

How to Construct Random Unitaries

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Berkeley → NYU

joint work with Hsin-Yuan Huang

Haar measure: uniform distribution on unitaries

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Property: for any unitary W , if $U \sim \text{Haar}$, $W \cdot U \sim \text{Haar}$

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Haar-random unitaries show up everywhere:

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Haar-random unitaries show up everywhere:

black hole
information
scrambling

entanglement

quantum learning
algorithms

...

quantum
crypto

random
quantum
circuits

unitary
complexity

quantum
error
correction

Challenge:

Haar-random unitaries are exponentially complex

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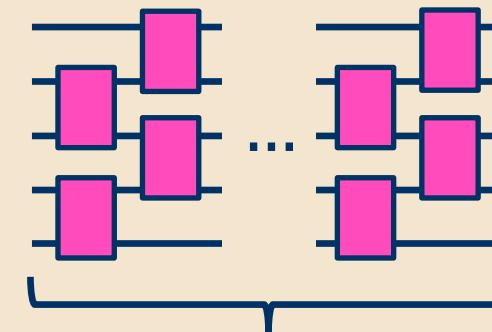


Challenge:

Haar-random unitaries are exponentially complex



minimal circuit for U

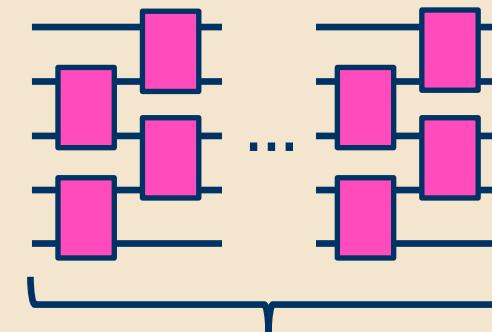


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This makes them impractical for most applications!

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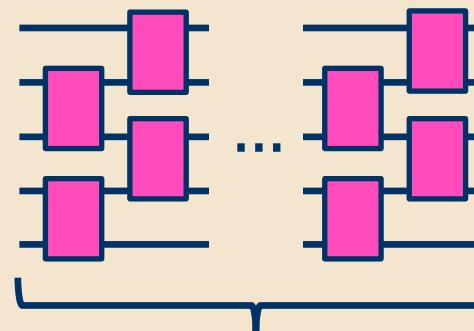
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$\exp(n)$

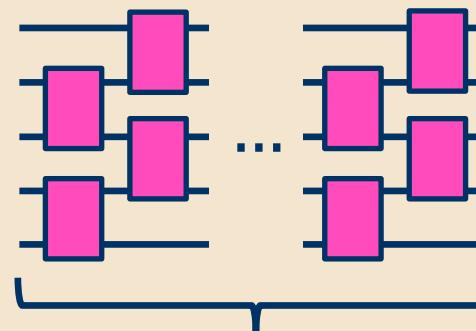
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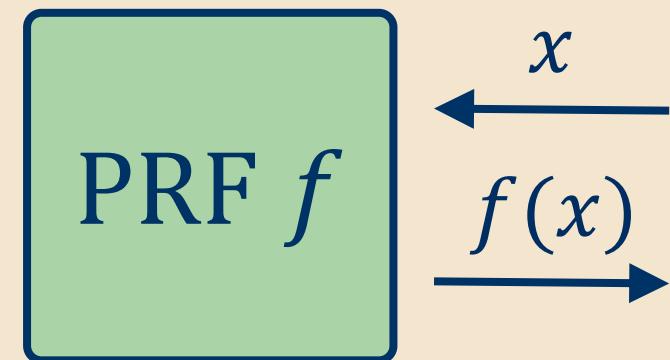


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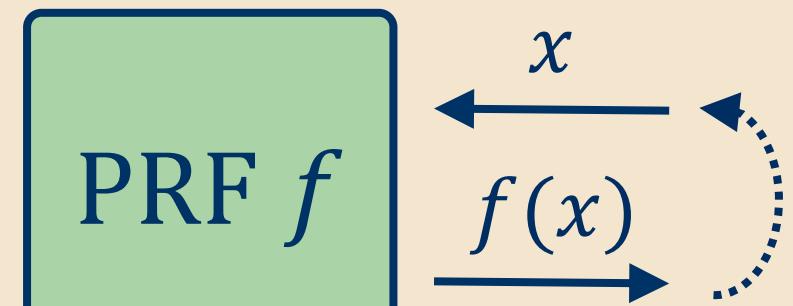


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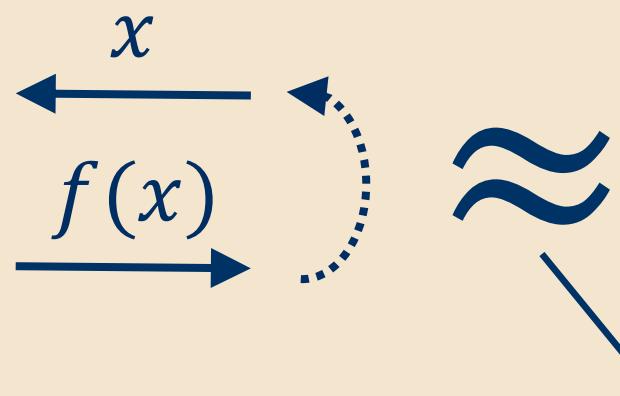


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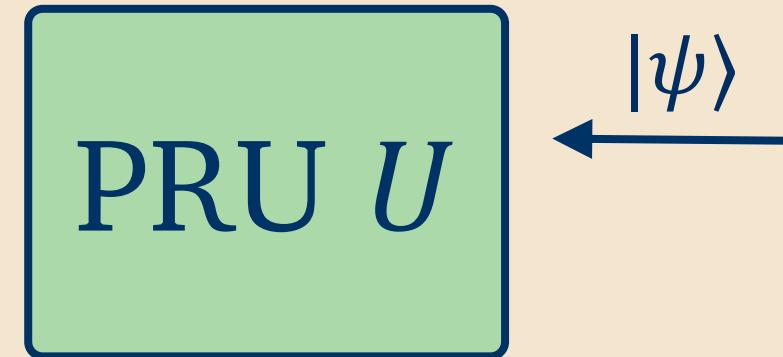
No efficient algorithm
can tell the difference!

Pseudorandom unitaries (PRUs) [JLS18]

efficiently-computable unitaries that appear Haar-random

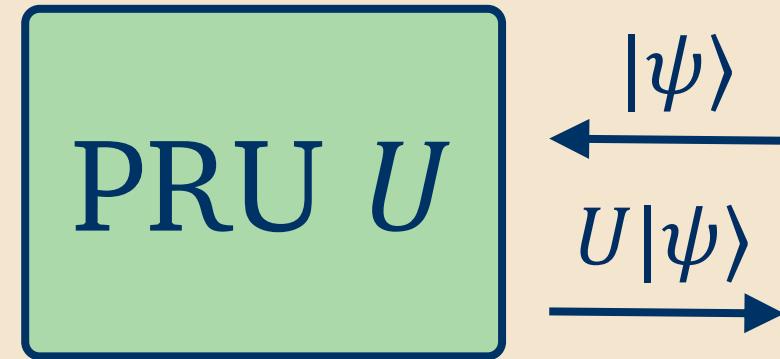
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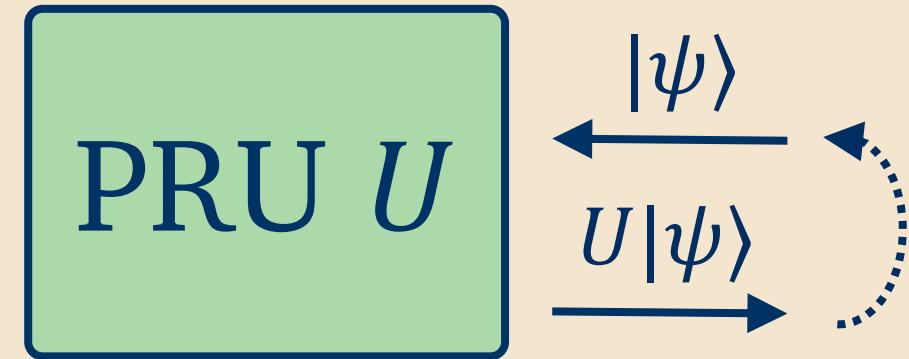
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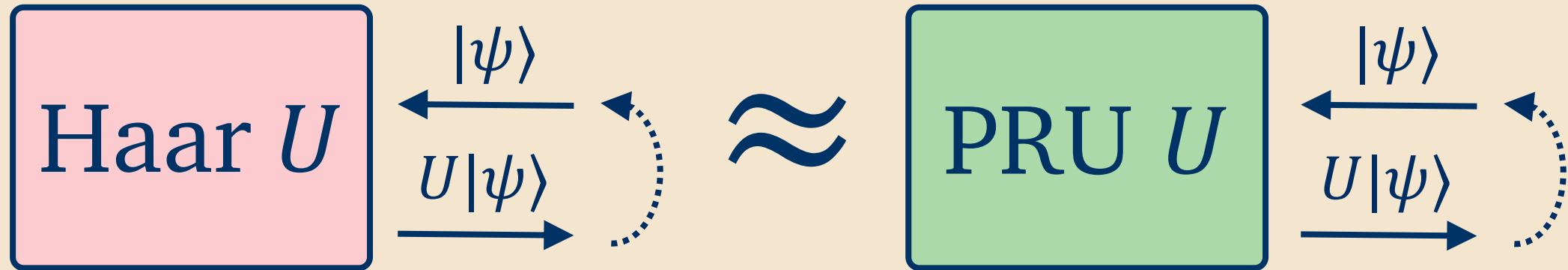
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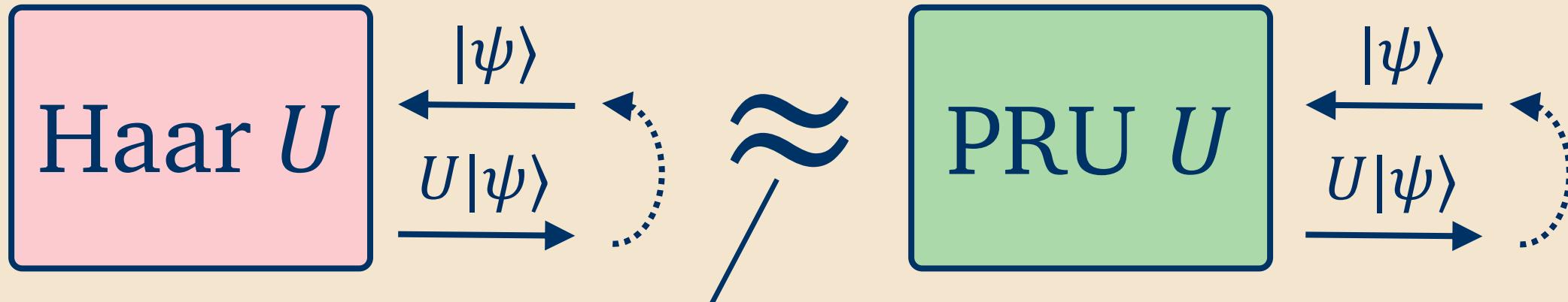
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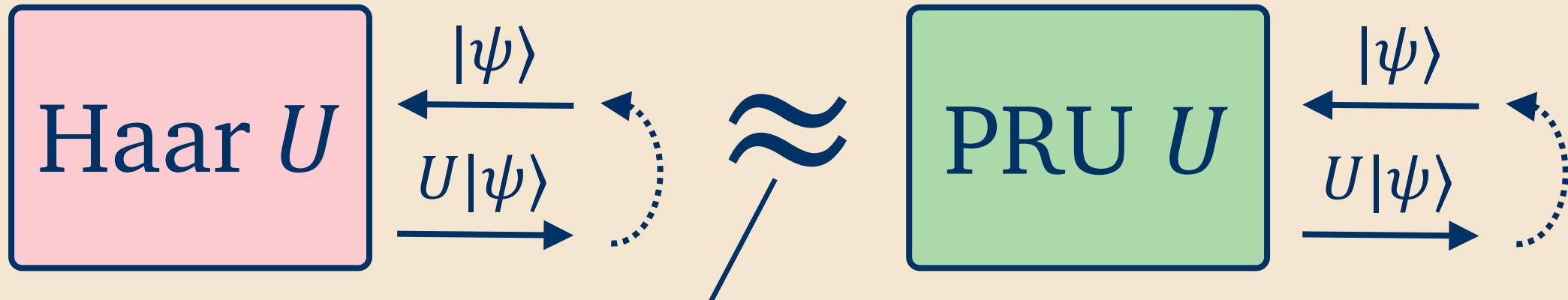
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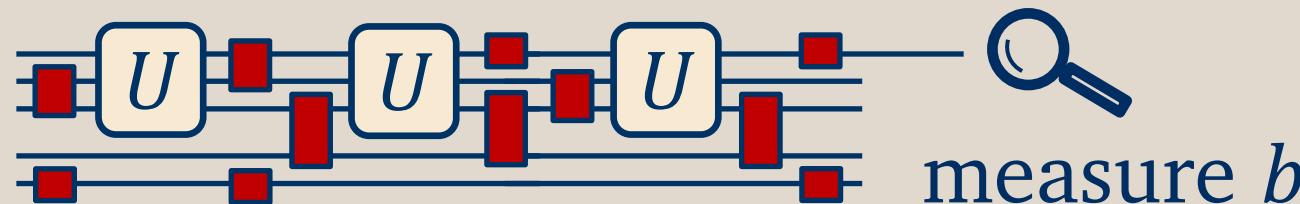
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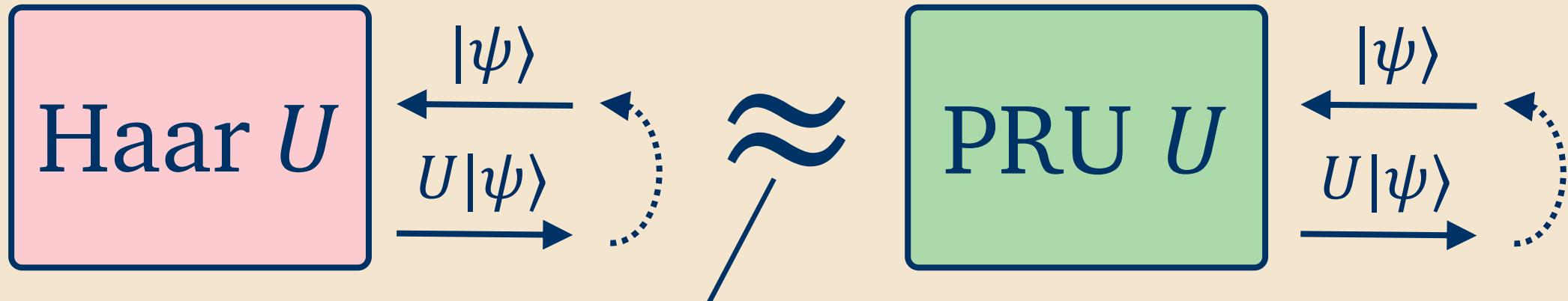


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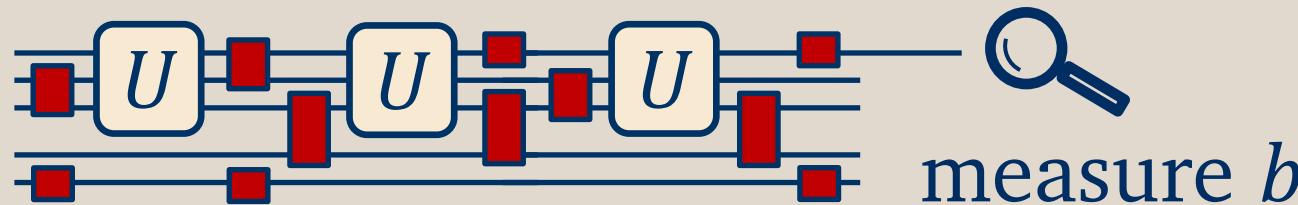


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For any efficient algorithm A :



$$\Pr[b = 1 \mid U \leftarrow \text{Haar}] \approx \Pr[b = 1 \mid U \leftarrow \text{PRU}]$$

Application: modeling black hole dynamics

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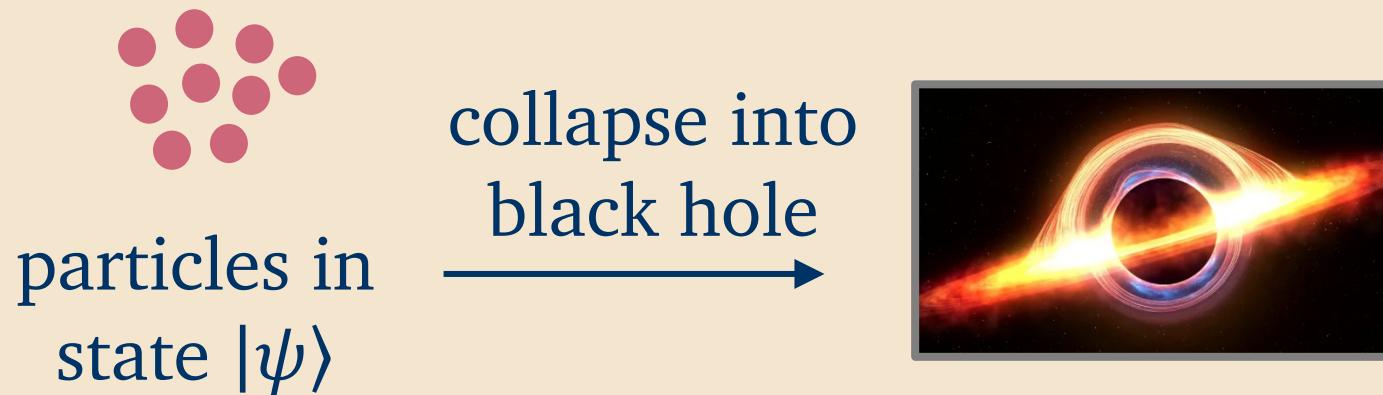
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particles in
state $|\psi\rangle$

