

# Classical Shadow Tomography with Locally Scrambled Quantum Dynamics

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[1] H.-Y. Hu, YZY. arXiv:2102.10132

[2] H.-Y. Hu, S. Choi, YZY. arXiv:2107.04817

[3] A. Akhtar, H.-Y. Hu, YZY. arXiv:2209.02093

[4] A. Akhtar, H.-Y. Hu, YZY. arXiv:2308.01653

*PQTC, Princeton, Oct. 2023*

# Acknowledgement

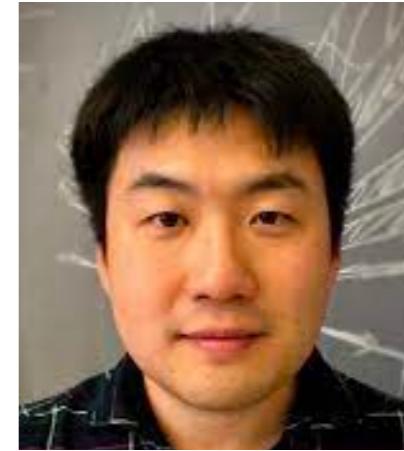
- Collaborators



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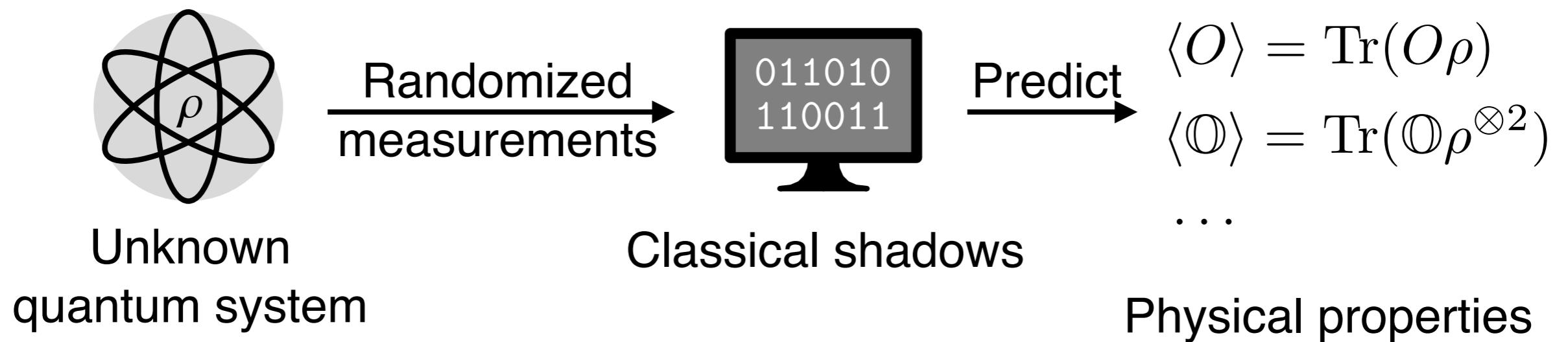
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(MIT)

- UC Hellman Fellow (2021)



# Quantum State Tomography

- Quantum state tomography: using repeated measurements to extract information about a quantum system.
- **Classical shadow tomography:** a general-purpose tomography scheme with **superior sample efficiency**



- Key idea: use **randomized measurement** to sample classical shadows  $\hat{\sigma}$ , without reconstructing  $\rho$  explicitly.

Huang, Kueng, Preskill, Nat. Phys. arXiv:2002.08953 (2020)

Elben et.al. *The randomized measurement toolbox*, Nat. Rev. Phys. (2022)

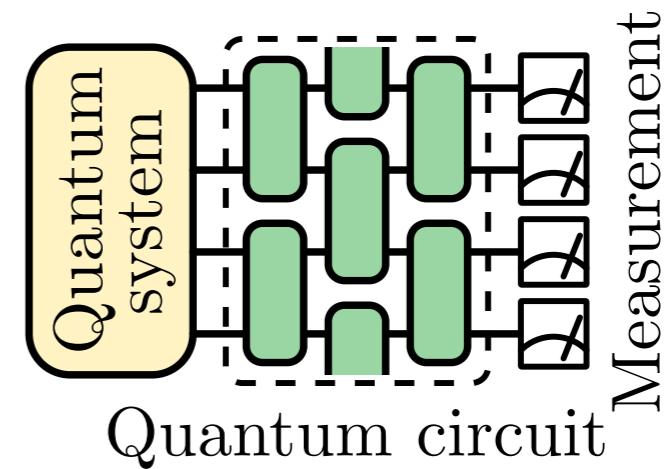
# Randomized Measurement

- Randomized measurement protocol
  - Prepare  $\rho_{\text{full}} = \rho \otimes (|0\rangle\langle 0|)^{\otimes N_{\text{anc}}}$
  - Pick  $U \sim p(U)$ , perform
$$\rho_{\text{full}} \rightarrow \rho'_{\text{full}} = U \rho_{\text{full}} U^\dagger$$
  - Measure a subset of qubits in the computational ( $Z$ ) basis

$$\rho'_{\text{full}} \rightarrow \frac{\Pi_b \rho'_{\text{full}} \Pi_b}{\text{Tr}(\Pi_b \rho'_{\text{full}} \Pi_b)}$$

with measurement outcome  $b \in \{0, 1\}^{\times N_{\text{msr}}}$

- Record the **measurement event**  $(U, b)$ , then repeat
- Goal: predict properties of  $\rho$  from  $\{(U, b)\}$



