# Hypercontractivity, XOR games and the Aaronson-Ambainis conjecture 

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## Introduction

In this talk, I will discuss how so-called hypercontractive inequalities can be used to give a new(ish) proof of a bound on the bias of multiplayer XOR games, which implies a (very) special case of a conjecture about quantum query algorithms.

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Outline:

- Introduction to hypercontractivity
- XOR games
- The Bohnenblust-Hille inequality and its proof
- The Aaronson-Ambainis conjecture.


## Hypercontractive inequalities: a CS perspective

Hypercontractive inequalities have been much used in the quantum field theory literature:

- introduced (in the form of log-Sobolev inequalities) by [Gross '75];
- for detailed reviews see e.g. [Davies, Gross and Simon '92], [Gross '06].


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In the computer science literature, first used by [Kahn, Kalai and Linial '88] in an important paper proving that every boolean function has an influential variable.

The hypercontractive inequality they used is a particularly simple and clean special case due to [Bonami '70], [Gross '75], and often known as the Bonami-Beckner inequality.

## Noise

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- For $\epsilon \in[0,1]$, define the noise operator $T_{\epsilon}$ as follows:

$$
\left(T_{\epsilon} f\right)(x)=\mathbb{E}_{y \sim{ }_{\epsilon} x}[f(y)]
$$

- Here the expectation is over strings $y \in\{ \pm 1\}^{n}$ obtained from $x$ by negating each element of $x$ with independent probability $(1-\epsilon) / 2$.


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- Here the expectation is over strings $y \in\{ \pm 1\}^{n}$ obtained from $x$ by negating each element of $x$ with independent probability $(1-\epsilon) / 2$. So...
- If $\epsilon=1, T_{\epsilon} f=f$;
- If $\epsilon=0, T_{\epsilon} f$ is constant.
- Fairly easy to show that $T_{\epsilon}$ is a contraction, i.e.

$$
\left\|T_{\epsilon} f\right\|_{p} \leqslant\|f\|_{p}
$$

where $\|f\|_{p}:=\left(\frac{1}{2^{n}} \sum_{x \in\{ \pm 1\}^{n}}|f(x)|^{p}\right)^{1 / p}$.

## Noise and polynomials

- Any function $f:\{ \pm 1\}^{n} \rightarrow \mathbb{R}$ can be expanded as a multilinear polynomial:

$$
f\left(x_{1}, \ldots, x_{n}\right)=\sum_{S \subseteq[n]} \hat{f}(S) x_{S}
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- Parseval's equality: $\|f\|_{2}^{2}=\sum_{S \subseteq[n]} \hat{f}(S)^{2}$.
- The noise operator has a nice "Fourier-side" description in terms of polynomials: for $g(x)=x_{S}$,

$$
\left(T_{\epsilon} g\right)(x)=\epsilon^{|S|} x_{S}
$$

and by linearity, for any $f:\{ \pm 1\}^{n} \rightarrow \mathbb{R}$,

$$
\left(T_{\epsilon} f\right)(x)=\sum_{S \subseteq[n]} \epsilon^{|S|} \hat{f}(S) x_{S} .
$$

## Hypercontractivity of $T_{\epsilon}$

## The Bonami-Beckner inequality [Bonami '70] [Gross '75]

For any $f:\{ \pm 1\}^{n} \rightarrow \mathbb{R}$, and any $p$ and $q$ such that
$1 \leqslant p \leqslant q \leqslant \infty$ and $\epsilon \leqslant \sqrt{\frac{p-1}{q-1}}$,

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\left\|T_{\epsilon} f\right\|_{q} \leqslant\|f\|_{p}
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Intuition: usually $\|f\|_{p} \leqslant\|f\|_{q}$ for $p \leqslant q$, but applying noise to $f$ smoothes out its peaks and makes the norms comparable.

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## Corollary

Let $f:\{ \pm 1\}^{n} \rightarrow \mathbb{R}$ be a polynomial of degree $d$. Then:

- for any $p \leqslant 2,\|f\|_{p} \geqslant(p-1)^{d / 2}\|f\|_{2}$;
- for any $q \geqslant 2,\|f\|_{q} \leqslant(q-1)^{d / 2}\|f\|_{2}$.

Intuition: low-degree polynomials are smooth.

