not communicate
- Alice sends answer a Bob sends answer b back to the referee, who then
 decides whether both win or both lose

Nonlocal games



Mathematically speaking

- Four finite sets $\mathcal{A}, \mathcal{B}, \mathcal{S}, \mathcal{T}$
- probability distribution π over $\mathcal{S} \times \mathcal{T}$ $\pi: \mathcal{S} \times \mathcal{T} \to [0,1]$
- The referee sends with probability $\pi(s,t)$ s to Alice and t to Bob
- ullet They answer with an element $a\in\mathcal{A}$ and $b\in\mathcal{B}$ respectively
- $\bullet \ \mathsf{A} \ \mathsf{map} \ \mathsf{V} : \mathcal{S} \times \mathcal{T} \times \mathcal{A} \times \mathcal{B} \to \{0,1\}$
- They win if V(s, t, a, b) = 1 and lose otherwise

Classical strategies

- ullet All players know π and V and the information they received but not what the other players received
- They are allowed to agree on a strategy beforehand

Classical strategies

- ullet All players know π and V and the information they received but not what the other players received
- They are allowed to agree on a strategy beforehand but must not communicate once the game started
- A deterministic strategy is a map $a: \mathcal{S} \to \mathcal{A}$ for Alice and $b: \mathcal{T} \to \mathcal{B}$ for Bob The winning probability then is:

$$\mathbb{E}_{s,t\sim\pi}\left[V(a(s),b(t),s,t)\right]$$

Quantum case

- Suppose Alice and Bob have a subsystem X_A, X_B of a quantum system X which is in state ρ , i.e. Alice and Bob share state ρ
- If the state is entangled measurements can give correlated measurement outcomes
- Alice and Bob may gain information by performing measurements
- Answering according to measurement outcomes could increase winning probability

Mathematically speaking

- A quantum system X consisting of two n-dimensional subsystems X_A, X_B in some entangled state ρ
- Alice performs a measurement $\{F_s^a\}_{a\in\mathcal{A}}\subseteq\mathbb{C}^{n\times n}$ on her subsystem X_A and Bob performs a measurement $\{G_t^b\}_{b\in\mathcal{B}}\subseteq\mathbb{C}^{n\times n}$ on his subsystem X_B
- They send their measurement outcome as their answer to the referee
- Their winning probability is:

$$\mathbb{E}_{s,t \sim \pi} \left[\sum_{a \in \mathcal{A}} \sum_{b \in \mathcal{B}} \mathsf{Tr}(\rho \mathsf{F}^{a}_{s} \otimes \mathsf{G}^{b}_{t}) \mathsf{V}(a,b,s,t) \right]$$

•	Since states are convex combinations of pure states and the trace function is	
linear we only need to consider pure entangled states		

Two player XOR games

- Let the sets $\mathcal A$ and $\mathcal B$ be $\{0,1\}$, so Alice and Bob both answer either with 1 or 0
- The predicate V is defined as $V(a,b,s,t)=[a\oplus b=f(s,t)]$, where $f:\mathcal{S}\times\mathcal{T}\to\{0,1\}$
- A truth table for $a \oplus b$ looks like this

\oplus	0	1
0	0	1
1	1	0

Bias and violation ratio

- \bullet Alice and Bob can always win with probability $\frac{1}{2}$ by flipping an unbiased coin
- The classical bias of an XOR game G is defined as the difference of the probabilities of winning and losing for an optimal strategy and denoted by $\beta(G)$
- The bias $\beta^*(G)$ of entangled strategies is calculated the same way
- ullet It is twice the amount by which the maximal winning probability exceeds $\frac{1}{2}$
- The violation ratio is defined as $\frac{\beta^*(G)}{\beta(G)}$

Signs and observables

- It is convenient to use the $\{-1,1\}$ -basis instead of the $\{0,1\}$ -basis for boolean valued objects.
- Let $a: \mathcal{S} \to \{0,1\}$ and $b: \mathcal{T} \to \{0,1\}$ be classical strategies and π the probability distribution the referee uses to pick s,t
- The bias is given by the probability under π that $a(s) \oplus b(t) = f(s,t)$ minus the probability under π that $a(s) \oplus b(t) \neq f(s,t)$