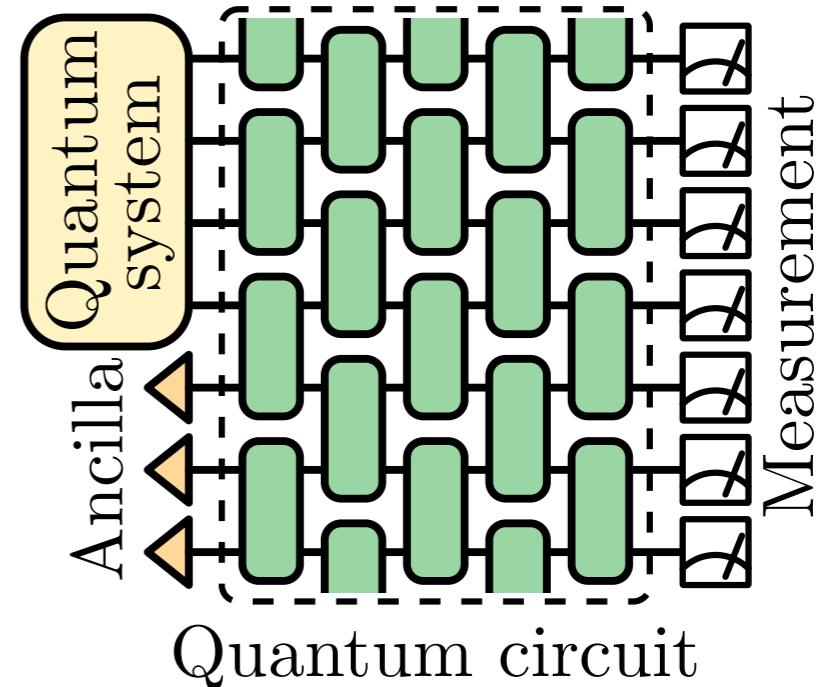
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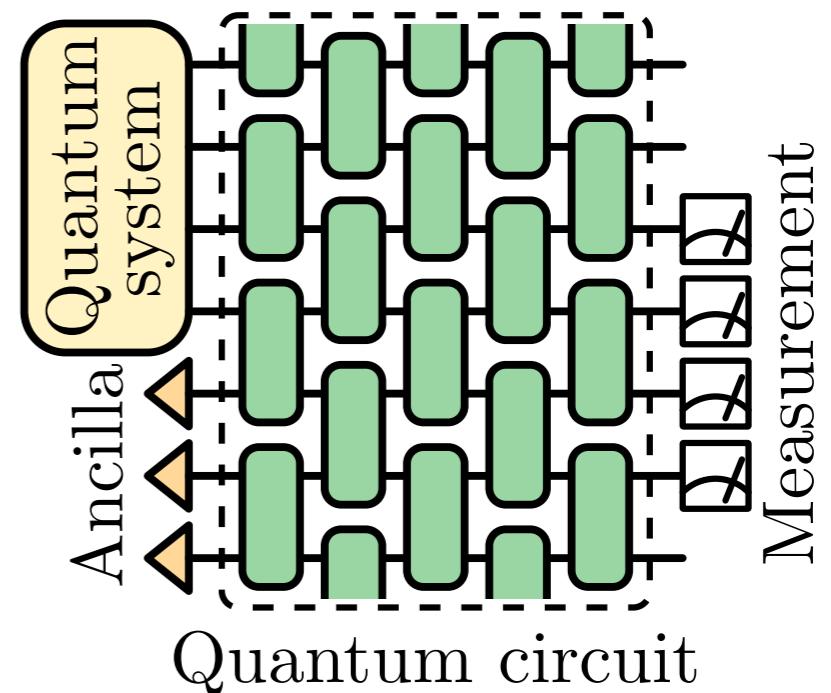
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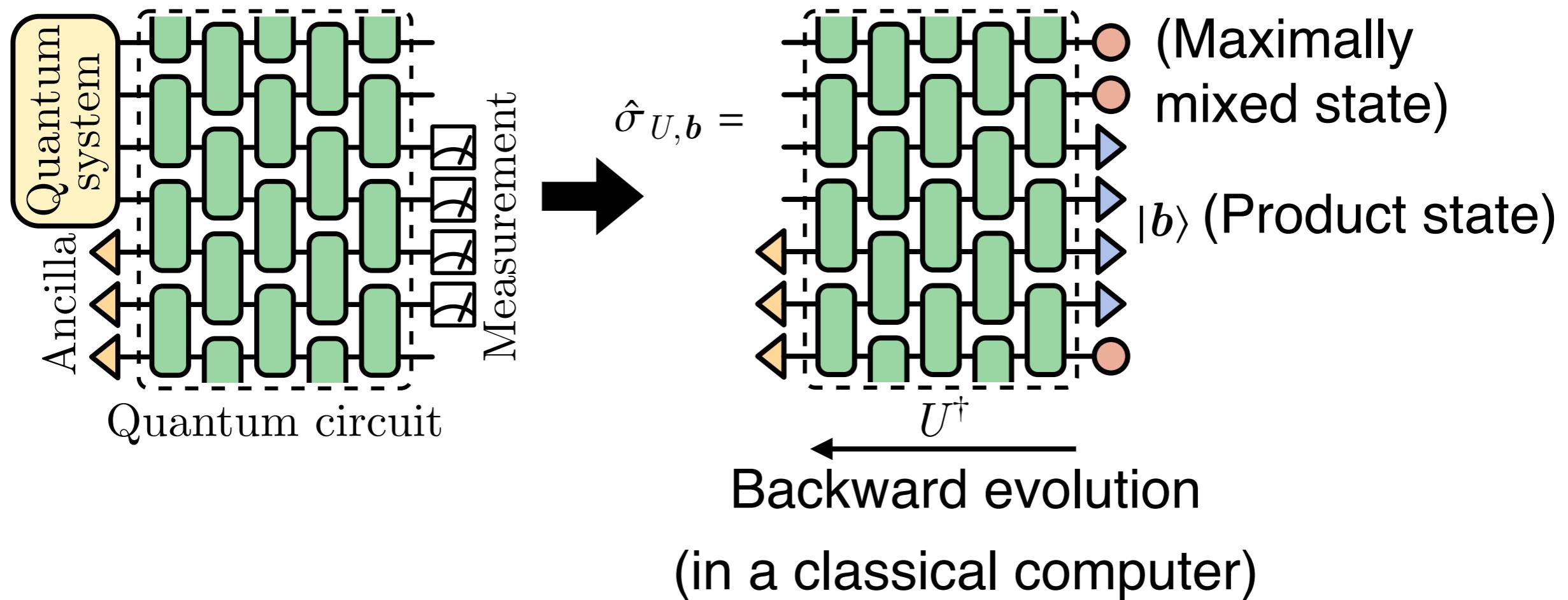
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# Randomized Measurement

- Each randomized measurement event  $(U, b)$  is characterized by a **classical shadow** state  $\hat{\sigma}_{U,b}$ , such that

$$p(\hat{\sigma}_{U,b} | \rho) \propto \text{Tr}(\hat{\sigma}_{U,b} \rho)$$



# Measurement Channel

- **Measurement channel:** measure  $\rho$  prepare  $\hat{\sigma}$

$$\rho \rightarrow \sigma := \mathbb{E}_{\hat{\sigma} \sim p(\hat{\sigma}|\rho)} \hat{\sigma} = \mathcal{M}[\rho]$$

- $\mathcal{M}$  is invertible, if the scheme is **tomographically complete**
- **Reconstruction map:** reconstruct  $\rho$  from  $\hat{\sigma}$

$$\rho = \mathcal{M}^{-1}[\sigma] = \mathbb{E}_{\hat{\sigma} \sim p(\hat{\sigma}|\rho)} \mathcal{M}^{-1}[\hat{\sigma}]$$

- $\mathcal{M}^{-1}$  is *not* a physical quantum process, and can only be implemented by classical computing.
- It specifies how to post-process the classical data to make predictions, e.g.

$$\langle O \rangle_\rho = \text{Tr } O \rho = \mathbb{E}_{\hat{\sigma} \sim p(\hat{\sigma}|\rho)} \text{Tr}(O \mathcal{M}^{-1}[\hat{\sigma}])$$

# Operator Shadow Norm

- **Sample complexity:** the number  $M$  of samples needed to control the estimation error within the level of  $\epsilon$ ,

$$M \sim \frac{1}{\epsilon^2} \|O\|_{\mathcal{E}_\sigma}^2 \quad \text{Huang, Kueng, Preskill (2020)}$$

- The scaling  $\epsilon \sim 1/\sqrt{M}$  follows the large number theorem
- The coefficient is set by the **operator shadow norm**

$$\begin{aligned} \|O\|_{\mathcal{E}_\sigma}^2 &:= \mathbb{E}_{\hat{\sigma} \sim p(\hat{\sigma})} \left( \mathrm{Tr}(O\mathcal{M}^{-1}[\hat{\sigma}]) \right)^2 \\ &= \frac{\mathrm{Tr}(O\mathcal{M}^{-1}[O])}{\mathrm{Tr} \mathbf{1}} \end{aligned}$$

which depends on

- The observable  $O$  of interest
- The randomized measurement channel  $\mathcal{M}$

# Pauli and Clifford Measurements

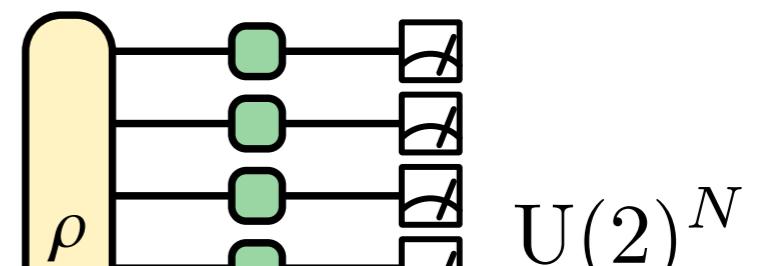
- Everything boils down to computing  $\mathcal{M}^{-1}$ .

- Know results

- Randomized Pauli measurement

$$\mathcal{M}^{-1}[\sigma] = \bigotimes_i (3\sigma_i - \mathbb{1}_i)$$

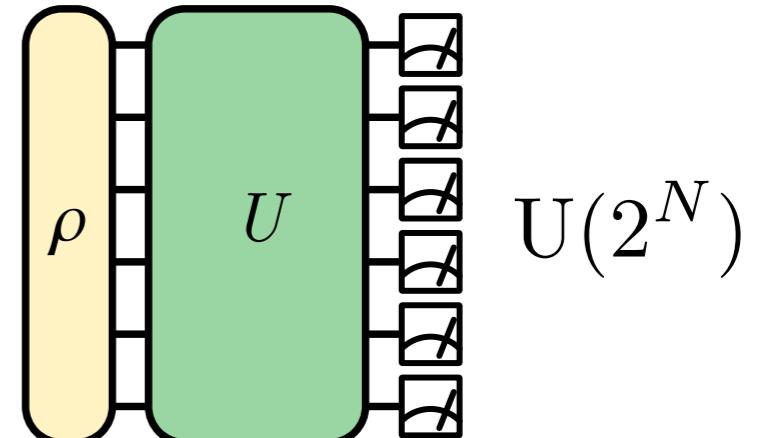
$$\|O\|_{\mathcal{E}_\sigma}^2 = 3^{|\text{supp } O|}$$



- Randomized Clifford measurement

$$\mathcal{M}^{-1}[\sigma] = (2^N + 1)\sigma - \mathbb{1}$$

$$\|O\|_{\mathcal{E}_\sigma}^2 \simeq \text{Tr } O^2$$



- What about other more general randomized unitary ensemble (beyond Haar measure of unitary groups)?

