



THE UNBEARABLE DIFFICULTY OF FINDING NEAR-TERM QUANTUM ADVANTAGES

JENS EISERT, FU BERLIN

MUNICH 2025

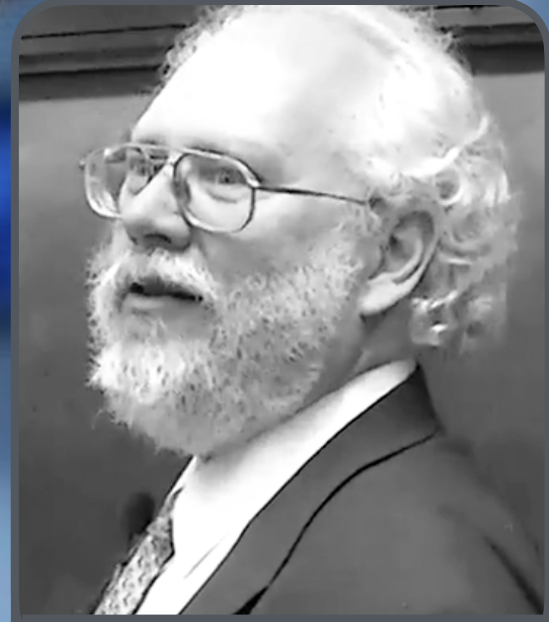
- **Quantum computing** was a purely conceptual idea



Feynman

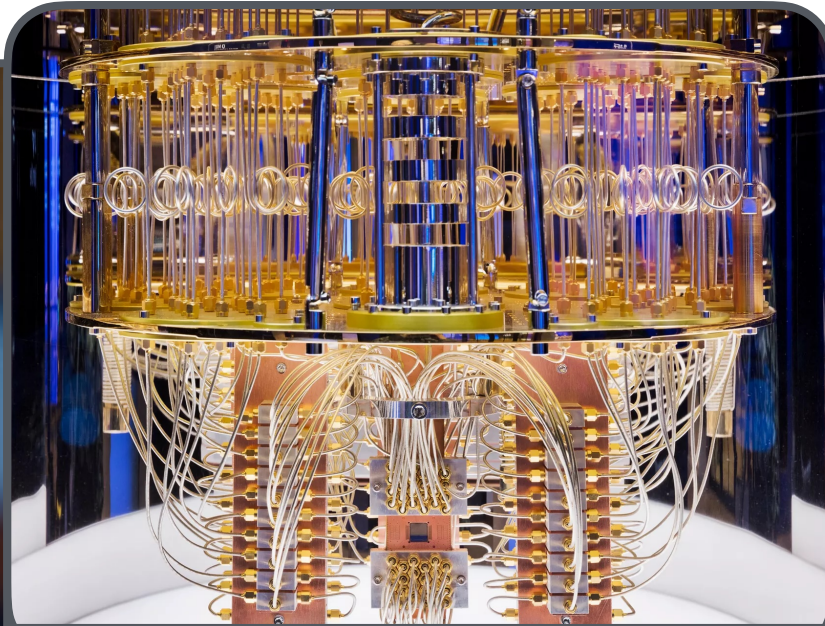


Benioff, Deutsch

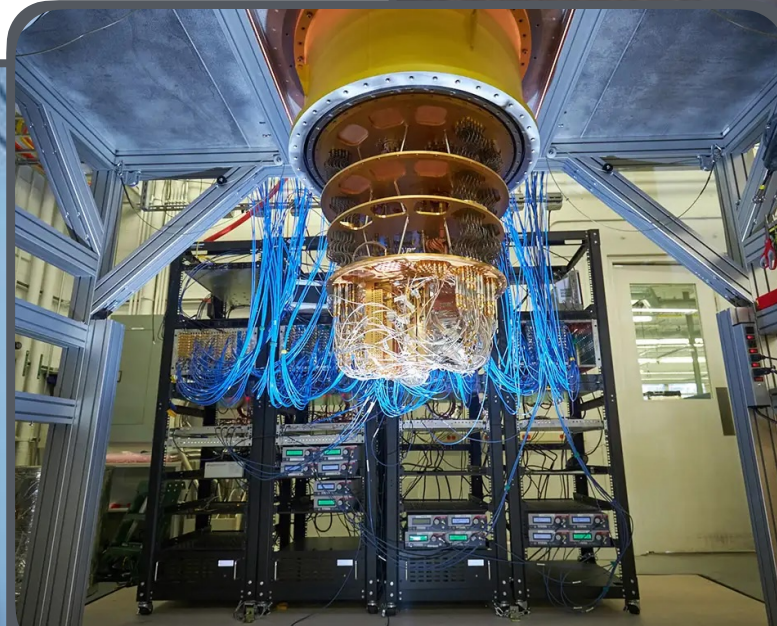


Shor

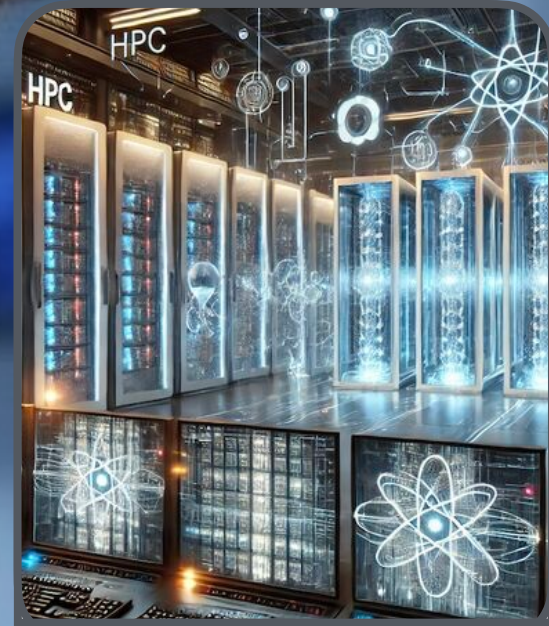
- Seems to be moving closer to **reality**



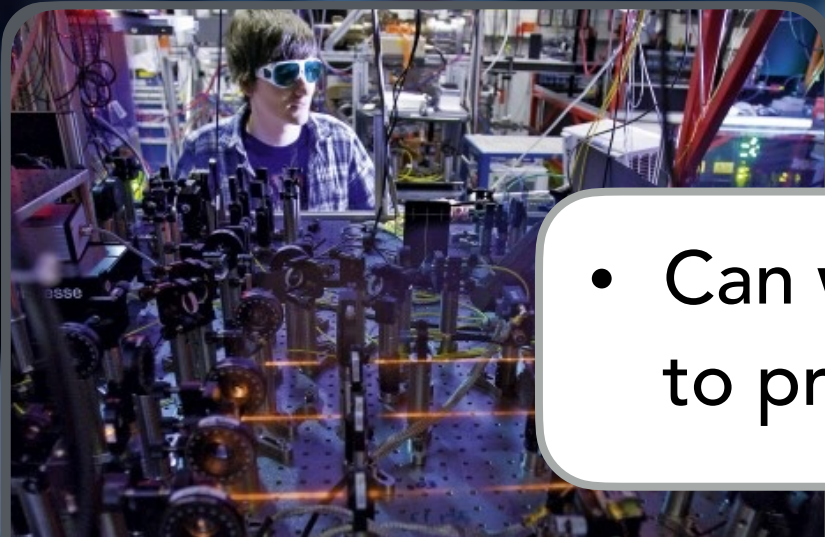
IBM



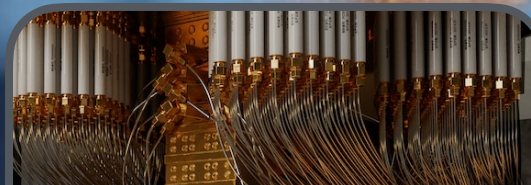
Google



QuERA



European efforts, MQV,
planqc, Pasqal, etc



Chinese efforts

- Can we reasonably hope realistic quantum devices to provide a **speedup over classical computers**?

- Quantum algorithms

Problem	Group	Complexity	Cryptosystem
Factorisation	$\mathbb{Z}$	Polynomial [89]	RSA
Discrete log	$\mathbb{Z}_{p-1} \times \mathbb{Z}_{p-1}$	Polynomial [89]	Diffie-Hellman, DSA, ...
Elliptic curve discrete log	Elliptic curve	Polynomial [76]	ECDH, ECDSA, ...
Principal ideal	$\mathbb{R}$	Polynomial [46]	Buchmann-Williams
Shortest lattice vector	Dihedral group	Subexponential [58, 80]	NTRU, Ajtai-Dwork, ...
Graph isomorphism	Symmetric group	Exponential	—

Montanaro, arXiv:1511.04206 (2015)

- Potentially controversial:** We do not know as much about **practically minded** problems and **near-term** algorithms as we would like to



OBSTRUCTIONS FOR NOISY QUANTUM CIRCUITS

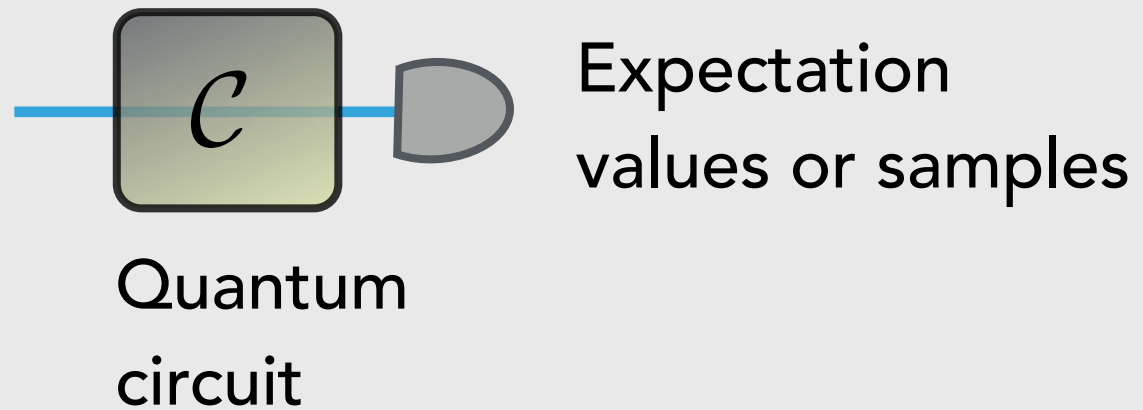
arXiv:2502.14252 (2025)

arXiv:2403.13927 (2024)

Nature Communications 15, 434 (2024)