## Proof of the corollary

Given a degree $d$ (multilinear) polynomial

$$
f\left(x_{1}, \ldots, x_{n}\right)=\sum_{S \subseteq[n],|S| \leqslant d} \hat{f}(S) x_{S}
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where $x_{S}=\prod_{i \in S} x_{i}$, write $f=k=\sum_{S,|S|=k} \hat{f}(S) x_{S}$.

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\|f\|_{q}^{2}=\left\|\sum_{k=0}^{d} f^{=k}\right\|_{q}^{2}=\left\|T_{1 / \sqrt{q-1}}\left(\sum_{k=0}^{d}(q-1)^{k / 2} f=k\right)\right\|_{q}^{2}
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& \leqslant(q-1)^{d} \sum_{S \subseteq[n]} \hat{f}(S)^{2}=(q-1)^{d}\|f\|_{2}^{2} .
\end{aligned}
$$

(last two equalities: Parseval's equality)

## Applications in quantum computation

The above inequality has recently found a number of applications in quantum computation:

- Separations between quantum and classical communication complexity [Gavinsky et al '07]
- Limitations on quantum random access codes [Ben-Aroya, Regev and de Wolf '08]
- Bounds on non-local games [Buhrman '11]
- Lower bounds on quantum query complexity [Ambainis and de Wolf '12]
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Today: one more application.

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The players are allowed to communicate before the game starts, to agree a strategy, but cannot communicate during the game.

## Multiplayer XOR games

For example, consider the CHSH game:

- Two players, two possible inputs, chosen uniformly ( $k=2$, $n=2, \pi$ is uniform).
- $A=\left(\begin{array}{cc}1 & 1 \\ 1 & -1\end{array}\right)$ : the players win if their outputs are the same, unless $i_{1}=i_{2}=2$, when they win if their outputs are different.


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In general, the maximal bias (i.e. difference between probability of success and failure) achievable by deterministic strategies is

$$
\beta(G):=\max _{x^{1}, \ldots, x^{k} \in\{ \pm 1\}^{n}}\left|\sum_{i_{1}, \ldots, i_{k}=1}^{n} \pi_{i_{1}, \ldots, i_{k}} A_{i_{1}, \ldots, i_{k}} x_{i_{1}}^{1} \ldots x_{i_{k}}^{k}\right| .
$$

It's easy to see that shared randomness doesn't help.

## Why care about XOR games?

- In some cases (e.g. the CHSH game), if the players are allowed to share entanglement they can beat any possible classical strategy.


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- XOR games thus provide a clean, mathematically tractable way of studying the power of entanglement.


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- XOR games thus provide a clean, mathematically tractable way of studying the power of entanglement.
- XOR games are also interesting in themselves classically:
- Applications in communication complexity, e.g. [Ford and Gál '05]
- Known to be NP-hard to compute bias
- Connections to combinatorics and coding theory.


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## Today's question

What is the hardest $k$-player XOR game for classical players?
i.e. what is the game which minimises the maximal bias achievable?

## Previously known results

Until recently, there was a big gap between lower and upper bounds on $\min _{G} \beta(G)$ :

- There exists a game $G$ for which $\beta(G) \leqslant n^{-(k-1) / 2}[$ Ford and Gál '05].
- Any game $G$ has $\beta(G) \geqslant 2^{-O(k)} n^{-(k-1) / 2}$ [Bohnenblust and Hille '31].


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A recent and significant improvement:
Theorem [Defant, Popa and Schwarting '10] [Pellegrino and Seoane-Sepúlveda '12]
There exists a universal constant $c>0$ such that, for any XOR game $G$ as above, $\beta(G)=\Omega\left(k^{-c} n^{-(k-1) / 2}\right)$.

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We will show how this result can be proven using hypercontractivity (as a small step in the proof).

## XOR games and multilinear forms

A homogeneous polynomial $f:\left(\mathbb{R}^{n}\right)^{k} \rightarrow \mathbb{R}$ is said to be a multilinear form if it can be written as

$$
f\left(x^{1}, \ldots, x^{k}\right)=\sum_{i_{1}, \ldots, i_{k}} \hat{f}_{i_{1}, \ldots, i_{k}} x_{i_{1}}^{1} x_{i_{2}}^{2} \ldots x_{i_{k}}^{k}
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for some multidimensional array $\hat{f} \in \mathbb{R}^{n} \times \mathbb{R}^{n} \times \cdots \times \mathbb{R}^{n}$.

