

Classical XOR games: of size n : $R = (R_{s,t})_{s,t \in [n]}, R_{s,t} \in \mathbb{R}$,

$$\sum_{s,t=1}^n |R_{s,t}| = 1, \equiv \pi \text{ - probab. distribution on } [n] \times [n].$$

Referee \rightarrow ~~message~~ $(s,t) \in [n] \times [n] \sim \pi$, $s \rightarrow$ Alice
 $t \rightarrow$ Bob

Response $\rightarrow A^s \rightarrow a, B^t \rightarrow b \quad a,b \in \{0,1\}$

result \rightarrow win if ~~and~~ $(-1)^{a \oplus b} = \text{sign}(R_{s,t})$

• Strategy: $A: s \mapsto p(s)$ probab. of $a(s)=0$ $p(0,s) \in [0,1]$
 $B: t \mapsto q(t)$ $q(t)=0$

• success probability:

$$P_{\text{succ}}(R, p, q) = \sum_{s,t} |R_{s,t}| P_{\text{succ}}(s,t, p, q), \text{ where}$$

$$P_{\text{succ}}(s,t, p, q) = \begin{cases} p(s)q(t) + (1-p(s))(1-q(t)) & \text{if } \text{sign } R_{s,t} = 1 \\ p(s)(1-q(t)) + (1-p(s))q(t) & \text{if } \text{sign } R_{s,t} = -1 \end{cases}$$

• random strategy: $p(s) = q(t) = \frac{1}{2} \quad \forall s, t$

$$\Rightarrow P_{\text{succ}}(R, p, q) = \sum_{s,t} |R_{s,t}| \cdot \frac{1}{2} \quad P_{\text{succ}}(s,t, p, q) = \frac{1}{2} \quad \forall s, t$$

$$\Rightarrow P_{\text{succ}}(R, p, q) = \frac{1}{2} \sum_{s,t} |R_{s,t}| = \frac{1}{2}$$

• $P_{\text{succ}}(R, p, q) - \frac{1}{2}$ bias: $\omega(R)$

$$\text{sign } R_{s,t} = 1 \Rightarrow P_{\text{succ}}(s,t, p, q) - \frac{1}{2} = 2p(s)q(t) + \frac{1}{2} - q(t) - p(s)$$

$$\text{sign } R_{s,t} = -1 \Rightarrow P_{\text{succ}}(s,t, p, q) - \frac{1}{2} = p(s) - 2p(s)q(t) + q(t) - \frac{1}{2}$$

$$\Rightarrow P_{\text{succ}}(s,t, p, q) - \frac{1}{2} = \text{sign } R_{s,t} (2p(s)q(t) - q(t) - p(s) + \frac{1}{2})$$

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 $t \rightarrow \text{Bob}$

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reward $\rightarrow \min$ if $(-1)^{a \oplus b} = \text{sgn}(R_{s,t})$

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$$\Rightarrow P_{\text{succ}}(s,t, p, q) - \frac{1}{2} = \text{sgn } R_{s,t} \left(2p(s)q(t) - q(t) - p(s) + \frac{1}{2} \right)$$

$$\begin{aligned}
 P_{\text{succ}}(s, t; p, \frac{1}{2}) - \frac{1}{2} &= \text{sgn } R_{s,t} \left((2p(s) - 1) 2(t) - \frac{1}{2} p(s) + \frac{1}{2} \right) \\
 &= \text{sgn } R_{s,t} \left((2p(s) - 1) (2(t) - \frac{1}{2}) \right) = \\
 &= \frac{1}{2} \text{sgn}(R_{s,t}) (2p(s) - 1) (2(t) - 1)
 \end{aligned}$$

Bias($R, p, \frac{1}{2}$)

$$\Rightarrow \text{Bias}(R) = \sum_{s,t} |R_{s,t}| \left(P_{\text{succ}}(s, t; p, \frac{1}{2}) - \frac{1}{2} \right) =$$

$$= \frac{1}{2} \sum_{s,t} R_{s,t} (2p(s) - 1) (2(t) - 1)$$

classical bias



→ max
p, 1/2

w(R)

CLASSICAL BIAS

$$= \frac{1}{2} \max_{\substack{x, y \\ \|x\|_\infty \leq 1 \\ \|y\|_\infty \leq 1}} \left| \sum_{s,t} R_{s,t} x_s y_t \right| = \frac{1}{2} \max_{x_s, y_t \in \{-1, 1\}} \left| \sum_{s,t} R_{s,t} x_s y_t \right|$$

replace real x_s, y_t by complex ~~bias~~



COMPLEX BIAS

$$w^C(R) = \max_{\substack{x, y \in \mathbb{C}^n \\ \|x\|_\infty, \|y\|_\infty \leq 1}} \left| \sum_{s,t} R_{s,t} x_s y_t \right|$$

$$x, y \in \mathbb{C}^n \quad x = \sum x_s |e_s\rangle \quad y = \sum y_t |e_t\rangle \quad \text{or } b$$

$$\begin{aligned}
 w^C(R) &= \sum_{s,t} R_{s,t} \langle e_s | x \rangle \langle y | e_t \rangle = \sum_{s,t} R_{s,t} |e_s\rangle \langle e_t| \\
 &= \langle x | R y \rangle
 \end{aligned}$$

$$w^C(R) = \max_{\substack{x, y \\ \|x\|_\infty, \|y\|_\infty \leq 1}} \langle x | R y \rangle$$

SDP relaxation:

$$w^{sdp}(R) = \sup_{\substack{d, x_s, y_t \in \mathbb{C}^d \\ \|x_s\|, \|y_t\| \leq 1}} \left| \sum_{s,t} R_{s,t} \langle \bar{x}_s, y_t \rangle \right|$$

(semidefinite program
(lin.))

