

Random unitaries from random quantum circuits

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Based on work with:

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Huang, Yunchao Liu, Tony Metger, Thomas Schuster,
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Part I: Why random unitaries? A positive outlook.

A quantum state's shadow



- Prediction of observables $\langle O \rangle$ from classical snapshots of state.

A quantum state's shadow



- ▶ **Prediction** of observables $\langle O \rangle$ from classical snapshots of state.
- ▶ **Rotate** state by random quantum operation (unitary) and measure.

Aaronson, STOC 2018;
Huang, Kueng, Preskill, Nat. Phys 2021

Build-A-Shadow

- ▶ Average measurement channel:

$$M(|\psi\rangle) := \underbrace{\mathbb{E}_U}_{\text{classical randomness}} \left(\underbrace{\sum_{b \in \{0,1\}^n} |\langle \psi | U | b \rangle|^2}_{\text{quantum randomness}} U | b \rangle \langle b | U^\dagger \right).$$

Build-A-Shadow

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- ▶ Classical shadow from samples and inverting M:

$$\hat{\rho} := M^{-1} \left(\frac{1}{N} \sum_{i=1}^N U_i | b_i \rangle \langle b_i | U_i^\dagger \right).$$

- ▶ Estimate $\langle O \rangle_\psi$ by $\text{Tr}[\hat{\rho} O]$.
- ▶ Guarantees from $\text{Var}(\text{Tr}[\hat{\rho} O])$.
- ▶ 3rd moments of Haar measure suffice!

Chapters

1. Why random unitaries? A positive outlook.
2. Unitary designs and pseudorandom unitaries
3. Random quantum circuits converge to designs.
4. Random unitaries in extremely low depth.
5. Outlook

Part II: Unitary designs and pseudorandom unitaries

Quantum pseudorandomness

- ▶ Haar random unitaries require exponentially **deep** circuits.

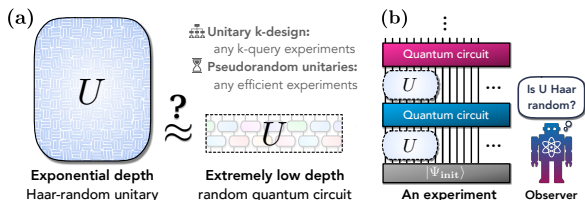
Quantum pseudorandomness

- ▶ Haar random unitaries require exponentially **deep** circuits.
- ▶ Haar random unitaries are **hard** to learn from counting arguments.



Quantum pseudorandomness

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- ▶ Haar random unitaries are **hard** to learn from counting arguments.



- ▶ Unitary k -design: Indistinguishable from k copies of U .
- ▶ PRU's: Indistinguishable in **polynomial** time.

Unitary designs

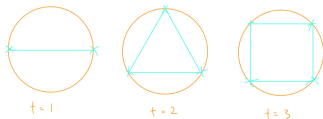


Unitary k -design ν matches the k th moments of Haar measure

$$\Phi_{\mu_H}^{(k)} = \Phi_{\nu}^{(k)}$$

for

$$\Phi_{\nu}^{(k)}(A) := \mathbb{E}_{U \sim \nu} \left[U^{\otimes k} A U^{\dagger, \otimes k} \right].$$



- ▶ Example: Moments of $U(1)$ $\mathbb{E}_{\phi} e^{it\phi} = 0$, but $e^{i2 \times 0} + e^{i2 \times \pi} \neq 0$.

Additive vs. multiplicative error approximate designs

Moment operator:

$$\Phi_\nu(A) := \mathbb{E}_{U \sim \nu} \left[U^{\otimes k} A U^{\dagger, \otimes k} \right].$$

Additive error approximate designs:

$$\|\Phi_{\mu_H} - \Phi_\nu\|_\diamond \leq \varepsilon.$$

Multiplicative error approximate designs:

$$(1 - \varepsilon) \Phi_H \preceq \Phi_\varepsilon \preceq (1 + \varepsilon) \Phi_H,$$

► CP ordering: $A \preceq B$ if $B - A$ is completely positive.

