

# How Much Information Is In A Quantum State?

$\rho$



Scott Aaronson  
Andrew Drucker

# Computer Scientist / Physicist Nonaggression Pact

You tolerate these complexity classes:

NP coNP BQP QMA BQP/qpoly QMA/poly

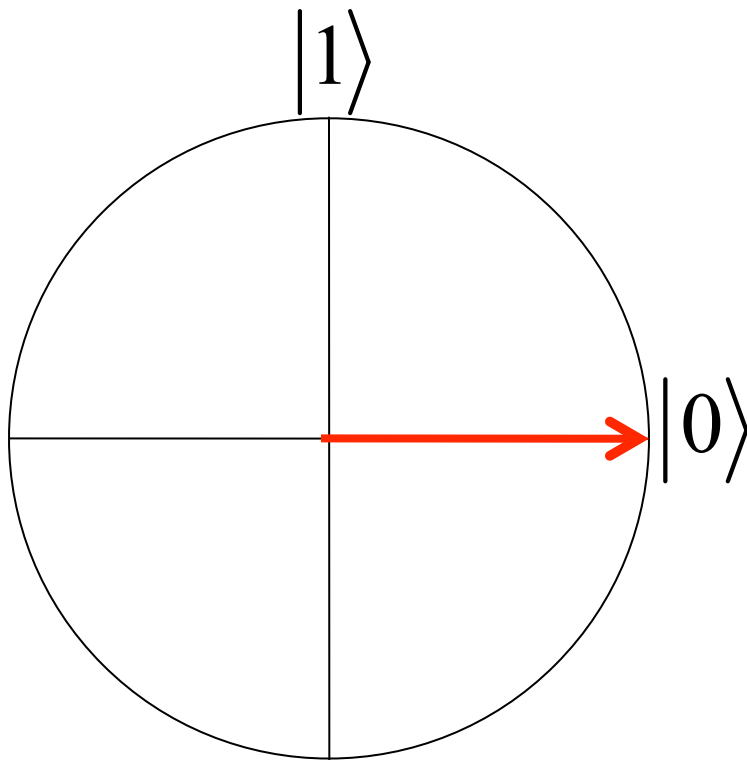
And I don't inflict these on you:

#P AM AWPP LWPP MA PostBQP PP CH  
PSPACE QCMA QIP SZK NISZK EXP NEXP UP  
PPAD PPP PLS TFNP  $\oplus P$  Mod<sub>k</sub>P

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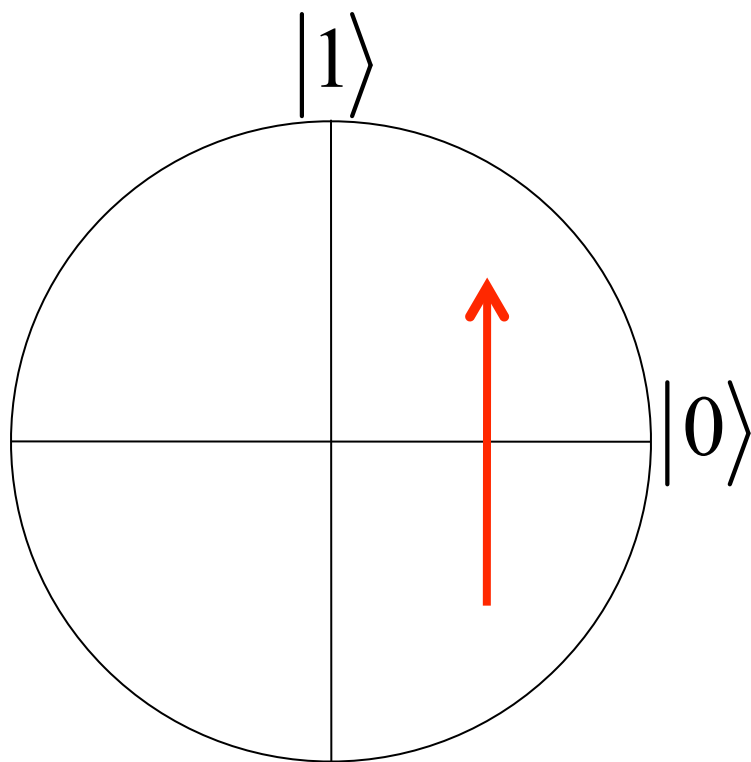
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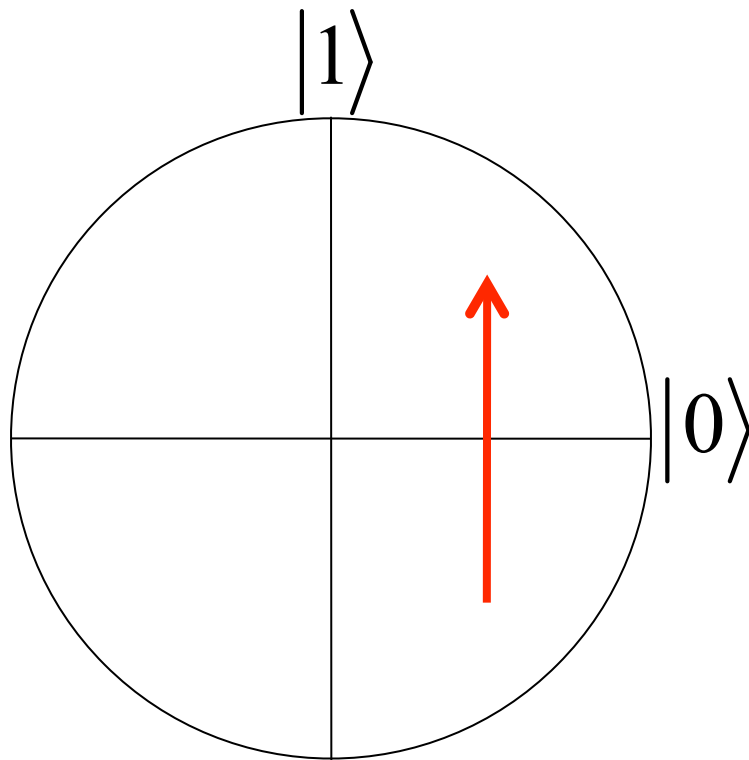
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**Life is too short for infinite precision**



