

Quantum Lower Bounds You probably Haven't Seen Before

(which doesn't imply that you don't know OF them)

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A History of Quantum Lower Bounds

QUANTUM
ARGUMENTS



POLYNOMIAL
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BBBV'94: $\Omega(\sqrt{n})$ lower bound for searching a list of n elements (i.e. Grover's algorithm is optimal)

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BBCMW'98: $\Omega\left(\sqrt{(k+1)(n-k)}\right)$ bound for *any* symmetric Boolean function $f(|X|)$ with $f(k) \neq f(k+1)$

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Ambainis'00: $\Omega(\sqrt{n})$ bounds for evaluating an AND-OR tree and for finding the '1' in a permutation

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A'02: $\Omega(n^{1/5})$ bound for the collision problem (deciding whether $f:\{1\dots n\}\rightarrow\{1\dots n\}$ is 1-to-1 or 2-to-1)

Shi'02: $\Omega(n^{1/3})$ bound for collision with large range,
 $\Omega(n^{2/3})$ for element distinctness

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Other results, including what I'll talk about today

True

Henceforth polynomial arguments shall be used for highly symmetric problems and for zero-error bounds, and quantum arguments otherwise.

Whosoever disobeys, must post to quant-ph.

Talk Outline

1. Quantum Certificate Complexity
2. Recursive Fourier Sampling
3. Query Complexity & Quantum Gravity
(special treat for Dave Bacon)

Quantum Certificate Complexity

Background

$f:\{0,1\}^n \rightarrow \{0,1\}$ is a total Boolean function

$D(f)$ (deterministic query complexity)

$\geq R_0(f)$ (zero-error randomized)

$\geq R_2(f)$ (bounded-error randomized)

$\geq Q_2(f)$ (bounded-error quantum)

$\leq Q_0(f)$ (zero-error quantum)

$\leq Q_E(f)$ (exact quantum)

Example

$$f = OR(x_1, K, x_n)$$

$$D(OR) = n$$

$$Q_E(OR) = \Theta(n)$$

$$R_0(OR) \approx n$$

$$Q_0(OR) = \Theta(n)$$

$$R_2(OR) \approx \frac{1-2\varepsilon}{1-\varepsilon} n$$

$$Q_2(OR) = \Theta(\sqrt{n})$$

Certificate Complexity $C(f) = \max_x C^x(f)$

$C^x(f) = \min \#$ of queries needed to distinguish X from every Y s.t. $f(Y) \neq f(X)$

Block Sensitivity $bs(f) = \max_x bs^x(f)$

$bs^x(f) = \max \#$ of disjoint blocks $B \subseteq \{x_1, \dots, x_n\}$ s.t. flipping B changes $f(X)$

Example: For $f = \text{MAJ}(x_1, x_2, x_3, x_4, x_5)$, letting $X = 11110$,

11110

$C^x(\text{MAJ}) = 3$

11**11**0

$bs^x(\text{MAJ}) = 2$

Randomized Certificate Complexity $RC(f) = \max_X RC^X(f)$

$RC^X(f) = \min$ # of *randomized* queries needed to distinguish X from any Y s.t. $f(Y) \neq f(X)$ with $\frac{1}{2}$ prob.

Quantum Certificate Complexity $QC(f)$

Example: For $f = \text{MAJ}(x_1, \dots, x_n)$, letting $X = 00 \dots 0$,

$$RC^X(\text{MAJ}) = 1$$

Different notions of nondeterministic quantum query complexity: Watrous 2000, de Wolf 2002

Adversary Method (special case)

Let D_0, D_1 be distributions over $f^{-1}(0), f^{-1}(1)$ s.t.
 D_0 looks “locally similar” to every 1-input, and
 D_1 looks “locally similar” to every 0-input:

$$\forall X \in f^{-1}(0), i \in \{1, \dots, n\} \quad \Pr_{Y \in D_1} [x_i \neq y_i] \leq \alpha$$