- Actual circuits are **noisy**

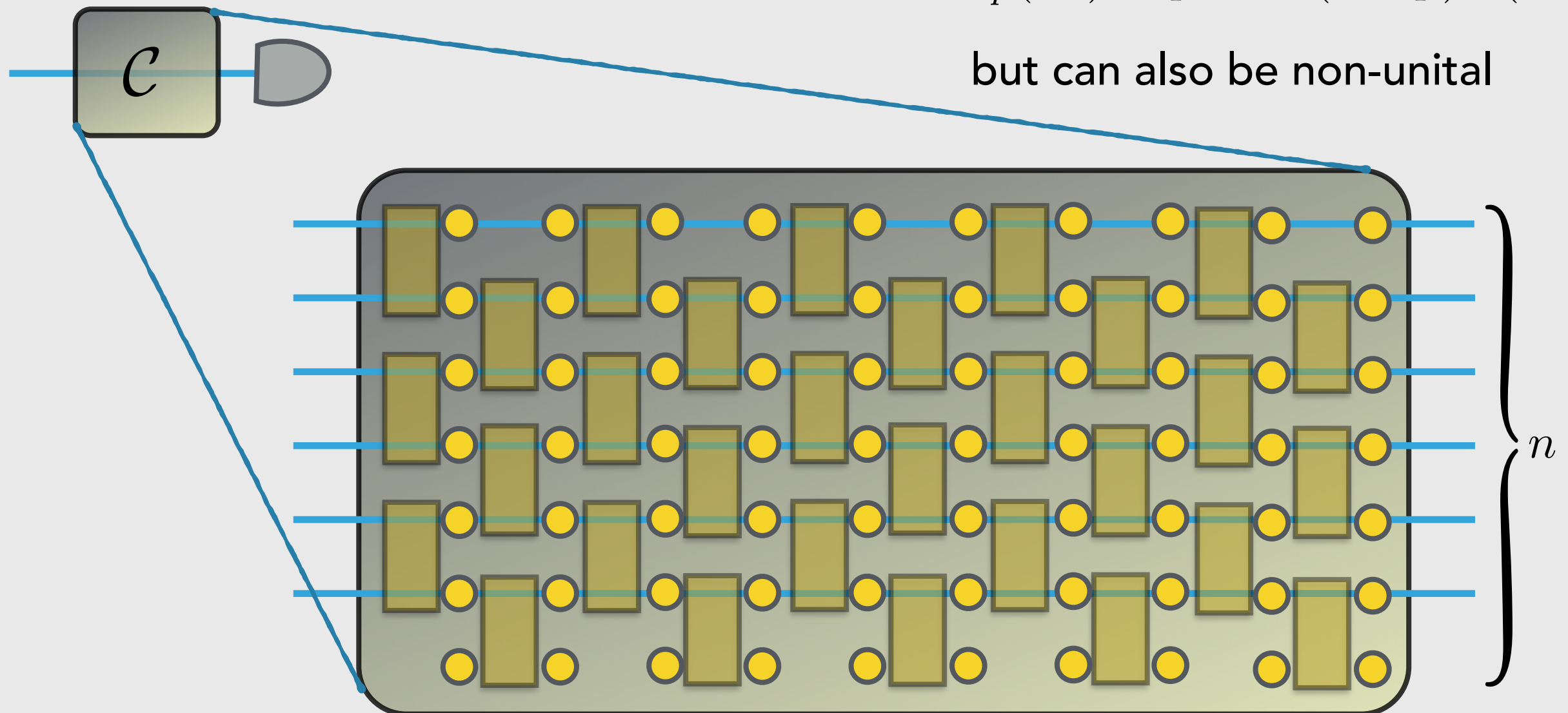


- Actual circuits are **noisy**

- Noise and decoherence,
e.g., **depolarizing**

$$\mathcal{D}_p(M) = pM + (1 - p)\text{tr}(M)\frac{\mathbb{I}}{2}$$

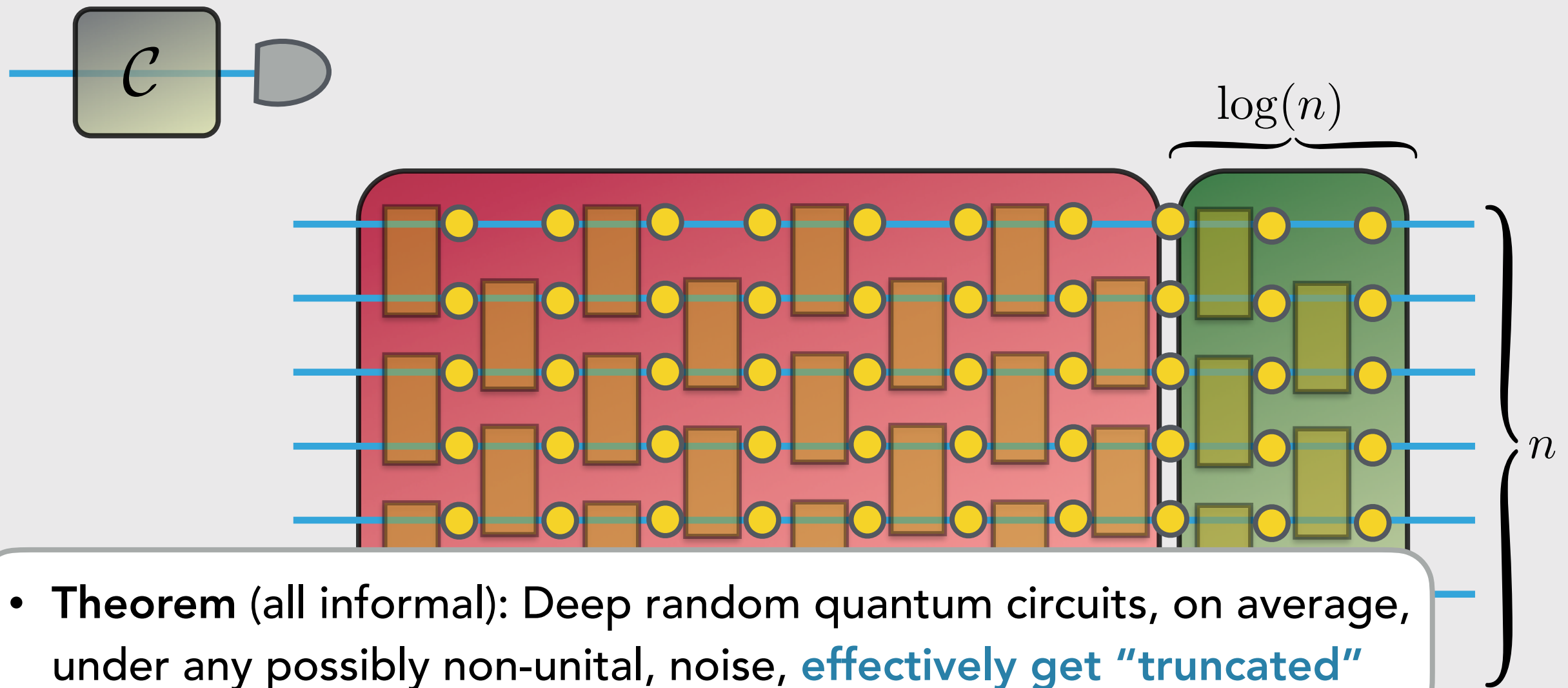
but can also be non-unital



EFFECTIVE LOGARITHMIC DEPTH



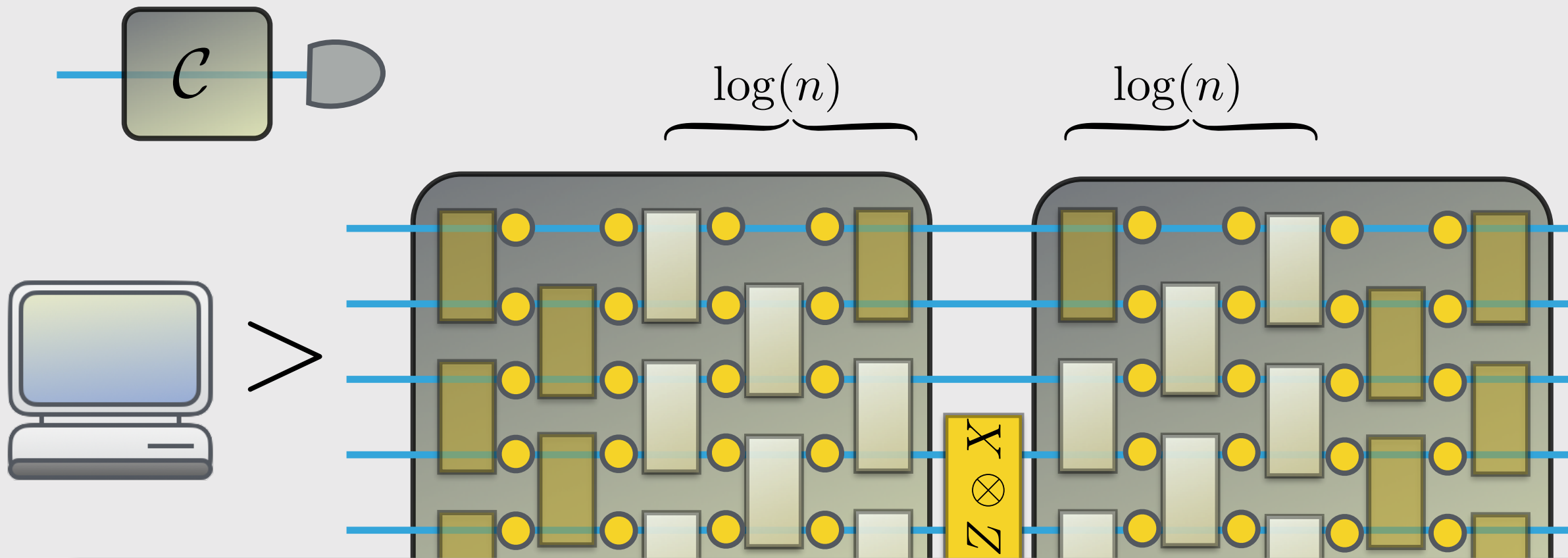
- Logarithmic depth is all we have



- **Theorem** (all informal): Deep random quantum circuits, on average, under any possibly non-unital, noise, **effectively get "truncated"**
- Influence of gates on Pauli expectation values decreases exponentially

Mele, Angrisani, Ghosh, Khatri, Eisert, Stilck França, Quek, arXiv:2403.13927 (2024)
Compare also Fawzi, Mueller-Herles, Shayeghi, ITCS (2022)

- Can often “dequantize”

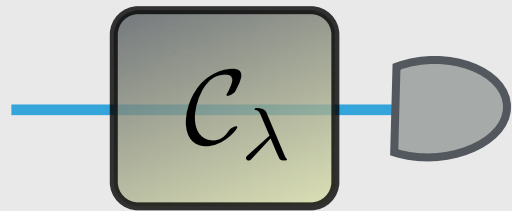


- **Theorem:** Can **classically simulate** on average expectation values observables to ε additive precision, with probability $1 - \delta$, at any depth, with runtime of $O(\exp(\log^D(\delta^{-1}\varepsilon^{-2})))$

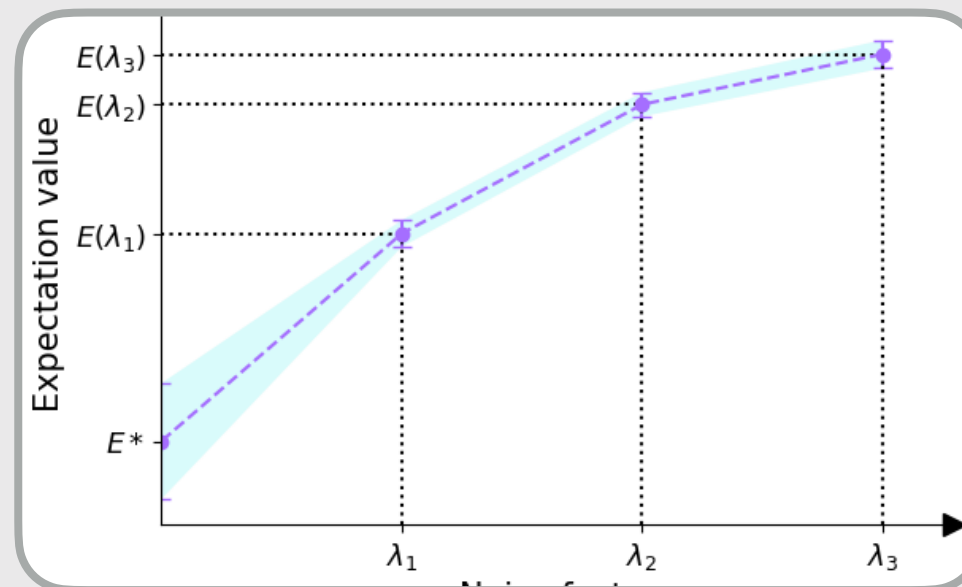
CAN QUANTUM ERROR MITIGATION HELP?