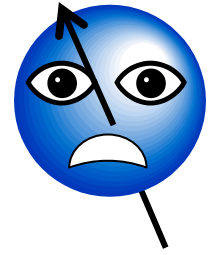
# A More Serious Point



$$|\psi\rangle = \sum_{x \in \{0,1\}^n} \alpha_x |x\rangle$$



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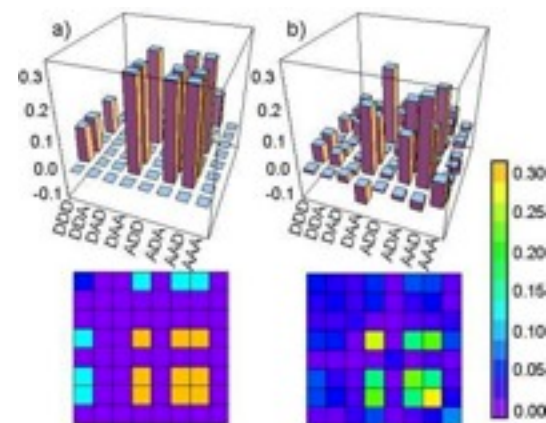
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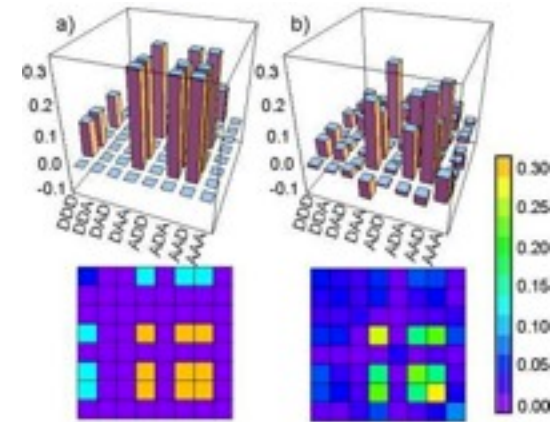
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But why should we worry about it?

# Answer 1: Quantum State Tomography

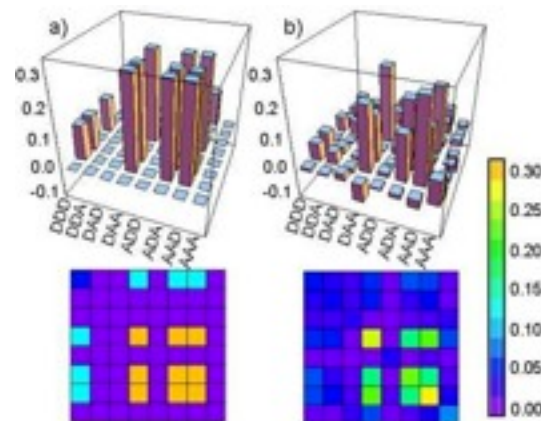


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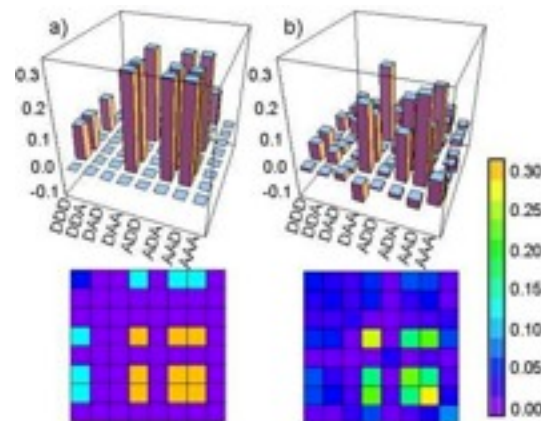


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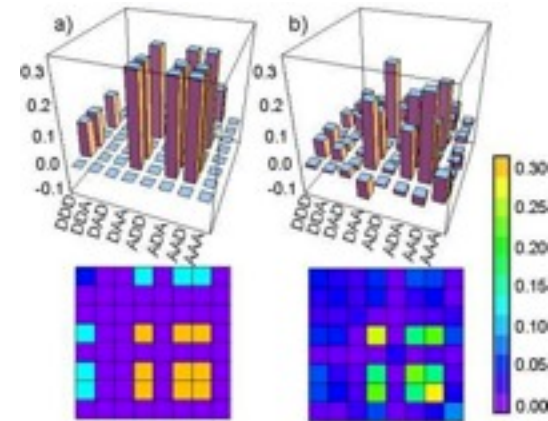
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This is a conceptual problem—not just a practical one!

# Answer 2: Quantum Computing Skepticism



Levin



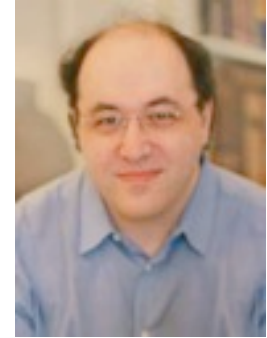
Goldreich



't Hooft



Davies



Wolfram

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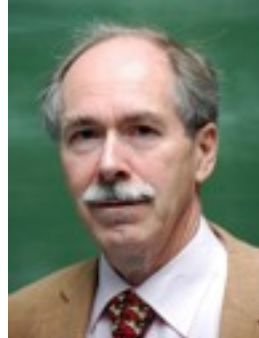
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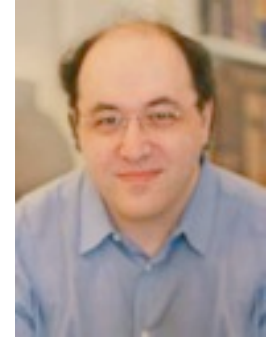
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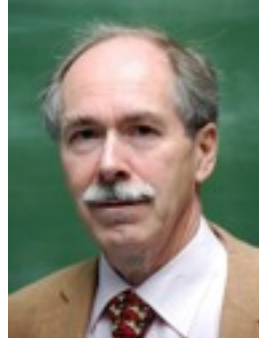
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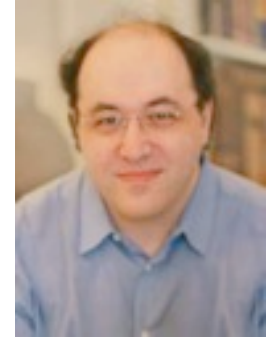
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But is it **really** an exponential amount?