This means the bias can be written as:

$$\mathbb{E}_{(s,t)\sim\pi} \left[(-1)^{[a(s)\oplus b(t)=f(s,t)]} \right] =$$

$$= \mathbb{E}_{(s,t)\sim\pi} \left[(-1)^{a(s)\oplus b(t)+f(s,t)} \right] =$$

$$= \mathbb{E}_{(s,t)\sim\pi} \left[(-1)^{a(s)} (-1)^{b(t)} (-1)^{f(s,t)} \right]$$

And we can define the sign matrix $\Sigma_{s,t}=(-1)^{f(s,t)}$ and functions $\chi(s)=(-1)^{a(s)}$ and $\psi(t)=(-1)^{b(t)}$. So the bias is

$$\mathbb{E}_{(s,t)\sim\pi}\left[\chi(s)\psi(t)\Sigma_{st}\right]$$

- ullet The outcomes in an XOR game are $\{0,1\}$
- Alice and Bob have measurements $\{F_s^0, F_s^1\}$ and $\{G_t^0, G_t^1\}$ and share an entangled state
- The probability of Alice and Bob answering with a,b upon receiving s,t respectively is $\langle \psi | F_s^a \otimes G_t^b | \psi \rangle$
- Lets calculate the expected value of $(-1)^{a \oplus b}$

$$\begin{split} &(1) \cdot \mathbb{P}\left[a = b\right] + (-1) \cdot \mathbb{P}\left[a \neq b\right] = \\ &= \langle \psi | F_s^0 \otimes G_t^0 | \psi \rangle + \langle \psi | F_s^1 \otimes G_t^1 | \psi \rangle \\ &- \langle \psi | F_s^1 \otimes G_t^0 | \psi \rangle - \langle \psi | F_s^0 \otimes G_t^1 | \psi \rangle \\ &= \langle \psi | (F_s^0 - F_s^1) \otimes (G_t^0 - G_t^1) | \psi \rangle \end{split}$$

- Define $\{-1,1\}$ -observables $F_s = F_s^0 F_s^1$ and $G_t = G_t^0 G_t^1$ with the property that its difference squared is the identity matrix
- Using this strategy the bias becomes

$$\mathbb{E}_{(s,t)\sim\pi}\left[\langle\psi|F_s\otimes G_t|\psi\rangle\Sigma_{s,t}\right]$$

More generally speaking

- For any XOR game the bias is defined as the difference of the probabilities of winning and loosing
- Which is, if considering the $\{-1,1\}$ basis, the expected value
- We are looking to maximize this quantity

Classical strategies

When using classical strategies this is

$$\max\{\mathbb{E}_{(s,t)\sim\pi}\left[\Sigma_{st}\chi(s)\psi(t)\right]:\chi:\mathcal{S}\to\{-1,1\},\\ \psi:\mathcal{T}\to\{-1,1\}\}$$

Entangled strategies

When using entangled strategies the winning probability might increase indefinitely with the dimensions, so we use the $\sup_{n\in\mathbb{N}}$

$$\sup_{n\in\mathbb{N}} \{\mathbb{E}_{(s,t)\sim\pi} \left[\Sigma_{st} \langle \psi | F_s \otimes G_t | \psi \rangle \right] : |\psi\rangle \in \mathbb{C}^n \otimes \mathbb{C}^n,
F_s, G_t \in O(\mathbb{C}^n) \}$$

The CHSH game

- The CHSH game (Clauser, Horner, Shimony, Holt) is a two player XOR game with $\mathcal{A}=\mathcal{B}=\mathcal{S}=\mathcal{T}=\{0,1\}$ and π being the uniform distribution
- $f(s,t) = s \wedge t$, i.e. f(1,1) = 1 and f(0,0) = f(0,1) = f(1,0) = 0
- Alice and Bob can win $\frac{3}{4}$ of the games by using deterministic strategies (0,0),(1,0) or (0,1)

Quantum strategy

- Let Alice and Bob share an EPR state
- Define

$$X = \begin{bmatrix} 0 & 1 \\ 1 & 0 \end{bmatrix}, Y = \begin{bmatrix} 0 & -i \\ i & 0 \end{bmatrix}$$

- XY + YX = 0 and $X^2 = Y^2 = I$
- For Alice define the observable for question 0 by $F_0 = X$ and for question 1 by $F_1 = Y$
- Bobs observables are going to be $G_0=(X-Y)/\sqrt{2}$ for question 0 and $G_1=(X+Y)/\sqrt{2}$ for question 1