- Zero-noise extrapolation



- Add noise levels to C_λ
- Measure $E(\lambda) = \text{tr}(C_\lambda(\rho_{\text{in}})O)$
- Extrapolate to $\lambda = 0$

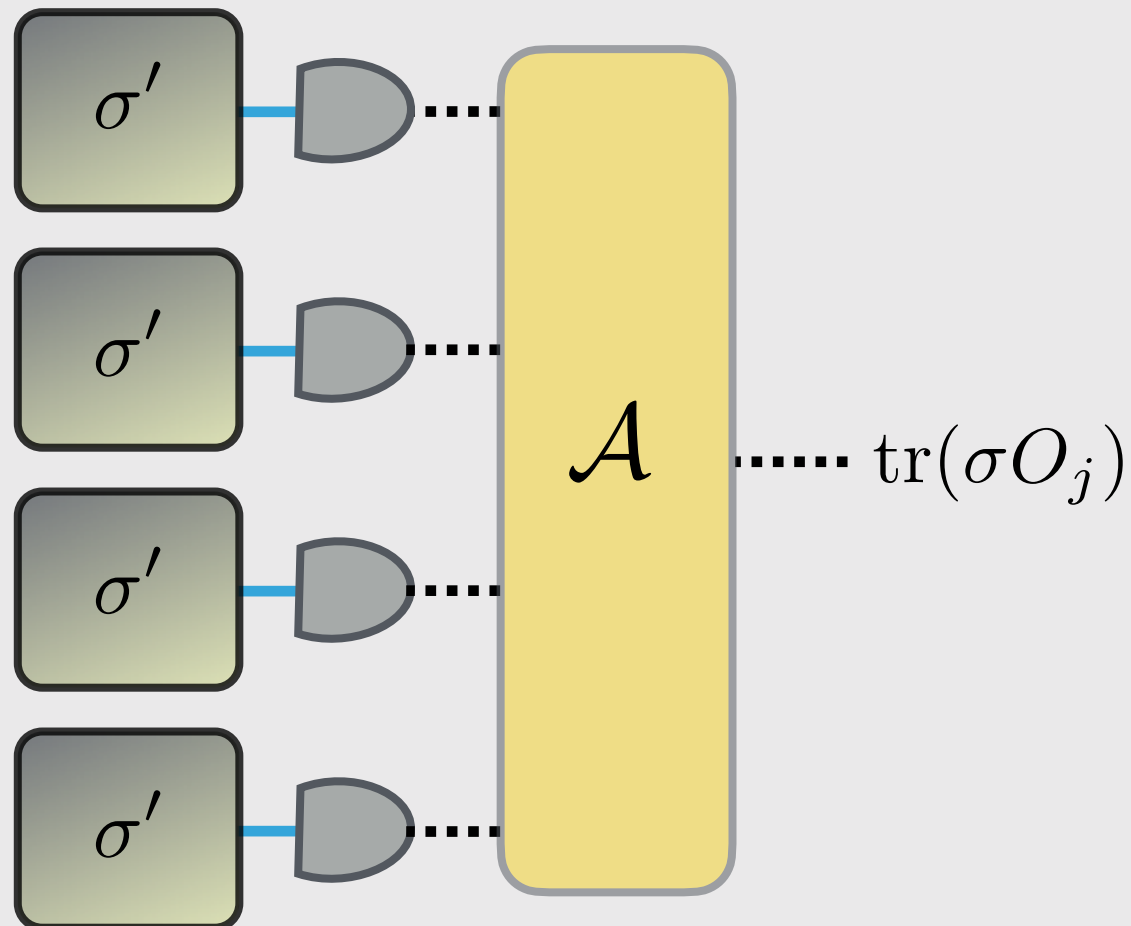


Majumdar, Rivero, Metz, Hasan, Wang, 2023 IEEE Int Conf Quant Comp Eng (2023)

Temme, Bravyi, Gambetta, Phys Rev Lett 119, 180509 (2017)

Cai, Babbush, Benjamin, Endo, Huggins, Li, McClean, O'Brien, arXiv:2210.00921 (2022)

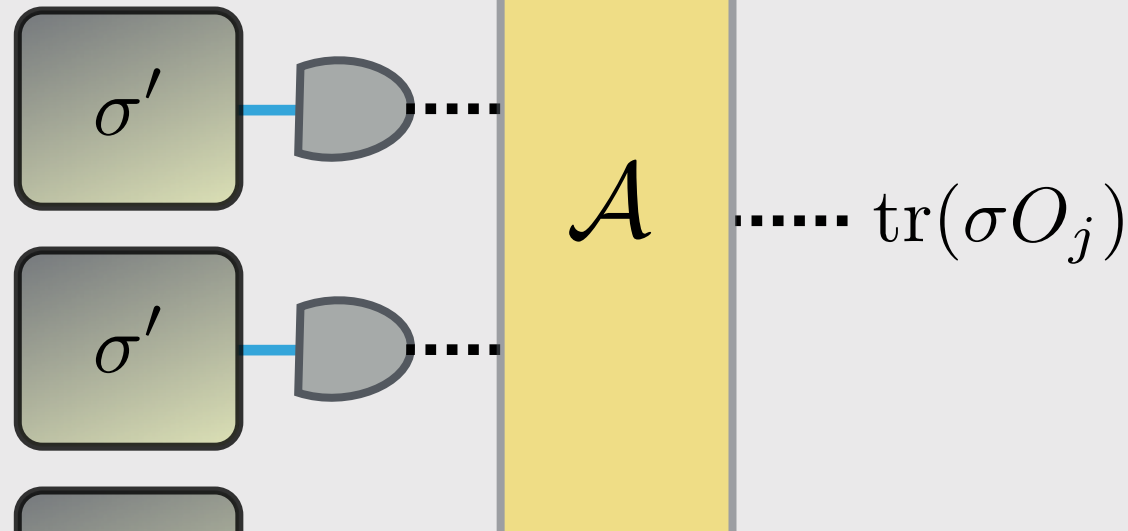
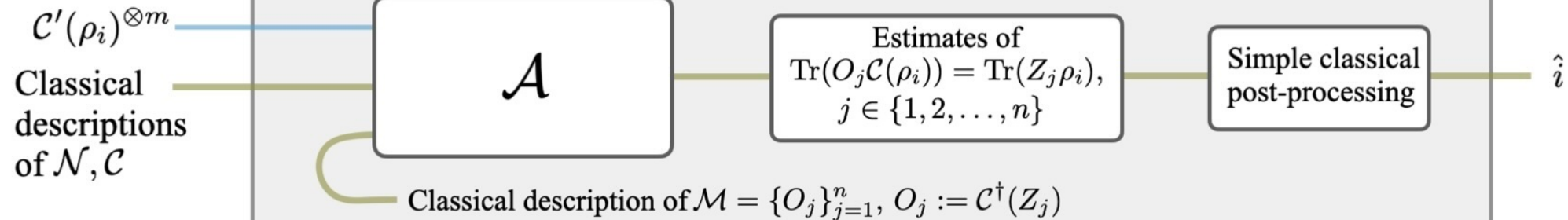
- How many copies of σ' are needed to estimate $\text{tr}(\sigma O_j)$ for $O_j \in \mathcal{M}$ to precision ε and probability $1 - \delta$?



- **Statistical inference problem**
 - in terms of circuit depth d
 - circuit width n
 - depolarizing noise strength p

Noisy state distinguishability

$$\rho_i \in \{|i\rangle\langle i|\}_{i=0}^{2^n-1} \cup \mathbb{I}/2^n$$

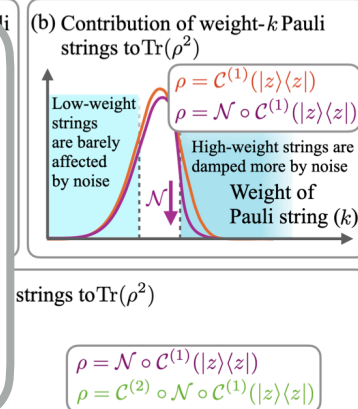


• Proof techniques

- Multi-hypothesis testing
- Fano's Lemma
- Log-depth Clifford circuits approximate unitary 2-designs

- Use **Fano's Lemma**: Any single-sample test distinguishing $N + 1$ distributions $P_1, \dots, P_N$ must fail with probability at least $1 - \alpha$,