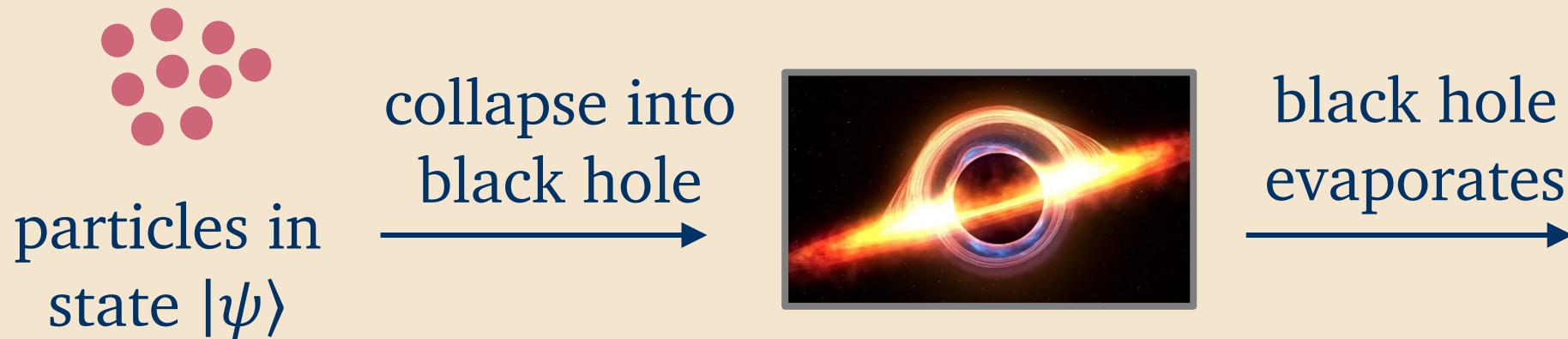
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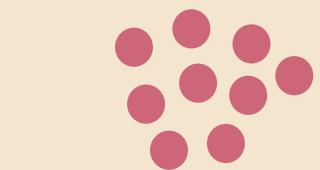
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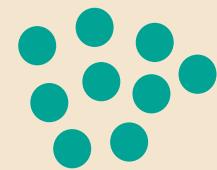


particles in
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collapse into
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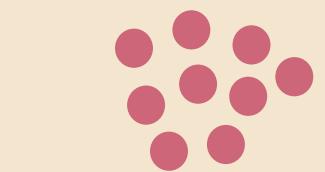
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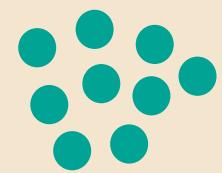


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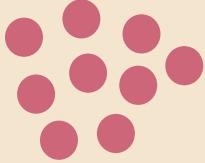


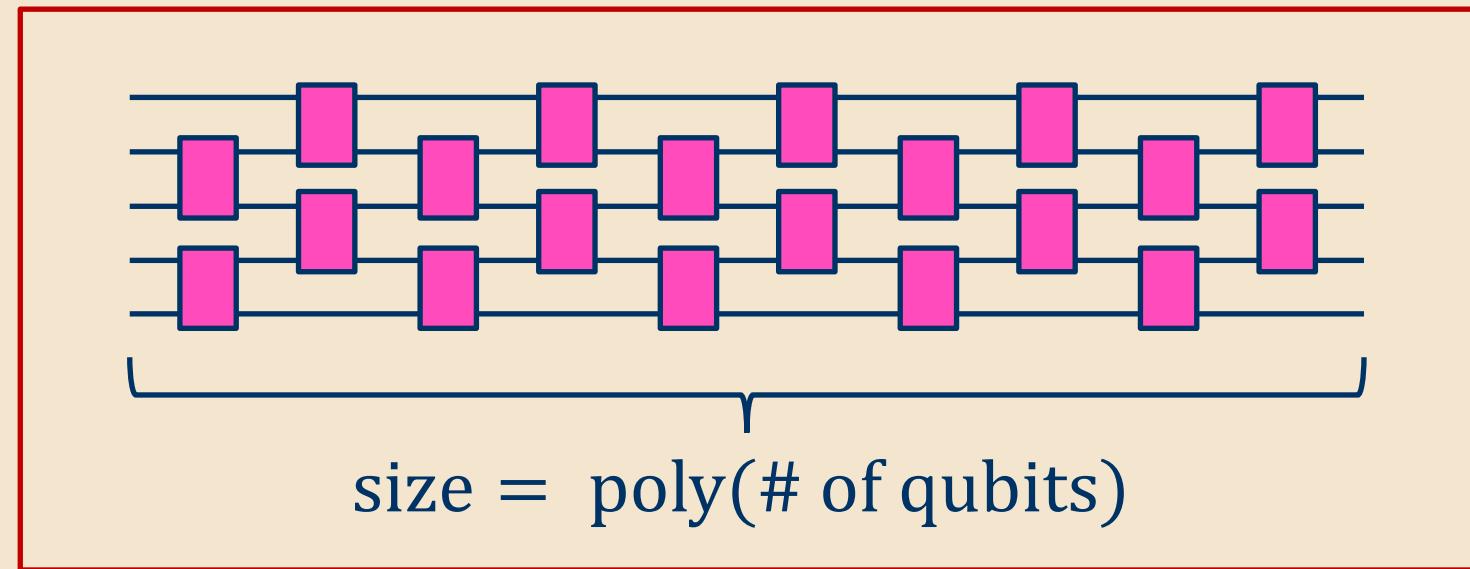
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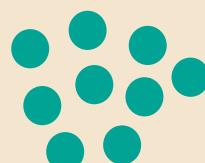
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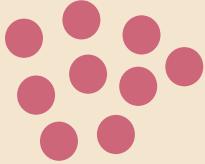


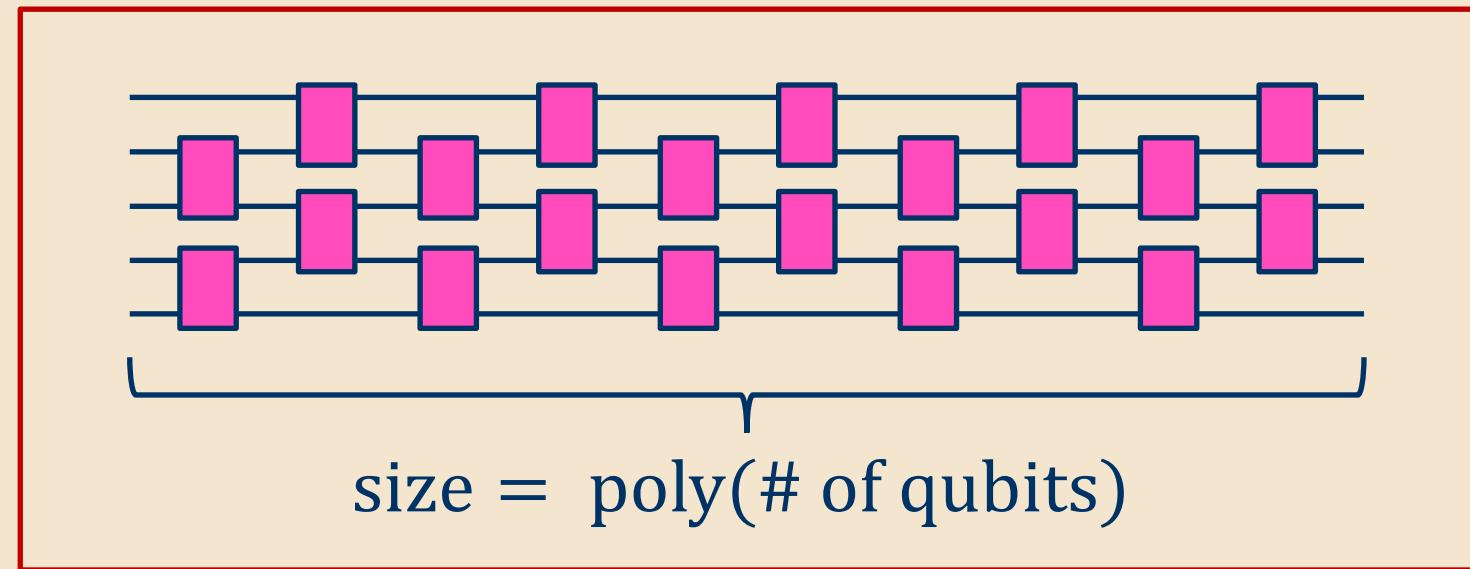

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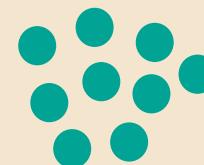
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Consequence: many physics results now rely on the assumption
that various physical processes are PRUs [KP23, YE23, EFLVY24]

But do PRUs exist?

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This was left as an open problem by [JLS18].

Prior work

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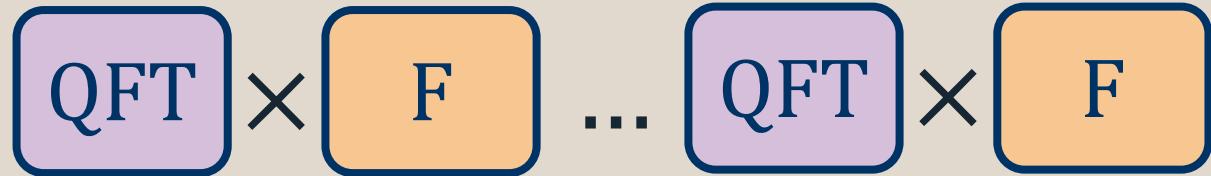
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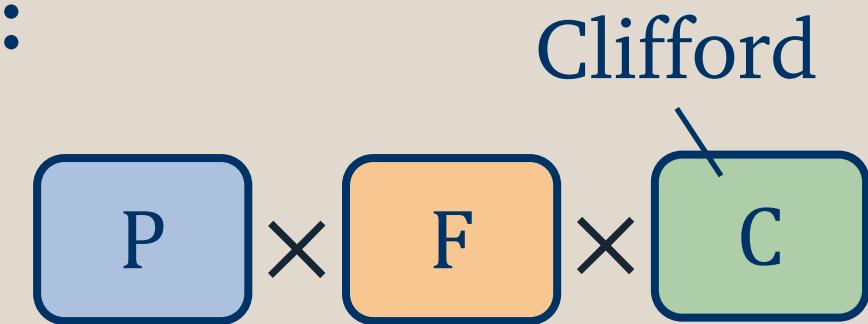
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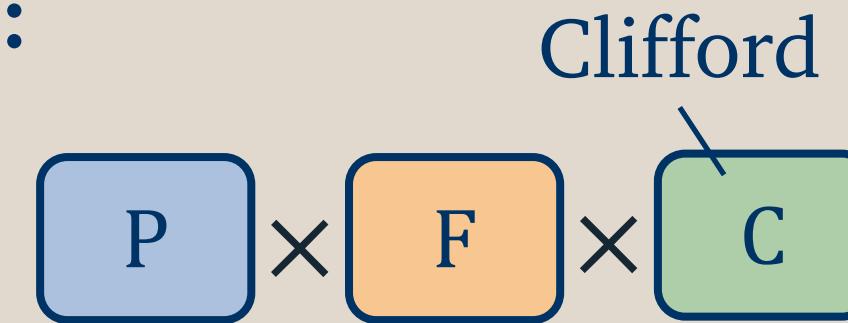
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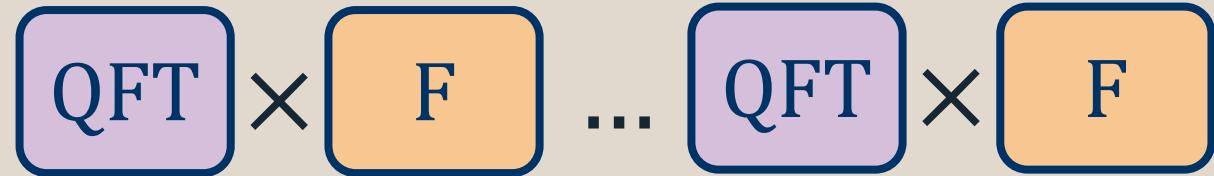


Clifford
[MPSY24]

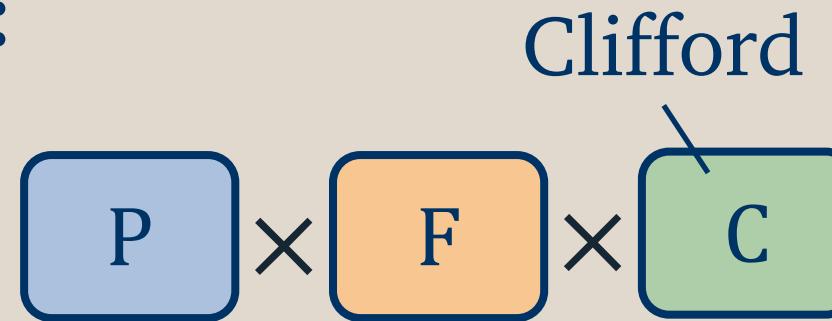
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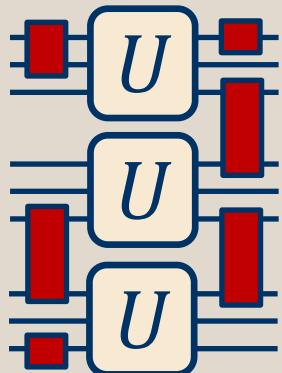
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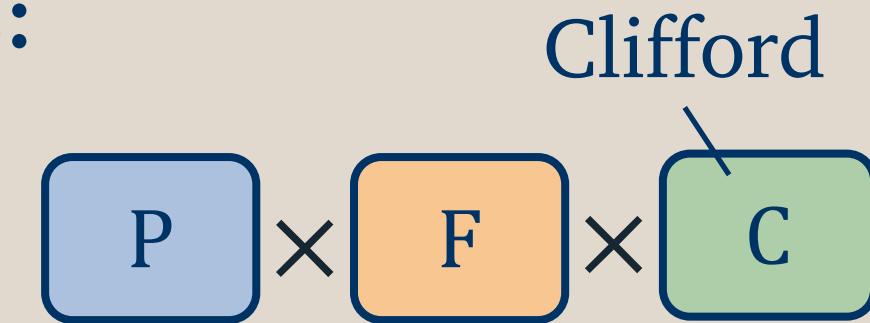


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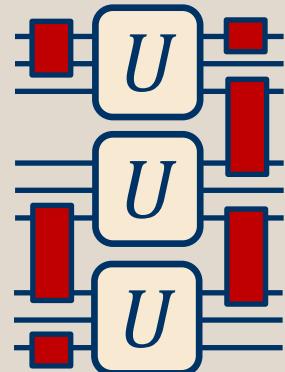
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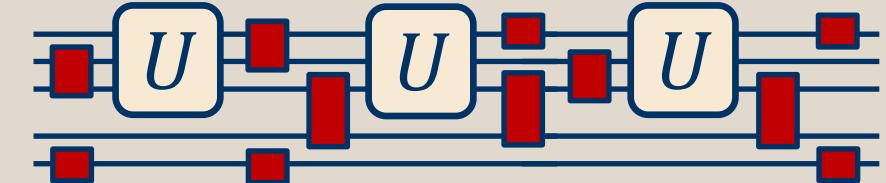
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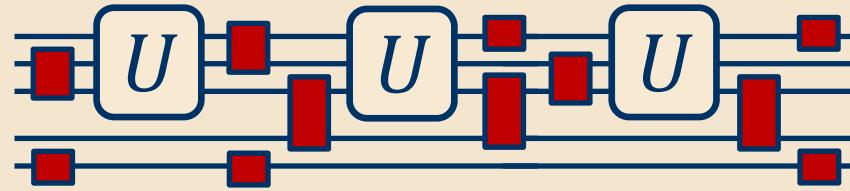
but not
this:



Why has it been hard to prove PRUs exist?

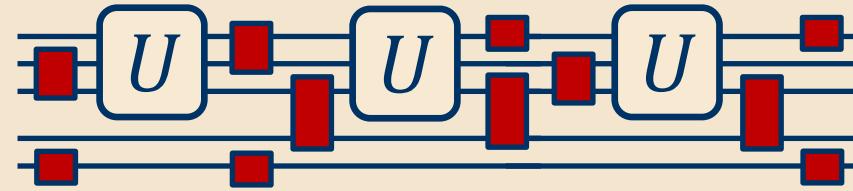
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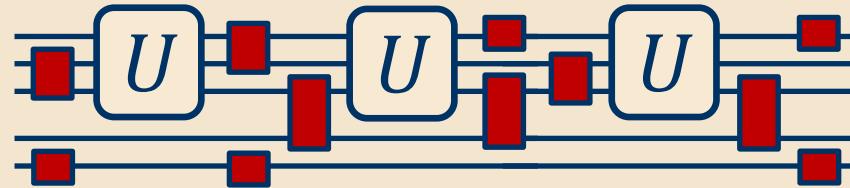
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2) Mathematics of random unitaries is complicated.

- Weingarten calculus
- free probability
- ???

Theorem 3.1. Let k be a positive integer. For any permutation $\sigma \in \mathcal{S}_k$ and nonnegative integer g , we have

$$(k-1)^g \#P(\sigma, |\sigma|) \leq \#P(\sigma, |\sigma| + 2g) \leq (6k^{7/2})^g \#P(\sigma, |\sigma|).$$

Theorem 3.2. For any $\sigma \in \mathcal{S}_k$ and $d > \sqrt{6}k^{7/4}$,

$$\frac{1}{1 - \frac{k-1}{d^2}} \leq \frac{(-1)^{|\sigma|} d^{k+|\sigma|} Wg^U(\sigma, d)}{\#P(\sigma, |\sigma|)} \leq \frac{1}{1 - \frac{6k^{7/2}}{d^2}}.$$

In addition, the l.h.s inequality is valid for any $d \geq k$.