The following auxiliary calculations will be helpful later:

$$\langle \mathsf{EPR} | X \otimes X | \mathsf{EPR} \rangle = rac{1}{2} \begin{pmatrix} 1 & 0 & 0 & 1 \\ 0 & 0 & 1 & 0 \\ 0 & 1 & 0 & 0 \\ 1 & 0 & 0 & 0 \end{pmatrix} \begin{pmatrix} 1 \\ 0 \\ 0 \\ 1 \end{pmatrix}$$

$$= rac{1}{2} \begin{pmatrix} 1 & 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} 1 \\ 0 \\ 0 \\ 1 \end{pmatrix} = rac{2}{2} = 1$$

$$\begin{split} \langle \mathsf{EPR} | Y \otimes Y | \mathsf{EPR} \rangle &= \frac{1}{2} \begin{pmatrix} 1 & 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} 0 & 0 & 0 & -1 \\ 0 & 0 & 1 & 0 \\ 0 & 1 & 0 & 0 \\ -1 & 0 & 0 & 0 \end{pmatrix} \begin{pmatrix} 1 \\ 0 \\ 0 \\ 1 \end{pmatrix} \\ &= \frac{1}{2} \begin{pmatrix} -1 & 0 & 0 & -1 \end{pmatrix} \begin{pmatrix} 1 \\ 0 \\ 0 \\ 1 \end{pmatrix} = -1 \end{split}$$

$$\begin{split} \langle \mathsf{EPR} | X \otimes Y | \mathsf{EPR} \rangle &= \frac{1}{2} \begin{pmatrix} 1 & 0 & 0 & 1 \end{pmatrix} \begin{pmatrix} 0 & 0 & 0 & -i \\ 0 & 0 & i & 0 \\ 0 & -i & 0 & 0 \\ i & 0 & 0 & 0 \end{pmatrix} \begin{pmatrix} 1 \\ 0 \\ 0 \\ 1 \end{pmatrix} \\ &= \frac{1}{2} \begin{pmatrix} i & 0 & 0 & -i \end{pmatrix} \begin{pmatrix} 1 \\ 0 \\ 0 \\ 1 \end{pmatrix} = 0 \\ \langle \mathsf{EPR} | Y \otimes X | \mathsf{EPR} \rangle &= 0 \end{split}$$

Lets calculate the expected values of the sign $a \oplus b$:

$$\langle \mathsf{EPR}|F_0\otimes G_0|\mathsf{EPR}\rangle = \langle \mathsf{EPR}|X\otimes \frac{1}{\sqrt{2}}(X-Y)|\mathsf{EPR}\rangle$$

$$= \langle \mathsf{EPR}|X\otimes \frac{1}{\sqrt{2}}X|\mathsf{EPR}\rangle - \langle \mathsf{EPR}|X\otimes \frac{1}{\sqrt{2}}Y|\mathsf{EPR}\rangle$$

$$= \frac{1}{\sqrt{2}} - 0 = \frac{1}{\sqrt{2}}$$

$$\begin{split} \langle \mathsf{EPR}|F_1 \otimes G_1|\mathsf{EPR}\rangle &= \langle \mathsf{EPR}|Y \otimes \frac{1}{\sqrt{2}}(X+Y)|\mathsf{EPR}\rangle \\ &= \langle \mathsf{EPR}|Y \otimes \frac{1}{\sqrt{2}}X|\mathsf{EPR}\rangle + \langle \mathsf{EPR}|Y \otimes \frac{1}{\sqrt{2}}Y|\mathsf{EPR}\rangle \\ &= 0 - \frac{1}{\sqrt{2}} = -\frac{1}{\sqrt{2}} \end{split}$$

$$\langle \mathsf{EPR}|F_0\otimes G_1|\mathsf{EPR}\rangle = \langle \mathsf{EPR}|X\otimes \frac{1}{\sqrt{2}}(X+Y)|\mathsf{EPR}\rangle$$

$$= \langle \mathsf{EPR}|X\otimes \frac{1}{\sqrt{2}}X|\mathsf{EPR}\rangle + \langle \mathsf{EPR}|X\otimes \frac{1}{\sqrt{2}}Y|\mathsf{EPR}\rangle$$