# Locally-Scrambled Shadow Tomography

- Much progress can be made by a mild assumption: the **prior** distribution of classical shadows is **local basis independent**
  - the random unitary is **locally scrambled** (twirled).

$$\forall V = \bigotimes_i V_i \text{ with } V_i \in U(2) :$$

$$p(\hat{\sigma}) = p(V^\dagger \hat{\sigma} V)$$

H.-Y. Hu, S. Choi, Y.-Z. You.  
arXiv:2107.04817

- The reconstruction map is given by

$$\boxed{\mathcal{M}^{-1}[O] = \sum_{P \in \mathcal{P}} \frac{\text{Tr } OP}{w_{\mathcal{E}_\sigma}(P) \text{Tr } \mathbf{1}} P}$$

- Only knowledge about Pauli weights is needed ( $P \in \mathcal{P}$ )

$$w_{\mathcal{E}_\sigma}(P) = \mathbb{E}_{\sigma \in \mathcal{E}_\sigma} (\text{Tr } P \sigma)^2$$

K. Bu, D. Enshan Koh, R. J. Gracia, A. Jaffe. arXiv: 2202.03272

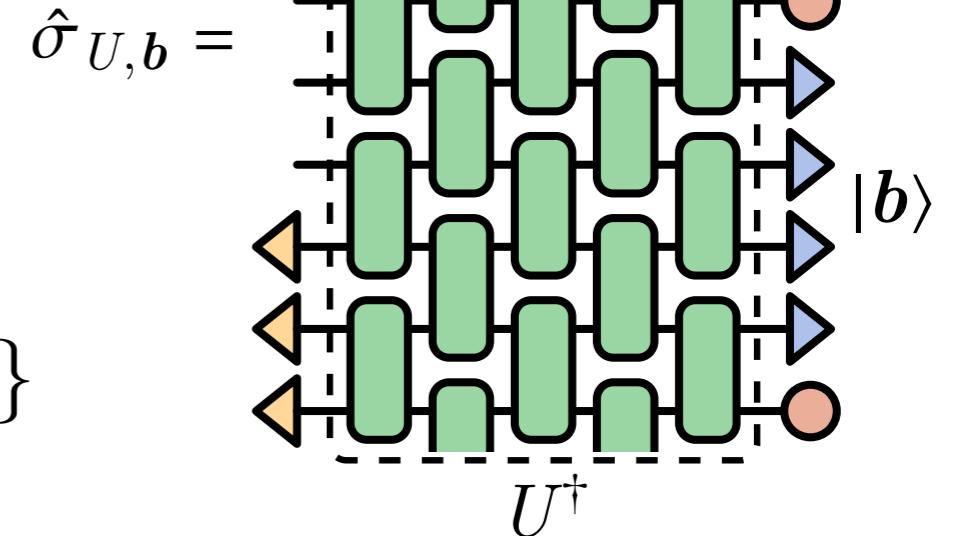
# Locally-Scrambled Quantum Dynamics

- Classical shadows are constructed from backward quantum dynamics

- Physical dynamics  $\sigma \rightarrow \mathcal{C}[\sigma]$
- Ensemble dynamics

$$\mathcal{E}_\sigma \rightarrow \mathcal{E}_{\mathcal{C}[\sigma]} = \{\mathcal{C}[\sigma] | \sigma \in \mathcal{E}_\sigma, \mathcal{C} \in \mathcal{E}_{\mathcal{C}}\}$$

- Pauli weight dynamics



$$w_{\mathcal{E}_{\mathcal{C}[\sigma]}}(P) = \sum_{P'} w_{\mathcal{E}_{\mathcal{C}}}(P, P') w_{\mathcal{E}_\sigma}(P')$$

Definitions:

$$w_{\mathcal{E}_\sigma}(P) = \mathbb{E}_{\sigma \in \mathcal{E}_\sigma} (\mathrm{Tr} P \sigma)^2$$

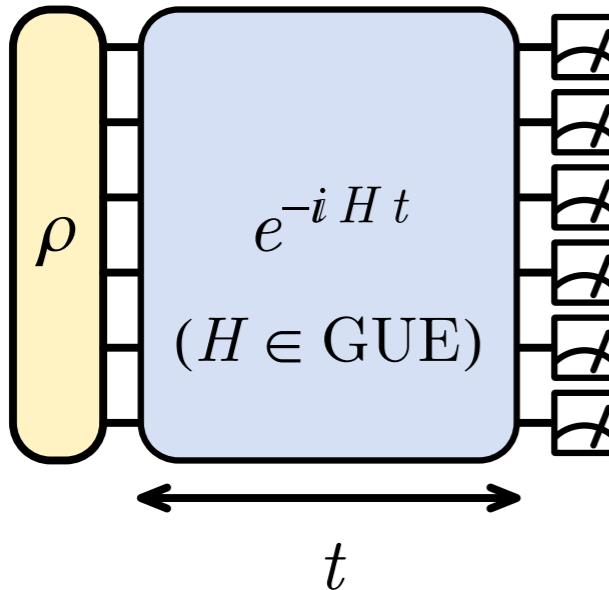
(State Pauli weight)

$$w_{\mathcal{E}_{\mathcal{C}}}(P, P') = \mathbb{E}_{\mathcal{C} \in \mathcal{E}_{\mathcal{C}}} \left( \frac{\mathrm{Tr}(P \mathcal{C}[P'])}{\mathrm{Tr} 1} \right)^2$$

(Channel Pauli weight)

# Locally-Scrambled Tomography Schemes

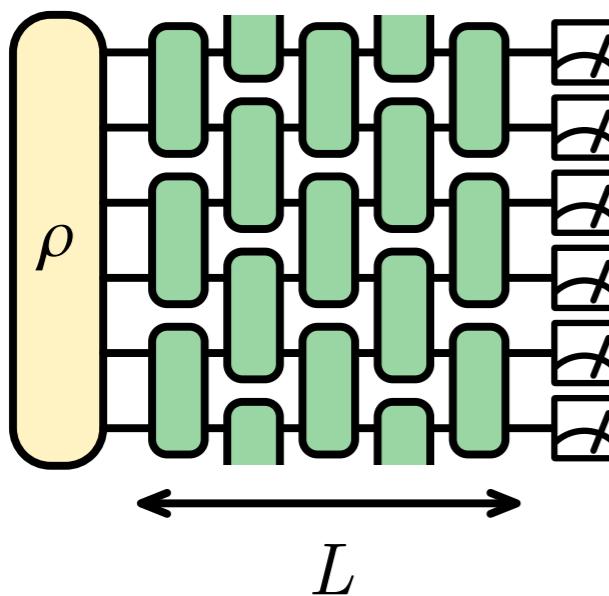
- Hamiltonian-driven shadow tomography



Hamiltonian  $H$  can be:  
GUE, GOE, GSE ...  
SYK, ETH, Quantum simulators ...