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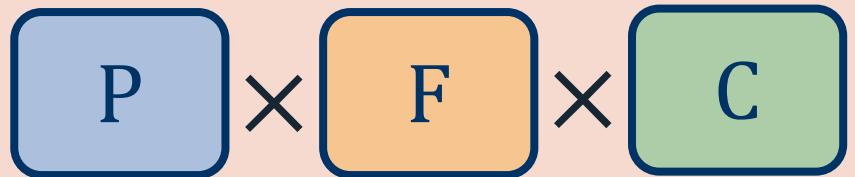


New technique: the path-recording oracle

- efficient simulation of Haar-random unitaries

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New technique: the path-recording oracle

- efficient simulation of Haar-random unitaries
- only uses basic quantum info (purification)

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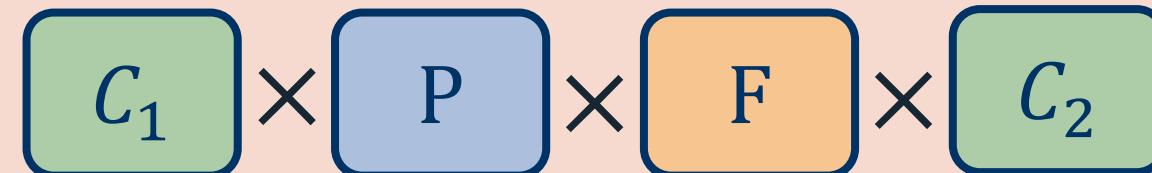
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But for this talk, I’ll focus on the weakest notion.

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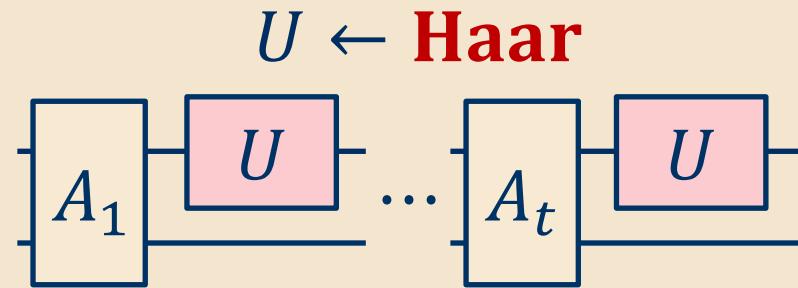
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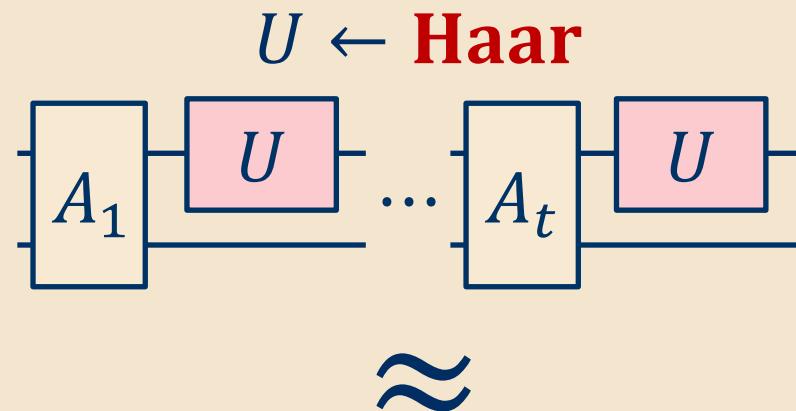
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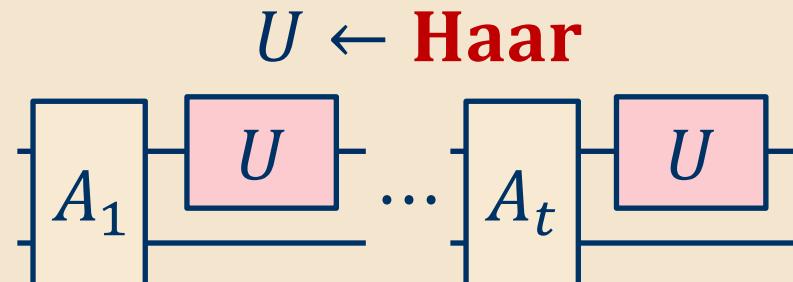
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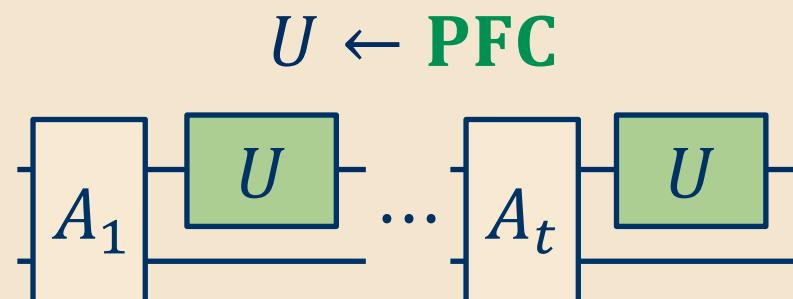
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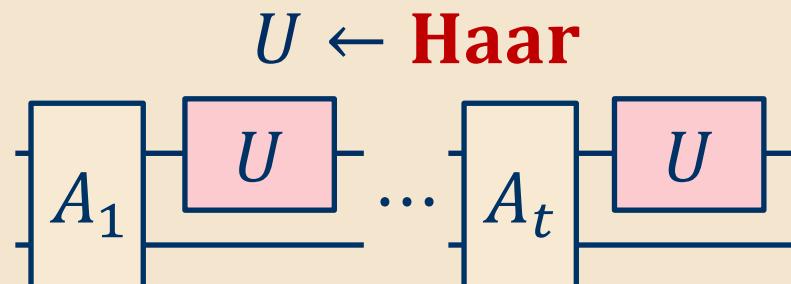


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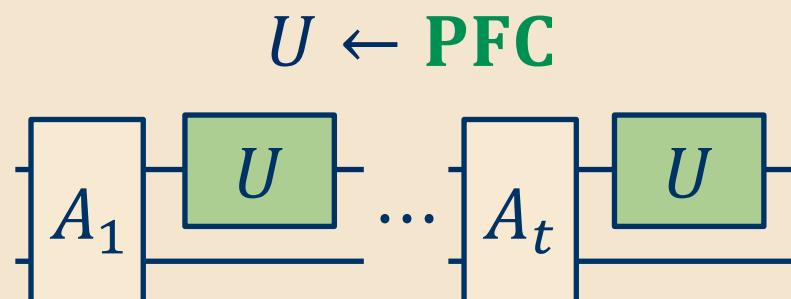
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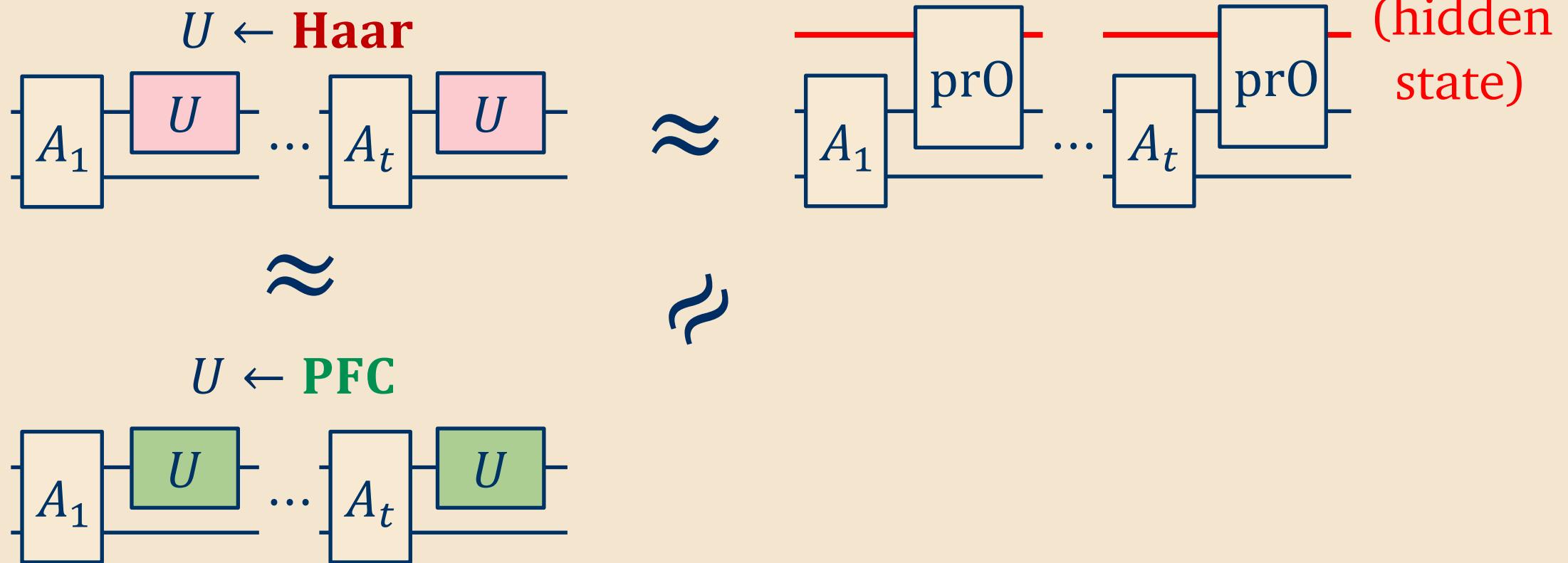


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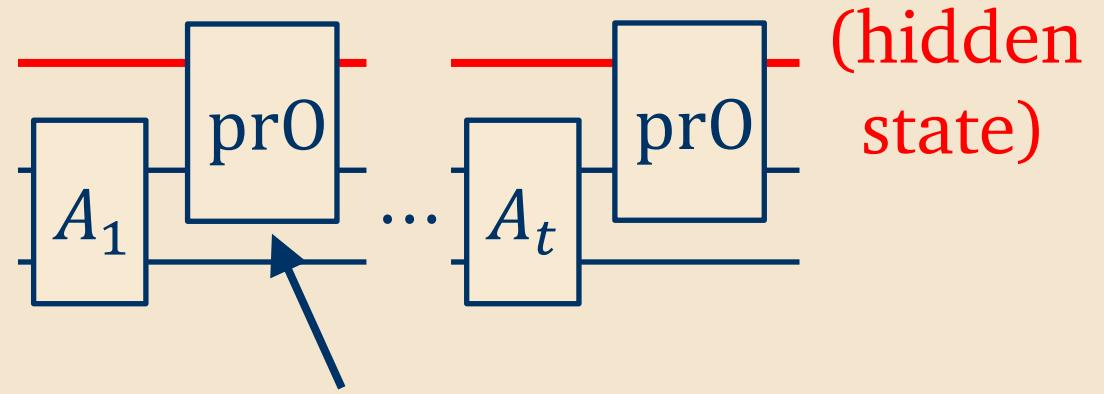
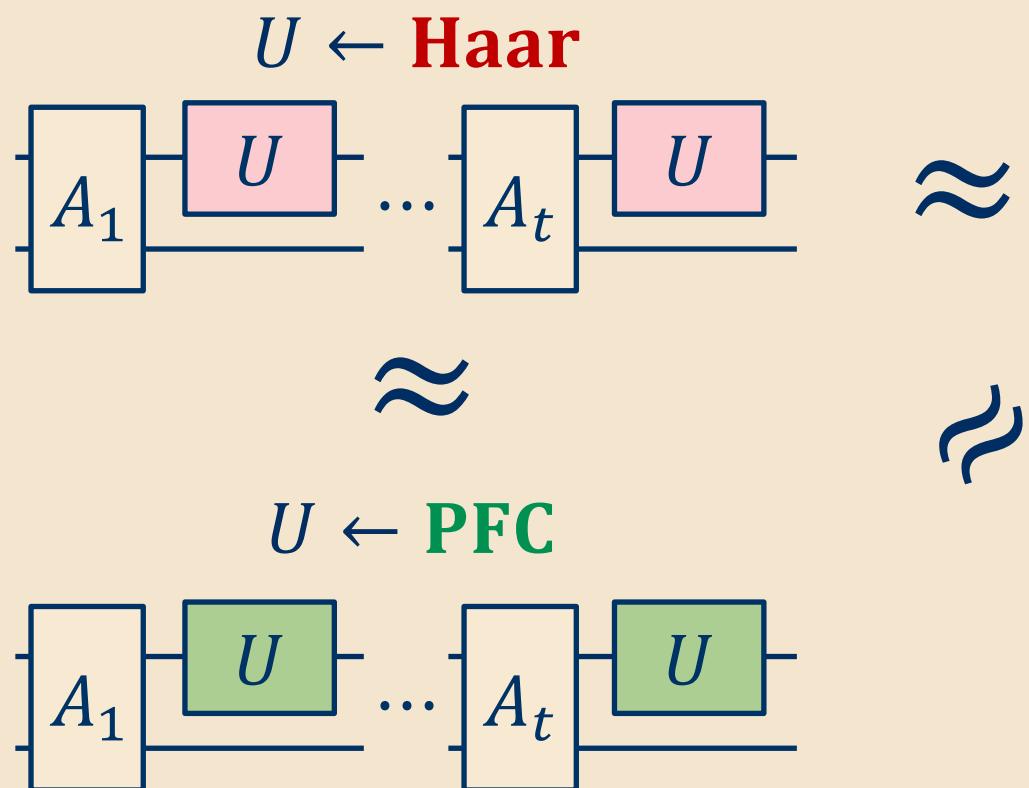


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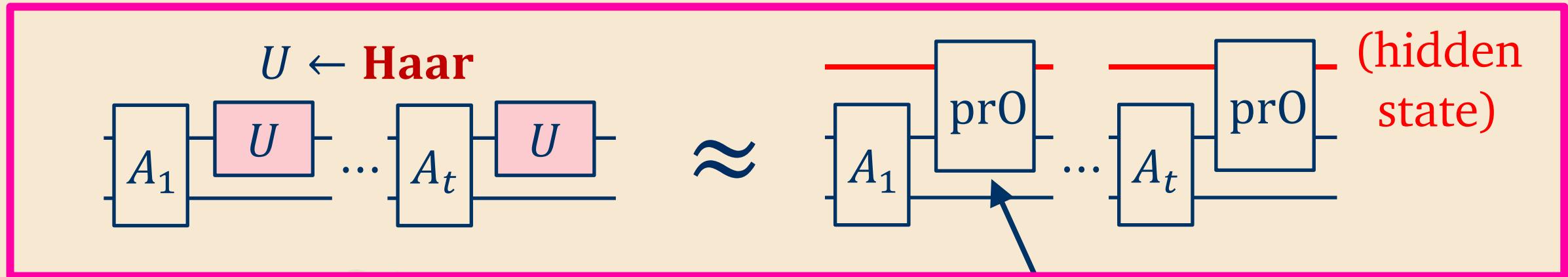
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most of the proof



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Rest of this talk

- **Lazy sampling of a random function**
- Lazy sampling of a random unitary
- Proving correctness + PRUs exist
- Applications

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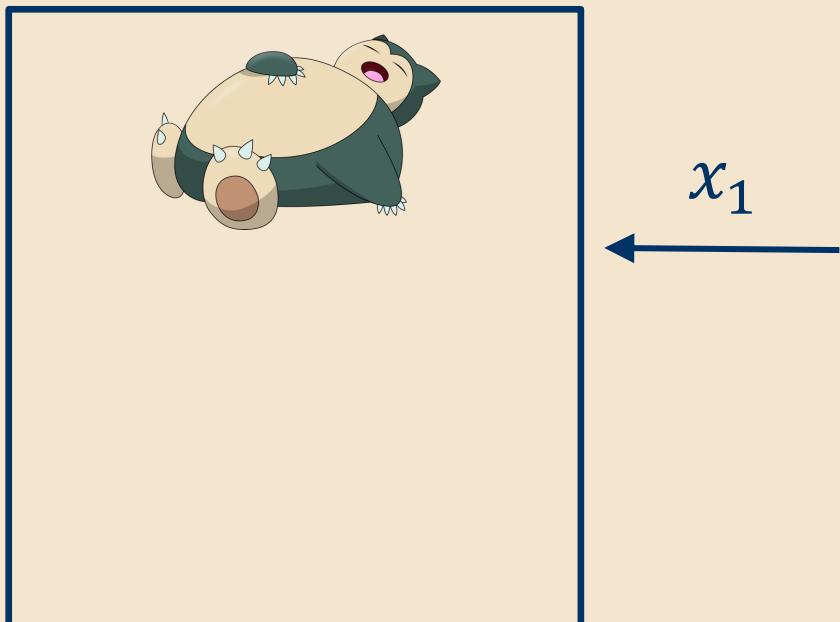
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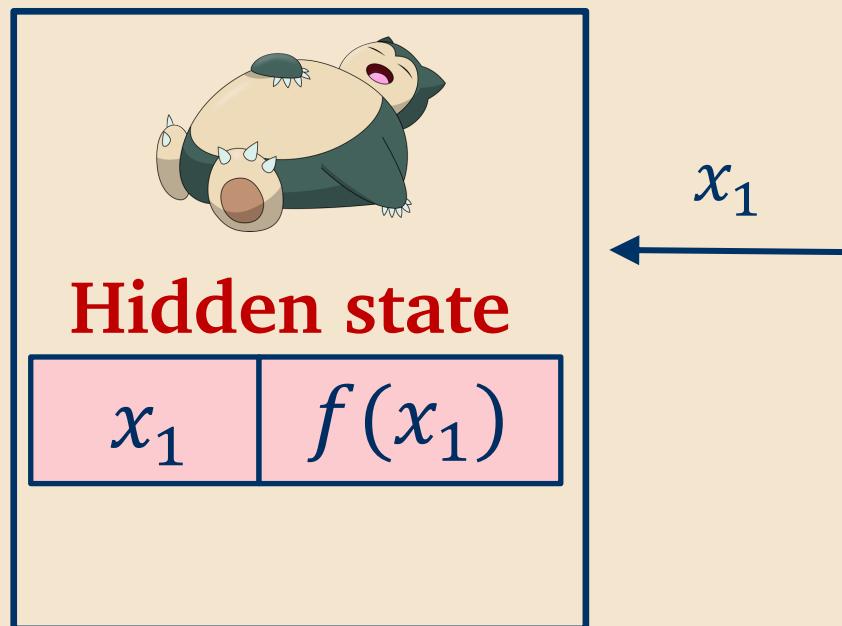