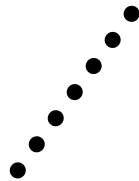
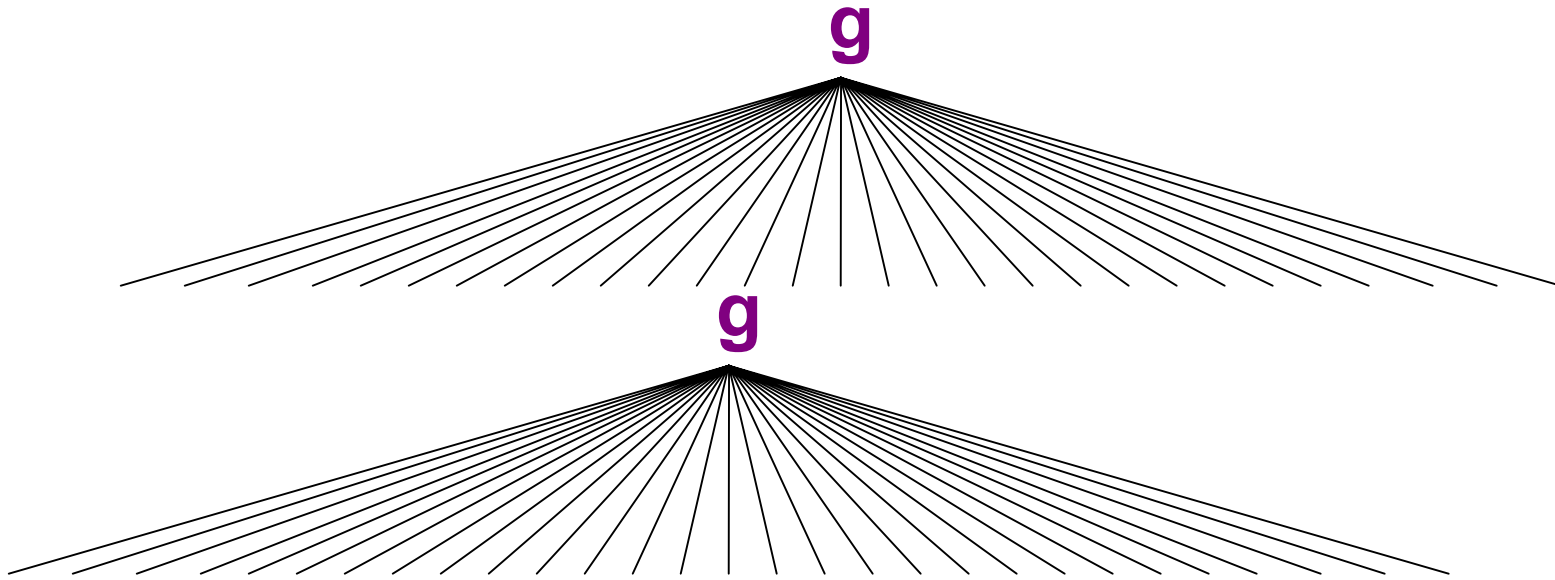
$$\forall Y \in f^{-1}(1), i \in \{1, \dots, n\} \quad \Pr_{X \in D_0} [x_i \neq y_i] \leq \beta.$$

Then
$$Q_2(f) = \Omega\left(\frac{1}{\sqrt{\alpha\beta}}\right).$$

Claim: $QC(f) = \Theta\left(\sqrt{RC(f)}\right)$

- Any randomized certificate for input X can be made *nonadaptive*
- By minimax theorem, exists distribution over $\{Y:f(Y)\neq f(X)\}$ s.t. for all i , $x_i\neq y_i$ w.p. $O(1/RC(f))$
- Adversary method then yields $\Omega\left(\sqrt{RC(f)}\right)$
- For upper bound, use “weighted Grover”

Example where $C(f) = \Theta(\text{QC}(f)^{2.205})$



$$k = x_1 + L + x_{29}$$

$$g(k) = \begin{cases} 0 & \text{if } k \leq 12 \\ 1 & \text{if } k = 13, 14, 15, 16 \\ 0 & \text{if } k \geq 17 \end{cases}$$

New Quantum/Classical Relation

For total f ,

$$\begin{aligned} R_0(f) &= O\left(RC(f) \text{ndeg}(f) \log n\right) \\ &= O\left(Q_2(f)^2 Q_0(f) \log n\right) \end{aligned}$$

where $\text{ndeg}(f) = \min \text{degree of poly } p \text{ s.t.}$
 $p(X) \neq 0 \Leftrightarrow f(X) = 1$

Previous: $D(f) = O(Q_2(f)^2 Q_0(f)^2)$ (de Wolf),
 $D(f) = O(Q_2(f)^6)$ (Beals et al.)