H.-Y. Hu, Y.-Z. You. arXiv:2102.10132

- Shallow shadow tomography



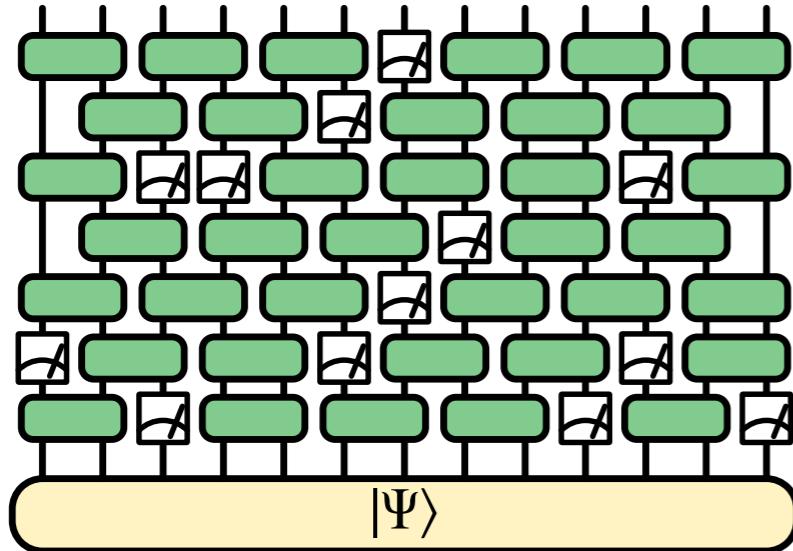
Finite-depth local random unitary  
Gate: Haar, Clifford ...  
Friendly to NISQ devices

H.-Y. Hu, S. Choi, Y.-Z. You. arXiv:2107.04817

# Locally-Scrambled Tomography Schemes

- Hybrid shadow tomography

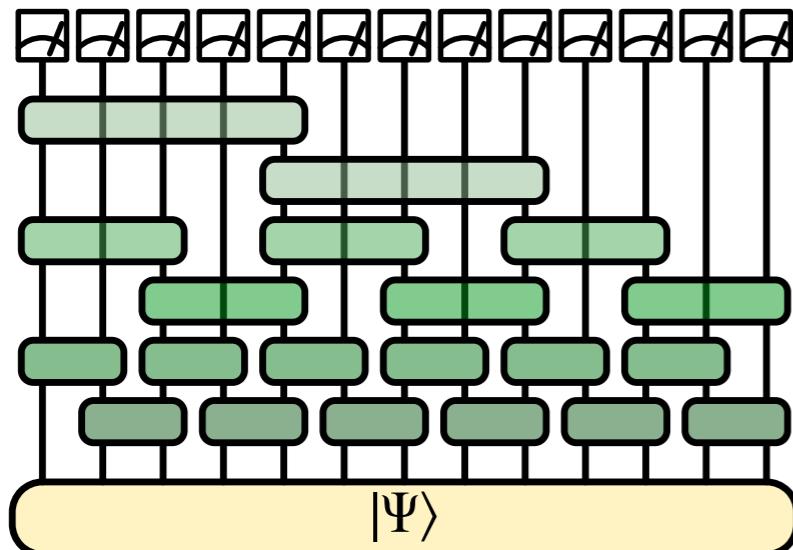
A.Akhtar, H.-Y. Hu, YZY. arXiv:2308.01653



Intermediate measurements with a measurement rate  $p$

- Optimal sample efficiency achieved at the measurement-induced phase transition.

- Holographic shadow tomography



Local measurements in the holographic bulk map to scale-free measurements on the holographic boundary.

X. Feng, S. Zhang, M. Ippoliti, YZY. (To appear)

# Scalable Tomography on Clifford Circuits

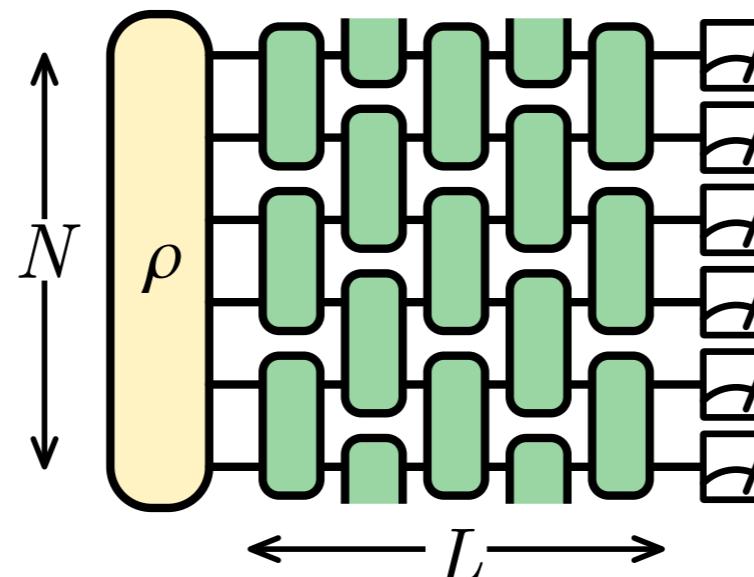
- Test states:

$$\rho_{\text{GHZ}} = |\text{GHZ}\rangle\langle\text{GHZ}| \quad |\text{GHZ}\rangle = \frac{1}{\sqrt{2}}(|00\cdots\rangle + |11\cdots\rangle)$$

$$\rho_{\text{ZXZ}} = |\text{ZXZ}\rangle\langle\text{ZXZ}| \quad Z_{i-1}X_iZ_{i+1}|\text{ZXZ}\rangle = |\text{ZXZ}\rangle$$

(Cluster state)

- Randomized measurement scheme:

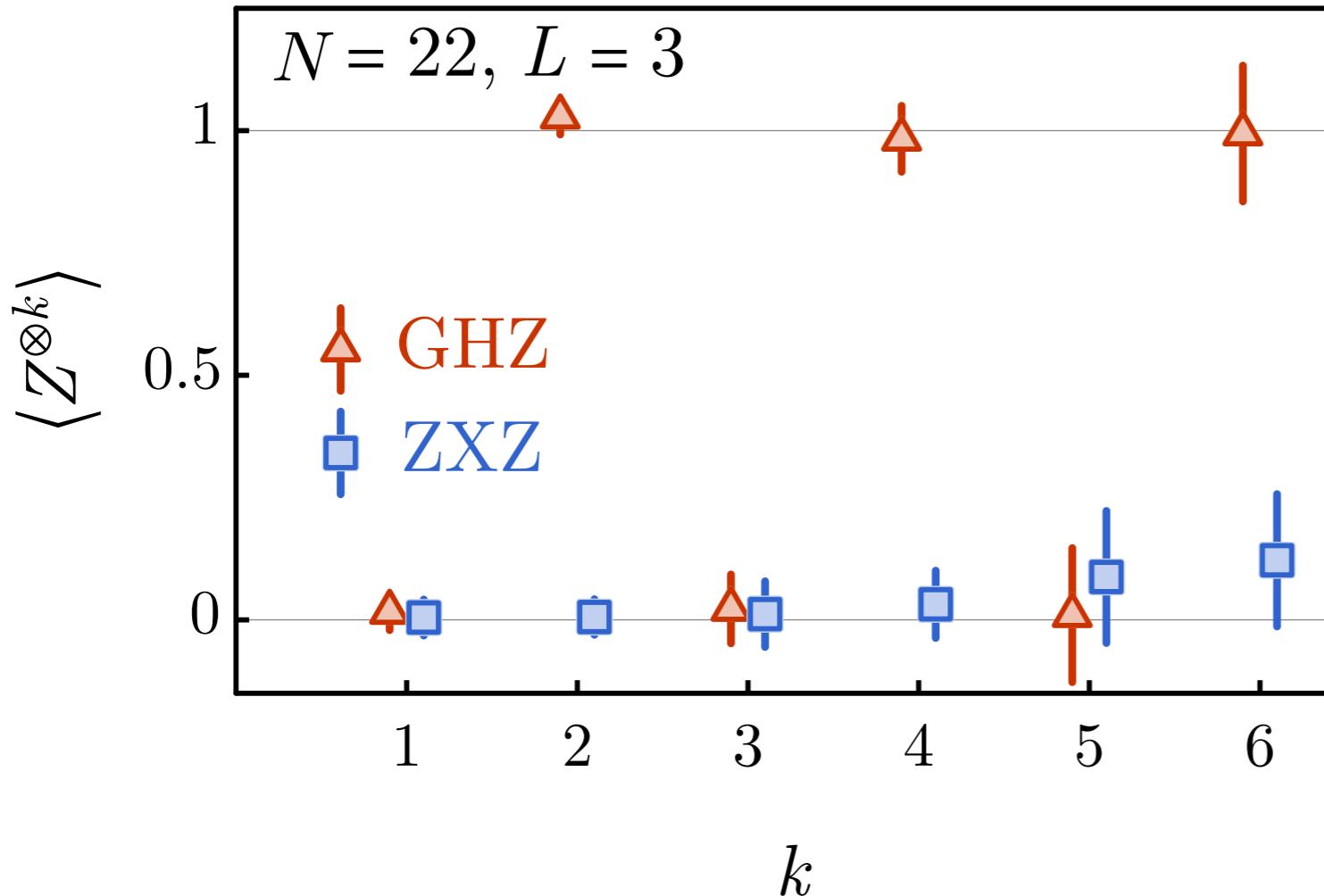


- Brick-wall arrangement of 2-qubit **random Clifford** gates
- Width (number of qubits):  $N$
- Depth (number of layers):  $L$