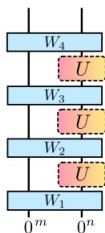
Information theoretical quantum pseudorandomness

- ▶ Approximate **designs** look Haar in any experiment that queries k -copies.

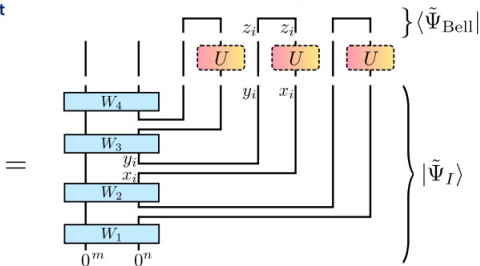
Information theoretical quantum pseudorandomness

- ▶ Approximate **designs** look Haar in any experiment that queries k -copies.
- ▶ **Additive** error $1/\text{superpoly}(n)$ -approximate designs property imply **non-adaptive** security.
- ▶ **Multiplicative** error $1/\text{superpoly}(n)$ -approximate designs property implies **adaptive** security.

Any adaptive quantum experiment



Parallel queries + Bell projection

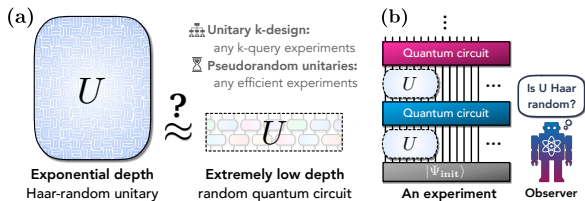


PRU's

Definition (Pseudorandom unitaries)

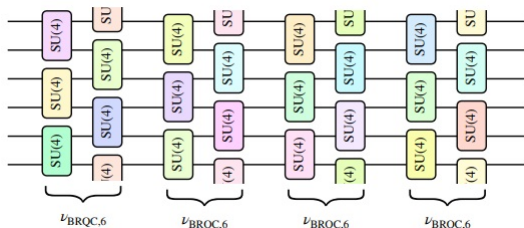
A family of **quantum** circuits $\{U_I\}_{I \in \{0,1\}^m}$ with $m = \text{poly}(n)$ that can be efficiently prepared. For any **poly-time** algorithm $A^U(1^n)$ that queries U any number of times, satisfies

$$\Pr_I[A^U(1^n) = 1] \approx \Pr_{U \sim \mu_H}[A^U(1^n) = 1].$$



Part II: Random quantum circuits converge to designs

Random quantum circuits



- ▶ Typically refers to circuits with gates drawn iid.
- ▶ Example: Brickwork circuits with gates drawn from Haar measure.
- ▶ Depth d corresponds to d -fold convolution: $\nu_{\text{BRQC},6}^{*4}$.
- ▶ Random walk on unitary group.

Unitary designs from spectral gaps

Gap of k -th moment operator:

$$\Delta_k(\nu) := 1 - \underbrace{\left\| \mathbb{E}_{U \sim \nu} U^{\otimes k} \otimes \bar{U}^{\otimes k} - \mathbb{E}_{U \sim \mu_H} U^{\otimes k} \otimes \bar{U}^{\otimes k} \right\|_{\infty}}_{=: g(\nu, k)}.$$

- ▶ Submultiplicative $g(\nu^{*d}, k) \leq g(\nu, k)^d$.
- ▶ ν is $g(\nu, k)2^{2nk}$ -approximate designs. Brandão, Harrow, Horodecki, Comm.