Idea (follows Buhrman-de Wolf):

Let p be s.t. $p(X) \neq 0 \Leftrightarrow f(X) = 1$

$x_1x_2 - x_2 + 2x_3$: $x_1x_2, 2x_3$ are “maxonomials”

Nisan-Smolensky: For every 0-input X and maxonomial M of p , X has a sensitive block whose variables are all in M

Consequence: Randomized 0-certificate must intersect each maxonomial w.p. $\geq \frac{1}{2}$

Randomized algorithm: Keep querying a randomized 0-certificate, until either one no longer exists or $p=0$

Lemma: $O(\text{ndeg}(f) \log n)$ iterations suffice w.h.p.

Proof: Let S be current set of monomials, and

$$\omega(S) = \sum_{M \in S} \deg(M)!$$

Initially $\omega(S) \leq n^{\text{ndeg}(f)} \text{ndeg}(f)!$

We're done when $\omega(S)=0$

Claim: Each iteration decreases $\omega(S)$ by expected amount $\geq \omega(S)/4e$

Reason: $\geq 1/e$ of $\omega(S)$ is concentrated on maxonomials, each of which decreases in degree w.p. $\geq 1/2$

Recursive Fourier Sampling

(quant-ph/0209060)

Fourier Sampling

Given $A:\{0,1\}^n \rightarrow \{0,1\}$

Promise: $A(x)=s \cdot x \pmod{2}$ for some s

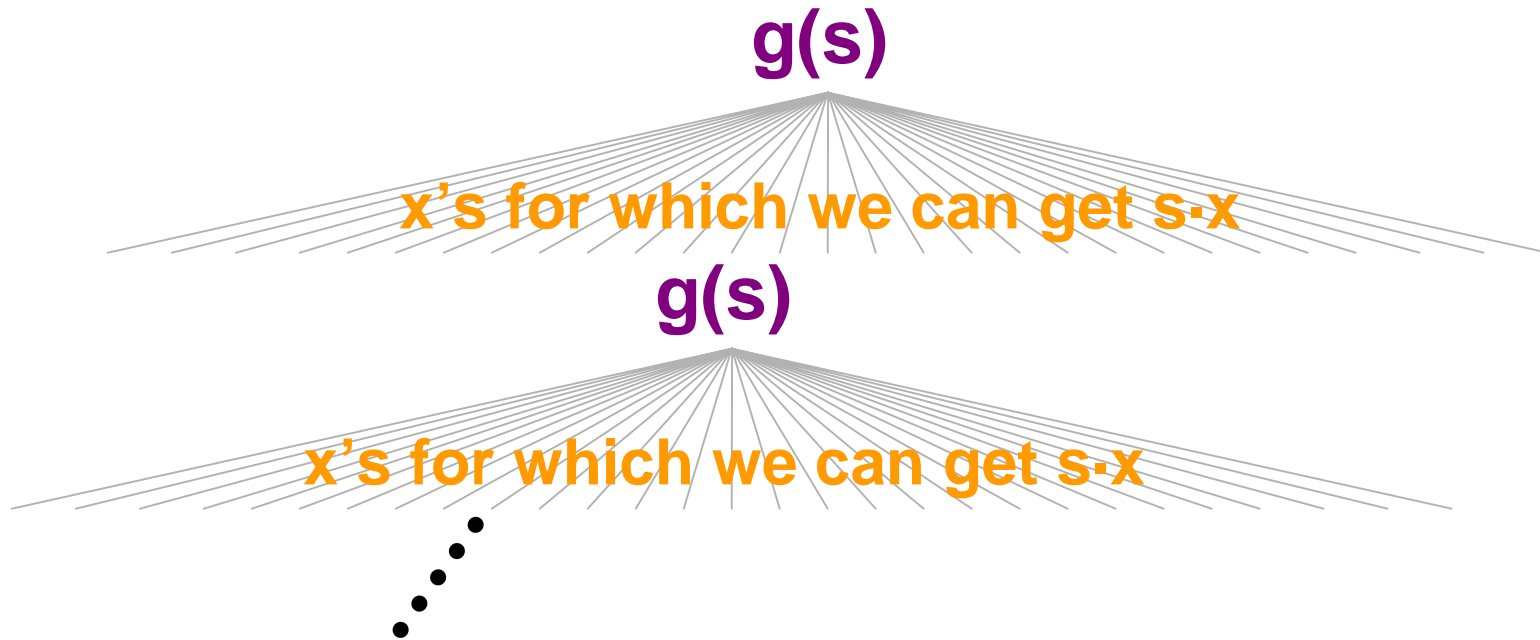
Return: $g(s)$, for some known $g:\{0,1\}^n \rightarrow \{0,1\}$
(possibly partial)