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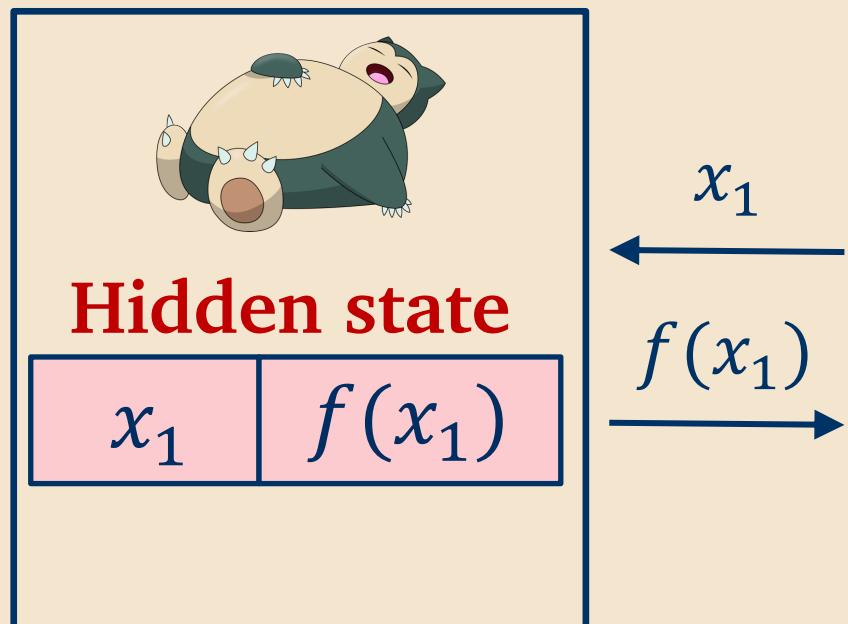


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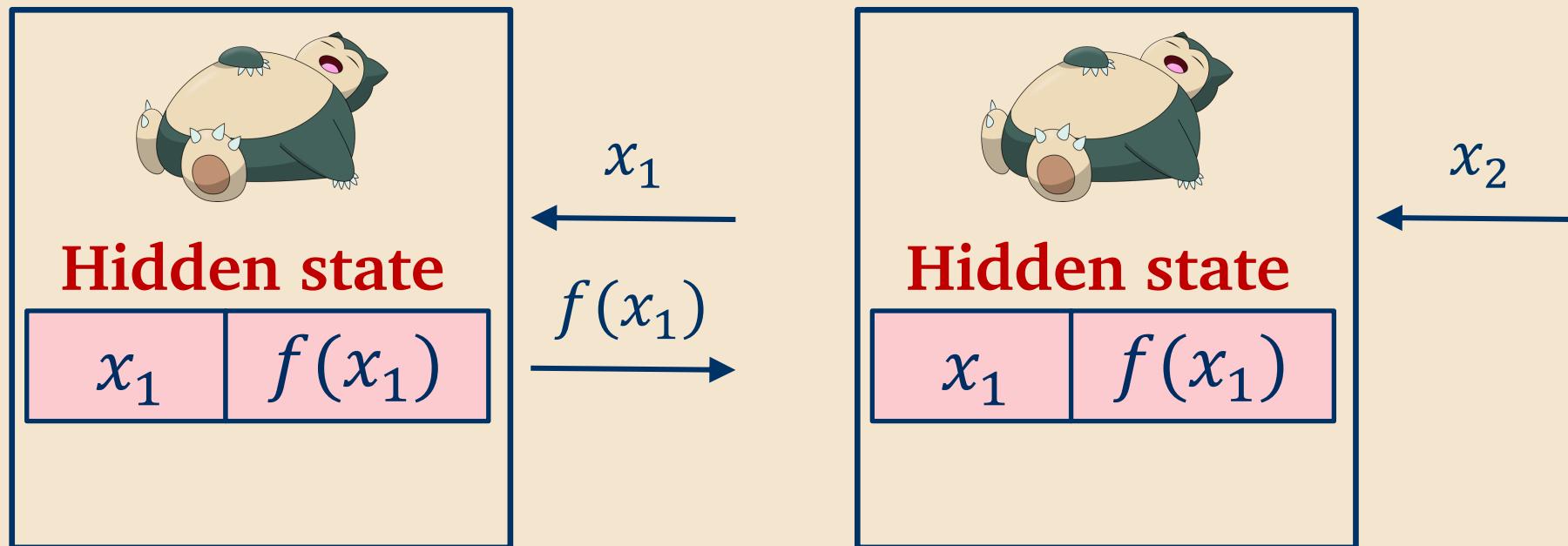


Lazy sampling of a random function

Goal: efficiently implement an algorithm that queries a random **function** f .

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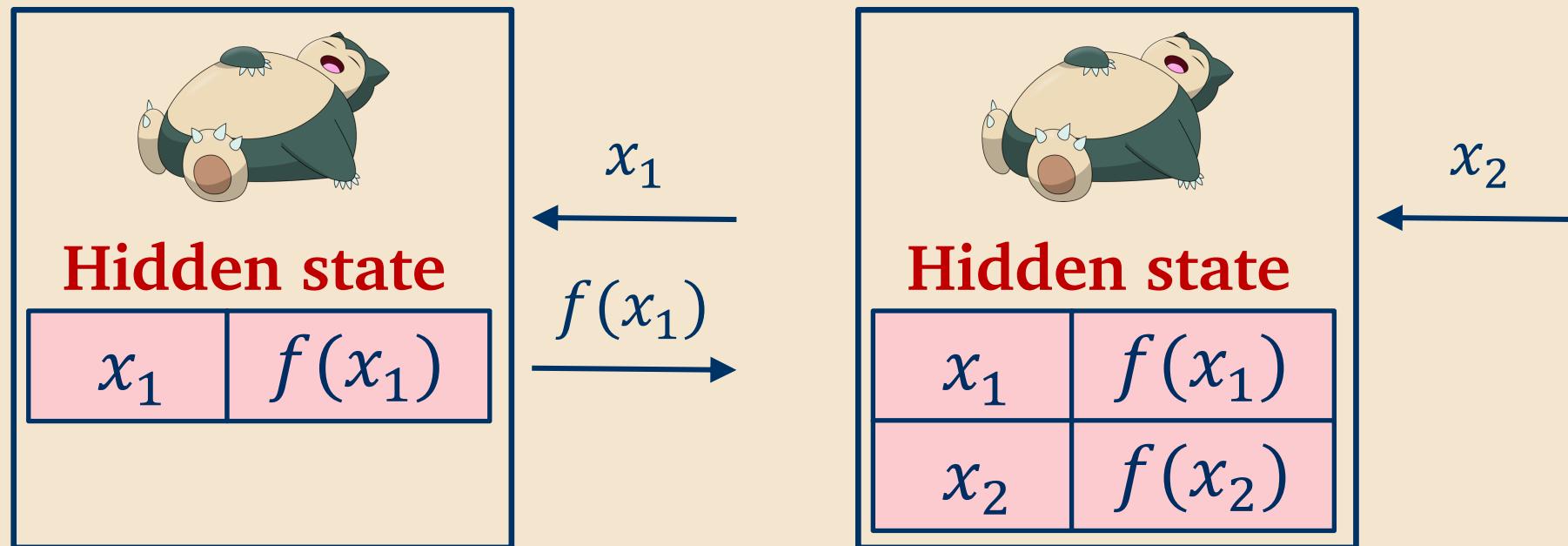


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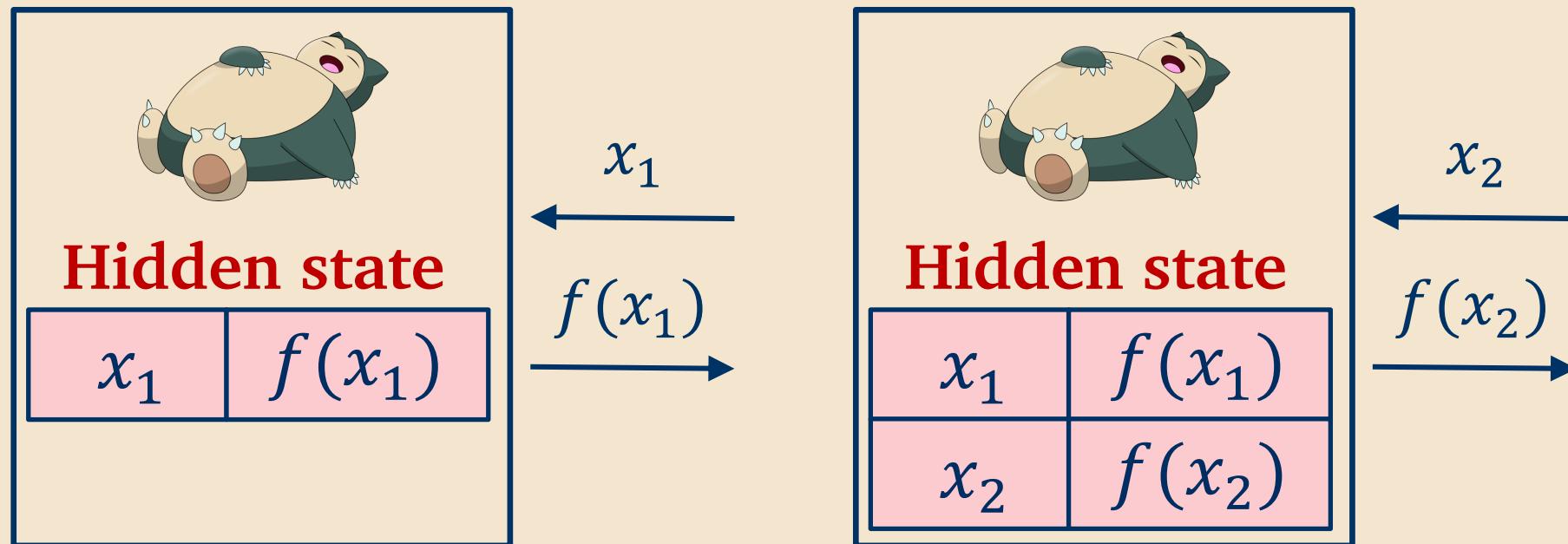


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Rest of this talk

- Lazy sampling of a random function
- **Lazy sampling of a random unitary**
- Proving correctness + PRUs exist
- Applications

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We use **entanglement** with a **hidden data structure** that succinctly “remembers” enough information to spoof a Haar-random U .



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Inspiration: compressed oracle technique [Zhandry19]

Up next:
“Derive” the path-recording oracle
through simple examples

Example 1: one query on $|0\rangle$

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The algorithm: $|0\rangle_A \xrightarrow{U} U|0\rangle_A$
 $(U \leftarrow \text{Haar})$

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The algorithm: $|0\rangle_A \xrightarrow{U} U|0\rangle_A$ Fact: this is the “maximally mixed” state
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How to “spoof” it:



$$\sum_y |y\rangle_A |y\rangle_S \quad (\text{S register is hidden})$$

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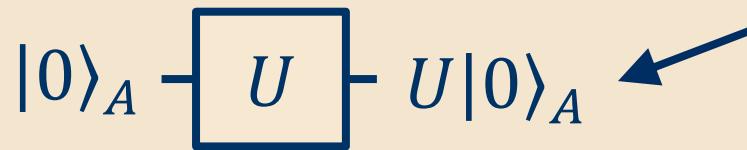
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Idea 1: entanglement with a **hidden register S** can simulate
one query to U .

Example 2: two queries on $|0\rangle$

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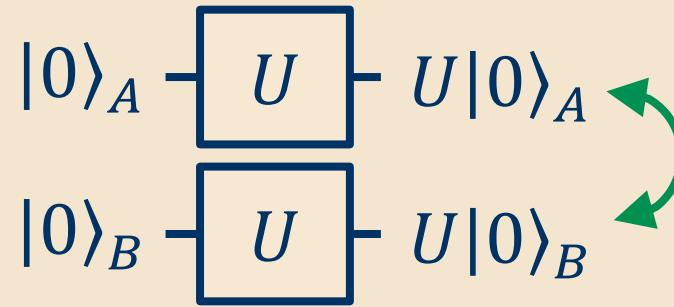
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$$|0\rangle_A \xrightarrow{U} U|0\rangle_A$$
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Example 2: two queries on $|0\rangle$

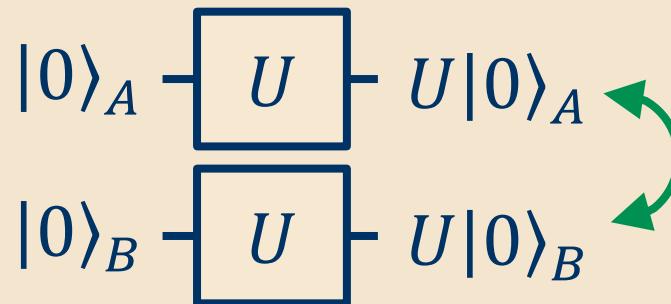
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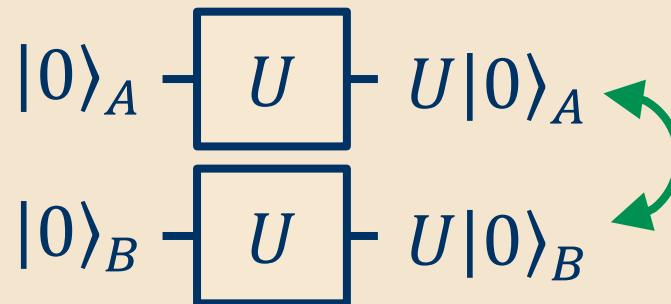


$$\sum_{y_1, y_2} |y_1\rangle_A |y_2\rangle_B |\{y_1, y_2\}\rangle_S$$

(S register is hidden)

Example 2: two queries on $|0\rangle$

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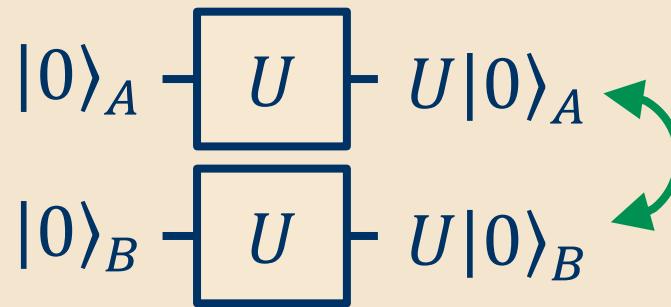


also symmetric!
$$\sum_{y_1, y_2} |y_1\rangle_A |y_2\rangle_B |\{y_1, y_2\}\rangle_S$$

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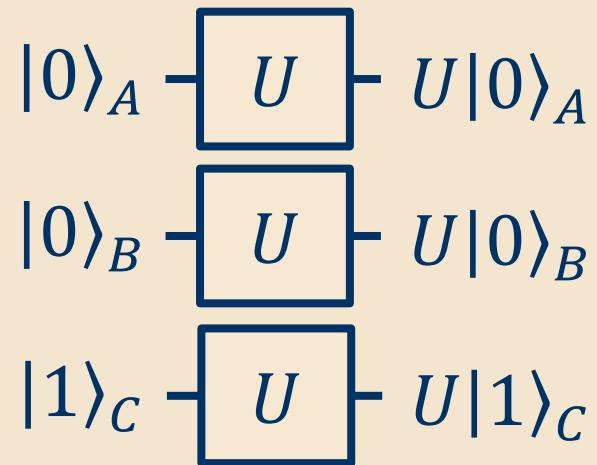
Idea 2: use an **unordered set** to spoof “swap-symmetry”.