$$= \frac{1}{\sqrt{2}} + 0 = \frac{1}{\sqrt{2}}$$

$$\langle \mathsf{EPR}|F_1\otimes G_0|\mathsf{EPR}\rangle = \langle \mathsf{EPR}|Y\otimes \frac{1}{\sqrt{2}}(X-Y)|\mathsf{EPR}\rangle$$

$$= \langle \mathsf{EPR}|Y\otimes \frac{1}{\sqrt{2}}X|\mathsf{EPR}\rangle - \langle \mathsf{EPR}|Y\otimes \frac{1}{\sqrt{2}}Y|\mathsf{EPR}\rangle$$

$$= 0 - (-\frac{1}{\sqrt{2}}) = \frac{1}{\sqrt{2}}$$

Thus, we have

$$\langle \mathsf{EPR} | F_s \otimes G_t | \mathsf{EPR}
angle = egin{cases} rac{1}{\sqrt{2}}, (0,0), (1,0), (0,1) \ -rac{1}{\sqrt{2}}, (1.1) \end{cases}$$

which is equivalent to

$$\langle \mathsf{EPR}|F_s\otimes G_t|\mathsf{EPR}
angle = rac{(-1)^{s\wedge t}}{\sqrt{2}}, s,t\in\{0,1\}$$

The bias of the entangled strategy equals

$$\begin{split} \mathbb{E}_{(s,t) \sim \pi} \left[\Sigma_{s,t} \langle \psi | F_s \otimes G_t | \psi \rangle \right] &= \\ &= \frac{1}{4} \sum_{s,t=0}^{1} (-1)^{s \wedge t} \langle \mathsf{EPR} | F_s \otimes G_t | \mathsf{EPR} \rangle \\ &= \frac{1}{4} \cdot \frac{4}{\sqrt{2}} = \frac{1}{\sqrt{2}} \end{split}$$

The bias is $\frac{1}{\sqrt{2}}$ from which follows that the winning probability is by definition:

$$\frac{1}{2} + \frac{1}{2} \cdot \frac{1}{\sqrt{2}} = \cos(\pi/8) \approx 0.85 \dots$$

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Nice slide to draw the connection between the games an LC

Let $(X_i)_{1 \le i \le m}$ and $(Y_j)_{1 \le j \le n}$ be families of random variables on a common probability space such that $|X_i|, |Y_j| \le 1$ almost surely. Then $A = (a_{ij})$ is the corresponding classical (or local) correlation matrix if

$$a_{ij} = \mathbb{E}[X_i Y_j]$$

for all $1 \le i \le m, 1 \le j \le n$.

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• Set of all local correlation matrices: $LC_{m,n}$

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Lemma

$$\mathsf{LC}_{m,n} = \mathsf{conv}\{\xi \eta^T \,|\, \xi \in \{-1,1\}^m, \eta \in \{-1,1\}^n\}$$

Let $(X_i)_{1 \le i \le m}$ and $(Y_j)_{1 \le j \le n}$ be families of random variables on a common probability space such that $|X_i|, |Y_j| \le 1$ almost surely. Then $A = (a_{ij})$ is the corresponding classical (or local) correlation matrix if

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Lemma

$$LC_{m,n} = conv\{\xi \eta^T \mid \xi \in \{-1,1\}^m, \eta \in \{-1,1\}^n\}$$

• No matter which probabilistic strategy there is a deterministic one which as at least as good as the one one chooses

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• $\xi \eta^T \in LC_{m,n}$ for all $\xi \in \{-1,1\}^m, \eta \in \{-1,1\}^n$ (Choose $X_i \equiv \xi_i, Y_i \equiv \eta_i$)

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- Let $a_{ij}^{(k)} = \mathbb{E}[X_i^{(k)}Y_j^{(k)}]$ for $k \in \{0,1\}$

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- Let $a_{ii}^{(k)} = \mathbb{E}[X_i^{(k)} Y_i^{(k)}]$ for $k \in \{0, 1\}$
- Find $(X_i), (Y_i)$ with $|X_i|, |Y_i| \le 1$ almost surely such that

$$\beta a_{ij}^{(0)} + (1 - \beta) a_{ij}^{(1)} = \mathbb{E}[X_i Y_j]$$

for
$$\beta \in [0,1]$$

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for $\beta \in [0, 1]$

• Define a Bernoulli random variable α such that $\mathbb{P}(\alpha = 0) = \beta$, $\mathbb{P}(\alpha = 1) = 1 - \beta$ and set $X_i = X_i^{(\alpha)}, Y_j = Y_j^{(\alpha)}$