Classically: n queries needed

Quantumly: 2 queries

$$2^{-n/2} \sum_{x \in \{0,1\}^n} (-1)^{A(x)} |x\rangle \xrightarrow{H_{2^n}} |s\rangle$$

Recursive Fourier Sampling (RFS)



Fourier sampling composed $\log n$ times

Classically: $n^{\log n}$ queries

Quantumly: $2^{\log n} = n$ queries—or fewer?

Overview

- **Bernstein-Vazirani 1993:** RFS puts $BQP \not\subseteq MA$ relative to oracle
- Candidate for $BQP \not\subseteq PH$
- Could it put (say) $BQP \not\subseteq PH[\text{polylog}]$?
- **Is uncomputing necessary? Why?**
- **Goal:** Show $Q_2(\text{RFS}) = \Omega(c^d)$ for $c > 1$
 $d = \text{tree depth}$
- **Trouble:** Suppose $g(s)$ is a parity function
Then $Q_2(\text{RFS}_g) = 1$

Plan of Attack

- We define a **nonparity coefficient** of the function g ,

$$\mu(g) \in [0, 3/4]$$

- Measures how uncorrelated g is with parity of any subset of input bits

Examples: $\mu(\text{Parity})=0$, $\mu(\text{Mod } 3)=3/4-O(1/n)$

- We then prove a lower bound:

$$Q_2(RFS_g) = \Omega\left(\left(\frac{1}{1-\mu(g)}\right)^{(\log n)/2}\right)$$

- If $\mu(g)$ is close to 0, this bound is useless. But we show that **if $\mu(g) < 0.146$ then g is a parity function**

The Nonparity Coefficient $\mu(g)$

Max μ^* s.t. for some distributions D_0 over $g^{-1}(0)$,
 D_1 over $g^{-1}(1)$,

for all $z \neq 0^n$, $t_0 \in g^{-1}(0)$, $t_1 \in g^{-1}(1)$,

$$\Pr_{s_0 \in D_0, s_1 \in D_1} \left[s_0 g z \equiv t_1 g z \pmod{2} \vee s_1 g z \equiv t_0 g z \pmod{2} \right] \geq \mu^*$$

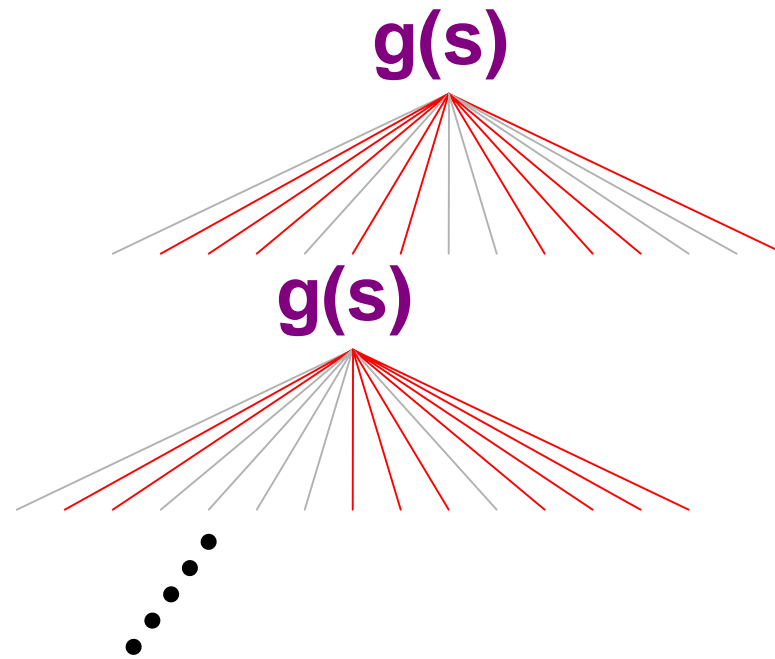
Theorem: $Q_2(RFS_g) = \Omega\left(\left(\frac{1}{1-\mu(g)}\right)^{(\log n)/2}\right)$

Proof Idea: Uses Ambainis' "most general" bound

Let $(x,y) \in R$ if $x \in f^{-1}(0)$, $y \in f^{-1}(1)$ "differ minimally"