(replace x_s, y_t by d -dim. vectors \bar{x}_s, \bar{y}_t)

Bounds from Grothendieck inequality

$$w(R) \leq w^{sdp}(R) \leq K_G^R(R) \quad \text{for real } R$$

$$w^{\mathbb{C}}(R) \leq w^{sdp}(R) \leq K_G^{\mathbb{C}}(R) \quad \text{complex } R$$

$K_G^R, K_G^{\mathbb{C}}$ Grothendieck constants

ENTANGLED BIAS

~~And B share a state~~
 And B have local quantum systems $\mathcal{H}_A, \mathcal{H}_B \subseteq \mathbb{C}^d$ and share a state $|\psi\rangle \in \mathcal{H}_{AB}$.

• strategy:

with some $D \in A_s, B_t \in I$, then
 $\rightarrow S \mapsto A$ measures $A_s, I - A_s$ on her part $\dots \rightarrow$ outcomes
 $\mapsto B \mapsto B_t, I - B_t \dots \rightarrow$

• bias $\sum_{s,t} R_{s,t} \langle \psi | A_s \otimes B_t | \psi \rangle$

$$w^b(R) = \sup_{d, A_s, B_t, |\psi\rangle} \left| \sum_{s,t} R_{s,t} \langle \psi | A_s \otimes B_t | \psi \rangle \right|$$

Entangled bias

$\omega^*(k) \leq \omega^s(k)$: choose $|\psi\rangle, A_s, B_t$ such that

$$\langle \psi | A_s \otimes B_t | \psi \rangle = x_s y_t \quad \text{for } x_s, y_t \in \mathbb{C}$$

$$|\psi\rangle = |\psi_1\rangle \otimes |\psi_2\rangle \Rightarrow \langle \psi_1 | A_s | \psi_1 \rangle \langle \psi_2 | B_t | \psi_2 \rangle$$

$|\psi_1\rangle = |\psi_2\rangle = |\psi\rangle$ unit vector

$$A_s = x_s |\psi\rangle\langle\psi|, \quad B_t = y_t |\psi\rangle\langle\psi|$$

Tsirelson :

For any real R , $\omega^*(k) \leq \omega^s(k)$.

Proof: Let U_s, V_t be unitaries $|\psi\rangle \in \mathbb{C}^d \otimes \mathbb{C}^d$ unit vector
 \rightarrow strategy

(the supremum can be taken over such strategies \rightarrow linearity!)

• we may write $|\psi\rangle = \sum_i \lambda_i |i\rangle |i\rangle$ (up to local unitaries)

$$\Rightarrow \langle \psi | U_s \otimes V_t | \psi \rangle = \sum_{i,j} \lambda_i \lambda_j \langle i | U_s | j \rangle \langle i | V_t | j \rangle$$

$$x_s, i, j = \{ \lambda_i \langle i | U_s | j \rangle \}$$

$$y_t, i, j = \{ \lambda_j \langle i | V_t | j \rangle \}$$

$$\text{then } \langle x_s, y_t \rangle = \sum_{i,j} \lambda_i \lambda_j \langle i | U_s | j \rangle \langle i | V_t | j \rangle = \langle \psi | U_s \otimes V_t | \psi \rangle$$

$$\text{and } \|x_s\|^2 = \sum_{i,j} \lambda_i^2 \langle i | U_s | j \rangle \langle i | U_s | j \rangle =$$

$$= \sum_{i,j} \lambda_i^2 \langle j | U_s | i \rangle \langle i | U_s | j \rangle =$$

$$= \text{Tr } U_s (\text{diag } \lambda_i^2) U_s = \sum_j \lambda_j^2 = 1$$

$$\text{similarly } \|y_t\|^2 = 1$$

In fact, Tsirelson proved $\omega^*(k) = \omega^s(k)$, with \nexists the \square
max. enlarged state

$$R_{11} = R_{12} = R_{22} = \frac{1}{4}, \quad R_{21} = -\frac{1}{4}$$

$$\bullet w(R) = \max_{x_1, y_1 \in \{\pm 1\}} \frac{1}{4} (x_1 y_1 + x_1 y_2 + x_2 y_1 - x_2 y_2)$$

$$= \max_{x_1, y_1 \in \{\pm 1\}} \frac{1}{4} (x_1(y_1 + y_2) + x_2(y_1 - y_2)) = \frac{1}{2}$$