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- Then

$$\mathbb{E}[X_i Y_j] = \mathbb{E}[X_i^{(\alpha)} Y_j^{(\alpha)} \mathbb{1}_{\{\alpha = 0\}}] + \mathbb{E}[X_i^{(\alpha)} Y_j^{(1)}] \mathbb{1}_{\{\alpha = 1\}}]$$
$$= \beta \mathbb{E}[X_i^{(0)} Y_j^{(0)}] + (1 - \beta) \mathbb{E}[X_i^{(1)} Y_j^{(1)}]$$

• Let $a_{ij} = \mathbb{E}[X_i Y_j]$ for \mathbb{R} -valued random variables $(X_i), (Y_j)$ defined on a common probability space Ω with $|X_i|, |Y_j| \leq 1$ almost surely.

$\mathsf{LC}_{m,n} \subset \mathsf{conv}\{\xi\eta^T \,|\, \xi \in \{-1,1\}^m, \eta \in \{-1,1\}^n\}$

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- Set $X = (X_1, ..., X_m)$ and $Y = (Y_1, ..., Y_n)$, then $X \in [-1, 1]^m$, $Y \in [-1, 1]^n$ almost surely.

$\mathsf{LC}_{\mathit{m},\mathit{n}} \subset \mathsf{conv}\{\xi\eta^{\mathsf{T}} \,|\, \xi \in \{-1,1\}^{\mathit{m}}, \eta \in \{-1,1\}^{\mathit{n}}\}$

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- Set $X = (X_1, ..., X_m)$ and $Y = (Y_1, ..., Y_n)$, then $X \in [-1, 1]^m$, $Y \in [-1, 1]^n$ almost surely.
- Hypercube description by its vertices: $[-1,1]^d = \operatorname{conv}\{\xi \,|\, \xi \in \{-1,1\}^d\}$

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- ullet Define random variables $\lambda_{\xi}^{(X)}:\Omega^m o [0,1]$ such that

$$X(\omega) = \sum_{\xi \in \{-1,1\}^m} \lambda_{\xi}^{(X)}(\omega)\xi$$

almost surely and $\sum_{\xi \in \{-1,1\}^m} \lambda_{\xi}^{(X)}(\omega) = 1$

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Some nice frame to connect QCs to the games

Let $(X_i)_{1\leq i\leq m}$ and $(Y_j)_{1\leq j\leq n}$ be self-adjoint operators on \mathbb{C}^{d_1} , respectively \mathbb{C}^{d_2} for some positive integers d_1,d_2 , satisfying $\|X_i\|_{\infty},\|Y_j\|_{\infty}\leq 1$. $A=(a_{ij})$ is called quantum correlation matrix if there exists a state on $\mathbb{C}^{d_1}\otimes\mathbb{C}^{d_2}$) such that

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Lemma

$$\mathsf{QC}_{m,n} = \{(\langle x_i, y_j \rangle)_{1 \le 1 \le m, 1 \le j \le n} \, | \, x_i, y_j \in \mathbb{R}^{\min\{m,n\}}, |x_i| \le 1, |y_j| \le 1\},$$

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• $a_{ij} = \operatorname{Tr} \rho X_i \otimes Y_j$, sate ρ on a Hilbert space $\mathcal{H} = \mathbb{C}^{d_1} \otimes \mathbb{C}^{d_2}$ and Hermitian operators $(X_i)_{1 \geq m}$, $(Y_j)_{1 \geq n}$ on \mathbb{C}^{d_1} , respectively \mathbb{C}^{d_2} satisfying $\|X_i\|_{\infty}$, $\|Y_i\|_{\infty} < 1$

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- Define a positive semidefinite symmetric bilinear form on the space of Hermitian operators on \mathcal{H} by $\beta: \mathcal{H} \times \mathcal{H} \to \mathbb{R}$ where $\beta(S,T) = \text{Re}(\text{Tr } \rho ST)$.