Weight inputs by D_0, D_1 from nonparity coefficient

Then for all i and $(x^*, y^*) \in R$,



$$\Pr_{x \in D_0 : (x, y^*) \in R} [x_i \neq y_i^*] \Pr_{y \in D_1 : (x^*, y) \in R} [x_i^* \neq y_i] \leq (1 - \mu)^h$$

“Pseudoparity” Functions Don’t Exist

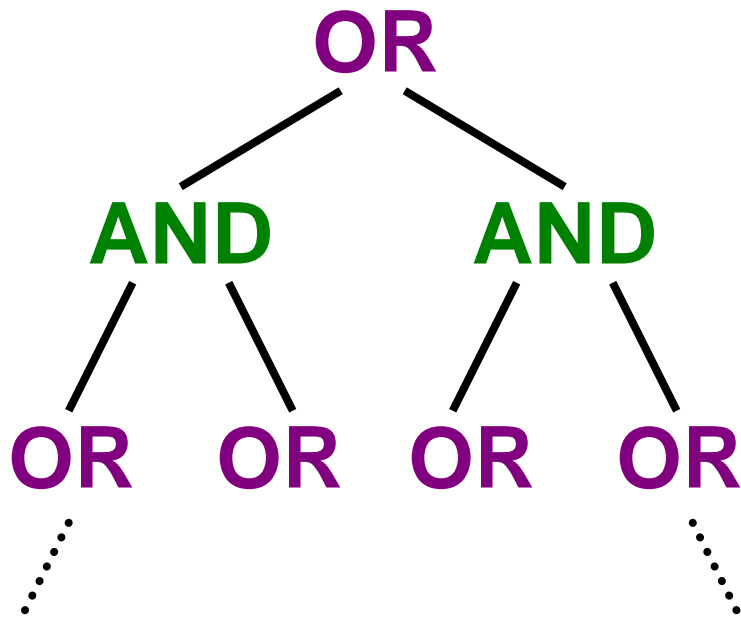
Theorem: If $\mu(g) < \frac{2 - \sqrt{2}}{4} \approx 0.146$

then $\mu(g)=0$ (i.e. g is a parity function)

So either

- (1) the adversary method gives a good quantum lower bound, or
- (2) there exists an efficient classical algorithm

In general, when can we do better for tree functions than by recursing on subtrees?

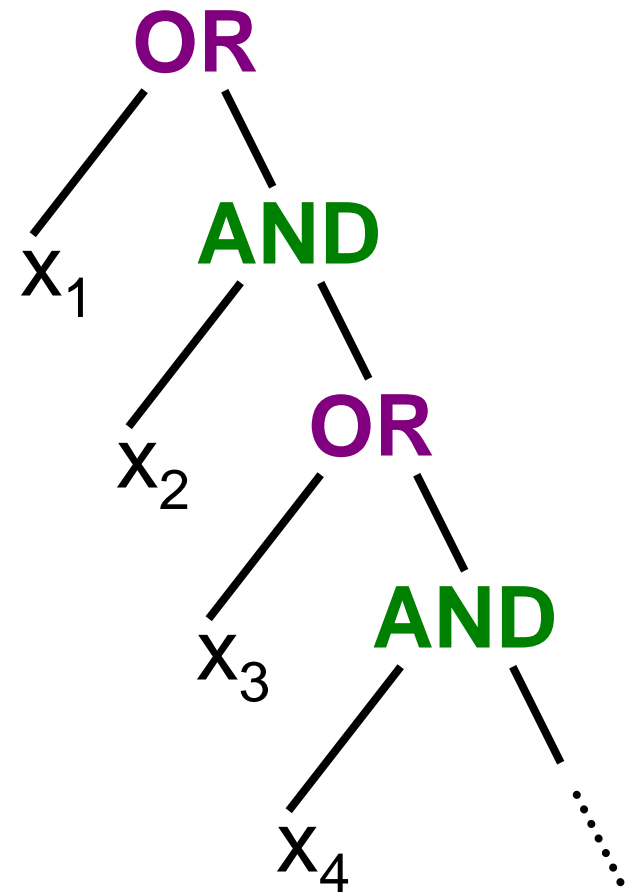


$$R_2(f) = \Theta(n^{0.753})$$

(Saks-Wigderson, Santha)

$$Q_2(f) = \Omega(\sqrt{n})$$

(Barnum-Saks: holds for any AND-OR tree)