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**Analogy:** A probability distribution over  $n$ -bit strings *also* takes  $\sim 2^n$  bits to specify. But that fact seems to be “more about the map than the territory”

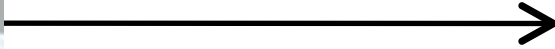
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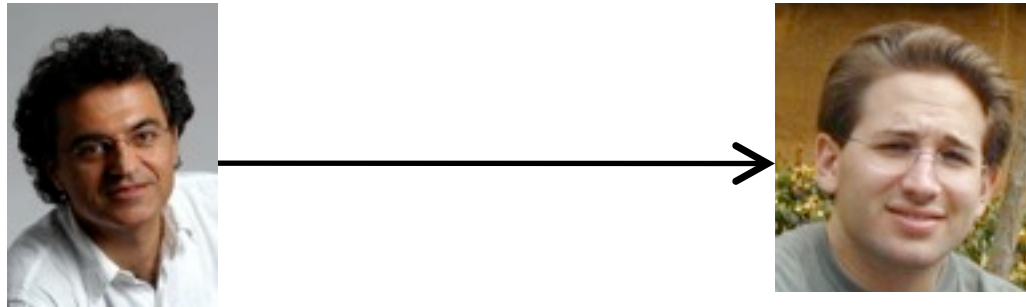
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- Describing a state by postselected measurements [A. 2004]
- “Pretty good tomography” using far fewer measurements [A. 2006]
  - Numerical simulation [A.-Dechter]
- Encoding quantum states as ground states of simple Hamiltonians [A.-Drucker 2009]

# The Absent-Minded Advisor Problem



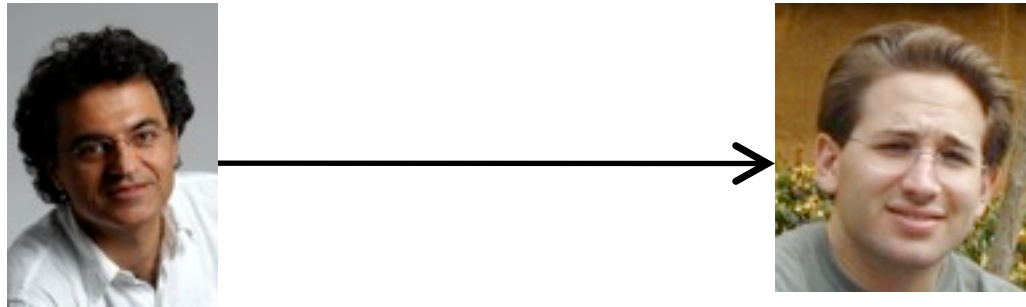
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Can you give your graduate student a quantum state  $\rho$  with  $n$  qubits (or  $10n$ , or  $n^3$ , ...)—such that by measuring  $\rho$  in a suitable basis, the student can learn your answer to any **one** yes-or-no question of size  $n$ ?



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**NO** [Ambainis, Nayak, Ta-Shma, Vazirani 1999]

Indeed, quantum communication is no better than classical for this problem as  $n \rightarrow \infty$ .

(Earlier, Holevo showed you need  $n$  qubits to send  $n$  bits)

**On the Bright Side...**



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Suppose Alice wants to describe an  $n$ -qubit state  $\rho$  to Bob, well enough that for any 2-outcome measurement  $E$ , Bob can estimate  $\text{Tr}(E\rho)$

Then she'll need to send  $\sim c^n$  bits, in the worst case.

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**Theorem (A. 2004):** In that case, it suffices for Alice to send  $\sim n \log n \cdot \log |S|$  bits



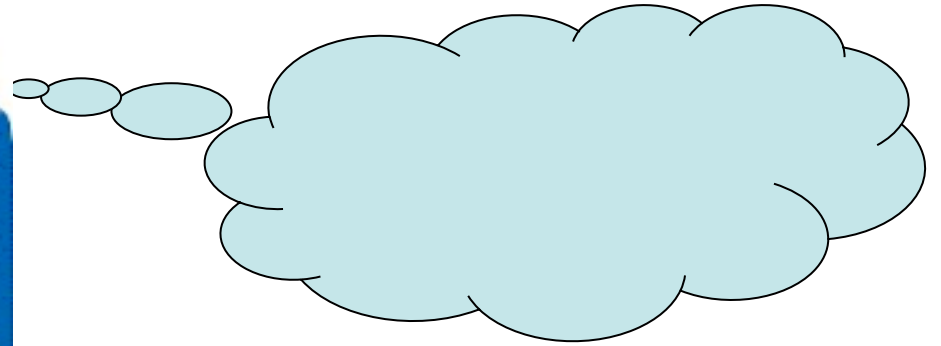
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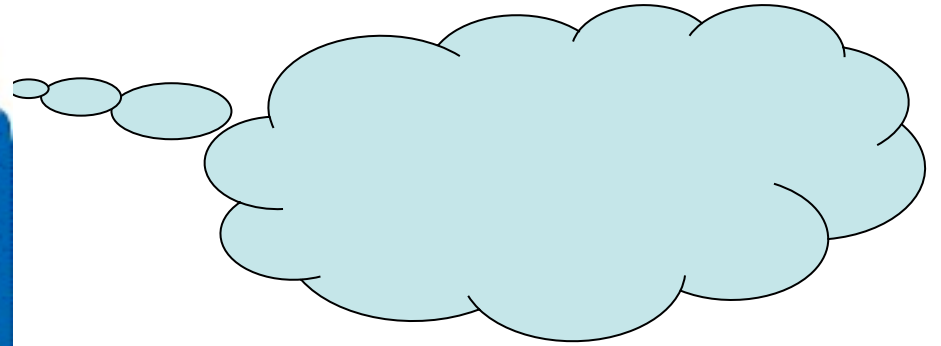
$|\psi\rangle$