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• Identify $X_i \otimes I$, $I \otimes Y_j$ with vectors x_i , y_j in U, then

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- The vectors have still not the right dimension but again, we can project them onto vectors in \mathbb{R}^m



In order to show

$$\mathsf{QC}_{m,n}\supset\{\big(\langle x_i,y_j\rangle\big)_{1\leq 1\leq m,1\leq j\leq n}\,|\,x_i,y_j\in\mathbb{R}^{\min\{m,n\}},|x_i|\leq 1,|y_j|\leq 1\}$$

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For all $n \ge 1$ there is a subspace of the $2^n \times 2^n$ Hermitian matrices where every vector is the multiple of a unitary matrix.

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Proposition

For all $n \ge 1$ there is a subspace of the $2^n \times 2^n$ Hermitian matrices where every vector is the multiple of a unitary matrix.

• The proof is based on n—fold tensor products of the Pauli matrices which are the three matrices

$$X = \begin{pmatrix} 0 & 1 \\ 1 & 0 \end{pmatrix}, \ Y = \begin{pmatrix} 0 & -i \\ i & 0 \end{pmatrix}, \ Z = \begin{pmatrix} 1 & 0 \\ 0 & -1 \end{pmatrix}, \ I = \begin{pmatrix} 1 & 0 \\ 0 & 1 \end{pmatrix}$$

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Proof.

Define

$$U_{i} = I^{\otimes (i-1)} \otimes X \otimes Y^{\otimes (n-i)},$$

$$U_{n+i} = I^{\otimes (i-1)} \otimes Z \otimes Y^{\otimes (n-i)}, i = 1, \dots n$$

- U_i 's are anti-commuting traceless Hermitian unitaries, i.e. $U_iU_j=-U_jU_i$ for $i\neq j$ and $U_i^2=I$
- For $X = \sum_{i=1}^{2n} \xi_i U_i$, $Y = \sum_{i=1}^{2n} \eta_i U_i$ we can calculate

$$XY = \sum_{i=1}^{2n} \xi_{i} \eta_{i} I + \sum_{1 \leq i, j \leq \leq 2n} \xi_{i} \eta_{j} U_{i} U_{j}$$

$$= \sum_{i=1}^{2n} \xi_{i} \eta_{i} I + \sum_{1 \leq i < j \leq \leq 2n} \xi_{i} \eta_{j} U_{i} U_{j} - \sum_{1 \leq i < j \leq \leq 2n} U_{i} U_{j} = \sum_{i=1}^{2n} \xi_{i} \eta_{i} I$$

$$= \langle \xi, \eta \rangle I.$$

• The result follows by setting X = Y.

 $\mathsf{QC}_{m,n} \supset \{(\langle x_i, y_j \rangle)_{1 \le 1 \le m, 1 \le j \le n} \, | \, x_i, y_j \in \mathbb{R}^{\min\{m,n\}}, |x_i| \le 1, |y_j| \le 1\}$

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- Tr $(X_iY_j^T) = d \cdot \langle x_i, y_j \rangle$ and $\|X_i\|_{\infty} \leq 1$ since $X_iX_i^* = |x_i|^2 I$ and $|x_i|^2 \leq 1$ (the same holds for Y_j)

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- Let $|\phi\rangle=\frac{1}{\sqrt{d}}\sum_{i=1}^{d}|ii\rangle$ and $\rho=|\phi\rangle\,\langle\rho|$. Note that we can write ρ as

$$\rho = |\phi\rangle\langle\phi| = \frac{1}{d}\sum_{1 \leq k,l \leq d} |kk\rangle\langle ll| = \frac{1}{d}\sum_{1 \leq k,l \leq d} |k\rangle\langle l| \otimes |k\rangle\langle l|$$



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Then

$$\operatorname{Tr}\left(\rho X_{i} \otimes Y_{j}\right) = \frac{1}{d} \sum_{1 \leq k,l \leq d} \operatorname{Tr}\left(\left|k\right\rangle \left\langle l\right| X_{i} \otimes \left|k\right\rangle \left\langle l\right| Y_{j}\right) = \frac{1}{d} \sum_{1 \leq k,l \leq d} \operatorname{Tr}\left(\left|k\right\rangle \left\langle l\right| X_{i}\right) \operatorname{T}$$

$$= \frac{1}{d} \operatorname{Tr} X_{i} Y_{j}^{T} = \left\langle x_{i}, y_{j}\right\rangle.$$

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- it holds $|\tilde{x}_i| \leq \lambda |(x_i, 0)| + (1 \lambda)|(0, \bar{x}_i)| \leq 1$ and $\langle \tilde{x}_i, \tilde{y}_i \rangle = \lambda \langle x_i, y_i \rangle + (1 \lambda)\langle \tilde{x}_i, \tilde{y}_i \rangle$.