# Pauli Operator Estimation

- Task 2: Estimate the expectation value of a Pauli string operator

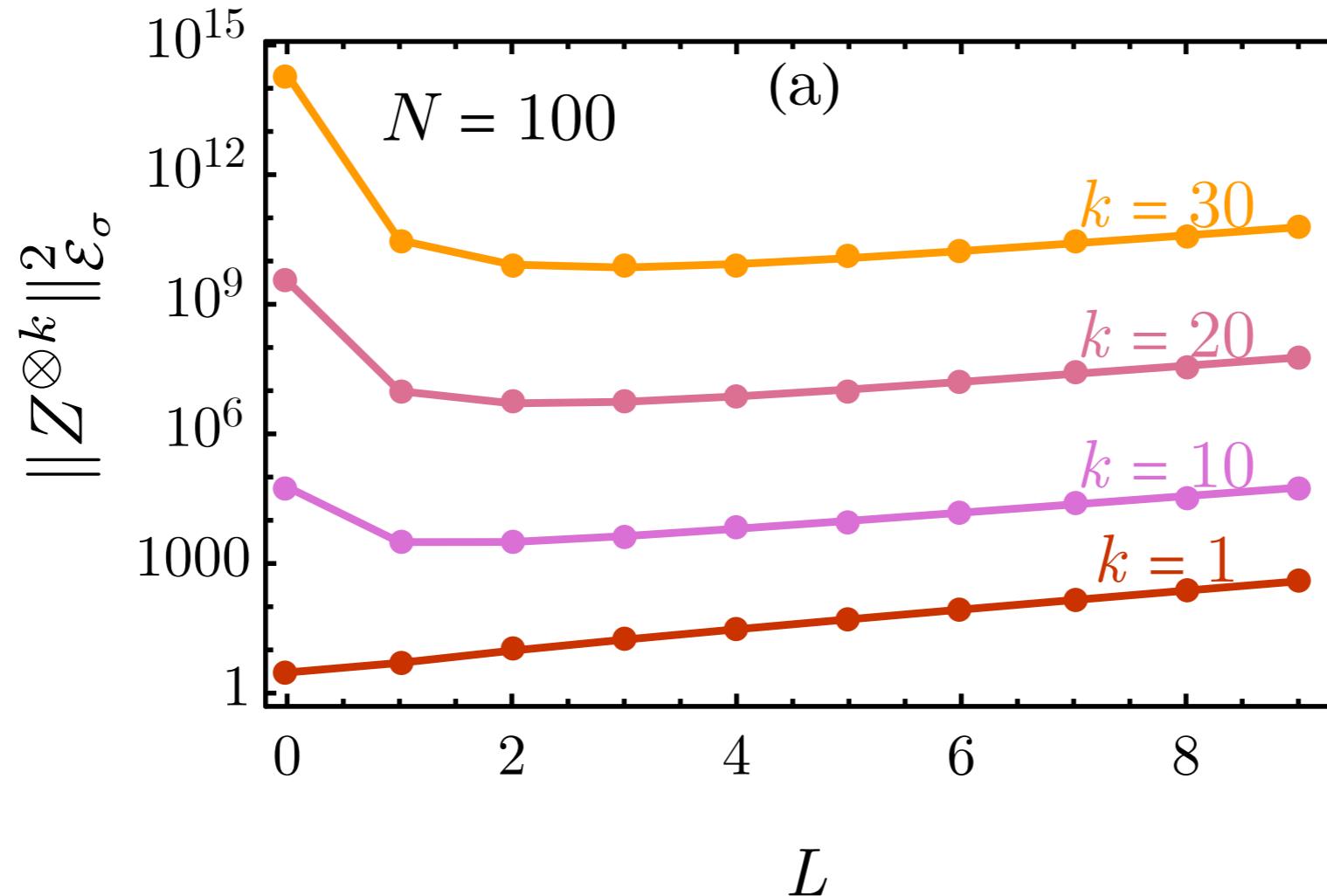
$$Z^{\otimes k} := \prod_{i=1}^k Z_i$$



- Estimations are unbiased (converge to the ground truth)
- Variance (shadow norm) increases with the weight (size) of the Pauli string
  - Note: For Pauli measurements ( $L = 0$ )
$$\|Z^{\otimes k}\|_{\mathcal{E}_\sigma}^2 \sim 3^k$$

# Pauli Operator Estimation

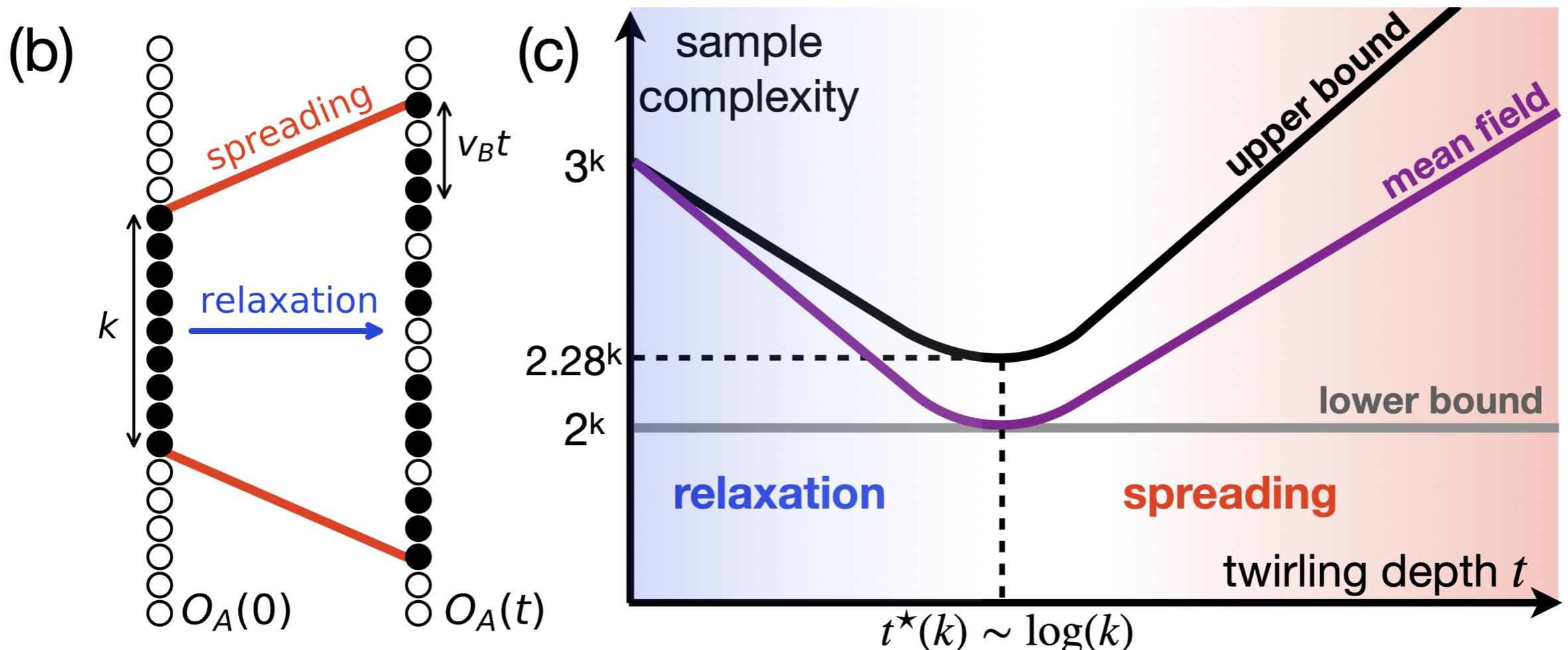
- Task 2: Estimate the expectation value of a Pauli string operator  $Z^{\otimes k} := \prod_{i=1}^k Z_i$