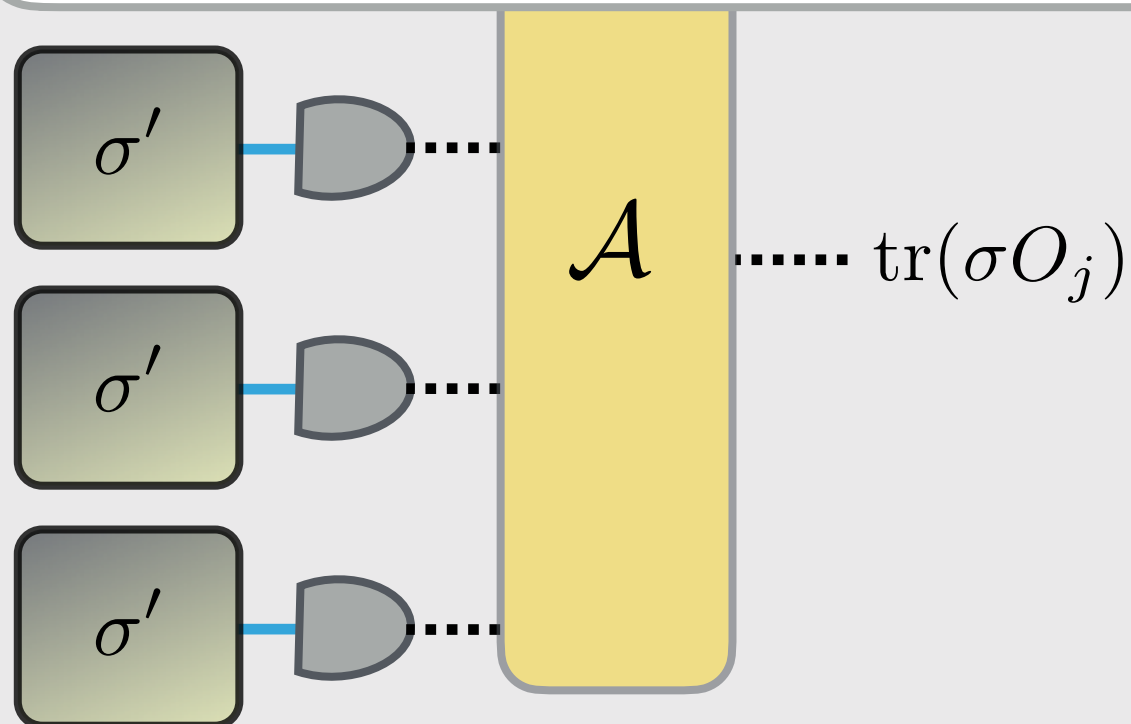
$$\frac{1}{\log(N)} \frac{1}{N + 1} \sum_{k=0}^N D(P_k || P_N) \leq \alpha$$



- **Theorem (informal):** One needs $\exp(\Omega(nd))$ **many samples**
- Previously thought: $\exp(\Omega(d))$, and since in NISQ regime $d = O(\log(n))$, our results are **exponentially** stronger

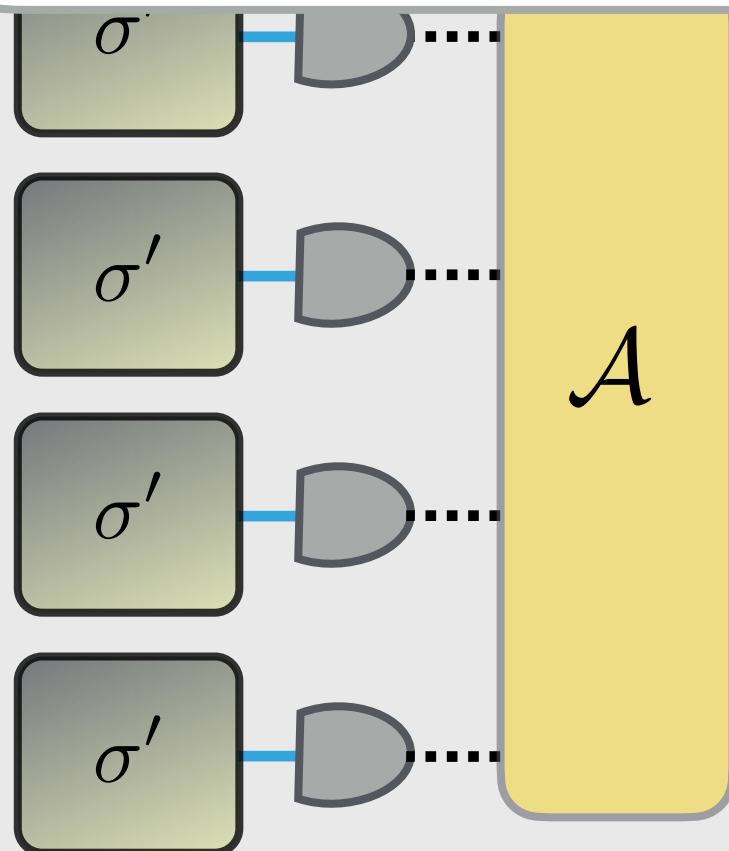
Takagi, Endo, Minagawa, Gu, npj Quant Inf 8, 114 (2022)

Takagi, Tajima, Gu, arXiv:2208.09178 (2022)





- There are **strong obstructions against error mitigation** already at log-log depth in worst case



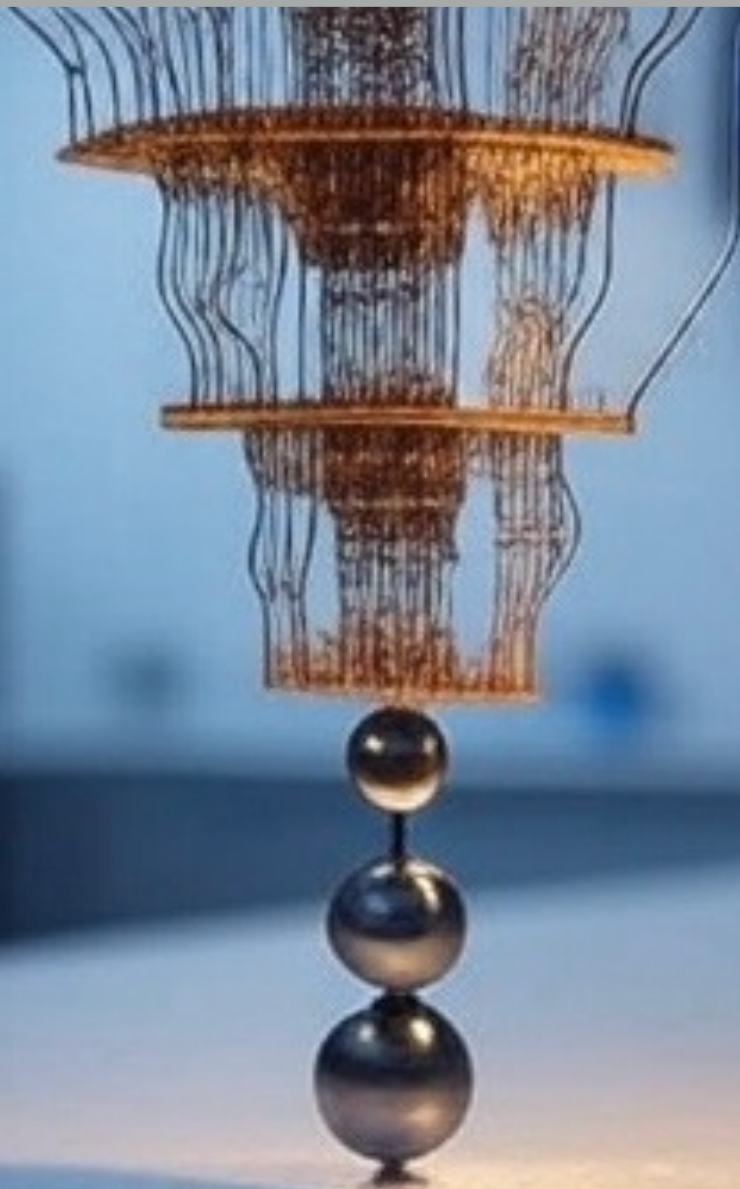
- Does **not** mean it does not work: It does not **scale** and not work well **for all circuits**

Compare also Gonzales-Garcia, Trivedi, Cirac, arXiv:2203.15632 (2022)

- Depolarizing noise catastrophic, non-unital noise allows **quantum refrigeration**

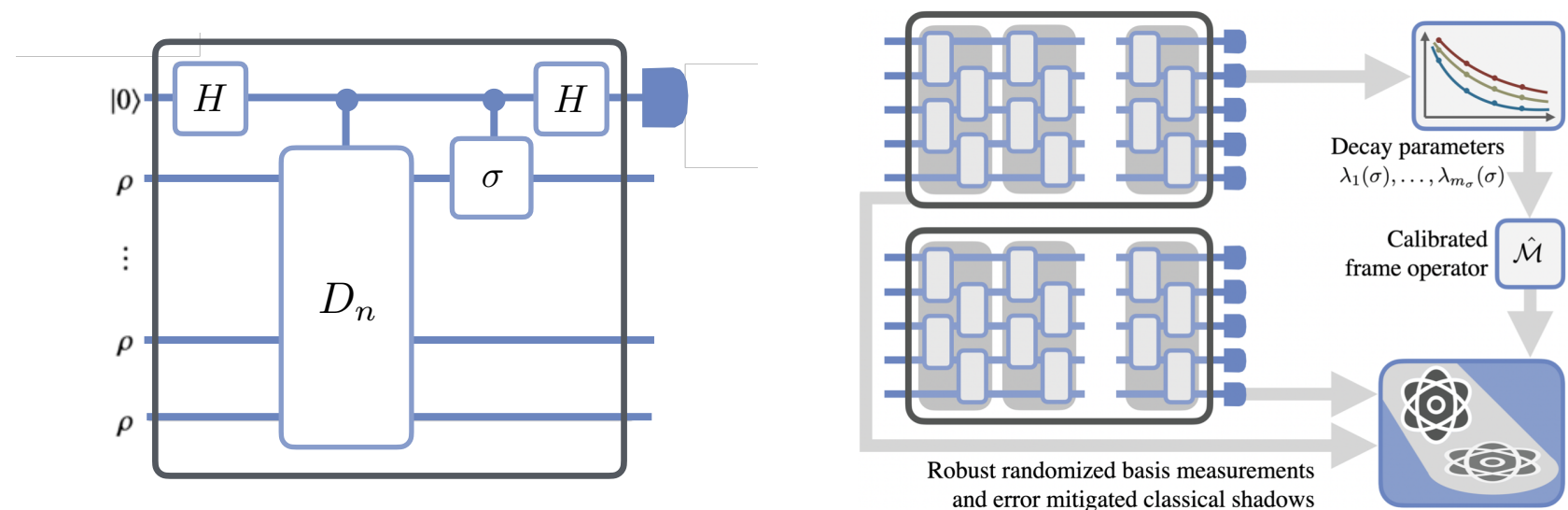
Ben-Or, Gottesman, Hassidim, arXiv:1301.1995 (2013)

- For small error rates, this allows for **reasonable depth** and QEM increases circuit volume: presently, common two-qubit average gate fidelities



- For small error rates, this allows for **reasonable depth** and QEM increases circuit volume: presently, common two-qubit average gate fidelities

- More intermediate steps: **coherent** quantum error mitigation



Onorati, Kitzinger, Helsen, Ioannou, Werner, Roth, Eisert, arXiv:2403.04751 (2024)

Seif, Cian, Zhou, Chen, Jiang, arXiv:2203.07309 (2022)

Koczor, Phys Rev X 11, 031057 (2021)

Huggins et al Phys Rev X 11, 041036 (2021)

- Steps **between** mitigation and fault tolerance, or virtual quantum channel purification - and will remain

Suzuki, Endo, Fujii Tokunaga, PRX Quantum 3, 010345 (2022)

Liu, Zhang, Fei, Cai, arXiv:2402.07866 (2024)



IS THERE HOPE FOR USEFUL NEAR-TERM ALGORITHMS?

arXiv:2502.14252 (2025)

arXiv:2505.04705 (2025)

arXiv:2505.15913, Phys Rev Lett, in press (2025)

arXiv:2411.15548 (2024)

Nature Comm 15, 9595 (2024)

