# A one-query lower bound for unitary synthesis and breaking quantum cryptography 

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joint work with Alex Lombardi and John Wright

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- Factoring: given a positive integer $N$, compute $f(N)=$ prime factorization of $N$.
- Local Hamiltonian: given a local Hamiltonian $H$, compute the bit $f(H)$ indicating whether $H$ has a low-energy ground state.

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Physics: computing AdS/CFT map, decoding black-hole radiation


## What can complexity theory say about the hardness of these inherently quantum problems?

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Not known how to solve this using any oracle, even an oracle for the halting problem!

## Before we continue:

1-minute detour for quantum computing 101

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poly $(n)$ gates


## Now back to:

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To apply complexity theory, we need to efficiently reduce the task of implementing a unitary $U$ to implementing a function $f$.

> The Unitary Synthesis Problem [AK06]: Is there a reduction for every unitary U?

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Note: [AK06] prove a 1-query lower bound for a very special class of oracle algorithms.

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(2) Even one-query algorithms are very powerful!

In fact, they can solve any classical input, quantum output problem. [Aar16, INNRY22, Ros23]

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Note: when $l=2^{2 n}$, possible to learn description of $U$ in one query.

## Rest of this talk

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Our answer: possibly harder than computing any function!

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Note: this result implies our unitary synthesis lower bound.

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given one query to a function $f$, which can depend on $R:=\left\{R_{k}\right\}$.

Next up: what does a one-query adversary look like?

# One-query adversaries 

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2) Apply matrix concentration

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## Rest of this talk

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## Part 2: <br> A special case of our proof

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Disclaimer: We can rule out these attacks with a counting argument, but today we'll see a different proof.

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Distinguishing advantage:

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\underset{k \leftarrow[K]}{\mathbb{E}}\left\langle\psi_{R_{k}}\right| \cdot O_{f} \cdot \Pi \cdot O_{f} \cdot\left|\psi_{R_{k}}\right\rangle-\underset{h}{\mathbb{E}}\left\langle\psi_{h}\right| \cdot O_{f} \cdot \Pi \cdot O_{f} \cdot\left|\psi_{h}\right\rangle
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Scalar Chernoff bound: If $X$ is a random scalar with bounded absolute value, then for i.i.d. $X_{1}, \ldots, X_{K}$

$$
\left|\frac{1}{K} \sum_{k} X_{k}-\mathbb{E}[X]\right| \approx 0\left(\frac{1}{\sqrt{K}}\right) \quad \text { (w.h.p.) }
$$

## Technical tool: matrix concentration

Scalar Chernoff bound: If $X$ is a random scalar with bounded absolute value, then for i.i.d. $X_{1}, \ldots, X_{K}$

$$
\begin{equation*}
\left|\frac{1}{K} \sum_{k} X_{k}-\mathbb{E}[X]\right| \approx O\left(\frac{1}{\sqrt{K}}\right) \tag{w.h.p.}
\end{equation*}
$$

Matrix Chernoff bound: If $X$ is a random Hermitian $L \times L$ matrix with bounded operator norm, then for i.i.d. $X_{1}, \ldots, X_{K}$

$$
\left\|\frac{1}{K} \sum_{k} X_{k}-\mathbb{E}[X]\right\|_{\mathrm{op}} \approx O\left(\frac{\sqrt{\log (L)}}{\sqrt{K}}\right) \quad \text { (w.h.p.) }
$$

Adversary's advantage (for this special class):
$\left.\max _{f:[\mathbb{N}] \rightarrow\{ \pm 1\}}\left|\frac{1}{K} \sum_{k}\left\langle\psi_{R_{k}}\right| \cdot O_{f} \cdot \Pi \cdot O_{f} \cdot\right| \psi_{R_{k}}\right\rangle-\mathbb{E}_{h}\left\langle\psi_{h}\right| \cdot O_{f} \cdot \Pi \cdot O_{f} \cdot\left|\psi_{h}\right\rangle \mid$

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$$

Matrix Chernoff:

$$
\left.\max _{|v\rangle}\left|\langle v| \cdot\left(\frac{1}{K} \sum_{k} X_{k}-\mathbb{E}[X]\right) \cdot\right| v\right\rangle \mid
$$

Adversary's advantage (for this special class):

$$
\max _{f:[\mathrm{N}] \rightarrow\{ \pm 1\}}|\frac{1}{K} \sum_{k}\left\langle\psi_{R_{k}}\right| \cdot \underbrace{O_{f} \cdot \Pi \cdot O_{f}}_{\text {max over matrices }} \cdot| \psi_{R_{k}}\rangle-\mathbb{E}_{h}\left\langle\psi_{h}\right| \cdot O_{f} \cdot \Pi \cdot O_{f} \cdot\left|\psi_{h}\right\rangle \mid
$$

Matrix Chernoff:

$$
\begin{aligned}
\max _{|v\rangle} \mid & \left.\langle v| \cdot\left(\frac{1}{K} \sum_{k} X_{k}-\mathbb{E}[X]\right) \cdot|v\rangle \right\rvert\, \\
& \text { random matrices } \quad \text { max over unit vectors }
\end{aligned}
$$