Math. Phys 2016

Lemma

ν^{*d} is ε -approximate k -design in depth

$$d \geq \frac{1}{\Delta_k} (2nk + \log(1/\varepsilon))$$

History of spectral gap estimates

Random **reversible** circuits (3-local all-to-all):

- ▶ Gowers 1996: $\Delta_k = \text{poly}^{-1}(nk) \implies$ Approximate **permutation designs** in depth $\text{poly}(nk)(2nk + \log(1/\epsilon))$.
- ▶ Hoory, Magen, Myers, Rackoff: $\Delta_k = \Omega(n^{-3}k^{-3})$
- ▶ Brodsky, Hoory : $\Delta_k = \Omega(n^{-2}k^{-1})$
- ▶ He and O'Donnell: $\Delta_k = \tilde{\Omega}(n^{-1}k^{-1})$

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Random **quantum** circuits (2-local all-to-all):

- ▶ Brown and Viola: First order expansion of the **gap**,
 $\Delta_k = \text{const.}/n + O_k(1/n^2)$.
- ▶ Harrow and Low/Brandao, Horodecki: $\Delta_{2/3} = \Omega(1/n)$.
- ▶ Brandão, Harrow, and Horodecki: $\Delta_k = \Omega(n^{-1}k^{-10.5})$
- ▶ **H**: $\Delta_k = \Omega(n^{-1}k^{-5-o(1)})$

Other constructions of approximate designs at FOCS 2024

All presented in session 2C, 1:30-2:30pm on Monday!!!

- ▶ Haah, Liu, and Tan: $\Delta_k(\text{Pauli rotations}) = \Omega(k^{-1})$. \implies multiplicative error approximate designs in (1D) depth $O(n^2k)$.
- ▶ Chen, Docter, Xu, Bouland, Brandão, Hayden: Construction of additive error designs in depth $O(k\text{poly}(n))$.
- ▶ Metger, Poremba, Sinha, Yuen, Additive error designs in depth $O(k\text{poly}(n))$.

Main result: k -independent gaps

Theorem (k -independent gap)

For 3-local all-to-all *random reversible* and 2-local all-to-all *random quantum* circuits $\Delta_k = \Omega(n^{-3})$ for all $k \leq 2^{n/2}$.

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For 3-local all-to-all *random reversible* and 2-local all-to-all *random quantum* circuits $\Delta_k = \Omega(n^{-3})$ for all $k \leq 2^{n/2}$.

Theorem (Quasi-optimal gap)

For all $k \leq 2^{n/6.1}$, we have $\Delta_k = \tilde{\Omega}(n^{-1})$, which is *optimal* (up to polylog factors) for random *reversible* and *quantum* circuits.

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Corollary (Designs in linear depth)

Random *brickwork* circuits form ε -approximate designs in depth $\text{polylog}(k)(2nk + \log(1/\varepsilon))$.

Elements of the proof

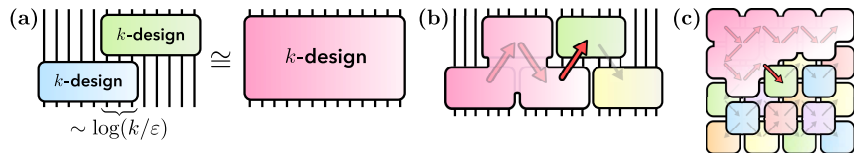
- ▶ Gap of random **reversible** circuits from **expanding Cayley graph** of symmetric group Kassabov, Inventionae 2010.
- ▶ Relate to gap of random **reversible** circuits via “**PFC**” ensemble Metger, Poremba, Sinha, Yuen, FOCS 2024:

$$\text{PFC} \cong \underbrace{\sum_{x \in \{0,1\}^n} |\pi(x)\rangle\langle x|}_{\text{random permutation, } \pi \in S_{2^n}} \times \underbrace{\sum_{x \in \{0,1\}^n} (-1)^{f(x)} |x\rangle\langle x|}_{\text{random phase, } f: \{0,1\}^n \rightarrow \{0,1\}} \times \underbrace{C}_{\text{random Clifford}}$$

- ▶ Approximate components by random walks in subgroups.
- ▶ Universality theorem for random **quantum** circuits H, Quantum 2022
- ▶ Techniques developed for frustration-free Hamiltonians: **Detectability lemma** and **quantum union bound**.