**ALL MEASUREMENTS PERFORMABLE  
USING  $\leq n^2$  QUANTUM GATES**

# How does the theorem work?



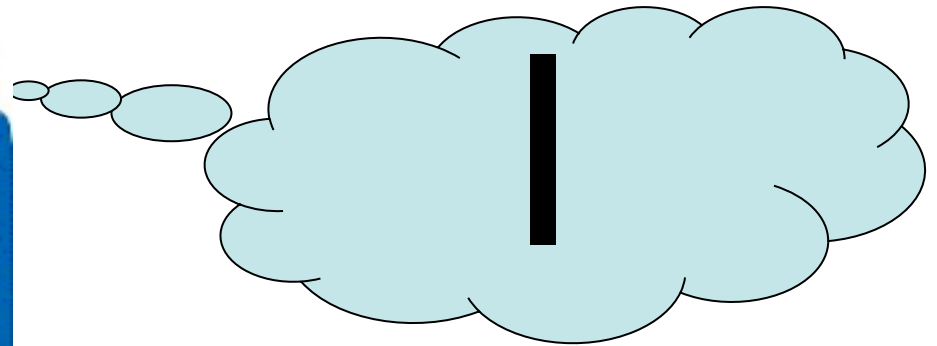
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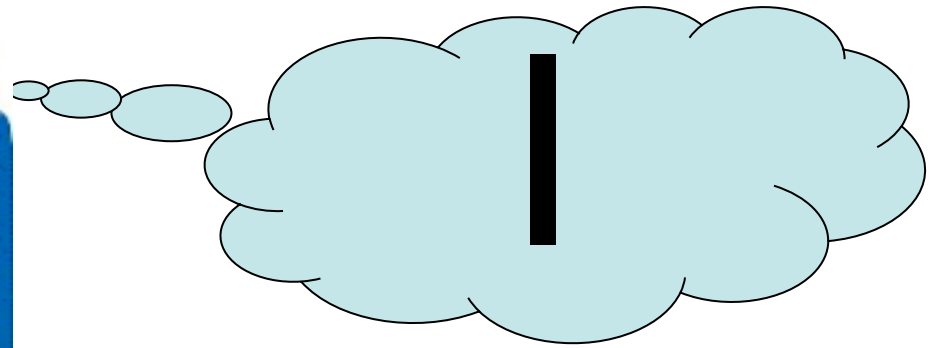
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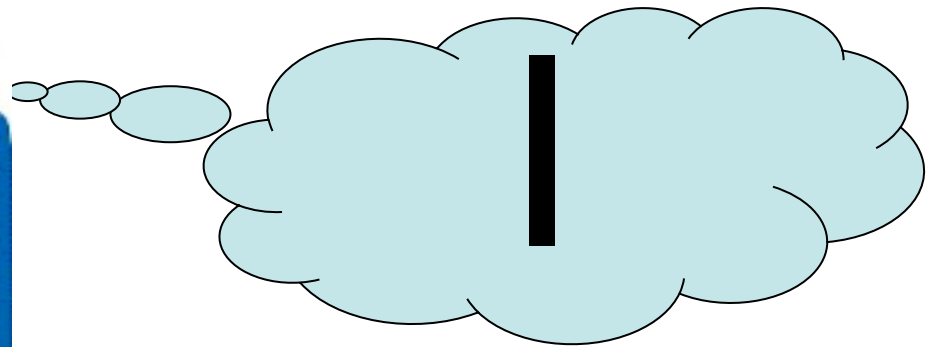


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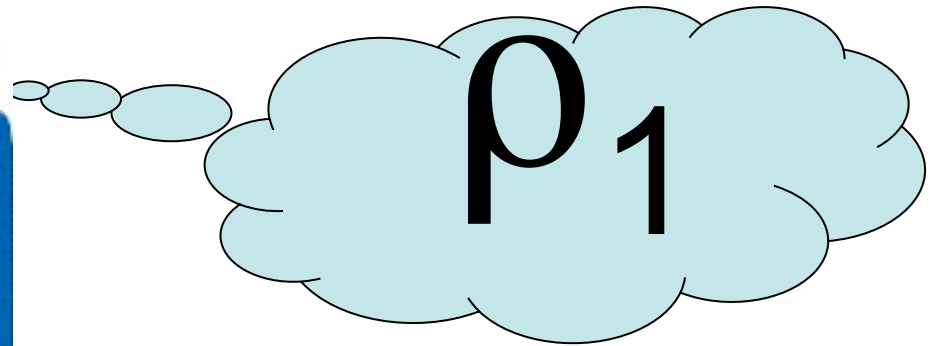
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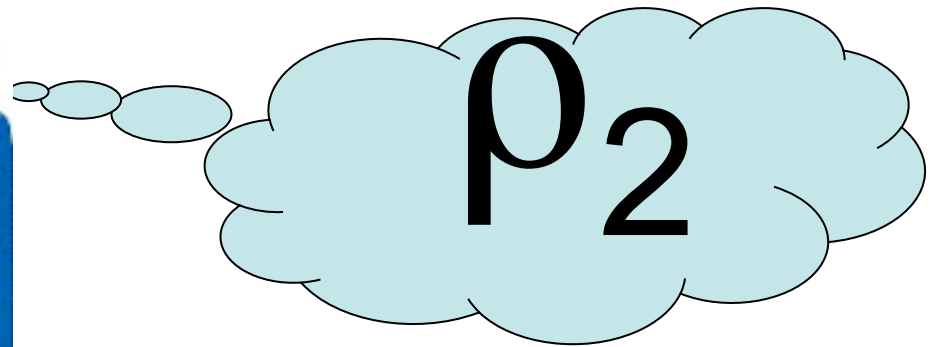
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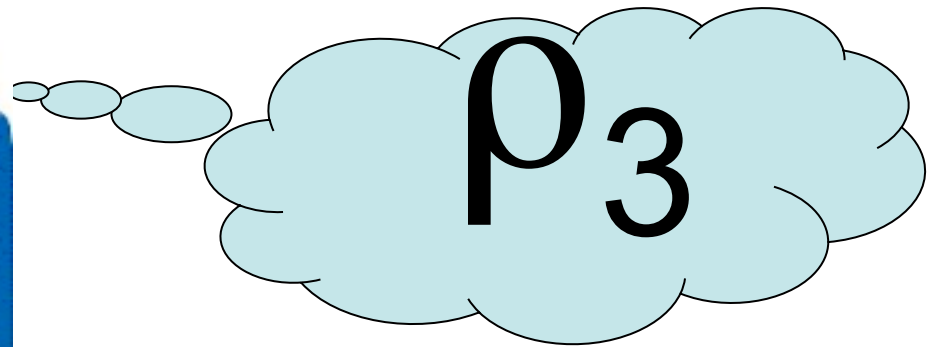
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1. Choose  $E_1, \dots, E_m$  independently from  $D$
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**“Quantum states are PAC-learnable”**