Example 3: mixed queries

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The algorithm:

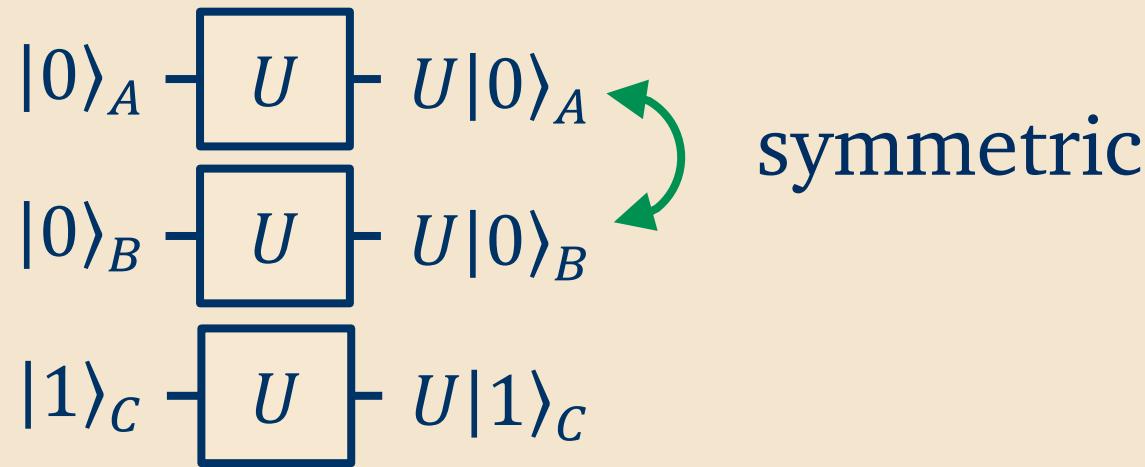
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Example 3: mixed queries

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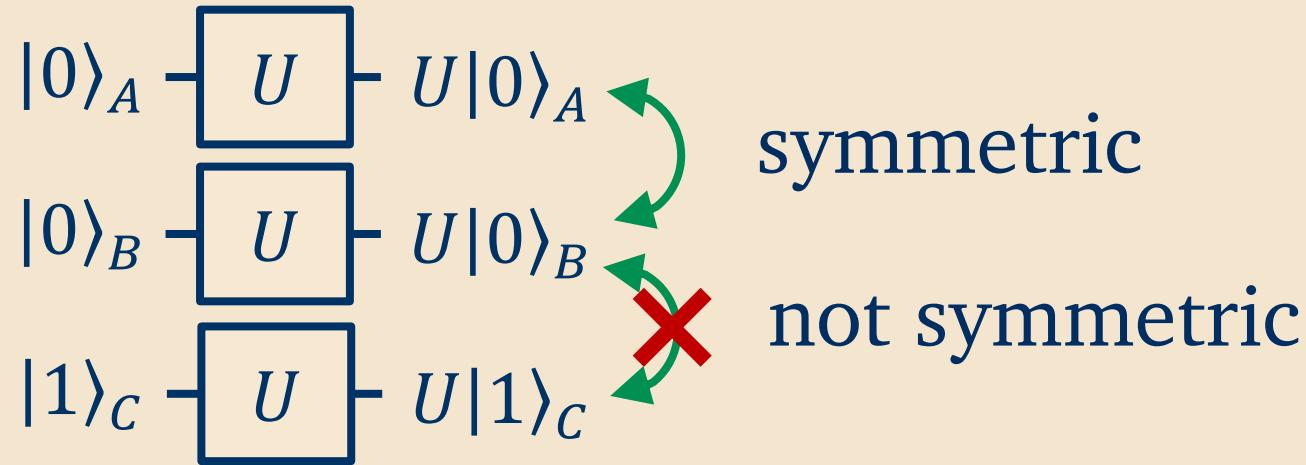


symmetric

Example 3: mixed queries

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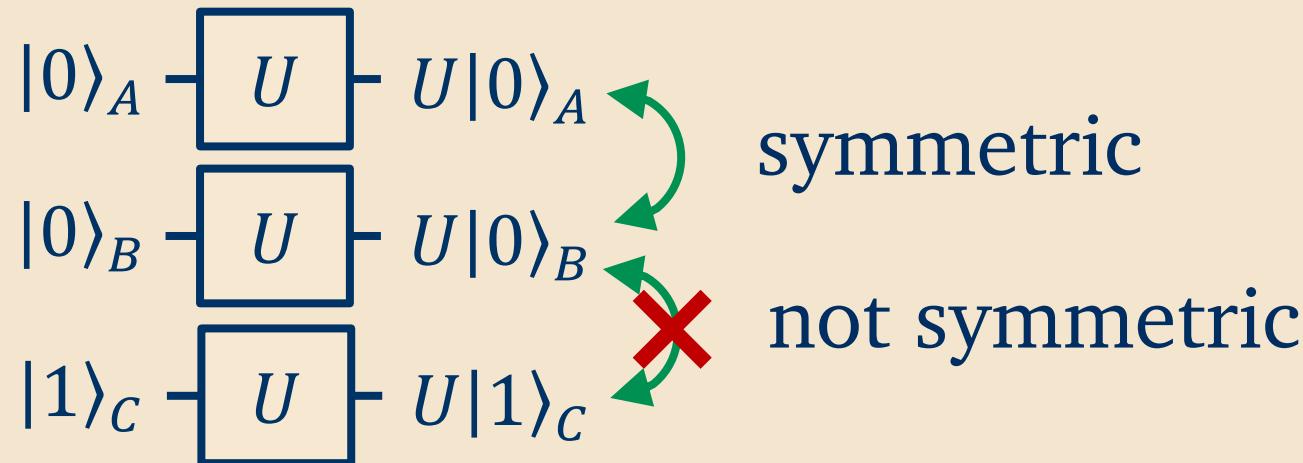
symmetric

not symmetric

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How to “spoof” it:

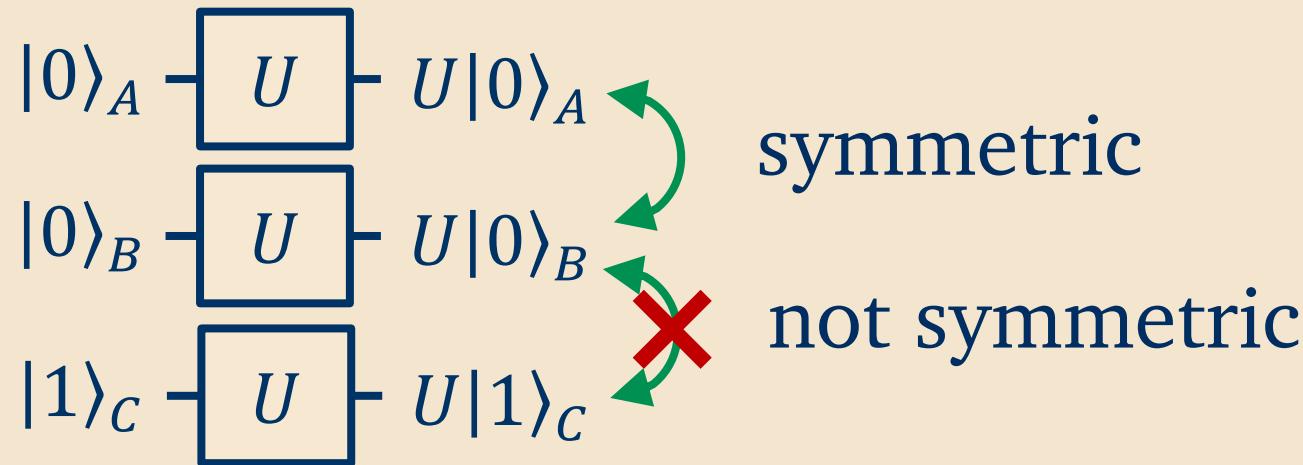


$$\sum_{y_1, y_2, y_3} |y_1\rangle_A |y_2\rangle_B |y_3\rangle_C |\{(0, y_1), (0, y_2), (1, y_3)\}\rangle_S$$

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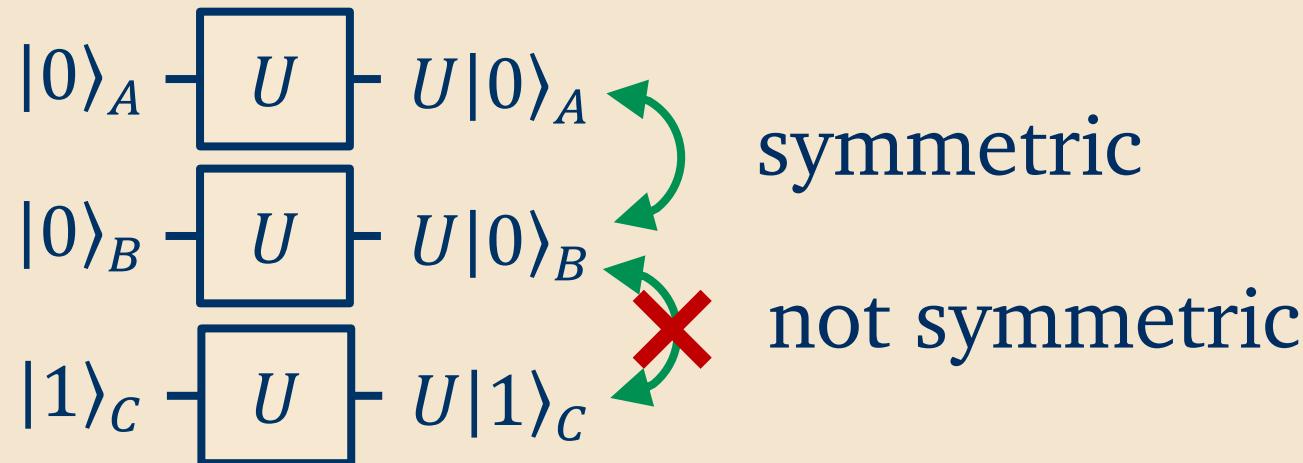


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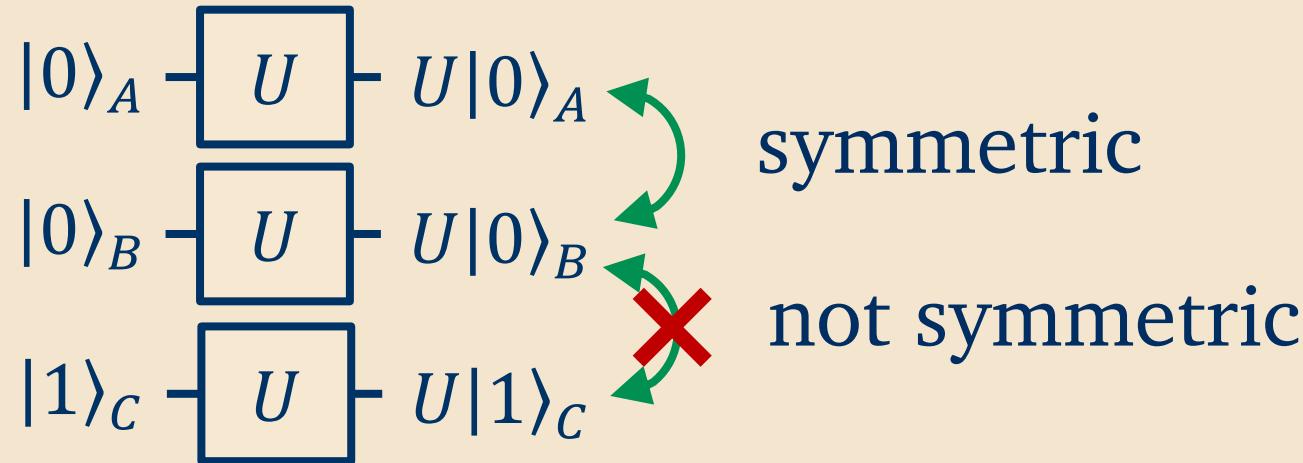


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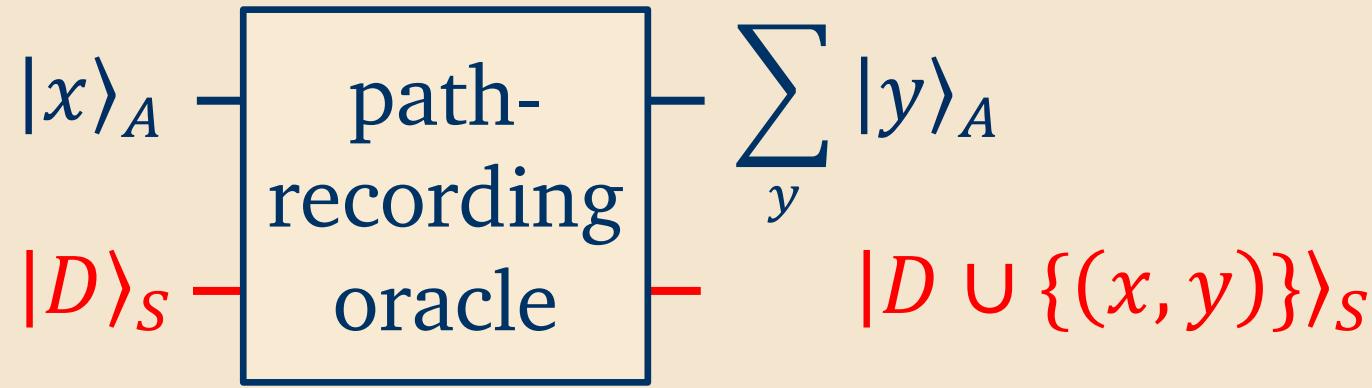


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Idea 3: use ordered pairs to simulate symmetry “structure”

We can generate all of these examples by simply replacing each query to U with this:

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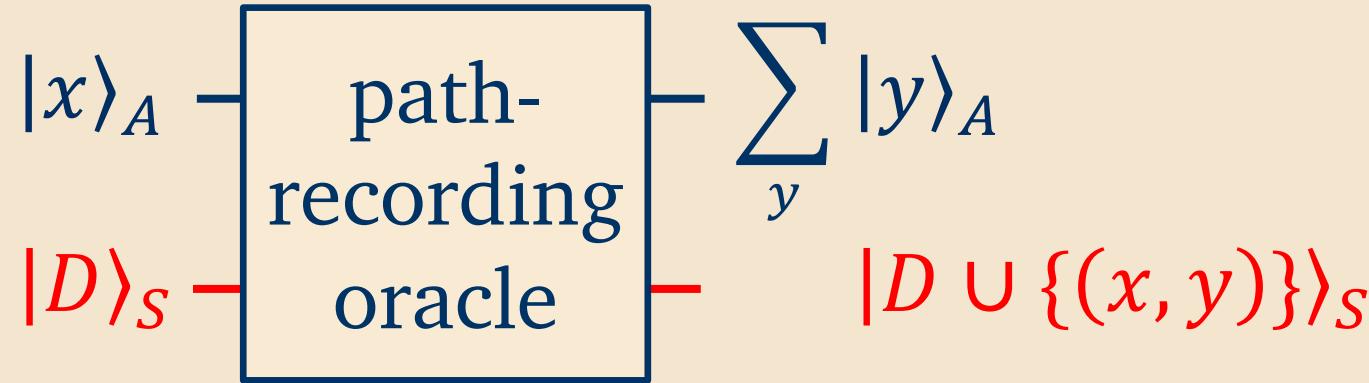
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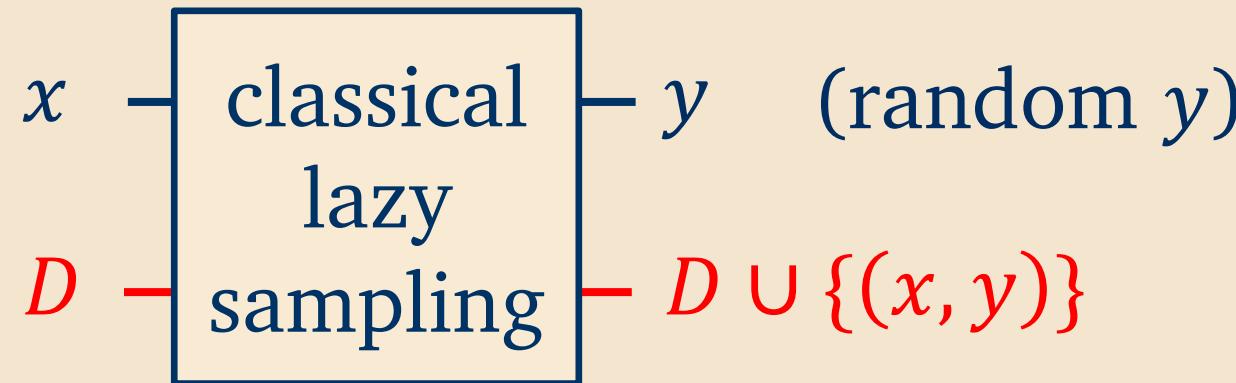
$$\begin{array}{c} |x\rangle_A \\ |D\rangle_S \end{array} \xrightarrow{\text{path-} \\ \text{recording} \\ \text{oracle}} \sum_y |y\rangle_A \quad |D \cup \{(x, y)\}\rangle_S$$

Note the similarity to classical lazy sampling:

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Note the similarity to classical lazy sampling:



Rest of this talk

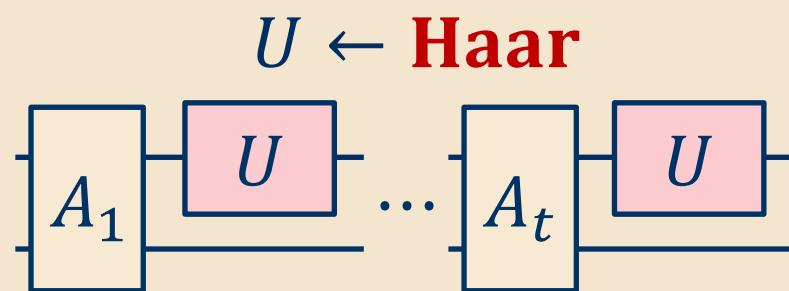
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Recall our cartoon proof overview

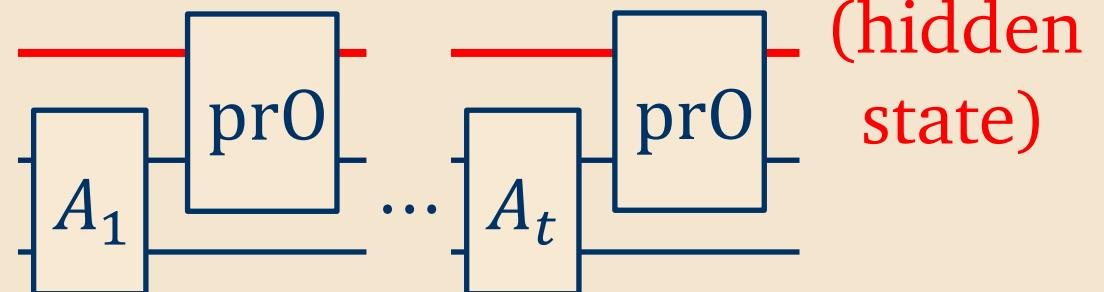
Want to show:

For all efficient adversaries A ,

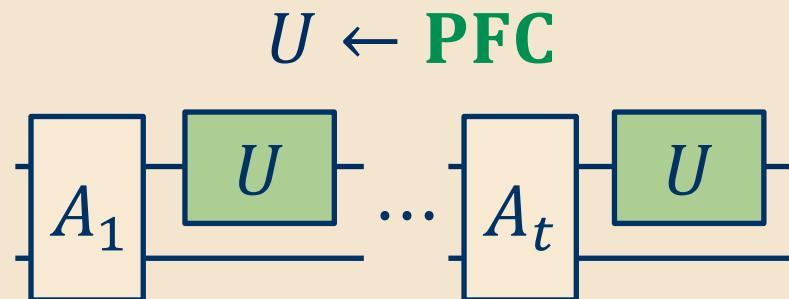
Proof strategy: show that both are indistinguishable from