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- Right dimension is obtained by projecting $(\tilde{x}_i)_{1 \leq i \leq m}$, $(\tilde{y}_j)_{1 \leq i \leq n}$ on span $\{x_1, \ldots, x_m\}$ or span $\{y_1, \ldots, y_n\}$, as in the proof before.

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- Inclusion is strict in general

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$$\mathsf{LC}_{2,2} = \mathsf{conv}\{\pm \begin{pmatrix} 1 & 1 \\ 1 & 1 \end{pmatrix}, \, \pm \begin{pmatrix} -1 & -1 \\ 1 & 1 \end{pmatrix}, \, \pm \begin{pmatrix} -1 & 1 \\ -1 & 1 \end{pmatrix}, \, \pm \begin{pmatrix} -1 & 1 \\ 1 & -1 \end{pmatrix}\}.$$

- Let us consider the case n=m=2.
- It holds

$$\mathsf{LC}_{2,2} = \mathsf{conv}\{\pm \begin{pmatrix} 1 & 1 \\ 1 & 1 \end{pmatrix}, \ \pm \begin{pmatrix} -1 & -1 \\ 1 & 1 \end{pmatrix}, \ \pm \begin{pmatrix} -1 & 1 \\ -1 & 1 \end{pmatrix}, \ \pm \begin{pmatrix} -1 & 1 \\ 1 & -1 \end{pmatrix}\}.$$

W can also write it as in intersections of halfspaces:

$$LC_{2,2} = \{ A \in \mathbb{R}^{2 \times 2} \mid -1 \le \operatorname{Tr} AM \le 1 \text{ for all } M \in \mathcal{K} \}, \tag{1}$$

where
$$\mathcal{K}=\{\frac{1}{2}\sigma(\begin{pmatrix} -1 & 1 \\ 1 & 1 \end{pmatrix}), \sigma(\begin{pmatrix} 1 & 0 \\ 0 & 0 \end{pmatrix}) \mid \sigma \in \{\mathsf{id}, (1\ 2), (1\ 3), (1\ 4)\}\}.$$

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$$\begin{aligned} \mathsf{Tr} \left(\mathsf{A} \mathsf{M} \right) &= \langle x_1, y_1 \rangle + \langle x_1, y_2 \rangle + \langle x_2, y_1 \rangle - \langle x_2, y_2 \rangle \\ &= \langle x_1 + x_2, y_1 \rangle + \langle x_1 - x_2, y_2 \rangle \leq |x_1 + x_2| |y_1| + |x_1 - x_2| |y_2| \\ &\leq |x_1 + x_2| + |x_1 - x_2|. \end{aligned}$$

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- Bound is achieved by $A=\frac{1}{\sqrt{2}}\begin{pmatrix}1&1\\1&1\end{pmatrix}$, induced by the vectors $x_1=x_2=\frac{1}{\sqrt{2}}(1,1)$ and $y_1=y_2=(1,0)$



Let $x, y \in \mathbb{R}^d$ be unit vectors. Let $r \in \mathbb{R}^d$ be a random unit vector chosen from O(d)-invariant probability distribution on the unit sphere. Then

- i, $\mathbb{P}[\operatorname{sign}(\langle x, r \rangle) \neq \operatorname{sign}(\langle y, r \rangle)] = \frac{\operatorname{arccos}(\langle x, y \rangle)}{\pi}$
- ii, $\mathbb{E}[\operatorname{sign}(\langle x, r \rangle) \operatorname{sign}(\langle y, r \rangle)] = \frac{2}{\pi} \arcsin(\langle x, y \rangle)$.

Proof.

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 - the normalized vector $n := s/\|s\|$ is uniformly distributed on the intersection of the unit sphere and span $\{x,y\}$ by the O(d)-invariance of the probability distribution