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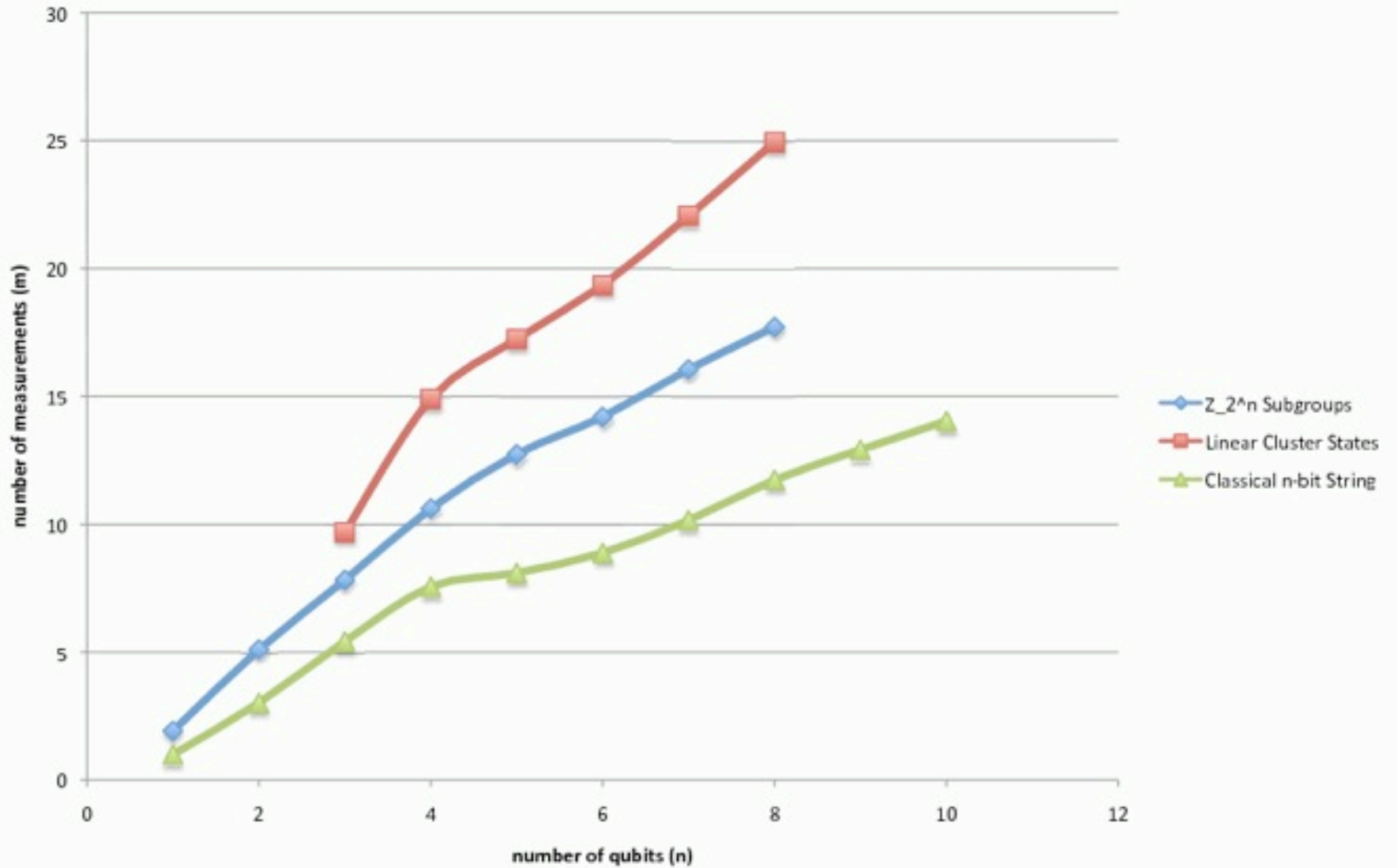
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**Result of experiment:** My theorem appears to be true

## Measurement Complexity of n







**Recap:** Given an unknown  $n$ -qubit entangled quantum state  $\rho$ , and a set  $S$  of two-outcome measurements...

**Learning theorem:** “Any hypothesis state  $\sigma$  consistent with a small number of sample points behaves like  $\rho$  on **most** measurements in  $S$ ”

**Postselection theorem:** “A particular state  $\rho_T$  (produced by postselection) behaves like  $\rho$  on **all** measurements in  $S$ ”

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**[A.-Drucker 2009]:** The dream theorem holds

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Any quantum state can be “simulated,” on all efficient measurements, by the ground state of a local Hamiltonian

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## IN OTHER WORDS...