\approx



\approx



\approx

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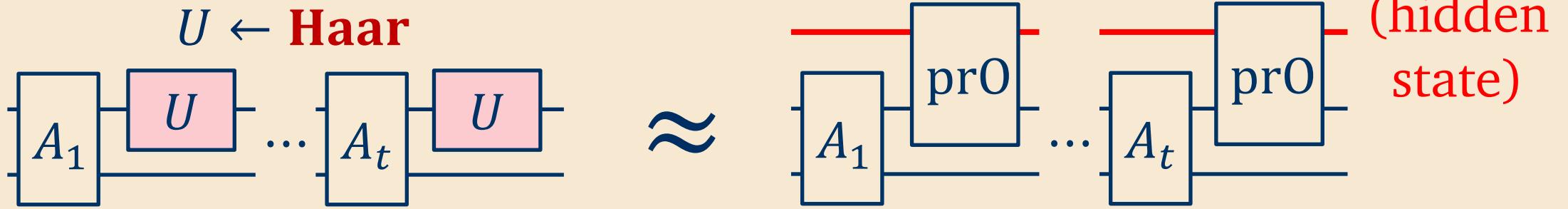
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$U \leftarrow \mathbf{PFC}$



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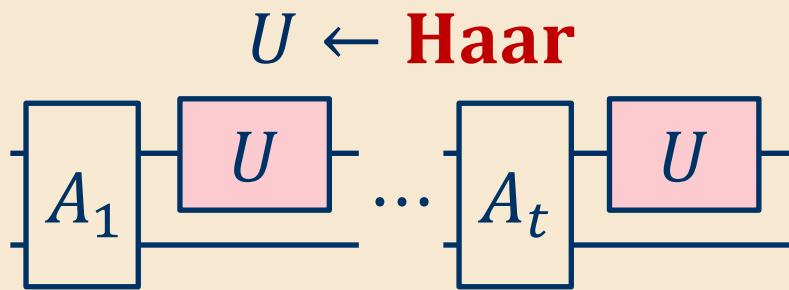
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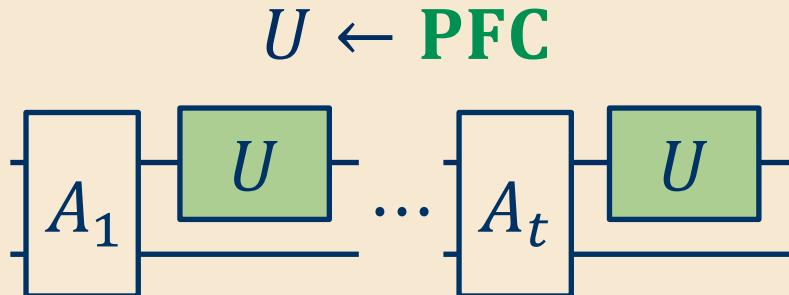
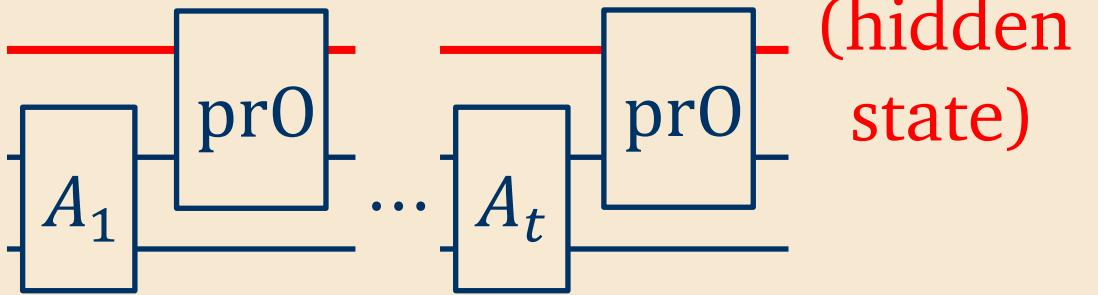
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\approx



$\not\approx$

The same proof
will show this!

$$U$$

$U \leftarrow \text{Haar}$

Hybrid 0



$U \leftarrow \text{Haar}$



$U \leftarrow \text{Haar}$

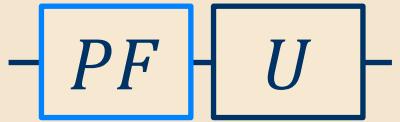
$$P \leftarrow S_N$$

$$F \leftarrow \{\pm 1\}^N$$

Hybrid 0 \equiv Hybrid 1



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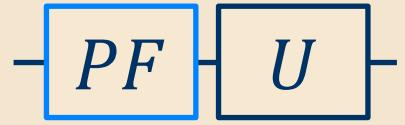
Step 1: insert random permutation P random ± 1 diagonal F .

$$P = \begin{pmatrix} & & 1 \\ 1 & & \\ & & 1 \end{pmatrix} \quad F = \begin{pmatrix} +1 & & \\ & -1 & \\ & & -1 \end{pmatrix}$$



$U \leftarrow \text{Haar}$

Hybrid 0



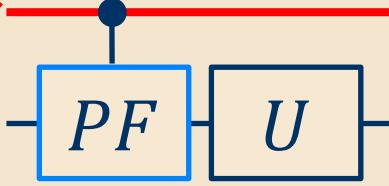
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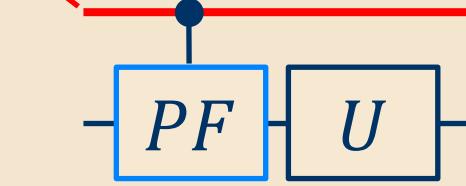
$F \leftarrow \{\pm 1\}^N$

Hybrid 1

$\sum_{P,F} |P, F\rangle$



$U \leftarrow \text{Haar}$



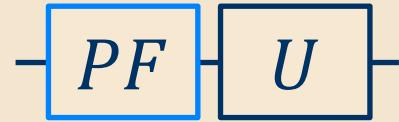
$U \leftarrow \text{Haar}$

Hybrid 2



$U \leftarrow \text{Haar}$

Hybrid 0



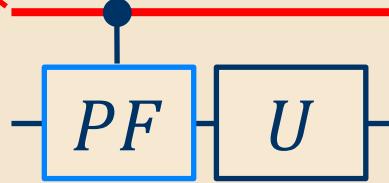
$U \leftarrow \text{Haar}$

$P \leftarrow S_N$

$F \leftarrow \{\pm 1\}^N$

\equiv Hybrid 1 \equiv Hybrid 2

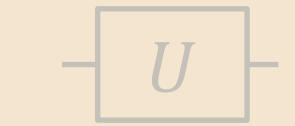
$\sum_{P,F} |P,F\rangle$



$U \leftarrow \text{Haar}$

Step 2: replace random P, F with a purification.

- Initialize external/ancilla system to $\sum_{P,F} |P,F\rangle$
- On each query, apply $P \cdot F$ **controlled** on $|P,F\rangle$



$U \leftarrow \text{Haar}$



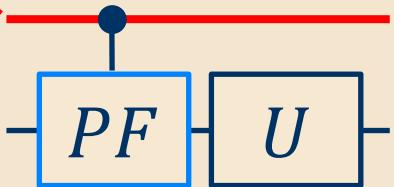
$U \leftarrow \text{Haar}$

$$P \leftarrow S_N$$

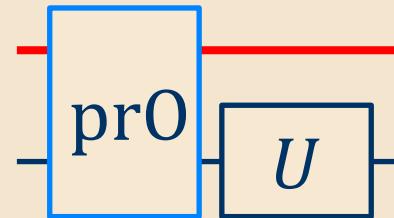
$$F \leftarrow \{\pm 1\}^N$$

$$\text{Hybrid 0} \equiv \text{Hybrid 1}$$

$$\sum_{P,F} |P,F\rangle$$



$U \leftarrow \text{Haar}$



$U \leftarrow \text{Haar}$

Hybrid 2 \approx Hybrid 3



$U \leftarrow \text{Haar}$

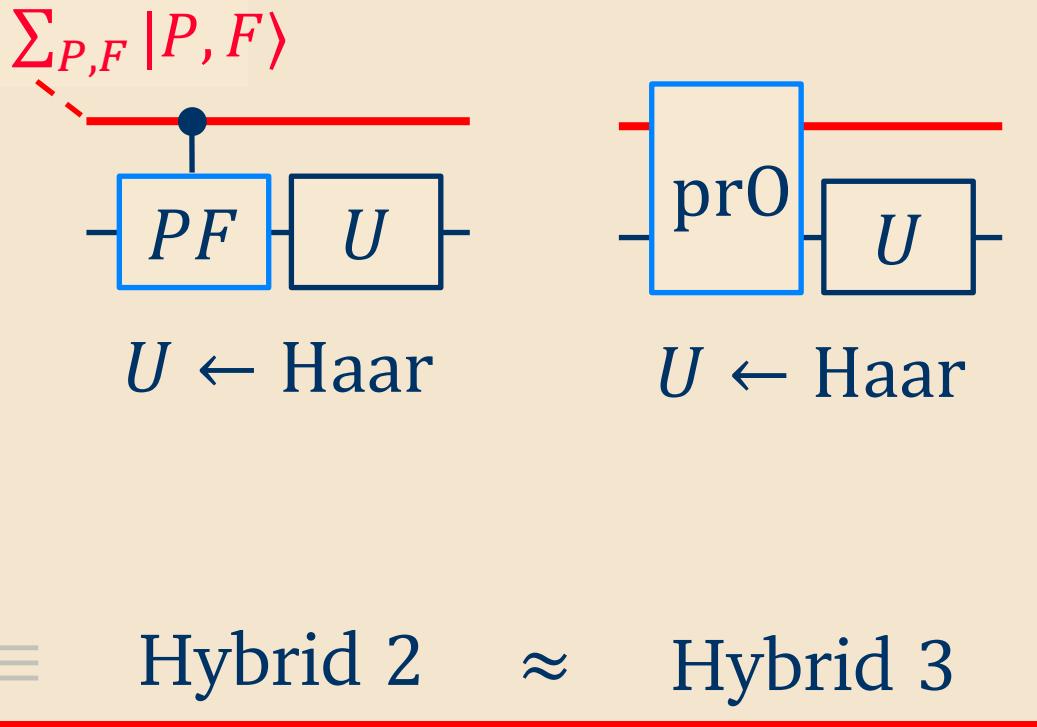


$U \leftarrow \text{Haar}$

$$P \leftarrow S_N$$

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Hybrid 0 \equiv Hybrid 1



Step 3: Key idea: analyze ctl-PF in a different basis.

Let's see how this works...

$$(1) \text{ ctl-}PF: \quad |x\rangle \otimes |P, F\rangle \mapsto (-1)^{F(x)} |P(x)\rangle \otimes |P, F\rangle$$

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Definition: for $D = \{(x_1, y_1), \dots, (x_t, y_t)\}$,

$$|\Phi_D\rangle := \sum_{P,F} (-1)^{F(x_1) + \dots + F(x_t)} \cdot \delta_{P(x_1)=y_1} \dots \delta_{P(x_t)=y_t} |P, F\rangle$$

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$$\text{pr0: } |x\rangle \otimes |D\rangle \mapsto \sum_y |y\rangle \otimes |D \cup \{(x, y)\}\rangle$$



$U \leftarrow \text{Haar}$

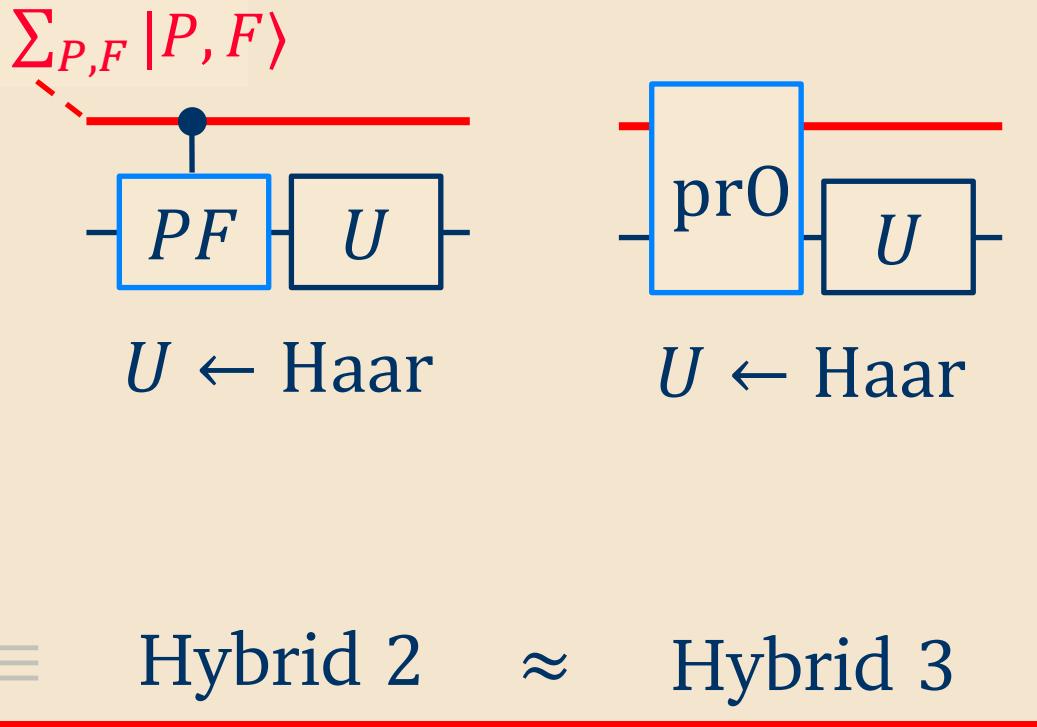


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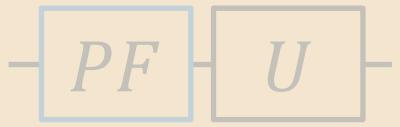


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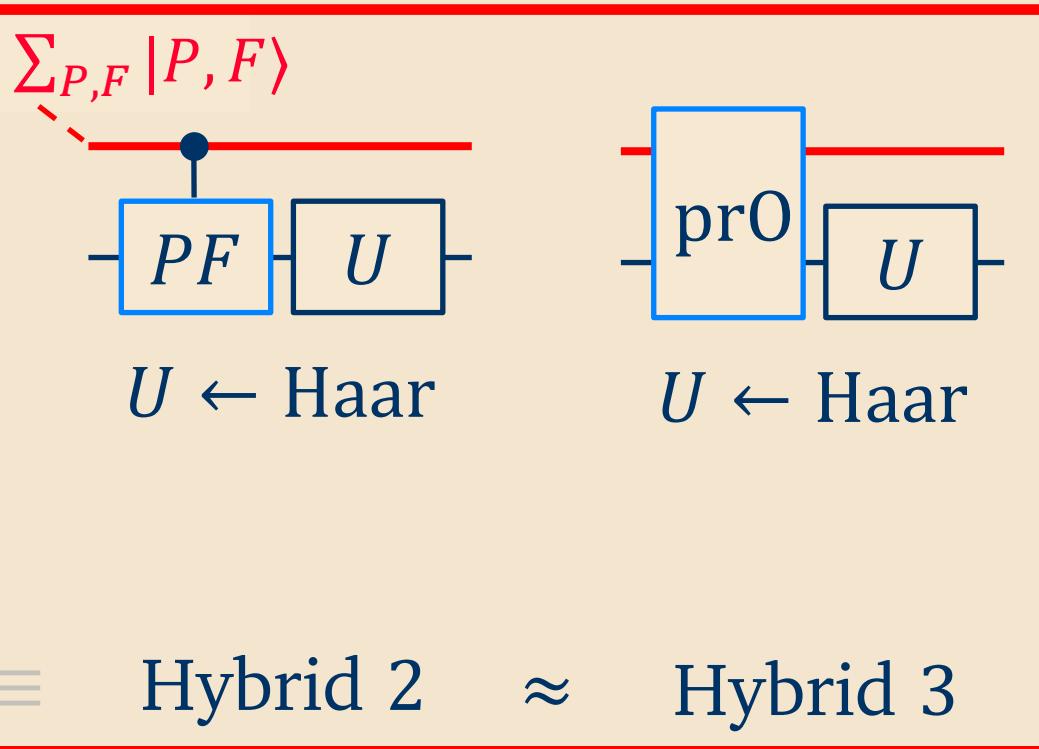


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- **Intuition:** ctl-PF behaves like pr0, up to relabeling $|\Phi_D\rangle \mapsto |D\rangle$



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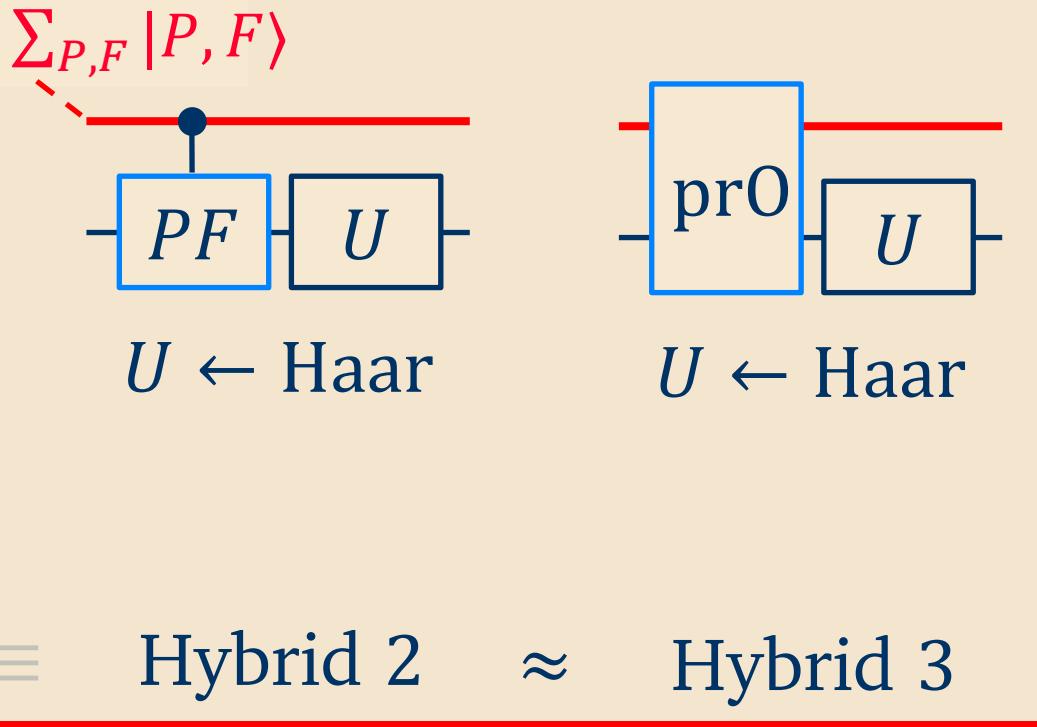


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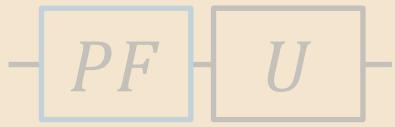
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- **Intuition:** ctl-PF behaves like pr0, up to relabeling $|\Phi_D\rangle \mapsto |D\rangle$
- Actually, $\{|\Phi_D\rangle\}_D$ aren't fully orthogonal. But composing with $U \leftarrow (2\text{-design})$ makes the “non-orthogonal” ones hard to find.