- Given the size  $k$  of the Pauli string, there is an **optimal circuit depth**  $L^*$  minimizing the shadow norm (sample complexity)

# Pauli Operator Estimation

- Considering continuous time limit, operator dynamics  
 $O_A \rightarrow O_A(t) = U O_A U^\dagger$  (with  $k = |A|$ )

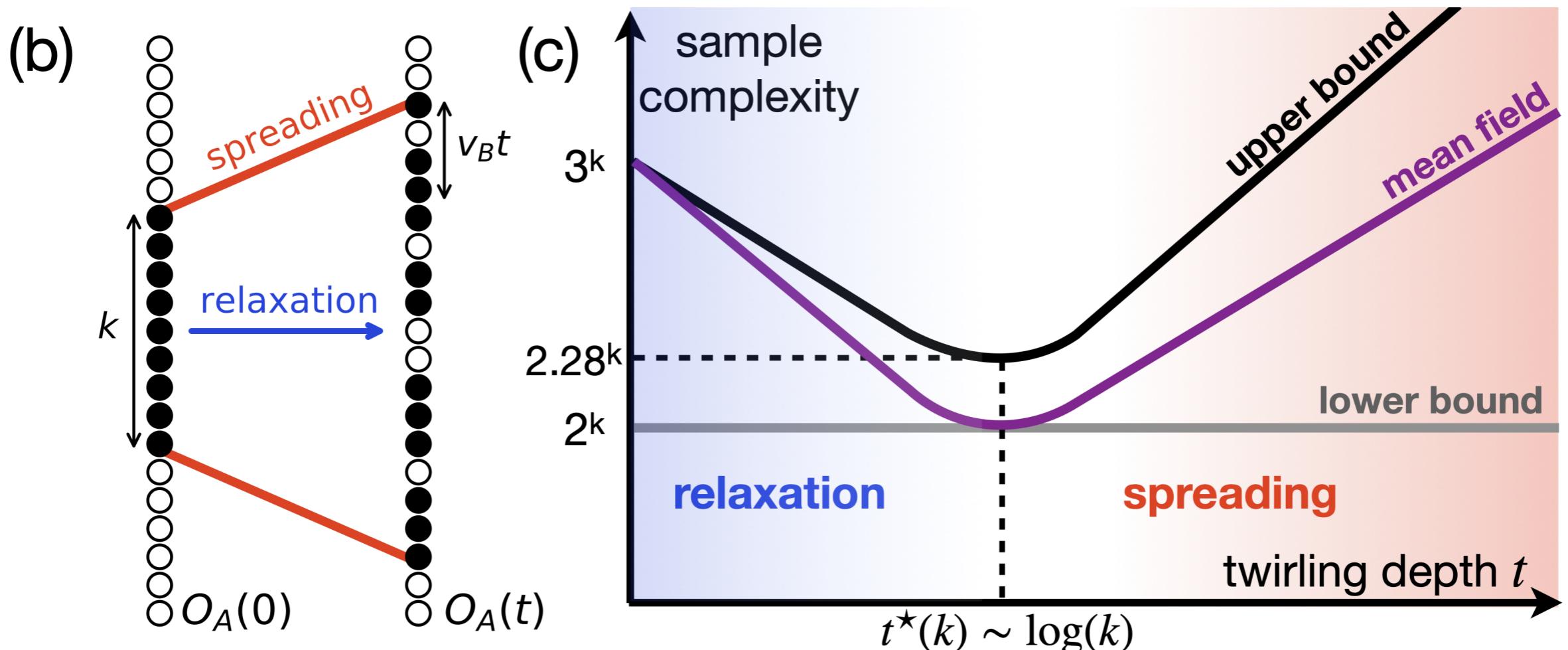


- Shadow norm  $\sim 3^{k_{\text{eff}}}$

$$k_{\text{eff}} \sim \left(\frac{3}{4} + \frac{1}{4}e^{-\gamma t}\right)(k + 2v_B t) \xrightarrow{\text{Min}} t^* \sim \gamma^{-1} \log k$$

# Pauli Operator Estimation

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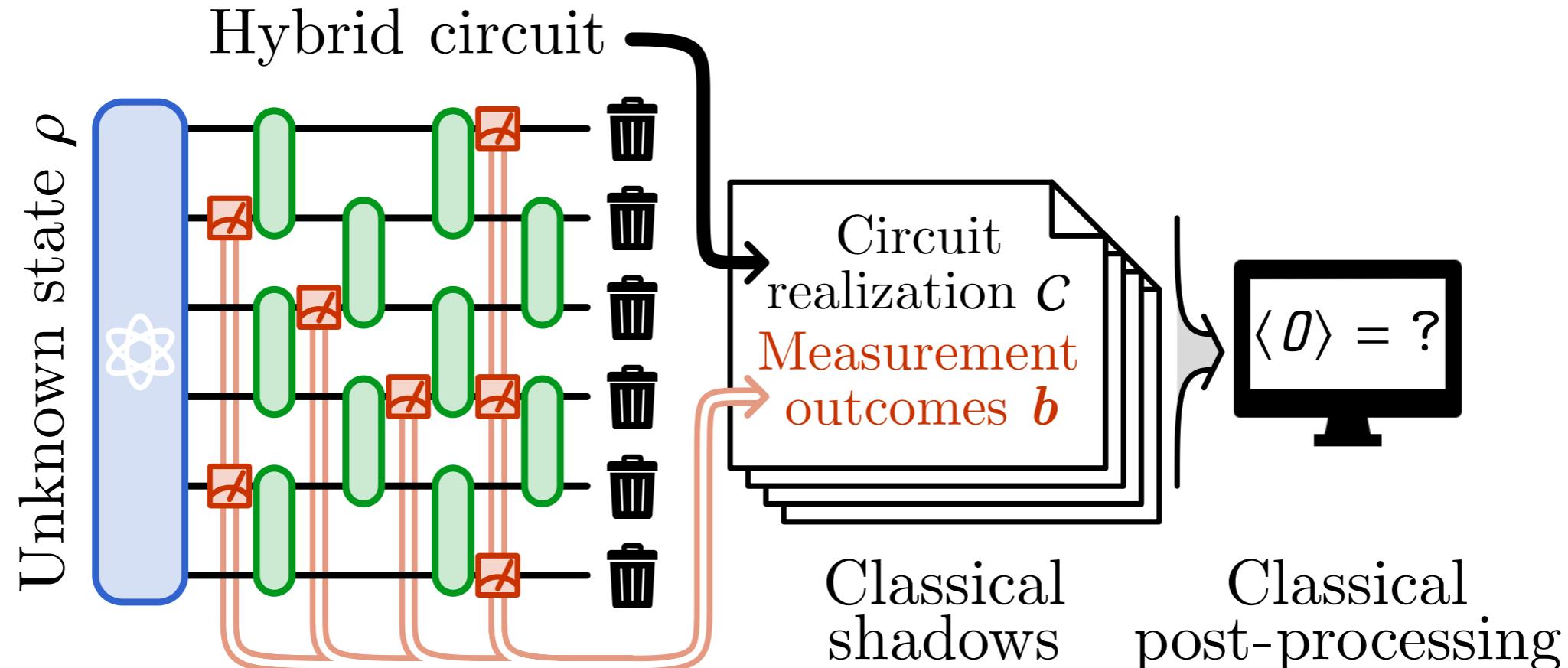


- At optimal circuit depth  $t^*$  shadow norm scales with  $k$  with a smaller base

$$2^k \lesssim \|O_A(t)\|_{\mathcal{E}_\sigma}^2 \lesssim 3^{\frac{3}{4}k} \approx 2.28^k$$