$$\bullet w^C(R) = \max_{\substack{x_1, y_1 \in \mathbb{C} \\ \max\{|x_1|, |y_1|\} \leq 1}} \frac{1}{4} |x_1(y_1 + y_2) + x_2(y_1 - y_2)| \leq \frac{\sqrt{2}}{2}$$

$$\bullet 1, w^C(R) \leq \frac{1}{4} \max_{\|y\|_2 \leq 1} (\|y_1 + y_2\| + \|y_1 - y_2\|)$$

$$\leq \frac{\sqrt{2}}{4} \left(\|y_1 + y_2\|^2 + \|y_1 - y_2\|^2 \right)^{1/2} \leq \frac{\sqrt{2}}{4} \left(2(\|y_1\|^2 + \|y_2\|^2) \right)^{1/2}$$

$$\leq \frac{\sqrt{2}}{2}$$

$$\begin{aligned} y_1 + y_2 &= a_1 + a_2 + i(b_1 + b_2) \\ \|y_1 + y_2\|^2 + \|y_1 - y_2\|^2 &= (a_1 + a_2)^2 + (b_1 + b_2)^2 + (a_1 - a_2)^2 + (b_1 - b_2)^2 \\ &= 2(a_1^2 + a_2^2 + b_1^2 + b_2^2) \end{aligned}$$

$$\|y_1 + y_2\|^2 + \|y_1 - y_2\|^2 \leq 4$$

$$\langle y_1 + y_2, y_1 + y_2 \rangle + \langle y_1 - y_2, y_1 - y_2 \rangle$$

$$= \langle y_1, y_1 \rangle + \langle y_1, y_2 \rangle + \langle y_2, y_1 \rangle + \langle y_2, y_2 \rangle$$

$$+ \langle y_1, y_1 \rangle + \langle y_1, y_2 \rangle -$$

$$= 2(\|y_1\|^2 + \|y_2\|^2)$$

conversely, put
 $w^{\text{odr}}(R) = \frac{\sqrt{2}}{2}$ the same argument.

$$\|x_1\| = \|x_2\| = 1$$

$$x_1 = 1$$

$$y_1 = \frac{1-i}{\sqrt{2}}, \quad y_2 = \frac{1+i}{\sqrt{2}}$$

$$|x_1(n+z) + x_2(n-z)| = \frac{1}{\sqrt{2}} (1 \cdot (1+i) + 1 \cdot (-1) + i \cdot (1-i - 1-i)) =$$

$$= \frac{1}{\sqrt{2}} (2 - i(2i)) = \frac{1}{\sqrt{2}} 4 \quad \cdot \frac{1}{4} = \frac{\sqrt{2}}{2} \quad \square$$

Classical XOR games:

bias:

For real matrix R :

$$\omega(R) \leq \omega^c(R) \leq \omega^e(R) = \omega^{SDP}(R) \leq K_G^R \omega(R)$$

↑
complex bias

↓
shared entanglement

↓
SDP relaxation

↗
Guthrie & Cozma

QUANTUM XOR GAMES

of size n : A Hermitian matrix $M \in \mathcal{M}(\mathbb{C}^n)$,

$\|M\|_1 \leq 1$. Strategy: $A \in \mathcal{O}_b(\mathbb{C}^n \otimes \mathcal{H}_A)$, $B \in \mathcal{O}_b(\mathbb{C}^n \otimes \mathcal{H}_B)$

$|\psi\rangle \in \mathcal{H}_A \otimes \mathcal{H}_B$.
(Symmetries) (Hermitian unitaries)

The bias: $\omega(A, B, |\psi\rangle, M) = \text{Tr}(A \otimes B)(M \otimes |\psi\rangle\langle\psi|)$

Classical? ~~Not classical~~ $R, A_s \in \mathcal{O}_b(\mathcal{H}_A), B_s \in \mathcal{O}_b(\mathcal{H}_B)$
 $|\psi\rangle \in \mathcal{H}_A \otimes \mathcal{H}_B$

$$\text{Tr}(A \otimes B)(M \otimes |\psi\rangle\langle\psi|) = \sum_{s,t} R_{s,t} \text{Tr}(A_s \otimes B_t) |\psi\rangle\langle\psi|$$

$$A = \sum_s |e_s\rangle\langle e_s| \otimes A_s$$

$$B = \sum_t |e_t\rangle\langle e_t| \otimes B_t$$

$$\Rightarrow \text{Tr}(A \otimes B)(M \otimes |\psi\rangle\langle\psi|) = \sum_{s,t} \text{Tr}(|e_s\rangle\langle e_s| \otimes A_s \otimes |e_t\rangle\langle e_t| \otimes B_t) (M \otimes |\psi\rangle\langle\psi|)$$

$$= \sum_{s,t} \underbrace{\langle e_s \otimes e_t | M | e_s \otimes e_t \rangle}_{R_{s,t}} \text{Tr}(A_s \otimes B_t) |\psi\rangle\langle\psi|$$

$$\|M\|_1 = \sum_{s,t} |R_{s,t}| = 1.$$

$$M = \sum_{s,t} R_{s,t} |e_s \otimes e_s\rangle\langle e_s \otimes e_s|$$

Diagonal matrix