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\|f\|_{p}:=\left(\frac{1}{2^{n k}} \sum_{x^{1}, \ldots, x^{k} \in\{ \pm 1\}^{n}}\left|f\left(x^{1}, \ldots, x^{k}\right)\right|^{p}\right)^{1 / p}
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Any XOR game $G=(\pi, A)$ corresponds to a multilinear form $f$ :

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and the bias $\beta(G)$ is precisely $\|f\|_{\infty}:=\max _{x \in\{ \pm 1\}^{n}}|f(x)|$.

## What we want to prove

Bohnenblust-Hille inequality [BH '31, DPS '10, PS '12]
For any multilinear form $f:\left(\mathbb{R}^{n}\right)^{k} \rightarrow \mathbb{R}$, and any $p \geqslant 2 k /(k+1)$,

$$
\|\hat{f}\|_{p}:=\left(\sum_{i_{1}, \ldots, i_{k}}\left|\hat{f}_{i_{1}, \ldots, i_{k}}\right|^{p}\right)^{1 / p} \leqslant C_{k}\|f\|_{\infty}
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where $C_{k}$ may be taken to be $O\left(k^{\log _{2} e}\right) \approx O\left(k^{1.45}\right)$.

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- As $\|\hat{f}\|_{p}$ is nonincreasing with $p$, it suffices to prove the claim for $p=2 k /(k+1)$.


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where $C_{k}$ may be taken to be $O\left(k^{\log _{2} e}\right) \approx O\left(k^{1.45}\right)$.
Implies $\beta(G)=\Omega\left(C_{k}^{-1} n^{-(k-1) / 2}\right)$ by choosing $p$ appropriately. We'll prove the claim by induction on $k$, for $k$ a power of 2 .

- As $\|\hat{f}\|_{p}$ is nonincreasing with $p$, it suffices to prove the claim for $p=2 k /(k+1)$.
- The base case $k=1$ is trivial $\left(C_{1}=1\right)$. So, assuming the theorem holds for $k / 2$, we prove it holds for $k$.


## Proof

We start with a matrix inequality [Defant, Popa and Schwarting '10]:

$$
\begin{aligned}
\|\hat{f}\|_{2 k /(k+1)} & \leqslant\left(\sum_{i_{1}, \ldots, i_{k / 2}}\left(\sum_{i_{k / 2+1}, \ldots, i_{k}} \hat{f}_{i_{k / 2+1}, \ldots, i_{k}}^{2}\right)^{k /(k+2)}\right)^{(k+2) / 4 k} \\
& \times\left(\sum_{i_{k / 2+1}, \ldots, i_{k}}\left(\sum_{i_{1}, \ldots, i_{k / 2}}\right)_{i_{1}, \ldots, i_{k}}^{2}\right)
\end{aligned}
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\end{aligned}
$$

We estimate the second term (the first follows exactly the same procedure).

## Proof

For each $i_{k / 2+1}, \ldots, i_{k} \in[n]$, define $f_{i_{k / 2+1}, \ldots, i_{k}}:\left(\mathbb{R}^{n}\right)^{k / 2} \rightarrow \mathbb{R}$ by

$$
f_{i_{k / 2+1}, \ldots, i_{k}}\left(x^{1}, \ldots, x^{k / 2}\right)=\sum_{i_{1}, \ldots, i_{k / 2}} \hat{f}_{i_{1}, \ldots, i_{k}} x_{i_{1}}^{1} x_{i_{2}}^{2} \ldots x_{i_{k / 2}}^{k / 2} .
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f_{i_{k / 2}+1, \ldots, i_{k}}\left(x^{1}, \ldots, x^{k / 2}\right)=\sum_{i_{1}, \ldots, i_{k / 2}} \hat{f}_{i_{1}, \ldots, i_{k}} x_{i_{1}}^{1} x_{i_{2}}^{2} \ldots x_{i_{k / 2}}^{k / 2} .
$$