Given any  $n$ -qubit state  $\rho$ , there exists a local Hamiltonian  $H$  (indeed, a sum of  $2D$  nearest-neighbor interactions) such that:

For any ground state  $|\psi\rangle$  of  $H$ , and measuring circuit  $E$  with  $\leq m$  gates, there's an efficient measuring circuit  $E'$  such that

$$\left| \langle \psi | E' | \psi \rangle - \text{Tr}(E\rho) \right| \leq \varepsilon.$$

Furthermore,  $H$  is on  $\text{poly}(n, m, 1/\varepsilon)$  qubits.

# What Does It Mean?

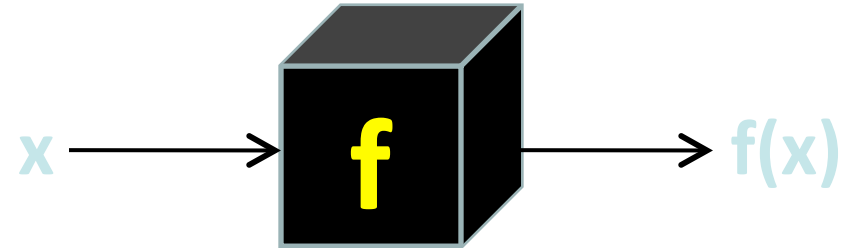
Without loss of generality, every quantum advice state is the ground state of a local Hamiltonian

$\text{BQP}/\text{qpoly} \subseteq \text{QMA}/\text{poly}$ . Indeed, trusted quantum advice is *equivalent* in power to trusted classical advice combined with untrusted quantum advice.

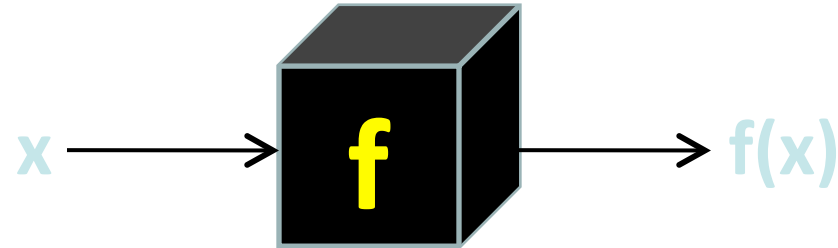
(“Quantum states never need to be trusted”)

**“Quantum Karp-Lipton Theorem”**:  $\text{NP}$ -complete problems are not efficiently solvable using quantum advice, unless some *uniform* complexity classes collapse

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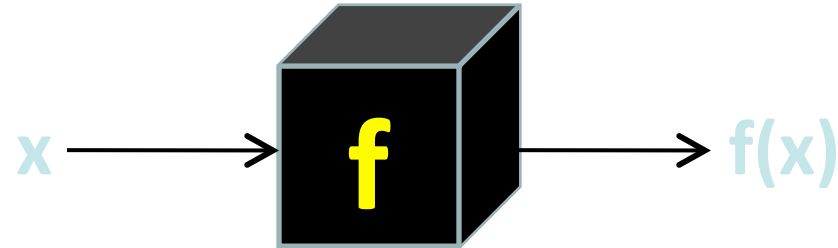
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that computes some Boolean function  $f:\{0,1\}^n \rightarrow \{0,1\}$  belonging to a “small” set  $S$  (meaning, of size  $2^{\text{poly}(n)}$ ). Someone wants to prove to us that  $f$  equals (say) the all-0 function, by having us check a polynomial number of outputs  $f(x_1), \dots, f(x_m)$ .



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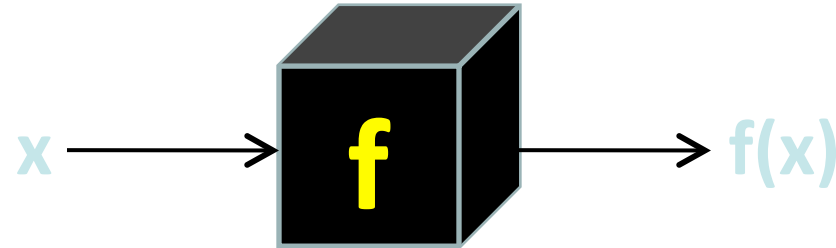
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This is trivially impossible!

↓

	$f_0$	$f_1$	$f_2$	$f_3$	$f_4$	$f_5$
$x_1$	0	1	0	0	0	0
$x_2$	0	0	1	0	0	0
$x_3$	0	0	0	1	0	0
$x_4$	0	0	0	0	1	0
$x_5$	0	0	0	0	0	1

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But ... what if we get **3** black boxes, and are allowed to simulate  $f=f_0$  by taking the point-wise MAJORITY of their outputs?

	$f_0$	$f_1$	$f_2$	$f_3$	$f_4$	$f_5$
$x_1$	0	1	0	0	0	0
$x_2$	0	0	1	0	0	0
$x_3$	0	0	0	1	0	0
$x_4$	0	0	0	0	1	0
$x_5$	0	0	0	0	0	1

Three blue arrows point down from the top of the table to the columns for  $f_1$ ,  $f_2$ , and  $f_3$ . The cells containing '1' in the  $f_1$ ,  $f_2$ , and  $f_3$  columns are circled in red.

# Majority-Certificates Lemma

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**Definitions:** A *certificate* is a partial Boolean function  $C:\{0,1\}^n \rightarrow \{0,1,*\}$ . A Boolean function  $f:\{0,1\}^n \rightarrow \{0,1\}$  is *consistent* with  $C$ , if  $f(x)=C(x)$  whenever  $C(x) \in \{0,1\}$ . The *size* of  $C$  is the number of inputs  $x$  such that  $C(x) \in \{0,1\}$ .