$U \leftarrow \text{Haar}$

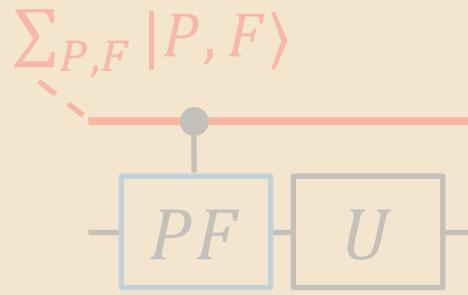


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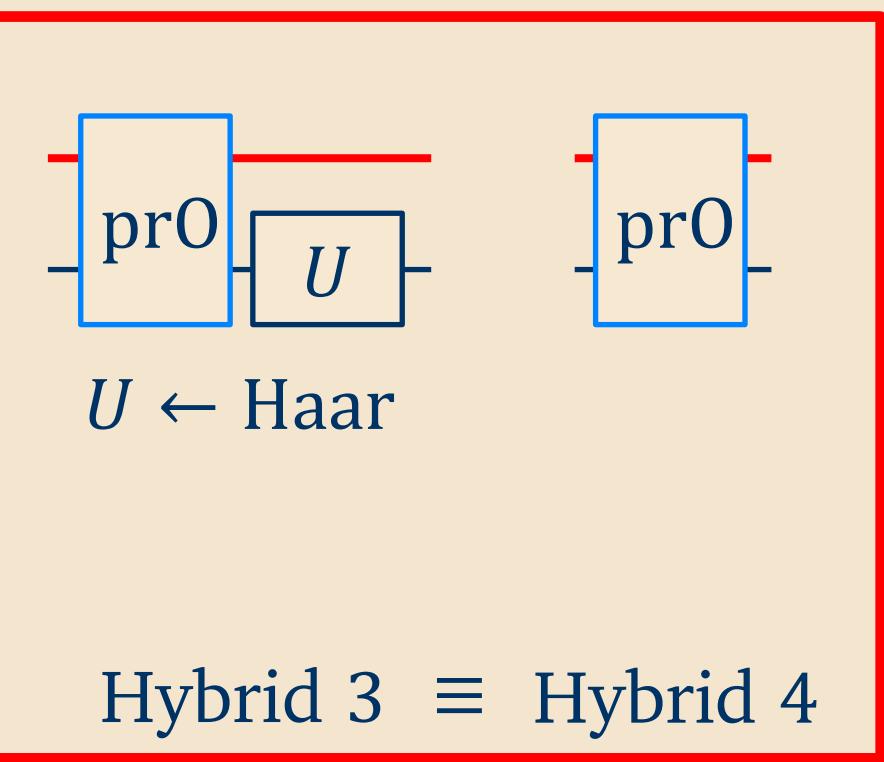
$P \leftarrow S_N$

$F \leftarrow \{\pm 1\}^N$

Hybrid 0 \equiv Hybrid 1 \equiv Hybrid 2



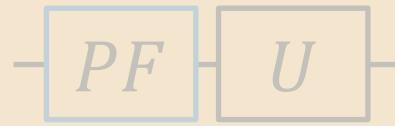
$U \leftarrow \text{Haar}$



Hybrid 3 \equiv Hybrid 4



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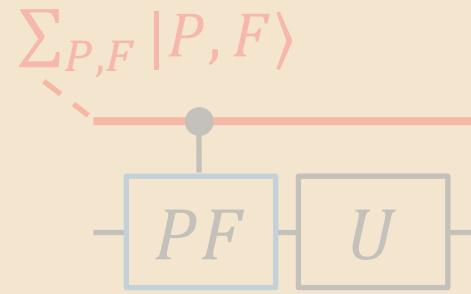


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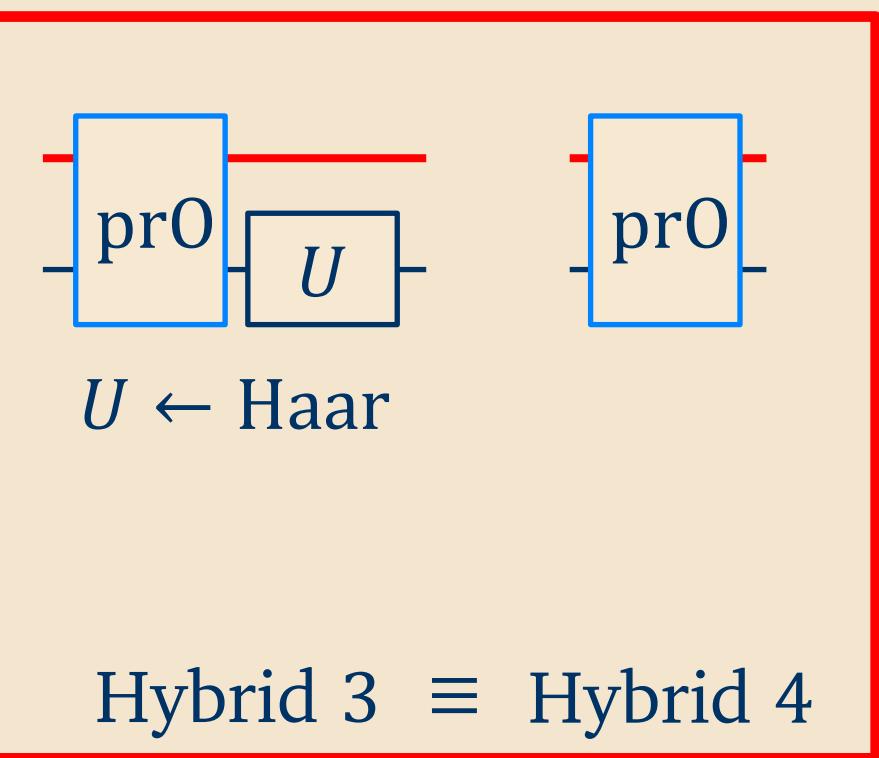
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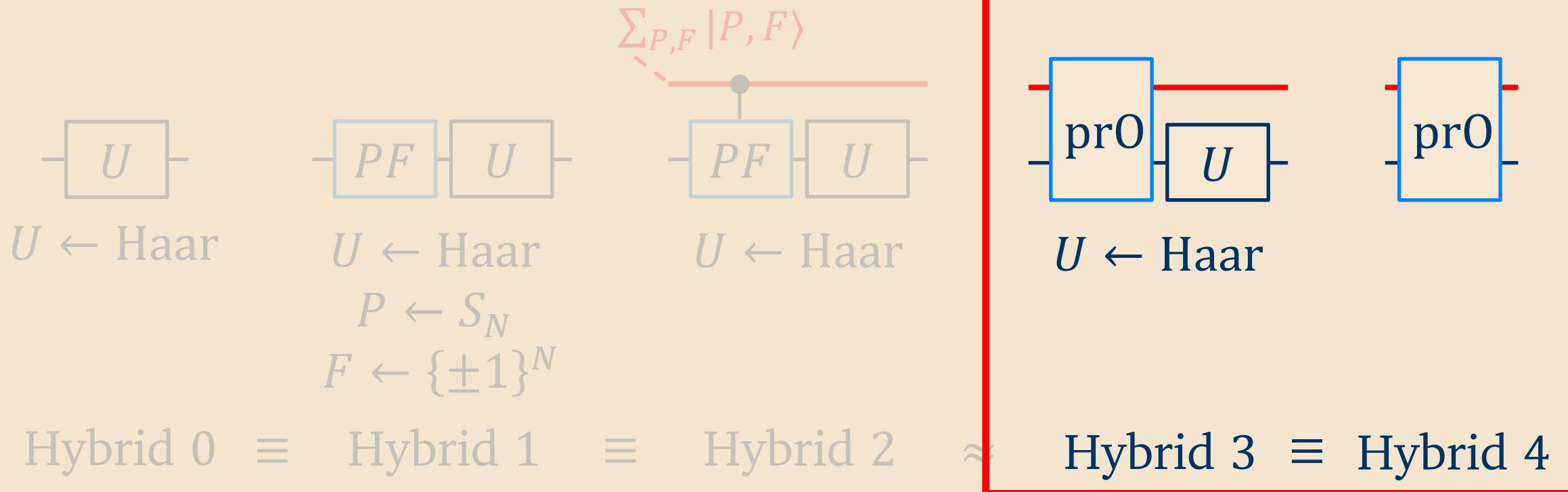
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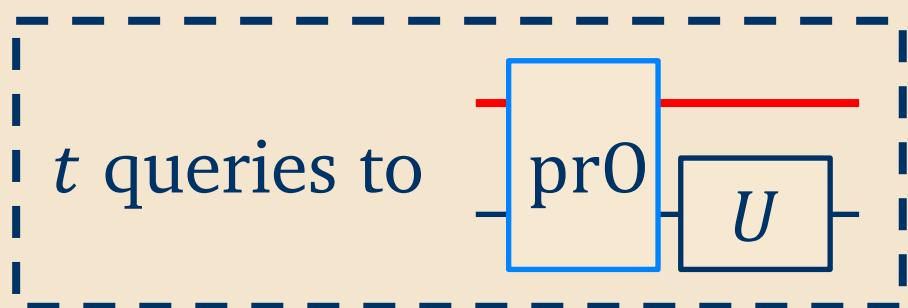
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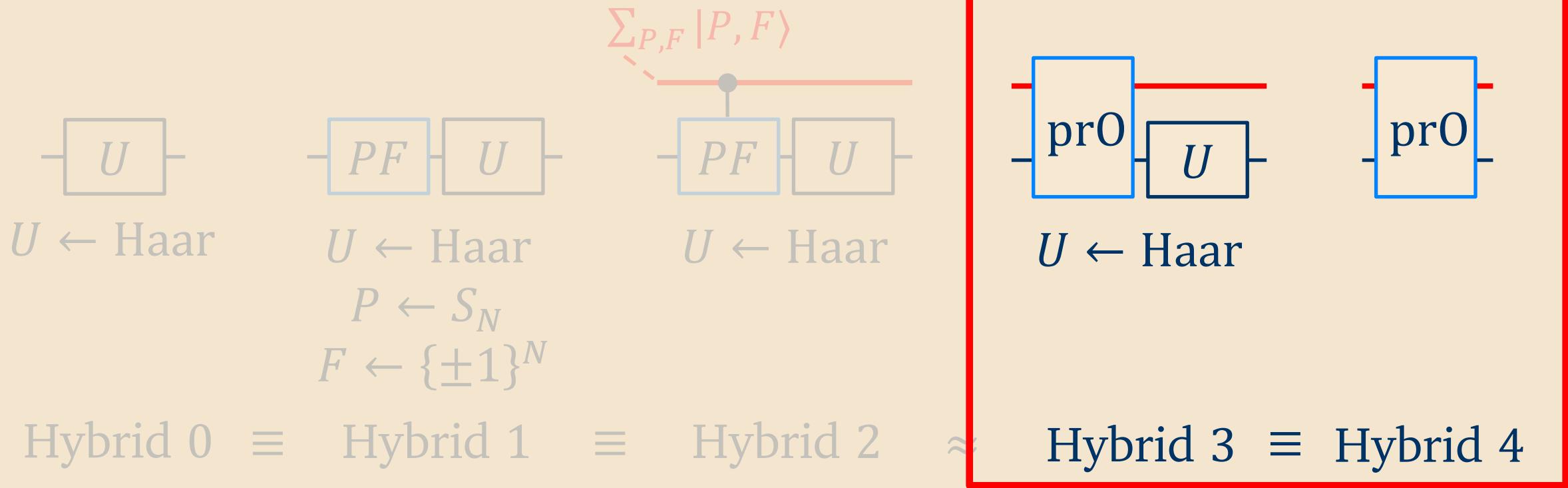
Hybrid 3 \equiv Hybrid 4

Step 4: Turns out pr0 has the following unitary invariance property:

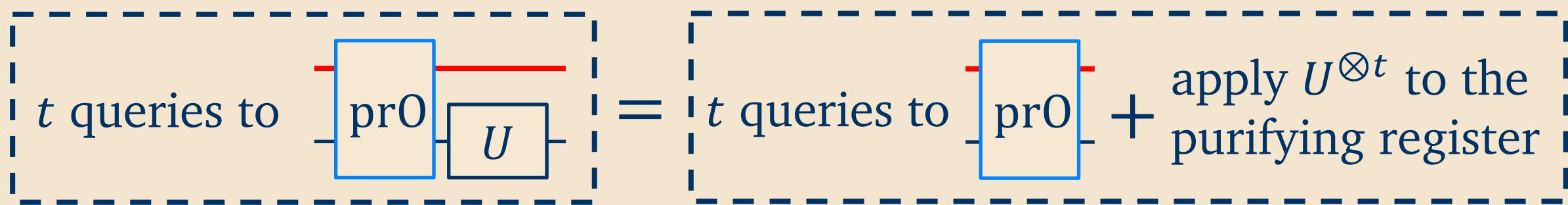


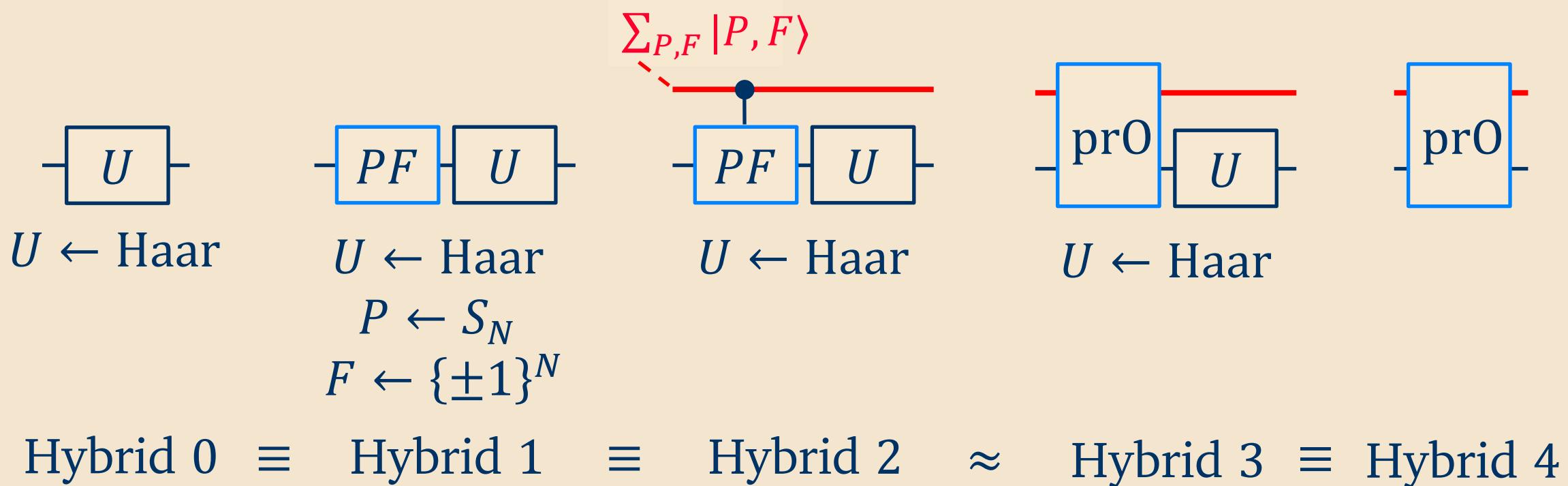
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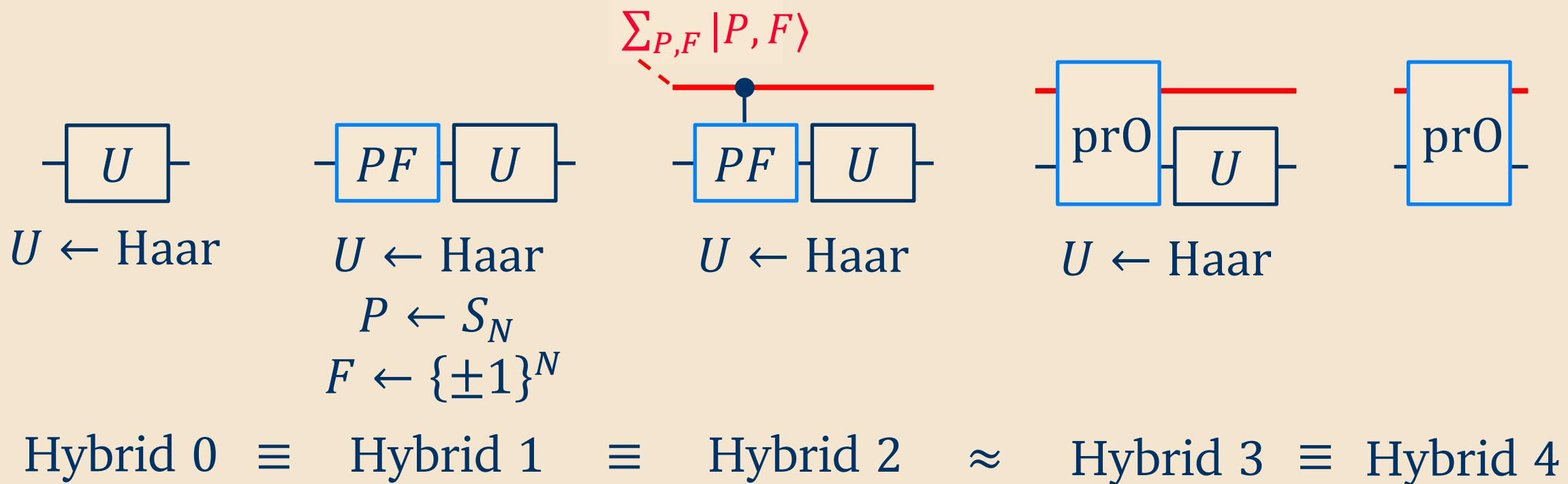