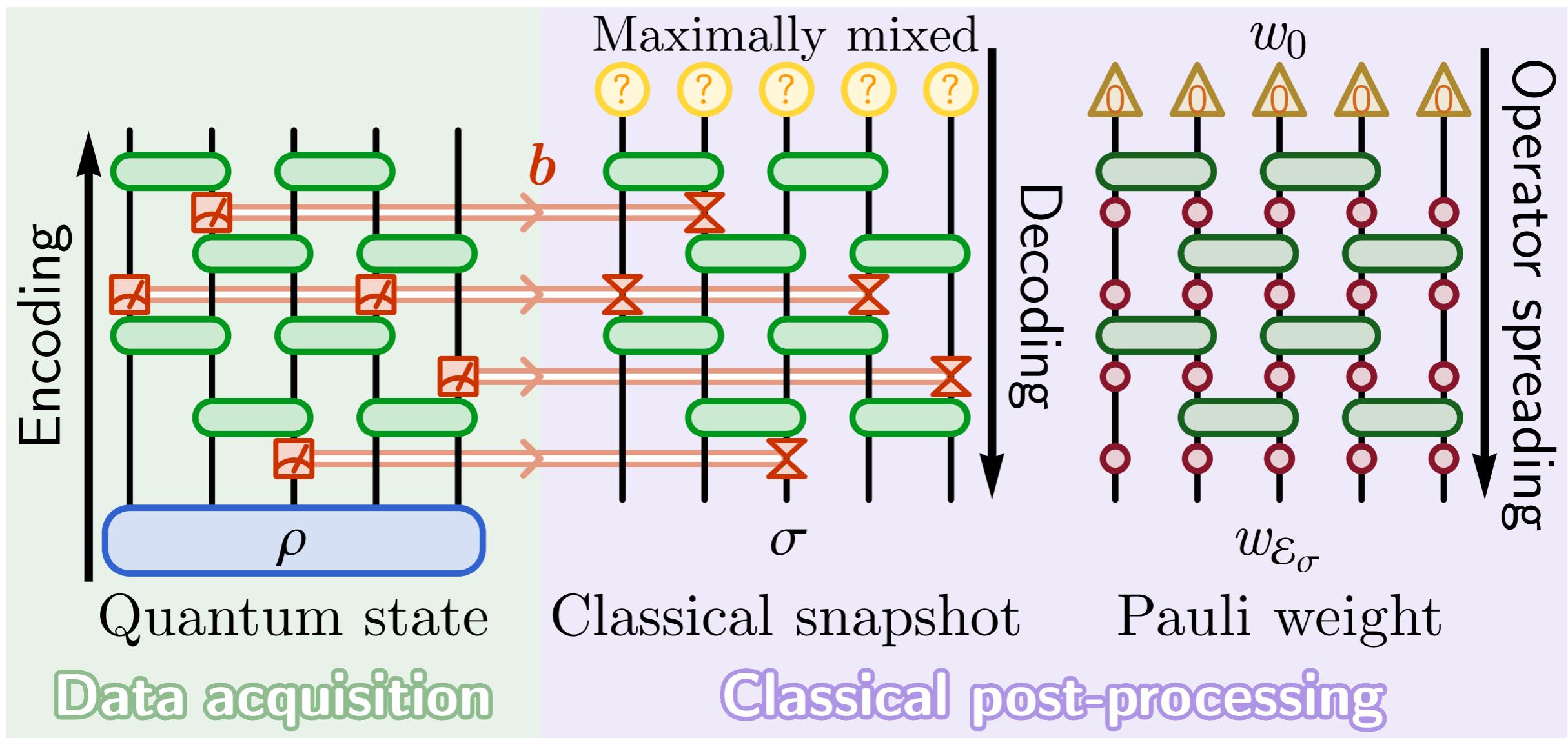
# Hybrid Shadow Tomography

- **Shallow shadow tomography** has an advantage in sample complexity scaling, but only achievable if the circuit depth is adjusted with the size of the observable.
- Can we perform the measurement on one circuit and make predictions for observables of all sizes optimally?
  - **Hybrid shadow tomography**



# Hybrid Shadow Tomography

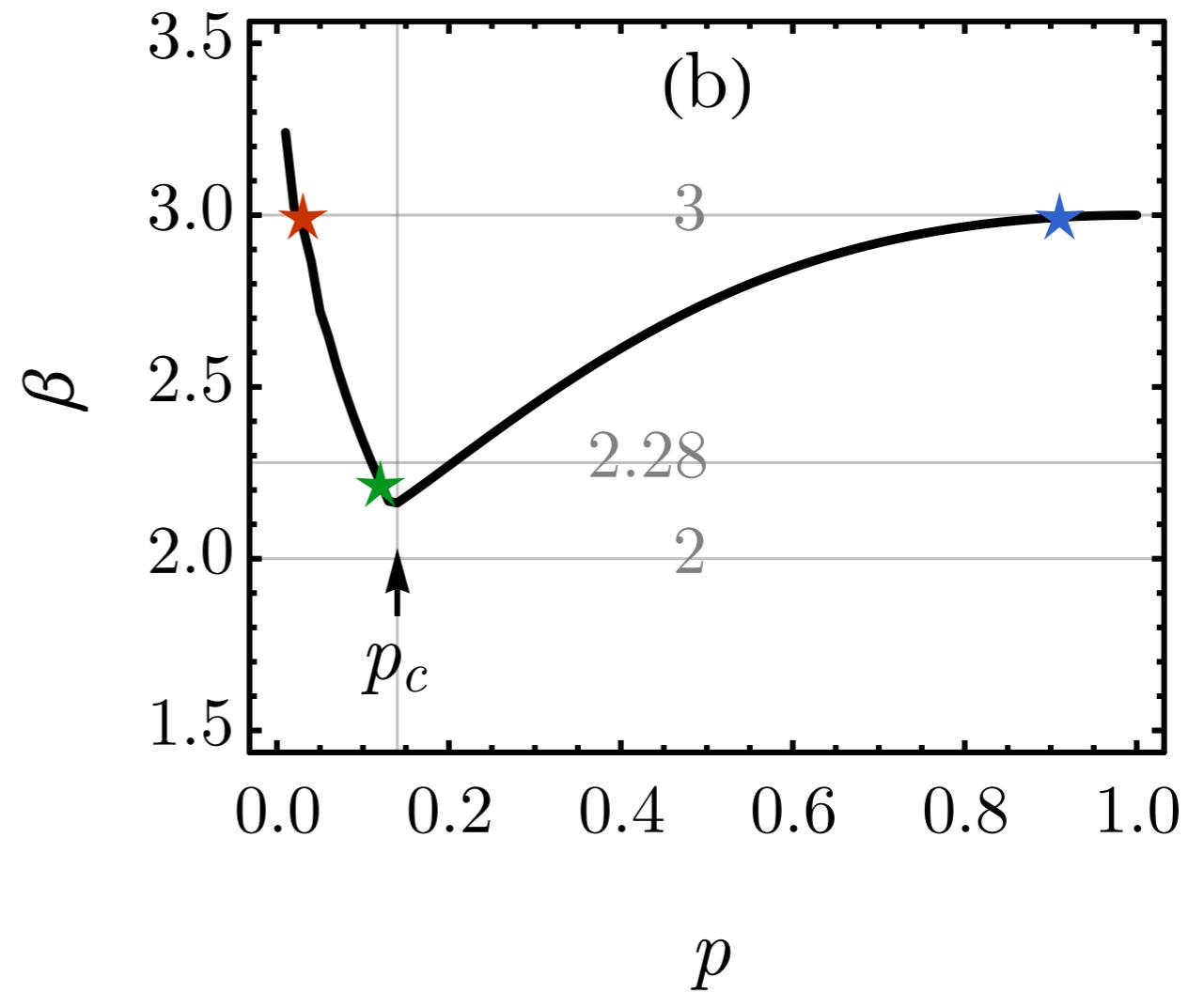
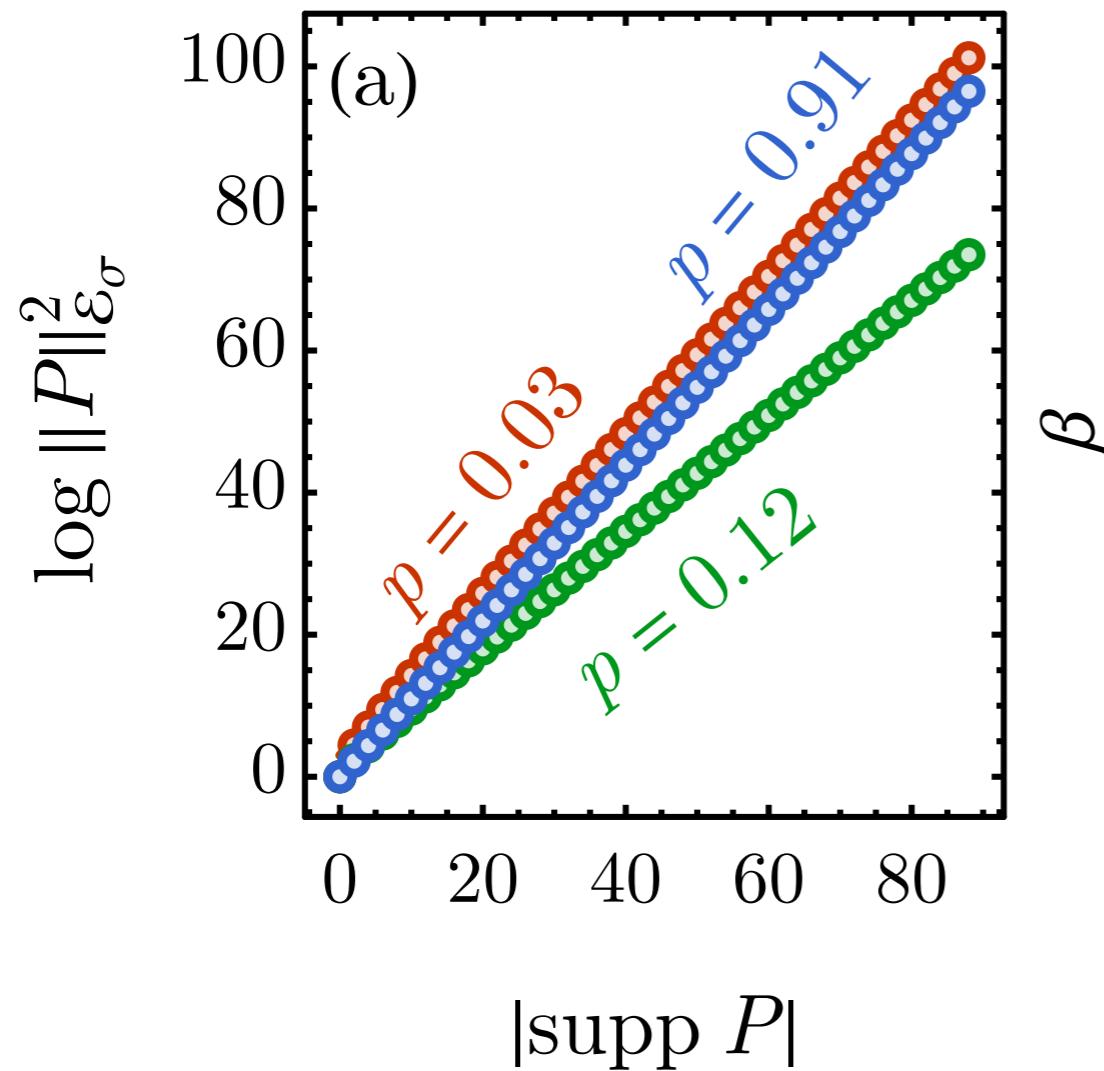
- Post-processing scheme