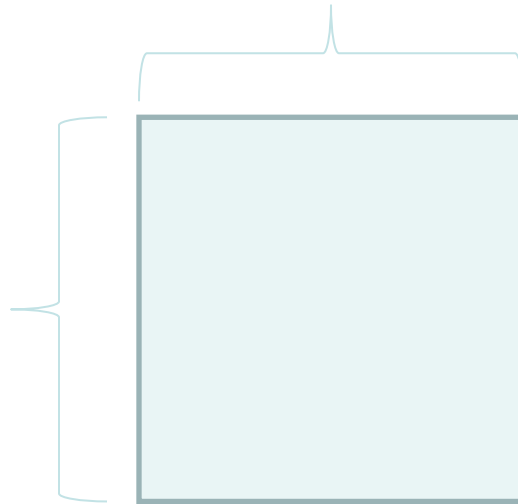
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**Lemma:** Let  $S$  be a set of Boolean functions  $f:\{0,1\}^n \rightarrow \{0,1\}$ , and let  $f^* \in S$ . Then there exist  $m=O(n)$  certificates  $C_1, \dots, C_m$ , each of size  $k=O(\log |S|)$ , such that

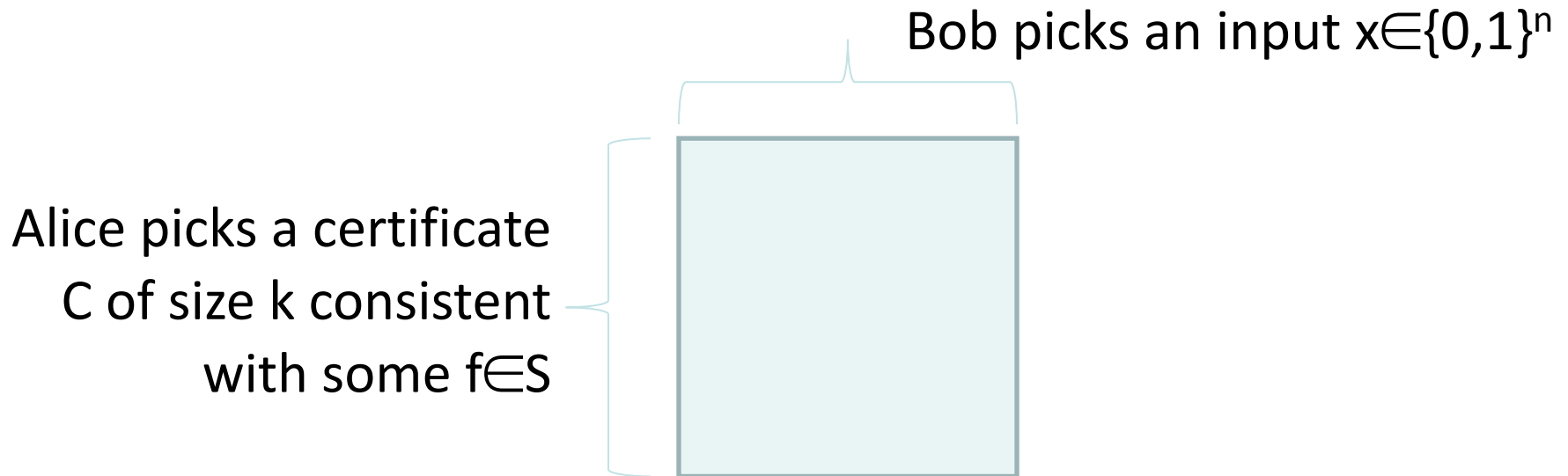
- (i) Some  $f_i \in S$  is consistent with each  $C_i$ , and
- (ii) If  $f_i \in S$  is consistent with  $C_i$  for all  $i$ , then  $\text{MAJ}(f_1(x), \dots, f_m(x)) = f^*(x)$  for all  $x \in \{0,1\}^n$ .

# Proof Idea



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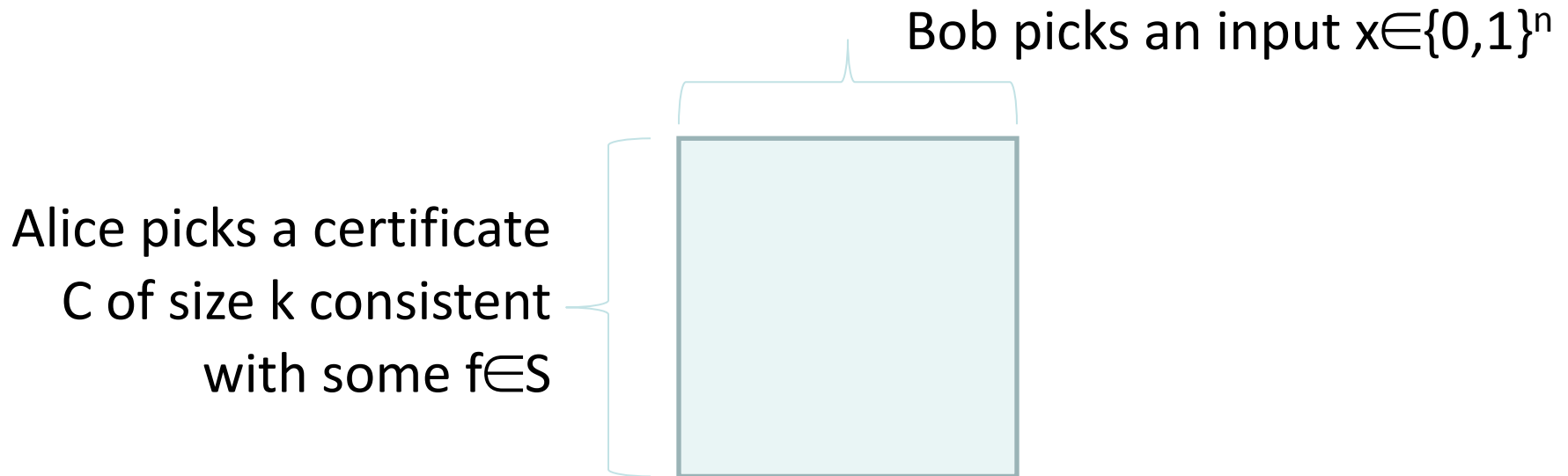
By symmetry, we can assume  $f^*$  is the all-0 function. Consider a two-player, zero-sum matrix game:



Alice wins this game if  $f(x)=0$  for all  $f \in S$  consistent with  $C$ .