Also define a "dual" function $f_{x^{1}, \ldots, x^{k / 2}}^{\prime}:\left(\mathbb{R}^{n}\right)^{k / 2} \rightarrow \mathbb{R}$ by

$$
f_{x^{1}, \ldots, x^{k / 2}}^{\prime}\left(x^{k / 2+1}, \ldots, x^{k}\right)=f\left(x^{1}, \ldots, x^{k}\right) .
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$$

We have
$f_{x^{1}, \ldots, x^{k / 2}}^{\prime}\left(x^{k / 2+1}, \ldots, x^{k}\right)=\sum_{i_{k / 2+1}, \ldots, i_{k}=1}^{n} f_{i_{k / 2+1}, \ldots, i_{k}}\left(x^{1}, \ldots, x^{k / 2}\right) x_{i_{k / 2+1}}^{k / 2+1} \ldots x_{i_{k}}^{k} ;$
of course $\left\|f_{x^{1}, \ldots, x^{k} / 2}^{\prime}\right\|_{\infty} \leqslant\|f\|_{\infty}$.

## Proof

For each tuple $i_{k / 2+1}, \ldots, i_{k}$ we have by Parseval's equality

$$
\sum_{i_{1}, \ldots, i_{k / 2}=1}^{n} \hat{f}_{i_{1}, \ldots, i_{k}}^{2}=\left\|f_{i_{k / 2+1}, \ldots, i_{k}}\right\|_{2}^{2}
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By hypercontractivity,

$$
\left\|f_{i_{k / 2+1}, \ldots, i_{k}}\right\|_{2}^{2 k /(k+2)} \leqslant\left(\frac{k+2}{k-2}\right)^{\frac{k^{2}}{2(k+2)}}\left\|f_{i_{k / 2+1}, \ldots, i_{k}}\right\|_{2 k /(k+2)}^{2 k /(k+2)}
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$$

We now observe that, for any $p \geqslant 1$,
$\begin{aligned} \sum_{i_{k / 2+1}, \ldots, i_{k}}\left\|f_{i_{k / 2+1}, \ldots, i_{k}}\right\|_{p}^{p} & =\mathbb{E}_{x^{1}, \ldots, x^{k / 2}}\left[\sum_{i_{k / 2+1}, \ldots, i_{k}}\left|f_{i_{k / 2+1}, \ldots, i_{k}}\left(x^{1}, \ldots, x^{k / 2}\right)\right|^{p}\right] \\ & =\mathbb{E}_{x^{1}, \ldots, x^{k / 2}}\left[\left\|\hat{f}^{\prime}{ }_{x^{1}, \ldots, x^{k / 2}}\right\|_{p}^{p}\right] .\end{aligned}$

## Proof

Hence, taking $p=2 k /(k+2)=2(k / 2) /(k / 2+1)$, we have

$$
\begin{aligned}
\sum_{i_{k / 2+1}, \ldots, i_{k}} & \left(\sum_{i_{1}, \ldots, i_{k / 2}} \hat{f}_{i_{1}, \ldots, i_{k}}^{2}\right)^{k /(k+2)} \\
& \leqslant\left(\frac{k+2}{k-2}\right)^{\frac{k^{2}}{2(k+2)}} \mathbb{E}_{x^{1}, \ldots, x^{k / 2}}\left[\left\|\hat{f}_{x^{1}, \ldots, x^{k / 2}}\right\|_{2 k /(k+2)}^{2 k /(k+2)}\right]
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& \leqslant\left(\frac{k+2}{k-2}\right)^{\frac{k^{2}}{2(k+2)}} C_{k / 2}^{2 k /(k+2)}\|f\|_{\infty}^{2 k /(k+2)}
\end{aligned}
$$

by the inductive hypothesis.

## Proof

Combining both terms in the first inequality,

$$
\left(\sum_{i_{1}, \ldots, i_{k}}\left|\hat{f}_{i_{1}, \ldots, i_{k}}\right|^{2 k /(k+1)}\right)^{(k+1) /(2 k)} \leqslant\left(\frac{k+2}{k-2}\right)^{k / 4} C_{k / 2}\|f\|_{\infty}
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Observing that $(1+4 /(k-2))^{k / 4} \leqslant(1+O(1 / k)) e$, we have $C_{k}=O\left(k^{\log _{2} e}\right)$ as claimed.

## A conjecture of Aaronson and Ambainis

The following beautiful conjecture is currently open:
Conjecture [Aaronson and Ambainis '11]
Every bounded low-degree polynomial on the boolean cube has an influential variable.