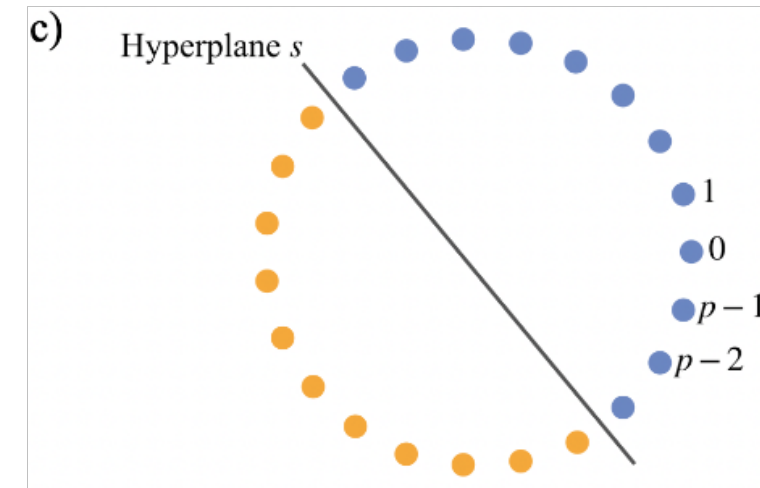
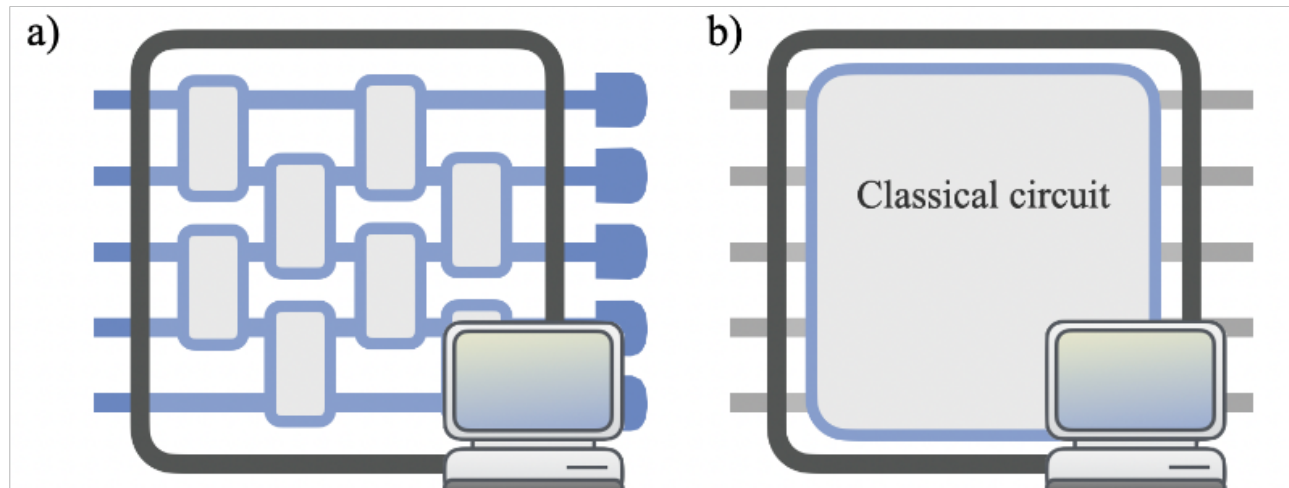


- Are there **constant depth** quantum advantages with a “practical feel”?

CONSTANT DEPTH ADVANTAGES?



- **Theorem:** There is a PAC learning advantage for **shallow quantum circuits**



- **Proof techniques**

- Construct concept classes from quantum advantages from shallow circuits

- **Features**

- Constant-depth quantum vs $\omega(\log \log(n))$ (NC^0) circuits
- Classical and artificial data
- Near-term quantum computer

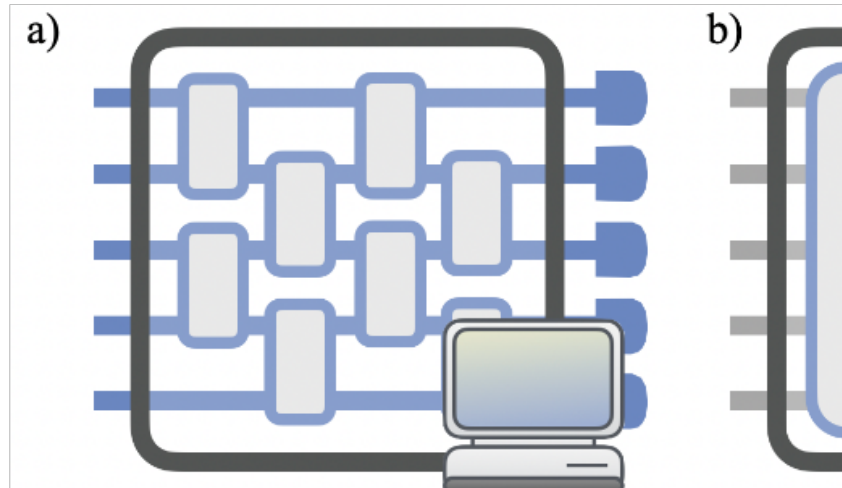
Pirnay, Jerbi, Seifert, Eisert, arXiv:2411.15548 (2024)

Compare Huang, Liu, Broughton, Kim, Anshu, Landau, McClean, arXiv:2401.10095 (2024)

Building on Bene Watts, Parham, arXiv:2301.00995 (2013)

Bravyi, Gosset, König, Science 362, 308 (2018)

- **Theorem:** There is a PAC learning advantage for **shallow quantum circuits**



- “Probably approximately correct” learning of distribution classes
- A distribution class $\mathcal{C}$ is efficiently PAC learnable w.r.t. distance d if there is an algorithm $\mathcal{A}$ which for every $D \in \mathcal{C}$ and every $\epsilon, \delta > 0$ given access to an oracle $O(D)$, outputs in time $\text{poly}(|D|, 1/\epsilon, 1/\delta)$
- with probability at least $1 - \delta$ (“probably”) a generator $\text{GEN}_{D'}$
- of a distribution D' such that
$$d(D, D') < \epsilon$$
(“approximately correct”)

- **Proof techniques**

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- **Features**

- Constant-depth quantum vs $\omega(\log \log(n))$ (NC^0) circuits
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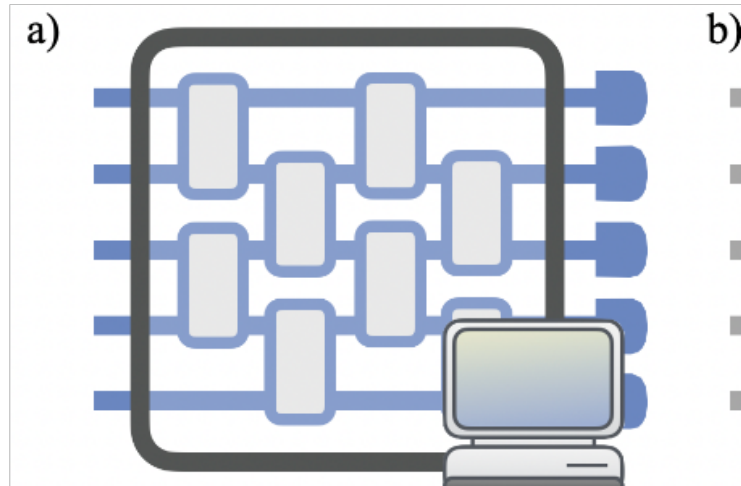
Pirnay, Jerbi, Seifert, Eisert, arXiv:2411.15548 (2024)

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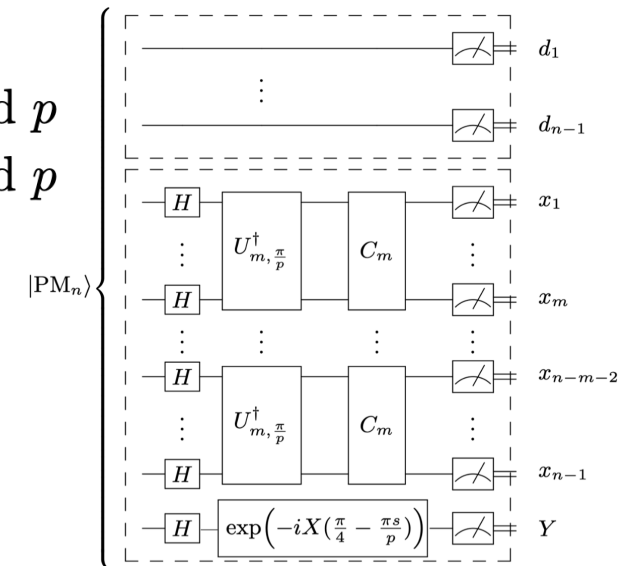
Bravyi, Gosset, König, Science 362, 308 (2018)

- **Theorem:** There is a PAC generator learning advantage for **shallow quantum circuits**



- **Core idea:** Devise a PAC generator learning advantage from an unconditional sampling advantage of QNC^0 over NC^0
- Encode hyperplane learning problem into the "majority mod p "

$$\text{majmod}_{p,s}(x) = \begin{cases} 0, & \text{if } |x| + s < p/2 \pmod{p} \\ 1, & \text{if } |x| + s > p/2 \pmod{p} \end{cases}$$



Building on Bene Watts, Parham, arXiv:2301.00995 (2013)
Bravyi, Gosset, König, Science 362, 308 (2018)

• Proof techniques

- Construct concept classes from quantum advantages from shallow circuits
- Constant-depth quantum vs $\omega(\log \log(n))$ (NC^0) circuits
- Classical and artificial data
- Near-term quantum computer

- **Lesson:** Think of first applications beyond sampling

Jerbi, Seifert, Eisert, arXiv:2411.15548 (2024)

Compton Huang, Liu, Broughton, Kim, Anshu, Landau, McClean, arXiv:2401.10095 (2024)

Building on Bene Watts, Parham, arXiv:2301.00995 (2013)

Bravyi, Gosset, König, Science 362, 308 (2018)



- Are there **simple learning algorithms**?
- Identifying the **symmetry properties** of quantum states
- **Definition** [State hidden subgroup problem] Let G be a finite group with a unitary representation $R : G \rightarrow \mathcal{H}$ acting on the Hilbert space $\mathcal{H}$ and let $H \leq G$ be a subgroup of G . Assume you have access to copies of an unknown state vector $|\psi\rangle \in \mathcal{H}$ that is promised to have the properties

$$1. \forall h \in H, \quad R(h) |\psi\rangle = |\psi\rangle \quad \square$$

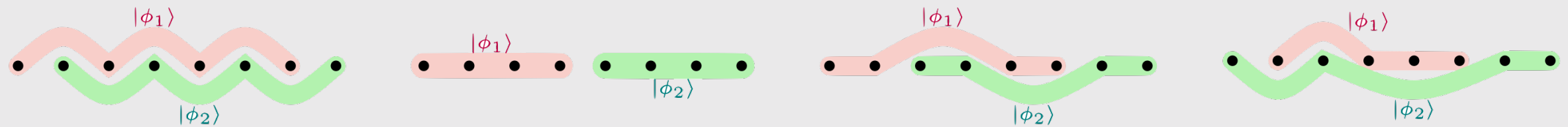
$$2. \forall g \notin H, \quad |\langle \psi | R(g) | \psi \rangle| \leq 1 - \epsilon \quad \square$$