Part IV. Random unitaries in extremely low depth (What the spectral gap can't see)

Random unitaries from gluing



Theorem

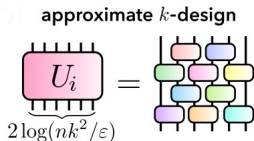
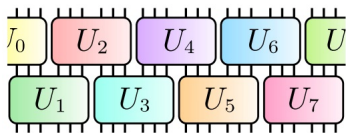
The product of two k -designs overlapping on $\xi \geq \log_2(nk/\epsilon)$ qubits form ϵ -approximate unitary design on the full space.

Approximate designs from gluing

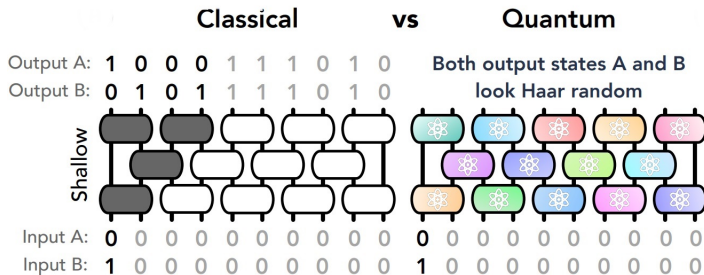
- ▶ Use random quantum circuits in the blocks.
- ▶ Plug in $\text{polylog}(k)(2nk + \log(1/\epsilon))$

Theorem

“Coarse-grained” random quantum circuits generate ϵ -approximate designs in depth $\text{polylog}(k)k \log(n/\epsilon)$.



How about classical circuits



- ▶ **Classical circuits** require **linear depth** to be approximately 2-wise independent.

Optimality of our results

Theorem

$\frac{1}{2}$ -approximate 2-designs require $\Omega(\log(n))$ depth.

- ▶ Lower bound on anticoncentration in any basis.

Theorem

PRU's require $\omega(\log(n))$ depth.

- ▶ 1D quantum circuits of depth $O(\log(n))$ can be efficiently learned. [Huang, Liu, Broughton, Kim, Anshu, Landau, McClean, STOC 2023](#)

PRUs from gluing

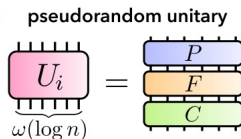
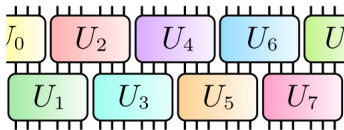
Use "PFC" in the blocks:

Theorem

If quantum secure one-way functions exist, PFC is a PRU. The depth of PFC is $\text{poly}(n)$.

Theorem

PRU's can be generated in depth $\text{polylog}(n)$ on any geometry, including a 1D line.



How to prove the gluing lemma?

Lemma (Unitary designs from EPR states)

A random unitary ensemble \mathcal{E} forms an ε -approximate unitary k -design with error

$$\varepsilon = \frac{4^{nk}}{k!} \cdot \left(1 + \frac{k^2}{2^{n+1}}\right) \cdot \left\| [(\Phi_{\mathcal{E}} - \Phi_H) \otimes \mathbb{1}](P^{\text{EPR}}) \right\|_{\infty}.$$

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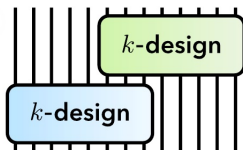
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- ▶ CP ordering $A \preceq B$ iff $A \otimes \mathbb{1}(P^{\text{EPR}}) \leq B \otimes \mathbb{1}(P^{\text{EPR}})$.
- ▶ Analyse spectrum of **Choi states**.