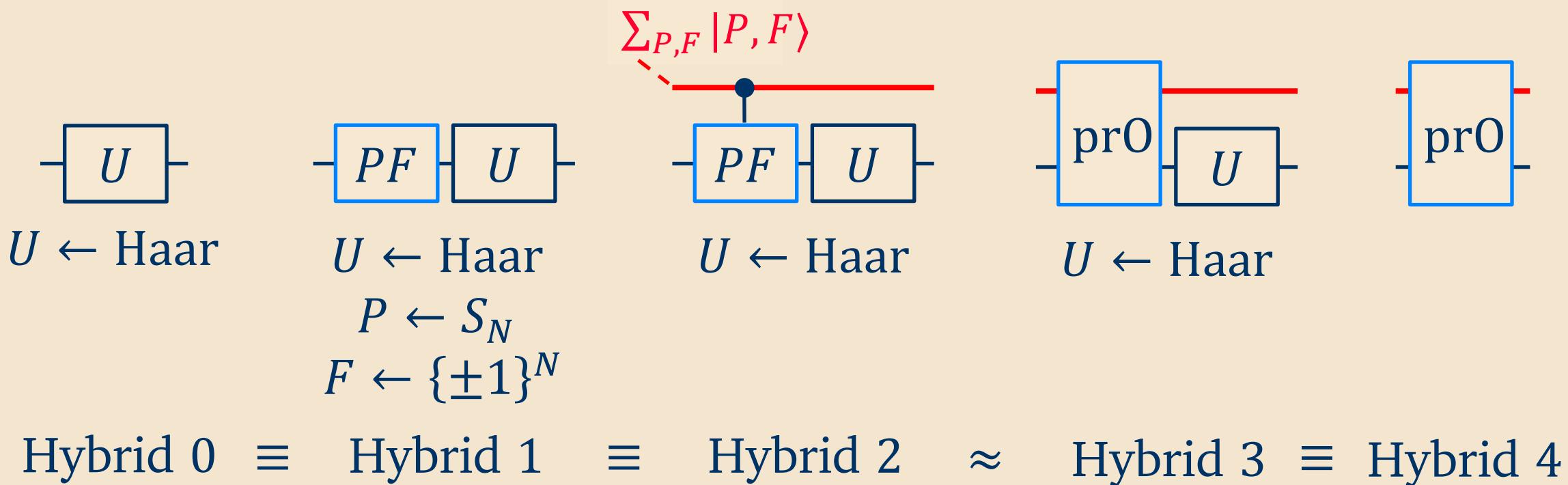
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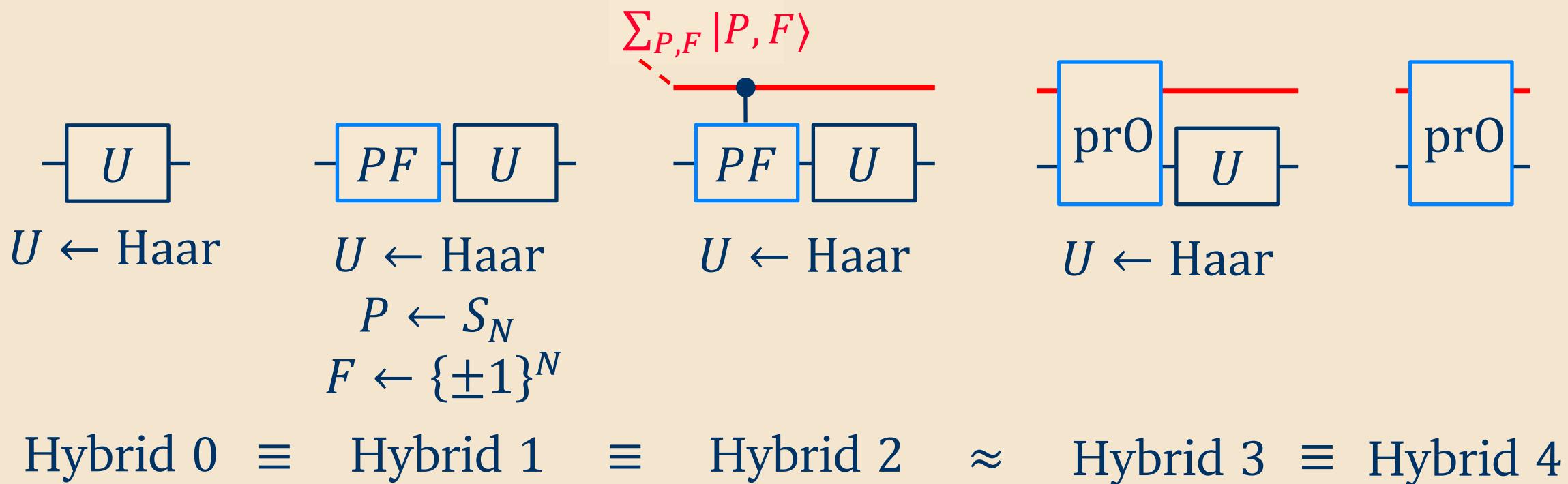


The PRU proof: Hybrid 2 \approx Hybrid 4 holds for any **2-design**.



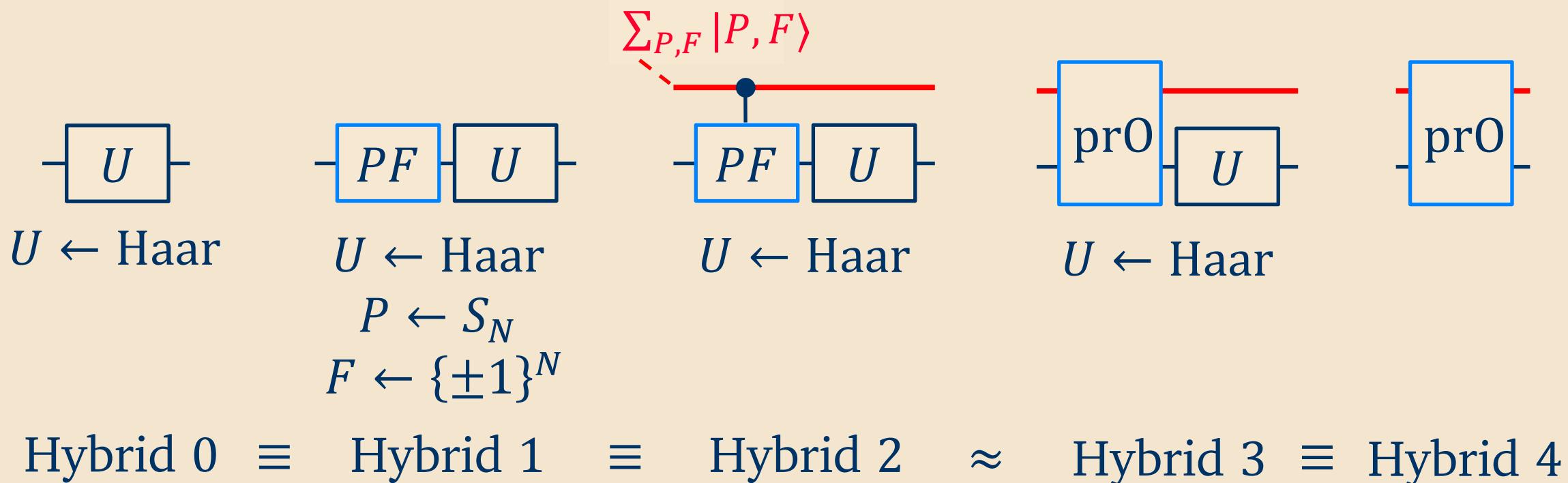
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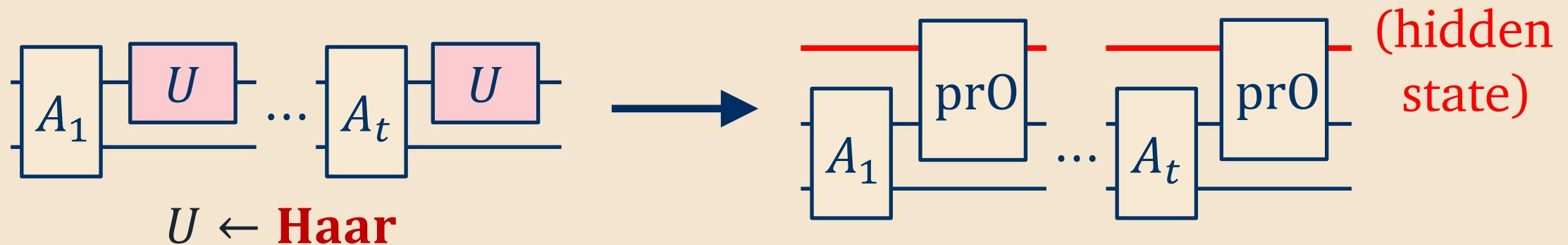
Finally, replace P and F with pseudorandom.

Rest of this talk

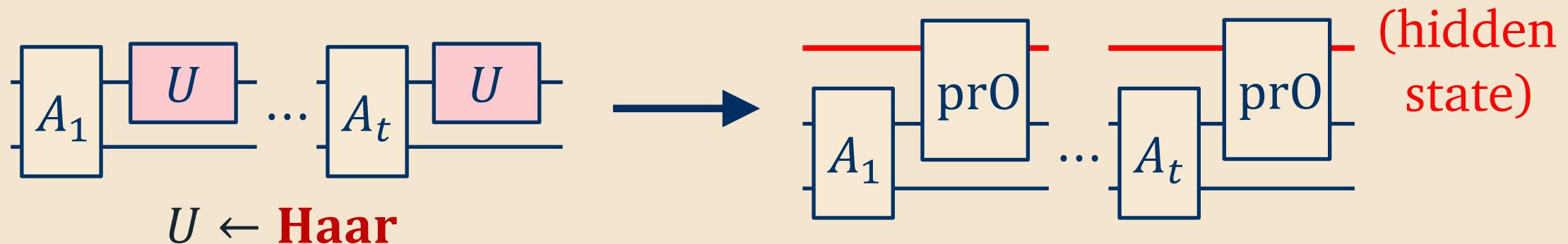
- Lazy sampling of a random function
- Lazy sampling of a random unitary
- Proving correctness + PRUs exist
- **Applications**

The path-recording oracle is a general-purpose tool for analyzing Haar-random unitaries.

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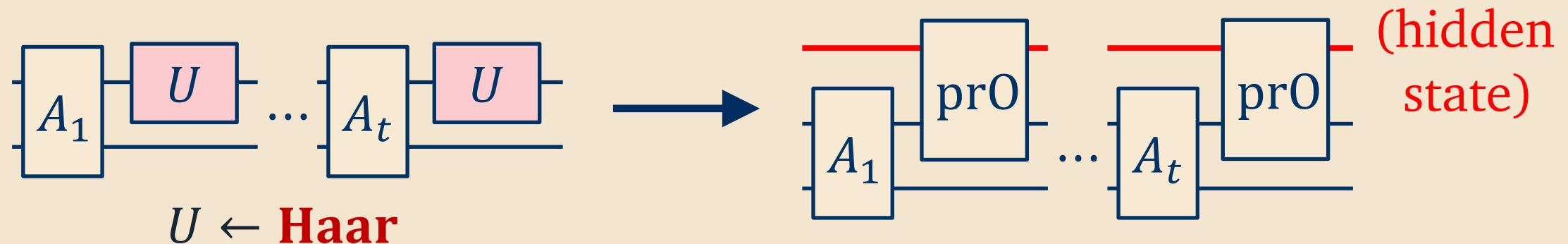


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Many statements about Haar-random U can be reduced to simple claims about this data structure

- [MH24]: elementary proof of [SHH24] gluing lemma
- [SMLBH25]: existence of low-depth PRUs

Let's see an example.

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Gluing lemma [SHH24]:

If U_1 and U_2 overlap on
 $|B| = \omega(\log n)$ qubits, then

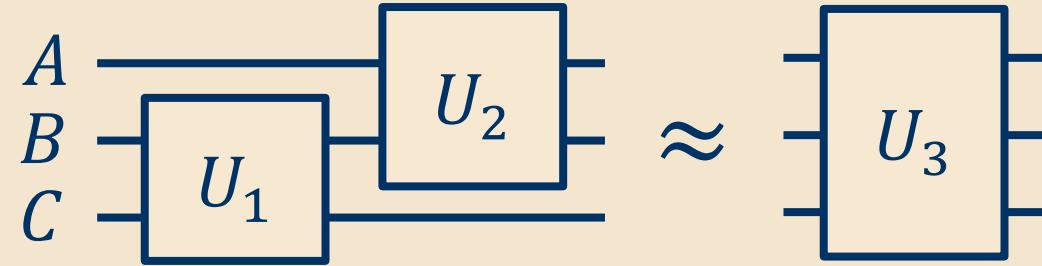
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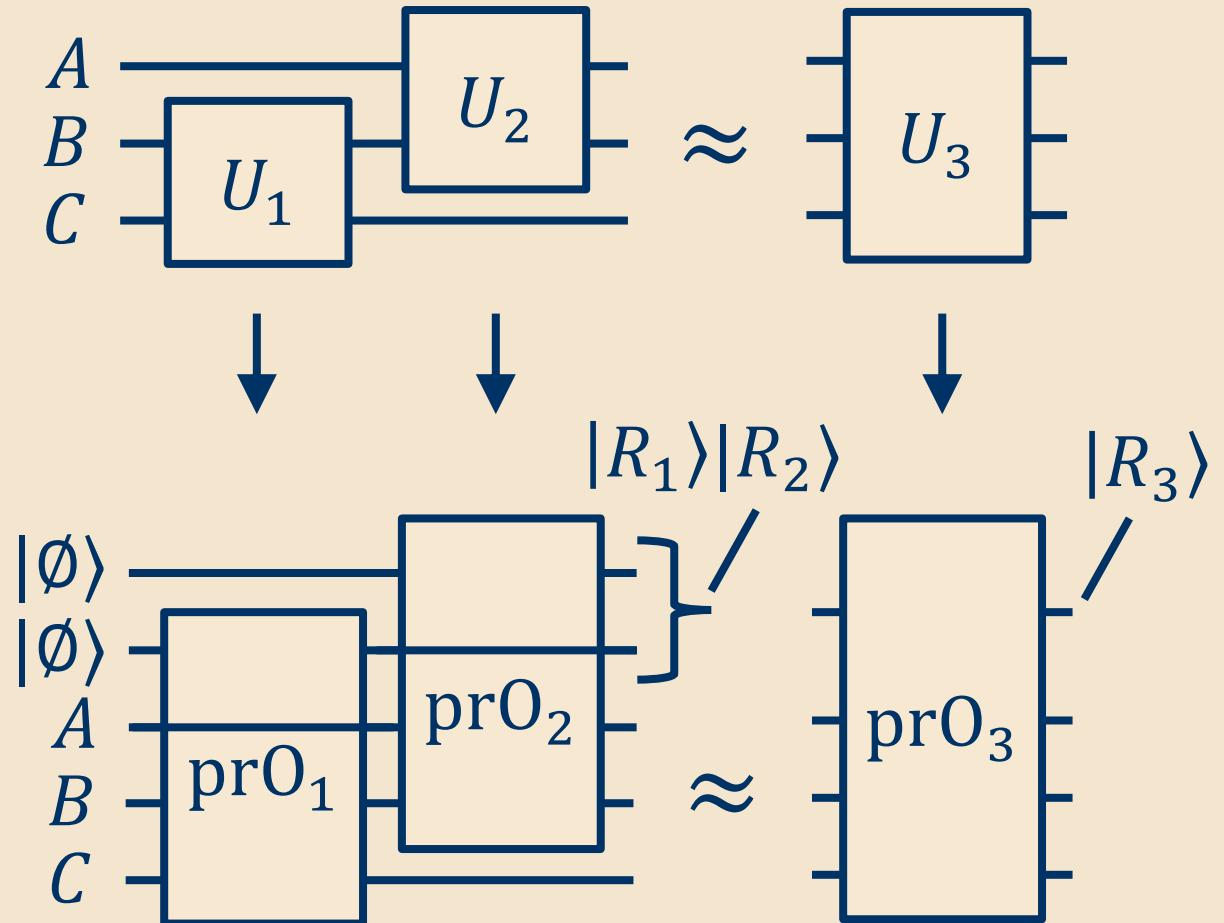
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New proof: combinatorial claim about path-recording oracle.



Future directions

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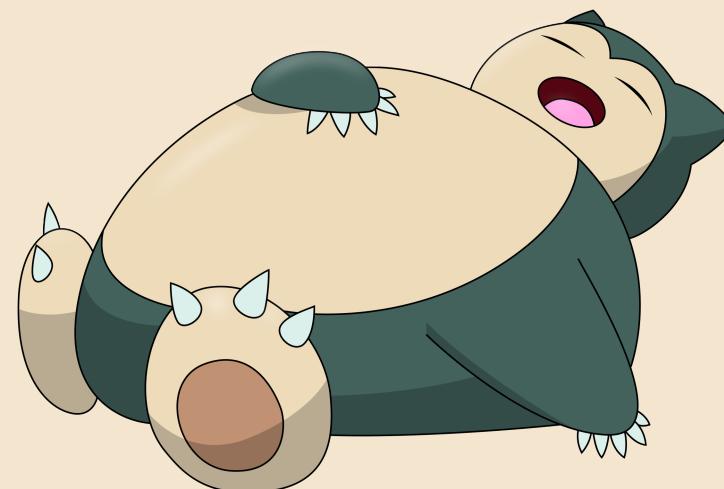
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Thanks!