The problem is to identify H

Bouland, Giurgica-Tiron, Wright, arXiv:2410.12706 (2024)



- Their application: Find “hidden cuts” along which pure states are unentangled



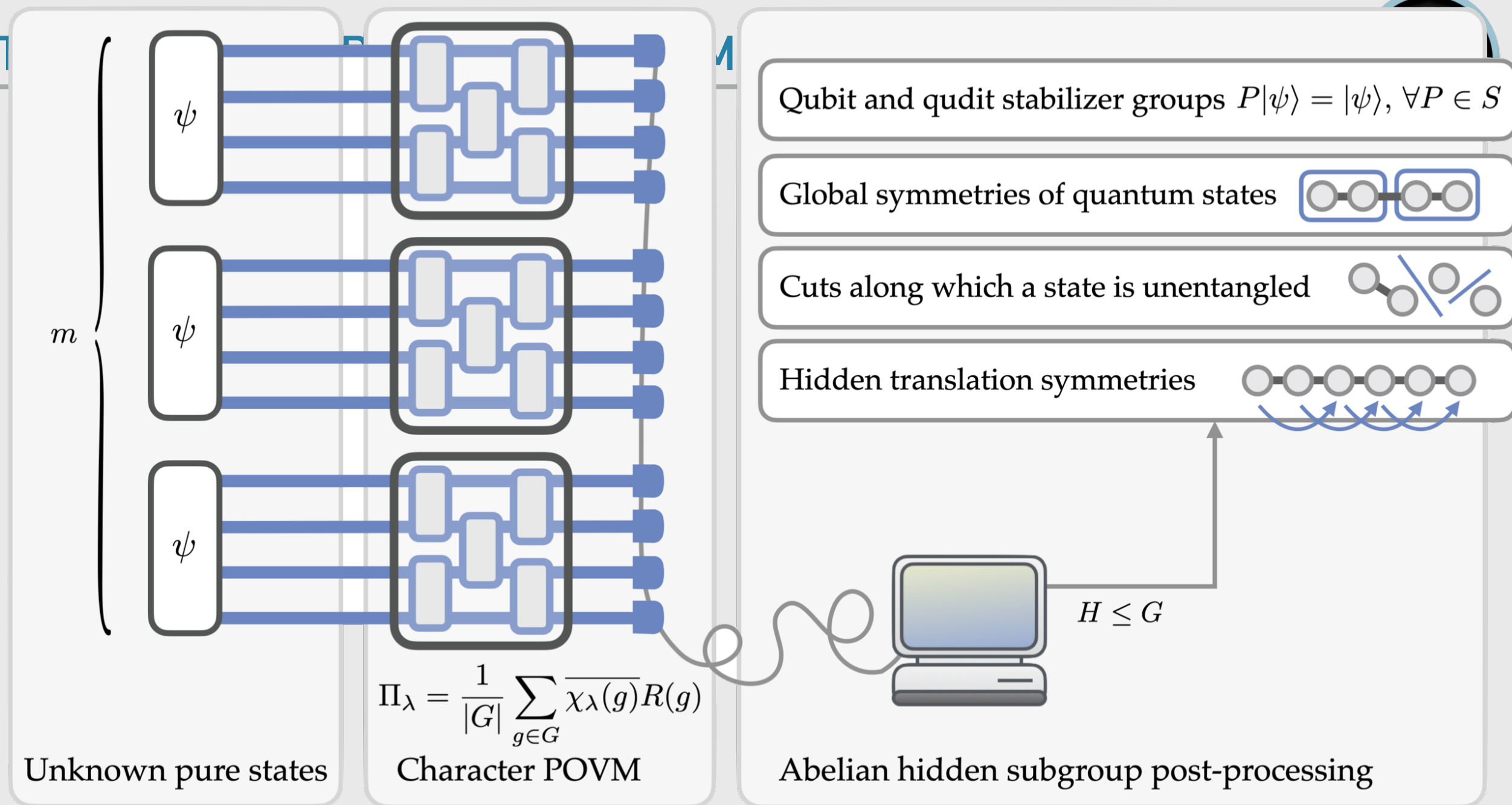
- **Our contribution:** Understand representation theory of problem and define “character POVM”

$$\Pi_{\lambda} = \frac{1}{|G|} \sum_{g \in G} \overline{\chi_{\lambda}(g)} R(g)$$

Hinsche, Eisert, Carrasco, arXiv:2505.15770 (2025)
 Perez-Salinas, Hinsche, Eisert, Carrasco,
 in preparation (2025)

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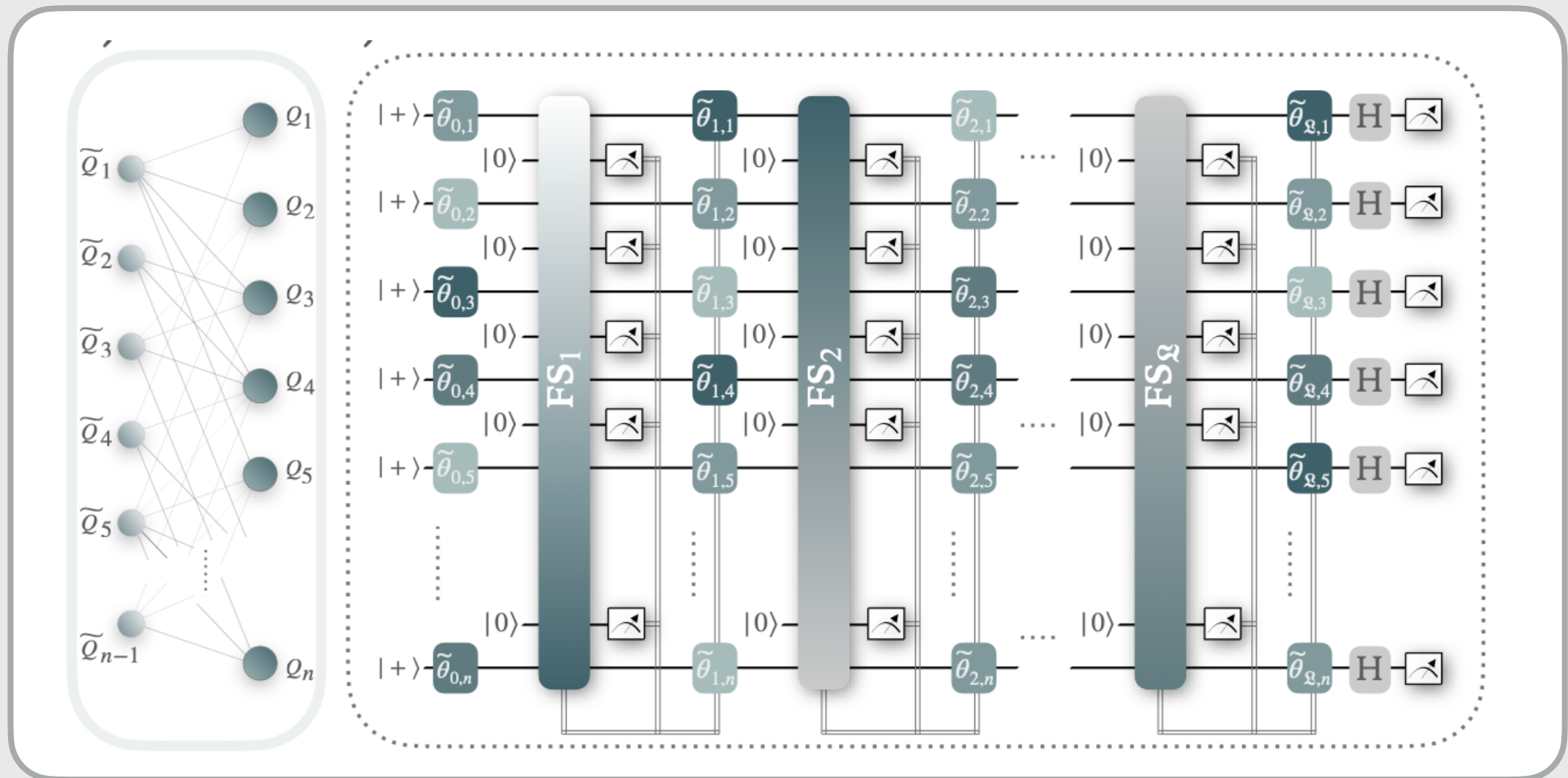
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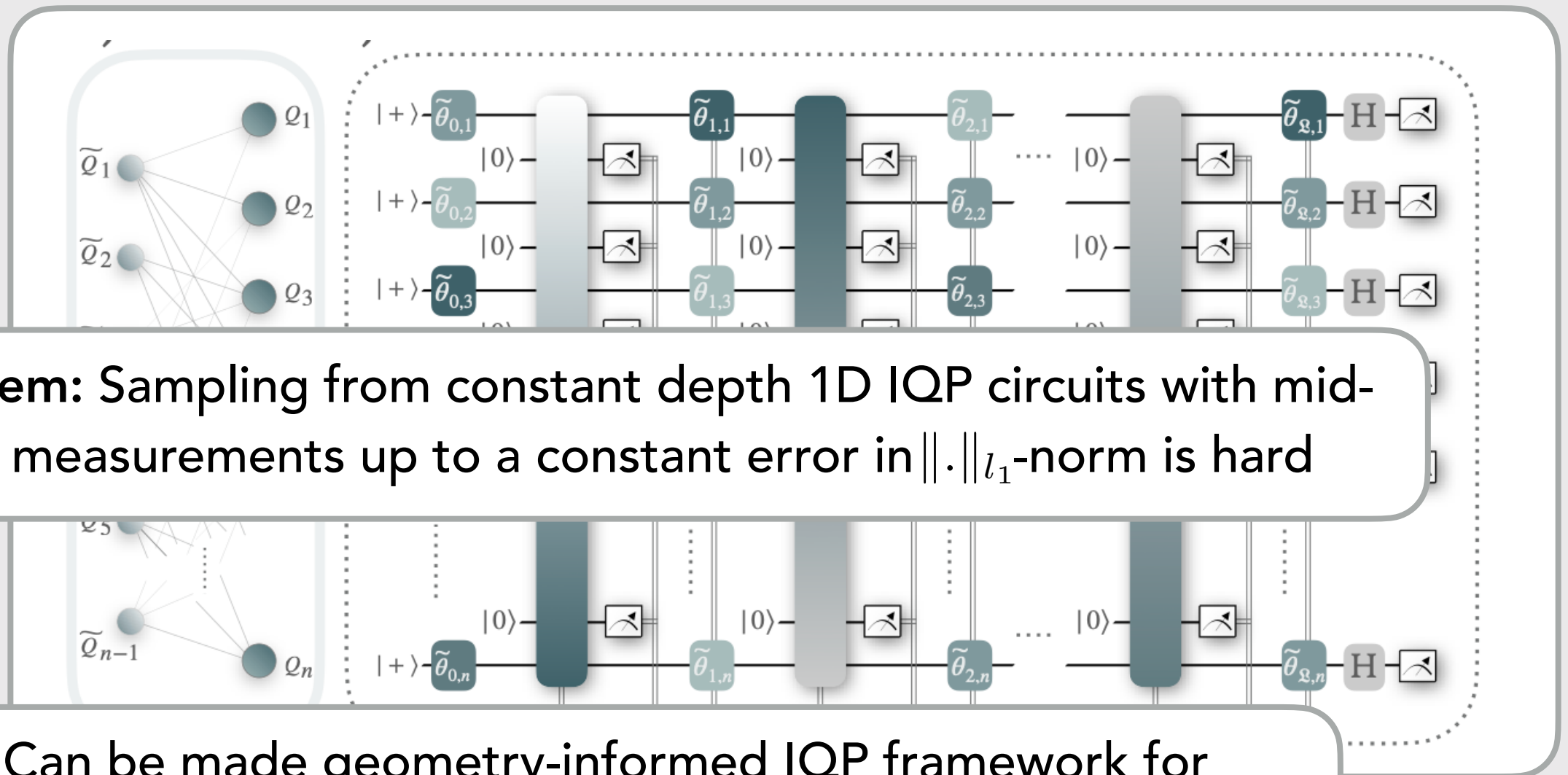
- **Lesson:** Is relatively near-term