Adversary's advantage (for this special class):

$$
\max _{f:[\mathrm{N}] \rightarrow\{ \pm 1\}} \left\lvert\, \frac{1}{K} \sum_{k}\left\langle\psi_{R_{k}}\right| \cdot \underbrace{O_{f} \cdot \Pi \cdot O_{f} \cdot\left|\psi_{R_{k}}\right\rangle-\mathbb{E}_{h}\left\langle\psi_{h}\right| \cdot O_{f} \cdot \Pi \cdot O_{f} \cdot\left|\psi_{h}\right\rangle \mid}_{\text {max over matrices }}\right.
$$

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$$

Key step: we can refactor this as $\left\langle v_{f}\right| \cdot\left(\right.$ random matrix) $\cdot\left|v_{f}\right\rangle$

$$
=\frac{1}{K} \sum_{k} X_{k}-E[X] \quad \begin{gathered}
f \text {-dependent } \\
\text { unit vector }
\end{gathered}
$$

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f \text {-dependent } \\
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Then matrix Chernoff will bound the max over all unit vectors.

Adversary's advantage (for this special class):
$\left.\max _{f:\{\mathbb{N} \mid \rightarrow\{ \pm 1\}}\left|\frac{1}{K} \sum_{k} \frac{\left\langle\psi_{R_{k}}\right| \cdot O_{f} \cdot \Pi \cdot O_{f} \cdot\left|\psi_{R_{k}}\right\rangle}{}-\mathbb{E}_{h}\left\langle\psi_{h}\right| \cdot O_{f} \cdot \Pi \cdot O_{f} \cdot\right| \psi_{h}\right\rangle \mid$
Since all the terms look identical, it suffices to just look at one term.

We'll rewrite this as $\left\langle v_{f}\right| \cdot$ (random matrix) $\cdot\left|v_{f}\right\rangle$
$\stackrel{\left\langle\psi_{R_{k}}\right| O_{f} \cdot \Pi \cdot O_{f}\left|\psi_{R_{k}}\right\rangle}{ }$

We'll rewrite this as $\left\langle v_{f}\right| \cdot$ (random matrix) $\cdot\left|v_{f}\right\rangle$
$\left\langle\psi_{R_{k}}\right| O_{f} \cdot \Pi \cdot O_{f}\left|\psi_{R_{k}}\right\rangle$
(1) Write the binary phase state $\left|\psi_{R_{k}}\right\rangle$ as

$$
\left|\psi_{R_{k}}\right\rangle=\left(\begin{array}{lll}
\ddots & & \\
& R_{k}(x) & \\
& & \ddots
\end{array}\right) \cdot \frac{1}{\sqrt{N}}\left(\begin{array}{c}
1 \\
\vdots \\
1
\end{array}\right)
$$

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1 \\
\vdots \\
1
\end{array}\right) \\
& \begin{array}{c}
N \times N \text { diagonal matrix, } \\
\\
\\
x \text {-th entry is } R_{k}(x)
\end{array} \\
& \text { superposition }
\end{aligned}
$$

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\end{array}\right)}_{:=D_{R_{k}}} \cdot \underbrace{\frac{1}{\sqrt{N}}\left(\begin{array}{c}
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1
\end{array}\right)}_{:=\left|+{ }_{N}\right\rangle}
$$

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$$
\begin{equation*}
\left.\stackrel{\langle }{\left\langle\psi_{R_{k}}\right| O_{f} \cdot \Pi \cdot O_{f} \mid \psi_{R_{k}}}\right\rangle=\left\langle+_{N}\right| D_{R_{k}} \cdot O_{f} \cdot \Pi \cdot O_{f} \cdot D_{R_{k}}\left|+_{N}\right\rangle \tag{1}
\end{equation*}
$$

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(2) $O_{f}$ is a diagonal matrix, so it commutes with $D_{R_{k}}$

$$
\begin{align*}
& \overbrace{\text { We'll rewrite this as }\left\langle v_{f}\right| \cdot \text { (random matrix) } \cdot\left|v_{f}\right\rangle}^{\left\langle\psi_{R_{k}}\right| O_{f} \cdot \Pi \cdot O_{f}\left|\psi_{R_{k}}\right\rangle}\rangle \\
& =\left\langle+_{N}\right| D_{R_{k}} \cdot O_{f} \cdot \Pi \cdot O_{f} \cdot D_{R_{k}}\left|+_{N}\right\rangle  \tag{1}\\
&  \tag{2}\\
& =\left\langle+_{N}\right| O_{f} \cdot\left(D_{R_{k}} \cdot \Pi \cdot D_{R_{k}}\right) \cdot O_{f}\left|+_{N}\right\rangle
\end{align*}
$$