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The following beautiful conjecture is currently open:
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- Generalises a prior result showing this for decision trees [ ${ }^{\prime}$ 'Donnell et al ${ }^{\prime} 05$ ].
- One reason this conjecture is interesting: it would imply that every quantum query algorithm can be approximated by a classical algorithm on "most" inputs.
- One special case known: when $f$ is symmetric, i.e. $f(x)$ depends only on $\sum_{i} x_{i}$ [Bačkurs '12].
- There are "L1" and "L2" versions of the conjecture [Bačkurs and Bavarian '13]; both are open. Here: the L2 version.


## A conjecture of Aaronson and Ambainis

A more formal version of the conjecture:
Conjecture [Aaronson and Ambainis '11]
For all degree $d$ polynomials $f:\{ \pm 1\}^{n} \rightarrow[-1,1]$, there exists $j$ such that $I_{j}(f) \geqslant \operatorname{poly}(\operatorname{Var}(f) / d)$.

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What does this mean?

- Write $\mathbb{E}[f]=\frac{1}{2^{n}} \sum_{x \in\{ \pm 1\}^{n}} f(x)$. Then the $\left(\ell_{2}\right)$ variance of $f$ is

$$
\operatorname{Var}(f)=\mathbb{E}\left[(f-\mathbb{E}[f])^{2}\right]
$$

- Define the influence of the $j^{\prime}$ th variable on $f$ as

$$
I_{j}(f)=\frac{1}{2^{n+2}} \sum_{x \in\{ \pm 1\}^{n}}\left(f(x)-f\left(x^{j}\right)\right)^{2}
$$

where $x^{j}$ is $x$ with the $j^{\prime}$ th variable negated.

## A conjecture of Aaronson and Ambainis

Using the above strengthening of the BH inequality, it is easy to prove a very special case of the Aaronson-Ambainis conjecture. Let

$$
f\left(x^{1}, \ldots, x^{k}\right)=\sum_{i_{1}, \ldots, i_{k}} \hat{f}_{i_{1}, \ldots, i_{k}} x_{i_{1}}^{1} x_{i_{2}}^{2} \ldots x_{i_{k}}^{k}
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- $f$ depends on $n k$ variables $x_{\ell}^{j}, 1 \leqslant j \leqslant k$ and $1 \leqslant \ell \leqslant n$.
- The influence of variable $(j, \ell)$ on $f$ is

$$
\operatorname{Inf}_{(j, \ell)}(f)=\sum_{i_{1}, \ldots, i_{j-1}, i_{j+1}, \ldots, i_{k}} \hat{f}_{i_{1}, \ldots, i_{j-1}, \ell, i_{j+1}, \ldots, i_{k}}^{2}=n^{k-1} \alpha^{2}
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$$

## Corollary

If $f$ is a multilinear form such that $\|f\|_{\infty} \leqslant 1$ and $\hat{f}_{i_{1}, \ldots, i_{k}}= \pm \alpha$ for some $\alpha$, then $I_{(j, \ell)}(f)=\Omega\left(\operatorname{Var}(f)^{2} / k^{3}\right)$ for all $(j, \ell)$.

## Summary

We have:

- ... used hypercontractivity to prove the Bohnenblust-Hille inequality;
- ... and hence give strong bounds on the worst-case classical bias in XOR games;
- ... and also prove a very special case of the Aaronson-Ambainis conjecture.

Open problems:

- Prove the Aaronson-Ambainis conjecture (using hypercontractivity!).


## Summary

On a more concrete level:

- Can one generalise the Bohnenblust-Hille inequality to polynomials? i.e. prove that for any degree $d$ multilinear polynomial $f:\{ \pm 1\}^{n} \rightarrow \mathbb{R}$, and any $p \geqslant 2 d /(d+1)$,

$$
\|\hat{f}\|_{p}:=\left(\sum_{S \subseteq[n]}|\hat{f}(S)|^{p}\right)^{1 / p} \leqslant C_{d}\|f\|_{\infty}
$$

where $C_{d}=\operatorname{poly}(d)$.

- This inequality holds for $C_{d}=2^{O(d)}$ (Andreas Defant, personal communication).
- Would this imply the Aaronson-Ambainis conjecture?


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Thanks!