Hinsche, Eisert, Carrasco, arXiv:2505.15770 (2025)
Perez-Salinas, Hinsche, Eisert, Carrasco,
in preparation (2025)

- Can **mid-circuit measurements** help?



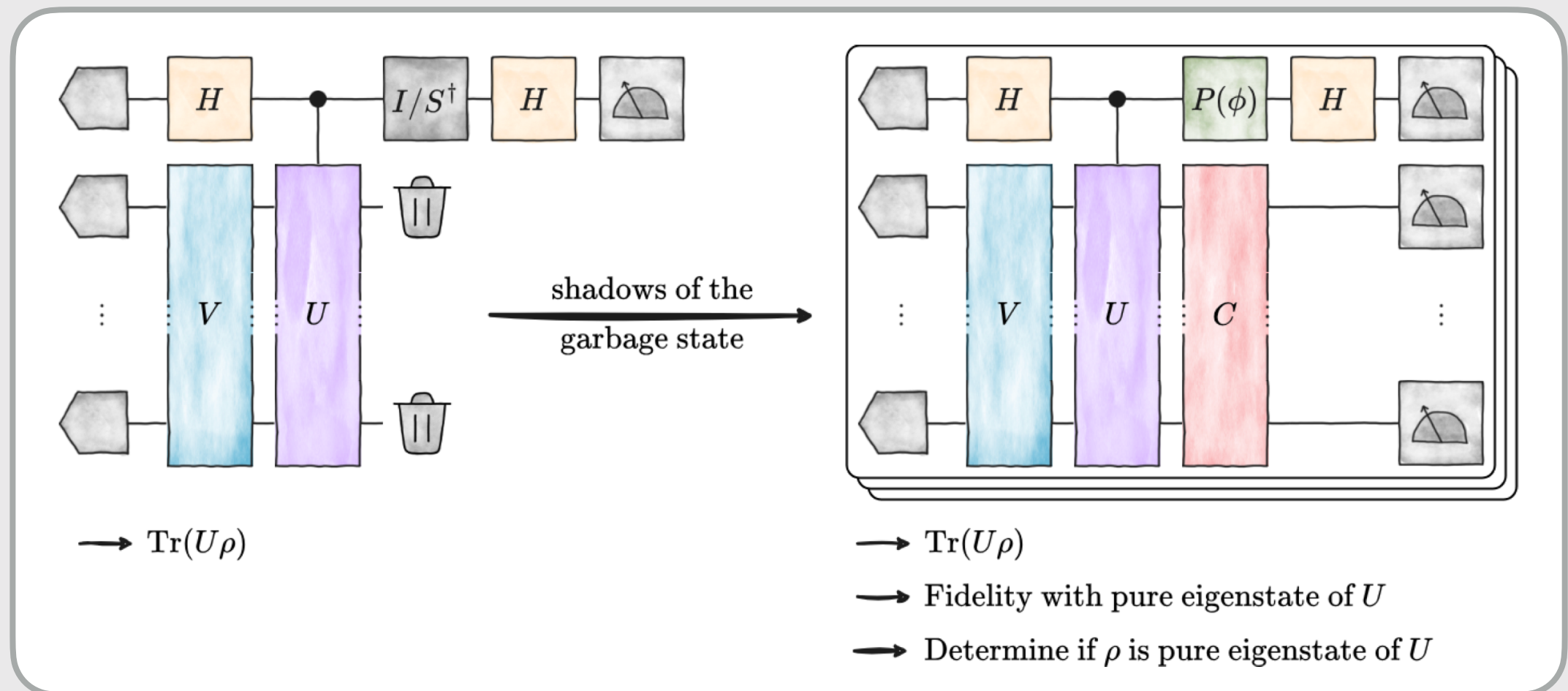
- Can **mid-circuit measurements** help?



- **Theorem:** Sampling from constant depth 1D IQP circuits with mid-circuit measurements up to a constant error in $\|\cdot\|_{l_1}$ -norm is hard

- **Lesson:** Can be made geometry-informed IQP framework for learning ground states - **mid-circuit measurements** seem to help

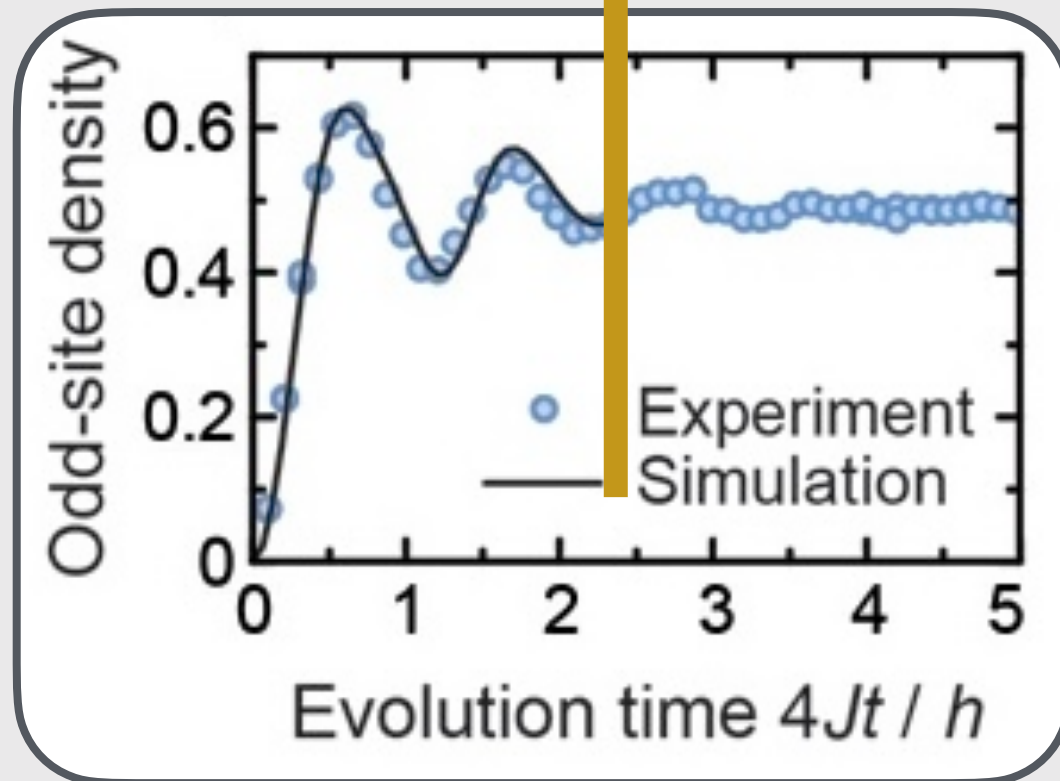
- Can we improve quantum algorithmic **primitives**?



- Lesson:** Hadamard test on a single auxiliary readout qubit with classical shadows on the remaining n -qubit work register improves primitive

- Can quantum simulation provide first **quantum utility**?

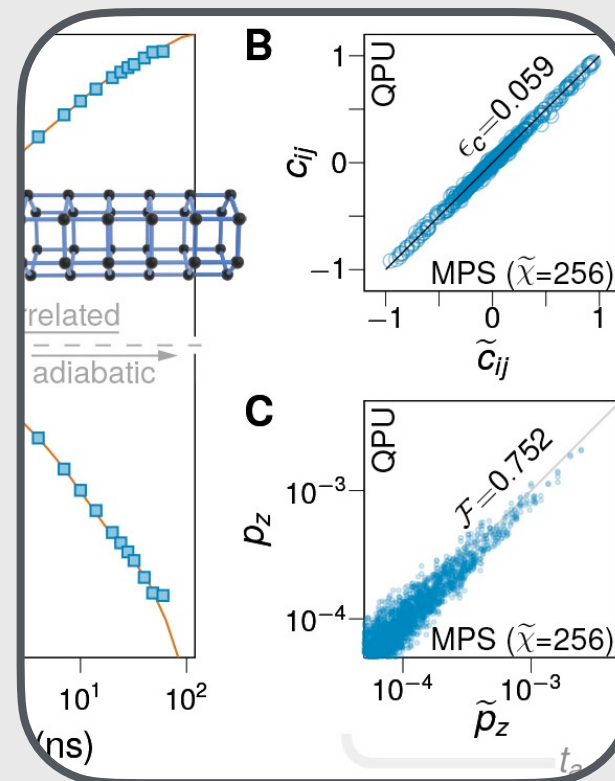
- Quenched ultra-cold atoms



vs matrix product states

Trotzky, Chen, Flesch, McCulloch, Schollwoeck, Eisert, Bloch, Nature Phys 8, 325 (2012)

