# 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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Even though this problem is "about" quantum states, the input and output are classical.

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

What can complexity theory say about 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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## 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 that works for every $U$ ?

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## Prior best-known bounds

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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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Main result: There is no efficient one-query oracle algorithm $A^{(\cdot)}$ for the Unitary Synthesis Problem.

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- unlimited space (number of qubits)
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Note: when $\ell=2^{2 n}$, possible to learn description of $U$ in one query.

## Rest of this talk

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Connect unitary synthesis to breaking quantum cryptography
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A special case of our proof (if time)

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Our answer: probably harder than computing any function.

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Main result \#2: Relative to a random oracle $R$, there exists a PRS secure against any efficient oracle adversary $A^{(\cdot)}$ making one query to an arbitrary function $f_{R}$, which can depend on $R$.

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Note: this result implies our unitary synthesis lower bound.

## Our PRS construction

For any function $h:[N] \rightarrow\{ \pm 1\}$, define the corresponding binary phase state $\left|\psi_{h}\right\rangle:=\frac{1}{\sqrt{N}} \sum_{x \in[N]} h(x)|x\rangle$. (recall $N=2^{n}$ )

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PRS construction: given random oracle $R:[K] \times[N] \rightarrow\{ \pm 1\}$, our PRS family is $\left\{\left|\psi_{R_{k}}\right\rangle\right\}_{k \in[K]}$ where $R_{k}(x):=R(k, x)$.

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given 1 query to a function $f$, which can depend on $R$.

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

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## We show:

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- We bound this norm by appealing to matrix concentration.

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## A special class of one-query adversaries

Assume adversary sets $\ell=n$ (no ancillas) and $U=\mathrm{Id}$.

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Assume adversary sets $\ell=n$ (no ancillas) and $U=$ Id.
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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\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.) }
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Matrix Chernoff bound: If $X$ is a random $L \times L$ matrix with bounded operator norm, then for i.i.d. $X_{1}, \ldots, X_{K}$

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\left\|\frac{1}{K} \sum_{k} X_{k}-\mathbb{E}[X]\right\|_{\mathrm{op}} \approx 0\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:

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

Adversary's advantage (for this special class):

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

Matrix Chernoff:

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

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$$
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$$

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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$$
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$$

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

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

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) $\cdot\left|v_{f}\right\rangle$
$\overbrace{\left\langle\psi_{R_{k}}\right| O_{f} \cdot \Pi \cdot O_{f} \mid \psi_{R_{k}}}\rangle$

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}
$$

We'll rewrite this as $\left\langle v_{f}\right| \cdot\left(\right.$ random matrix) $\cdot\left|v_{f}\right\rangle$

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

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

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

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

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So we can rewrite the distinguishing advantage as

$$
\left\langle+_{N}\right| O_{f}\left(\frac{1}{K} \sum_{k} D_{R_{k}} \cdot \Pi \cdot D_{R_{k}}-\mathbb{E}_{h}\left[D_{h} \cdot \Pi \cdot D_{h}\right]\right) O_{f}\left|+_{N}\right\rangle
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\begin{aligned}
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& \leq\left\|\frac{1}{K} \sum_{k} D_{R_{k}} \cdot \Pi \cdot D_{R_{k}}-\mathbb{E}_{h}\left[D_{h} \cdot \Pi \cdot D_{h}\right]\right\|_{\mathrm{op}}
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& \leq\left\|\frac{1}{K} \sum_{k} D_{R_{k}} \cdot \Pi \cdot D_{R_{k}}-\mathbb{E}_{h}\left[D_{h} \cdot \Pi \cdot D_{h}\right]\right\|_{\text {op }} \approx O\left(\sqrt{\frac{n}{K}}\right) \\
& \quad \\
& \quad \begin{array}{l}
\text { by Matrix Chernoff on the i.i.d. bounded } \\
\text { random matrices } D_{R_{k}} \cdot \Pi \cdot D_{R_{k}} .
\end{array}
\end{aligned}
$$

Extending this proof to general one-query adversaries requires more care.

## General one-query adversaries



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

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

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

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$$

Challenge: unclear how to commute $D_{h}$ and $O_{f}$ !
Our solution: factor $V\left|\psi_{h}\right\rangle=\widetilde{D_{h}} \cdot\left|w t_{V}\right\rangle$ w.r.t. a $V$-dependent unit vector $\left|\mathrm{wt}_{V}\right\rangle$ to obtain spectral relaxation.

## Conclusions

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Non-synthesis conjecture: our PRS distinguishing game is hard for any efficient oracle adversary $A^{f}$ that makes poly $(n)$ queries to $f$.


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## Thanks for listening!