# Hybrid Shadow Tomography

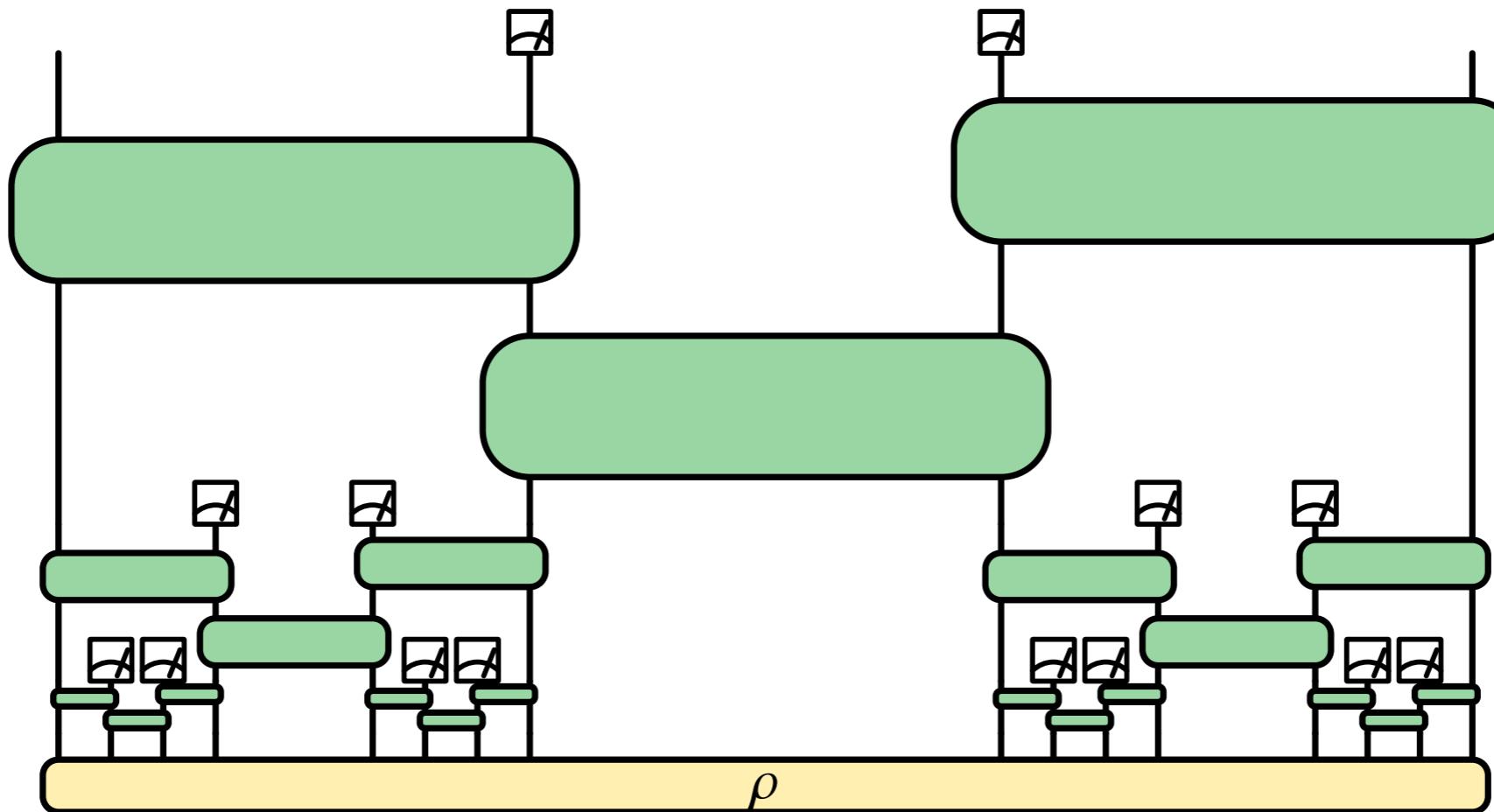
- Shadow norm scaling

$$\|P\|_{\mathcal{E}_\sigma}^2 \simeq \beta^k \text{poly}(k) \quad k = |\text{supp } P|$$



# Holographic Shadow Tomography

- However, **hybrid shadow tomography** requires fine-tuning the measurement-induced criticality.
- **Holographic shadow tomography** is automatically **critical**.



# Summary

- With locally-scrambled quantum dynamics, we extend the classical shadow tomography to a large class of quantum circuits, which is
  - **Scalable** (efficient classical post-processing)
  - **Flexible** (arbitrary circuit structure / quantum dynamics)
  - **NISQ friendly** (shallow circuits, simple gates, available devices)
- We expect our approach to have broad applications in many quantum information processing tasks (e.g. quantum error mitigation)

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Thanks for your attention!