- Annealers

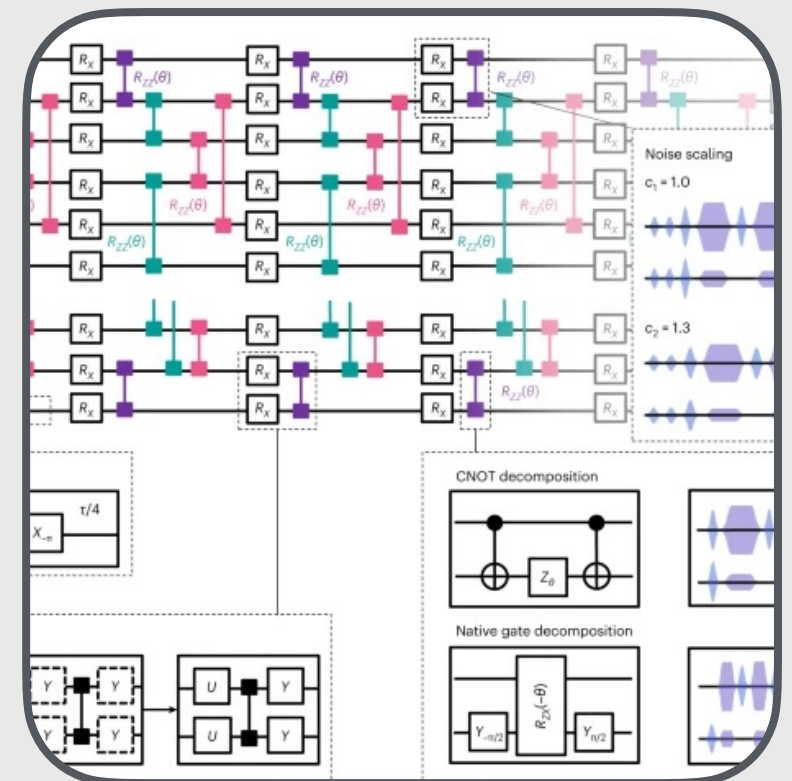


vs tensor networks, belief prop, neural network states

King et al, Science (2025)

Tindall, Mello, Fishman, Stoudenmire, Sels, arXiv:2503.05693 (2025)

- Superconducting circuits



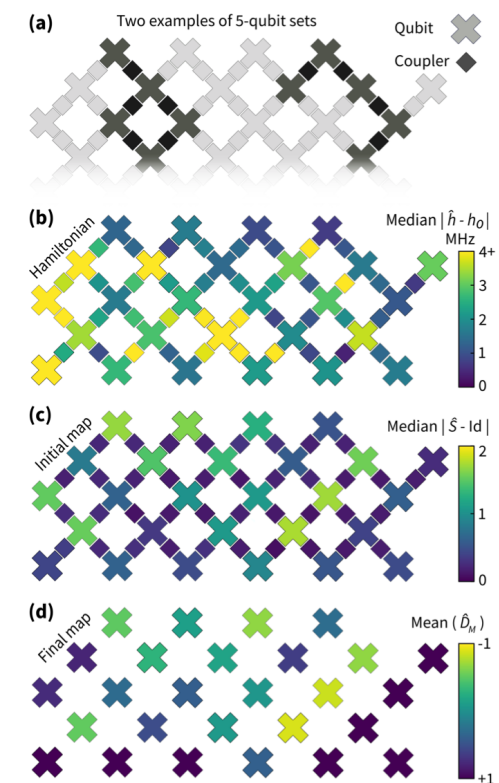
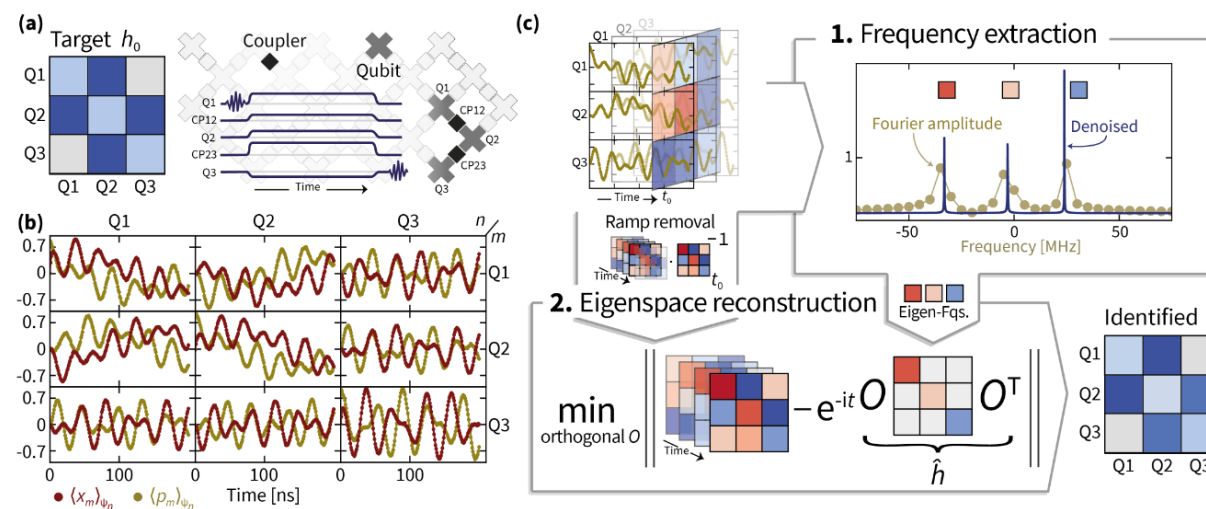
Kim et al, Nature, 618, 500–5 (2023)

Tindall, Fishman, Stoudenmire, Sels, PRX Quantum 5, 010308 (2024)

- Sweet spot** provided by **two-dimensional systems** in **non-equilibrium**

QUANTUM UTILITY IN QUANTUM SIMULATION?

- Hamiltonian and Liouvillian learning**



- Lesson:** Such techniques can increase predictive power



CODA

arXiv:2502.14252 (2025)

Science Advances 10, eadj5170 (2024)

Nature Communications 15, 434 (2024)

arXiv:2411.15548 (2024)

Quantum 5, 417 (2021)



- **Coda: Near-term** quantum advantages are not easy - constant factors matter



- **Coda:** Only part of the issue - need **robust advantages** for families of instances





- **Coda:** Only part of the issue - need **robust advantages** for families of instances

- **Our results on optimization:** • Approximate classically hard instances of integer programming

Pirnay, Ulitzsch, Wilde, Eisert, Seifert, arXiv:2212.08678, Science Advances 10, eadj5170 (2024)

- **Result by Google AI: Decoded quantum interferometry**

- Find $\mathbf{x} \in \mathbb{F}_2^n$ satisfying as many as possible among the m linear mod-2 equations $B\mathbf{x} = \mathbf{v}$, with $B \in \mathbb{F}_2^{m \times n}$, $\mathbf{v} \in \mathbb{F}_2^m$
- Approximate solutions of optimization by reduction to **decoding** of classical codes

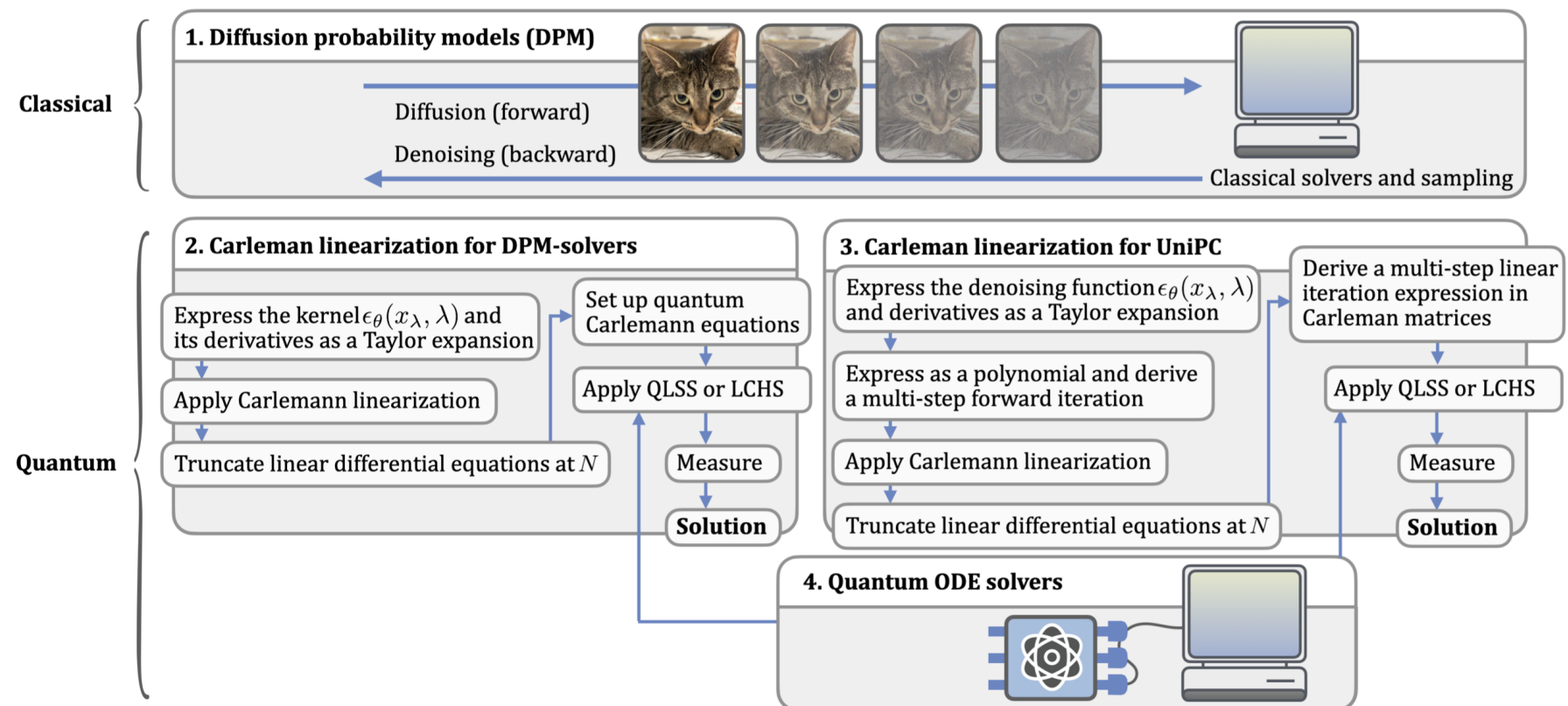
- **Max-k-XORSAT**

Jordan, Shutty, Wootters, Zalcman, Schmiedhuber, King, Isokov, Babbush, arXiv:2408.08292 (2024)

- Question to audience

- **Coda:** Only part of the issue - need **robust advantages** for families of instances

- Quantum **probability diffusion models**



Wang, Jiang, Fan, Jia, Eisert, Liu, Liu, arXiv:2502.14252 (2025)

Liu, Liu, Liu, Ye, Wang, Alexeev, Eisert, Jiang, Nature Communications 15, 434 (2024)



- **Coda:** As long as we stay away from the hype and take step by step, all will be well

Mind the gap: The fraught road to quantum advantage

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²Helmholtz-Zentrum Berlin für Materialien und Energie, 14109 Berlin, Germany

³Fraunhofer Heinrich Hertz Institute, 10587 Berlin, Germany

⁴Department of Computer Science, Virginia Tech, Alexandria, VA 22314, USA

⁵Phasecraft Inc., Washington, DC 20001, USA

⁶Institute for Quantum Information and Matter, California Institute of Technology, CA 91125, USA

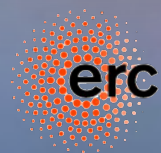
⁷QIS Center for Quantum Computing, Pasadena, CA 91125, USA



- **Coda:** Can we reasonably hope realistic quantum devices to provide a **speedup over classical computers**?

Definitely Maybe

THANKS FOR YOUR ATTENTION



Advanced
Grant

POSTDOCS AVAILABLE