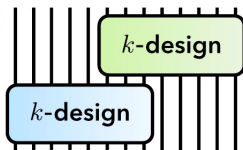
Weingarten and permutations



- ▶ Expand blocks in **permutations**:

$$\Phi_H(A) \equiv \mathbb{E}_{U \sim \mathcal{E}_H} [U^{\otimes k} A (U^\dagger)^{\otimes k}] = \sum_{\sigma, \tau \in \mathcal{S}_k} \text{Tr}(A \sigma^{-1}) \text{Wg}(\sigma \tau^{-1}; 2^{\xi/2}) \tau.$$

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- ▶ Exploit approximate **orthogonality** of **permutations**:

$$\|G - \mathbb{1}_{k! \times k!}\|_\infty \leq \frac{k^2}{2^{\xi/2}}, \quad G_{\pi, \sigma} \equiv \frac{1}{2^{\xi/2}} \text{Tr}[\pi \sigma].$$

Proof sketch

$$\Phi_H \approx \sum_{\pi \in \mathcal{S}_k} |\pi\rangle\langle\pi|, \quad \langle\pi| \equiv \frac{1}{2^{\xi/4}} \text{Tr}[\pi \bullet]$$

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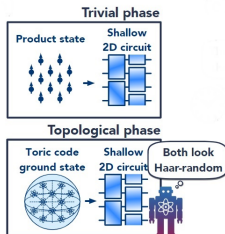
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► Use approximate **orthogonality** again:

$$\Phi_{H,1,2} \circ \Phi_{H,2,3} \approx \sum_{\pi \in \mathcal{S}_k} \begin{array}{cc} \boxed{\pi} & \boxed{\pi} \\ \boxed{\pi} & \boxed{\pi} \end{array} = \sum_{\pi \in \mathcal{S}_k} \boxed{\pi} \approx \Phi_{H,1,2,3} .$$

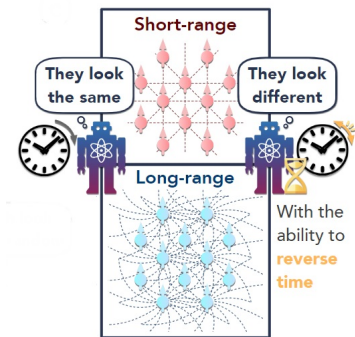
V. Applications

Hardness of quantum learning



- ▶ **Hard** to distinguish trivial order and toric code after applying PRU.
- ▶ **Topological order** up to circuits of subextensive depth.
- ▶ **Distinguishing** random pure states from **maximally mixed** using **single copy** measurements. [Chen, Cotler, Huang, and Li, FOCS 2022](#)

Power of time-reversal in quantum learning



- ▶ Distinguish 2D local circuit U_{2D} from U'_{2D} augmented with a long range interaction $e^{i\phi Z_i Z_j}$.
- ▶ Time-reversal allows to “see lightcones”.

Shallow shadows



Corollary

Classical shadows can be obtained with $\log(n)$ -depth circuits:

- ▶ Use the same inversion map as for Haar random measurement-channel.
- ▶ Learn M observables O with $O(\max_o \|O\|_1 \log(M))$ samples.

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- ▶ Sufficient for fidelity estimation.

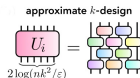
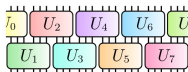
Outlook

Summary

- ▶ **Optimal** k -dependence for random **reversible** and random **quantum** circuits: resolved conjectures from 1996 and 2009.
- ▶ Also resolved robust **Brown-Susskind** conjecture.
- ▶ **Log-depth** convergence for random **quantum** circuits.
- ▶ Plenty of **hardness** results for learning.

Open problems

- ▶ **Log-depth** random **quantum** circuits with iid gates!
- ▶ Is the optimal scaling $\log(n) + k$ for approximate designs?
- ▶ Are random **quantum** circuits with iid gates PRU's?



Why is the purity not a counterexample?

- ▶ Unitary 2-design have maximal entanglement but shallow circuits do not!
- ▶ $\mathbb{E} \text{Tr}[\text{Tr}_A(|\psi\rangle\langle\psi|)^2] \leq (1 + \epsilon)2^{-\Omega(n)}$?

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- ▶ Relative errors only in the SWAP-test probability $\frac{1}{2} + \text{Tr}[\text{Tr}_A(|\psi\rangle\langle\psi|)^2]$.