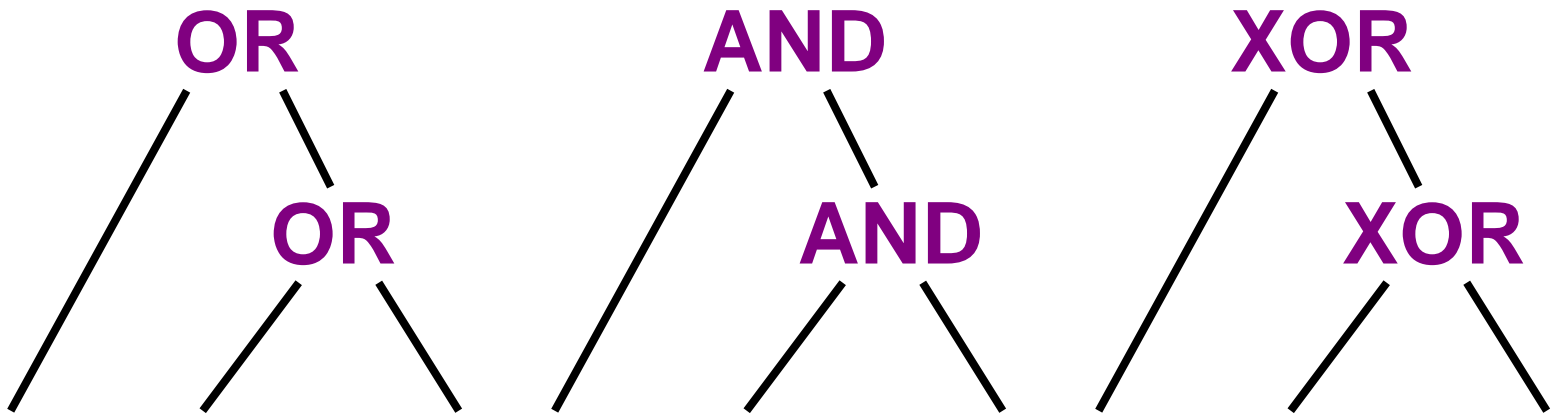


$$Q_2(f) = \Theta(\sqrt{n})$$

(Dürr-Høyer)

Does every Boolean function *have* a unique tree?

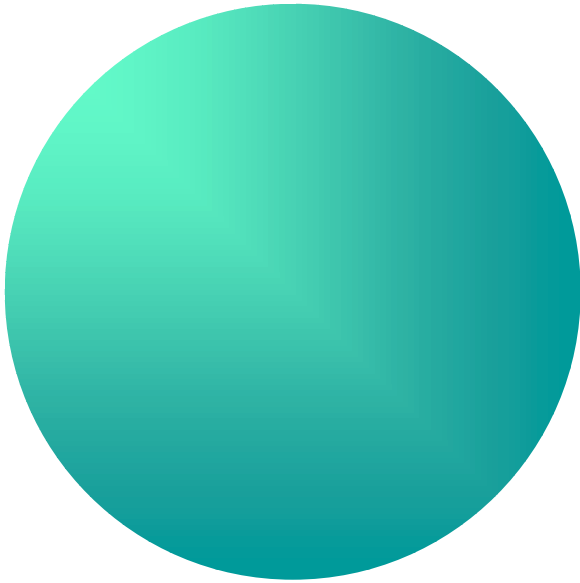
Theorem (A'2000): Yes, modulo three “degeneracies” ...



Query Complexity & Quantum Gravity

The Holographic Principle

('t Hooft, Susskind, Bekenstein, Bousso...)



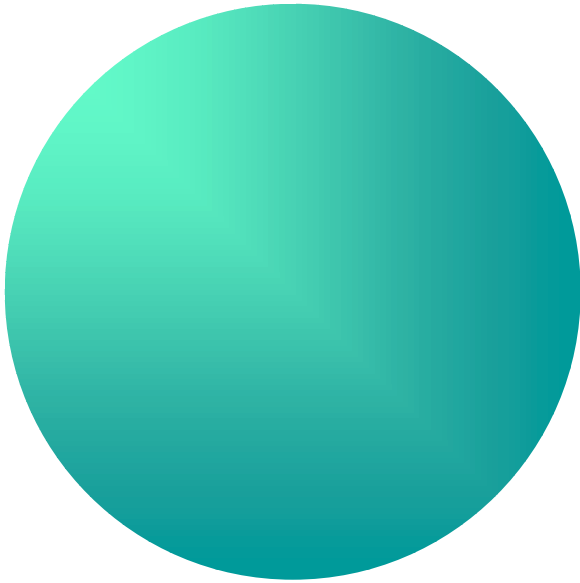
$$N \leq \frac{A}{4 \ln 2}$$

A = surface area of 3D region
(in Planck areas, $7.1 \times 10^{-70} \text{ m}^2$)