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Bob picks an input  $x \in \{0,1\}^n$

Alice  
C

The lemma follows from this claim! Just choose certificates  $C_1, \dots, C_m$  independently from Alice's winning distribution. Then by a Chernoff bound, almost certainly  $\text{MAJ}(f_1(x), \dots, f_m(x)) = 0$  for all  $f_1, \dots, f_m$  consistent with  $C_1, \dots, C_m$  respectively and all inputs  $x \in \{0,1\}^n$ . So clearly there *exist*  $C_1, \dots, C_m$  with this property.

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Use the Minimax Theorem! Given a distribution  $D$  over  $x$ , it's enough to create a *fixed* certificate  $C$  such that

$$\Pr_{x \in D} \left[ \exists f \text{ consistent with } C \text{ s.t. } f(x) = 1 \right] < \frac{1}{10}.$$

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**Stage I:** Choose  $x_1, \dots, x_t$  independently from  $D$ , for some  $t = O(\log |S|)$ . Then with high probability, requiring  $f(x_1) = \dots = f(x_t) = 0$

kills off every  $f \in S$  such that

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# “Lifting” the Lemma to Quantumland

Boolean Majority-	BQP/adv=QOP/
Set S of Boolean	Set S of $n(n)$ -qubit
“True” function	“True” advice state
Other functions $f_1, \dots, f_k$	Other states $\rho_1, \dots, \rho_k$
Certificate C. to	Measurement F. to

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New Difficulty	Solution
The class of p(n)-qubit quantum states is infinitely	Result of A.'06 on learnability of quantum
Instead of Boolean functions $f:\{0,1\}^n \rightarrow \{0,1\}$ , now we have	Learning theory has tools to deal with this: fat-
How do we verify a quantum	<b>QMA=QMA+</b> (Aharonov & ...)
What if a certificate asks us to	“Safe Winnowing Lemma”



# Majority-Certificates Lemma, Real Case

**Lemma:** Let  $S$  be a set of functions  $f: \{0,1\}^n \rightarrow [0,1]$ , let  $f^* \in S$ , and let  $\epsilon > 0$ . Then we can find  $m = O(n/\epsilon^2)$  functions  $f_1, \dots, f_m \in S$ , sets

$$X_1, \dots, X_m \subseteq \{0,1\}^n \text{ each of size } k = O\left(\frac{n}{\epsilon^3} \text{fat}_{\epsilon/48}(S)\right), \quad \text{and} \quad \alpha = \Omega\left(\frac{\epsilon^2}{n \text{fat}_{\epsilon/48}(S)}\right)$$

for which the following holds. All functions  $g_1, \dots, g_m \in S$  that

satisfy  $\max_{x \in X_i} |g_i(x) - f^*(x)| \leq \alpha$  for all  $i \in [m]$  also satisfy  $\max_{x \in \{0,1\}^n} \left| \frac{1}{m} [g_1(x) + \dots + g_m(x)] - f^*(x) \right| \leq \epsilon$ .



# Theorem: $\text{BQP}/\text{qpoly} \subseteq \text{QMA}/\text{poly}$ .

**Proof Sketch:** Let  $L \in \text{BQP}/\text{qpoly}$ . Let  $M$  be a quantum algorithm that decides  $L$  using advice state  $|\psi_n\rangle$ . Define

$$f_\rho(x) := \Pr[M(x, \rho) \text{ accepts}]$$

Let  $S = \{f_\rho : \rho\}$ . Then  $S$  has fat-shattering dimension at most  $\text{poly}(n)$ , by A.'06. So we can apply the real analogue of the Majority-Certificates Lemma to  $S$ . This yields certificates  $C_1, \dots, C_m$  (for some  $m = \text{poly}(n)$ ), such that any states  $\rho_1, \dots, \rho_m$

consistent with  $C_1, \dots, C_m$  respectively satisfy

$$\left| \frac{1}{m} \left( f_{\rho_1}(x) + \dots + f_{\rho_m}(x) \right) - f_{|\psi_n\rangle\langle\psi_n|}(x) \right| \leq \varepsilon$$

for all  $x \in \{0,1\}^n$  (regardless of entanglement). To check the  $C_i$ 's, we use the “**QMA+** super-verifier” of Aharonov & Regev.

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Our quantum analogue:

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**Proof Idea:** In  $\text{QMA}^{\text{PromiseQMA}}$ , first guess a local Hamiltonian  $H$  whose ground state  $|\psi\rangle$  lets us solve  $\text{NP}$ -complete problems in polynomial time, together with  $|\psi\rangle$  itself. Then pass  $H$  to the **PromiseQMA** oracle, which reconstructs  $|\psi\rangle$ , guesses the first quantified string of the  $\text{coNP}^{\text{NP}}$  statement, and uses  $|\psi\rangle$  to find the second quantified string.

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To check that  $|\psi\rangle$  actually works, use the self-reducibility of  $\text{NP}$ -complete problems (like in the original K-L Theorem)



# Summary

In many natural scenarios, the “exponentiality” of quantum states is an illusion

That is, there’s a short (though possibly cryptic) classical string that specifies how a quantum state  $\rho$  behaves, on any measurement you could actually perform

**Applications:** Pretty-good quantum state tomography, characterization of quantum computers with “magic initial states” ...

# Open Problems

Find classes of quantum states that can be learned in a **computationally** efficient way

[A.-Gottesman, in preparation]: Stabilizer states

Oracle separation between **BQP/poly** and **BQP/qpoly**

[A.-Kuperberg 2007]: Quantum oracle separation

Other applications of “isolatability” of Boolean functions?

“Experimental demonstration”?