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$\begin{aligned} & \text { Distinguishing } \\ & \text { advantage }\end{aligned} \quad \frac{1}{K} \sum_{k}\left\langle\psi_{R_{k}}\right| O_{f} \cdot \Pi \cdot O_{f}\left|\psi_{R_{k}}\right\rangle-\mathbb{E}_{h}\left\langle\psi_{h}\right| O_{f} \cdot \Pi \cdot O_{f}\left|\psi_{h}\right\rangle, ~$

## Distinguishing advantage <br> $$
\frac{1}{K} \sum_{k}\left\langle\psi_{R_{k}}\right| O_{f} \cdot \Pi \cdot O_{f}\left|\psi_{R_{k}}\right\rangle-\mathbb{E}_{h}\left\langle\psi_{h}\right| O_{f} \cdot \Pi \cdot O_{f}\left|\psi_{h}\right\rangle
$$

Rewrite as:

$$
=\left\langle+_{N}\right| O_{f}\left(\frac{1}{K} \sum_{k} D_{R_{k}} \cdot \Pi \cdot D_{R_{k}}-\mathbb{E}_{h}\left[D_{h} \cdot \Pi \cdot D_{h}\right]\right) O_{f}\left|+_{N}\right\rangle
$$

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Rewrite as:

$$
\begin{aligned}
& =\langle+_{N} \left\lvert\, O_{f}\left(\frac{1}{K} \sum_{k} D_{R_{k}} \cdot \Pi \cdot D_{R_{k}}-\mathbb{E}_{h}\left[D_{h} \cdot \Pi \cdot D_{h}\right]\right) \underbrace{O_{f} \mid+_{N}}_{\text {unit vector }}\right.\rangle \\
& \leq\left\|\frac{1}{K} \sum_{k} D_{R_{k}} \cdot \Pi \cdot D_{R_{k}}-\mathbb{E}_{h}\left[D_{h} \cdot \Pi \cdot D_{h}\right]\right\|_{\mathrm{op}}
\end{aligned}
$$

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\end{aligned}
$$

by Matrix Chernoff with $X_{k}=D_{R_{k}} \cdot \Pi \cdot D_{R_{k}}$

How do we handle general onequery adversaries?

## General one-query adversaries



Def: isometry $V=U \cdot(\operatorname{Id} \otimes|0\rangle)$, i.e. "add ancillas + apply $U$ "

## General one-query adversaries

## $\begin{aligned} n \text { qubit input: }\left|\psi_{h}\right\rangle & \equiv U \\ \text { ancilla: }|0\rangle & \equiv \\ & O_{f}\end{aligned}$

Def: isometry $V=U \cdot(\operatorname{Id} \otimes|0\rangle)$, i.e. "add ancillas + apply $U$ "

$$
\operatorname{Pr}\left[A^{f}\left(\left|\psi_{h}\right\rangle\right) \text { outputs 1] }=\left\langle+_{N}\right| D_{h} \cdot V^{\dagger} \cdot O_{f} \cdot \Pi \cdot O_{f} \cdot V \cdot D_{h}\left|+_{N}\right\rangle\right.
$$

## General one-query adversaries



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Challenge: unclear how to commute $D_{h}$ and $O_{f}$ !

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Our solution: Write $V \cdot D_{h}\left|+_{N}\right\rangle=\widetilde{D_{h}} \mid$ wt $\left.{ }_{V}\right\rangle$ w.r.t. a $V$-dependent unit vector $\left|w_{t}\right\rangle$.

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Our solution: Write $V \cdot D_{h}\left|+_{N}\right\rangle=\widetilde{D_{h}} \mid$ wt $\left.{ }_{V}\right\rangle$ w.r.t. a $V$-dependent unit vector $\left|w t_{V}\right\rangle$. Commute $\widetilde{D_{h}}, O_{f}$ to get spectral relaxation.

## Future directions

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Open problem \#1: prove that our PRS distinguishing game is hard even given $\operatorname{poly}(n)$ queries to an arbitrary $f$.

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## Future directions

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Open problem \#2:
$\left.\begin{array}{c}\left|0^{n}\right\rangle \\ \text { or }\left|1^{n}\right\rangle\end{array}\right\} \begin{aligned} & \text { random } \overline{\bar{E}} \\ & \text { circuit } C\end{aligned}$

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$\left|0^{n}\right\rangle$
or $\left|1^{n}\right\rangle$$\left[\begin{array}{l}\text { 首 } \begin{array}{l}\text { random } \\ \text { circuit } C\end{array} \\ \begin{array}{l}\overline{\bar{E}}\end{array} \begin{array}{l}\text { Task: given description of } C \text { and } \\ 2 n / 3 \text { qubits of } C\left|b^{n}\right\rangle \text {, determine } b . \\ \text { Is this easy given a halting oracle? }\end{array}\end{array}\right.$

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or $\left|1^{n}\right\rangle$$\left[\begin{array}{l}\text { random } \\ \text { circuit } C\end{array}\right] \begin{aligned} & \text { Task: given description of } C \text { and } \\ & 2 n / 3 \text { qubits of } C\left|b^{n}\right\rangle \text {, determine } b . \\ & \text { Is this easy given a halting oracle? }\end{aligned}$
Thanks for listening!