N = # of bits it contains

Tight for black holes

The Query Complexity Holographic Principle

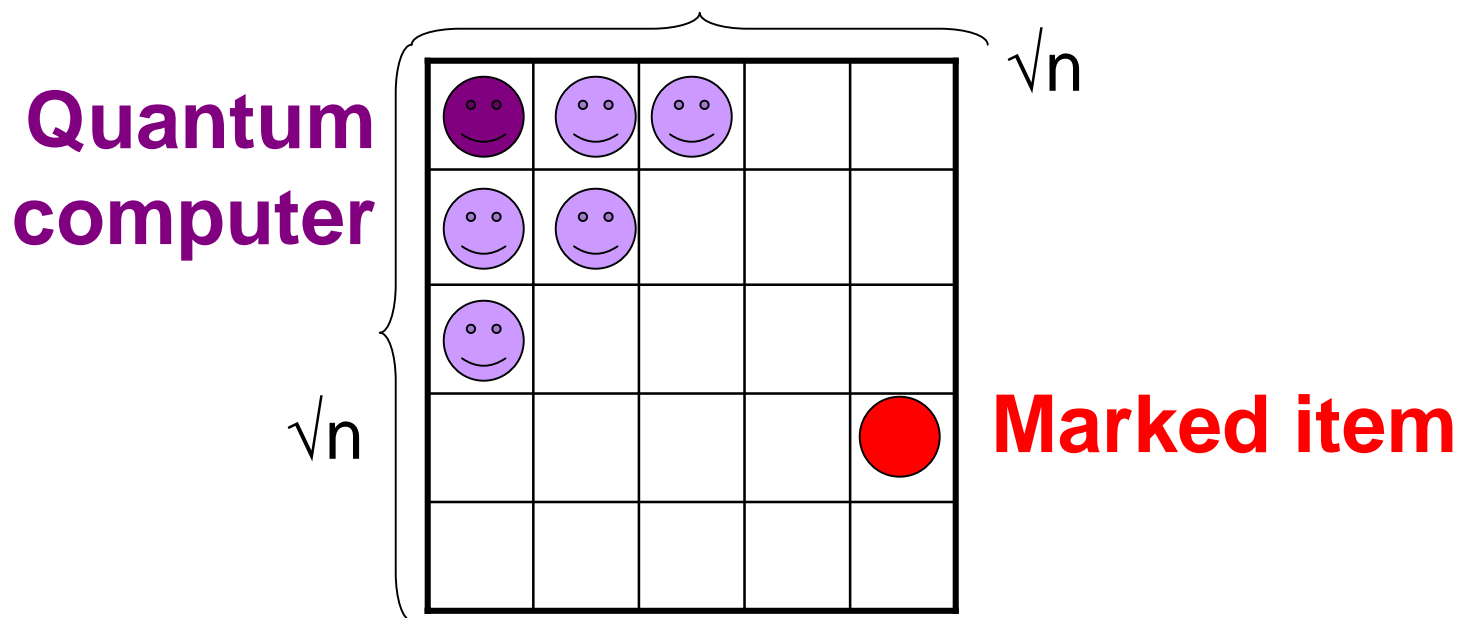


$$T = O(A)$$

A = surface area of 3D region

T = time needed to search it for a marked item (given finite speed of light)

Grover Search on a 2D Lattice



- Can do in $O(n^{3/4})$ time: searching a row classically takes \sqrt{n} time; combining the results using Grover takes $n^{1/4} \cdot \sqrt{n}$
- In d dimensions, can do in $O(n^{1/2+1/2d})$
- Implies “query complexity holographic principle”—when $d=3$, $n^{1/2+1/6}=n^{2/3}$ is $O(A)$, in the case where A is minimized (a sphere)
- **Conjecture:** $n^{1/2+1/2d}$ is optimal. Would imply “holographic” bound is tight for spheres (such as black holes...)