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Operator relaxation and the optimal depth of classical shadows

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Classical shadows are a powerful method for learning many properties of quantum states in a sample-efficient manner, by making use of randomized measurements. Here we study the sample complexity of learning the expectation value of Pauli operators via “shallow shadows”, a recently-proposed version of classical shadows in which the randomization step is effected by a local unitary circuit of variable depth t . We show that the shadow norm (the quantity controlling the sample complexity) is expressed in terms of properties of the Heisenberg time evolution of operators under the randomizing (“twirling”) circuit—namely the evolution of the *weight distribution* characterizing the number of sites on which an operator acts nontrivially. For spatially-contiguous Pauli operators of weight k , this entails a competition between two processes: *operator spreading* (whereby the support of an operator grows over time, increasing its weight) and *operator relaxation* (whereby the bulk of the operator develops an equilibrium density of identity operators, decreasing its weight). From this simple picture we derive (i) an upper bound on the shadow norm which, for depth $t \sim \log(k)$, guarantees an exponential gain in sample complexity over the $t = 0$ protocol in any spatial dimension, and (ii) quantitative results in one dimension within a mean-field approximation, including a universal subleading correction to the optimal depth, found to be in excellent agreement with infinite matrix product state numerical simulations. Our work connects fundamental ideas in quantum many-body dynamics to applications in quantum information science, and paves the way to highly-optimized protocols for learning different properties of quantum states.

Introduction. The development of controllable quantum simulators has enabled the creation of complex and highly entangled quantum states in laboratory settings, leading to exciting new developments in quantum information science and many-body physics [1–8]. These advances raise the issue of how to efficiently characterize such quantum states. Full quantum state tomography requires exponentially many measurements in the size of the system [9], motivating the need for more scalable and efficient state-learning protocols. Recent progress in this direction has come from the development of *classical shadows* [10–22], a method to extract many physical properties of states with a dramatically smaller number of measurements. In this work, we shed light on the inner workings of classical shadows by making connections to foundational ideas in quantum dynamics on the spreading and equilibration of operators.

Classical shadows use *randomized measurements* [23–25] to form a compact representation of a many-body quantum state, Fig. 1(a). The state ρ is first transformed by a random unitary operation U (chosen from a suitable “twirling ensemble”), then projectively measured, yielding a computational basis state $|b\rangle$. The measured basis state is then rotated backwards (on a classical computer), giving a “snapshot” $\hat{\sigma}_{U,b} = U^\dagger |b\rangle\langle b| U$. The average of these snapshots (over twirling unitaries and measurement outcomes) is related to the true state ρ by a quantum channel, $\mathbb{E}_{U,b}[\hat{\sigma}_{U,b}] = \mathcal{M}(\rho)$. If the measurements are *tomographically complete* [11], the channel \mathcal{M} can be inverted (again on a classical computer) to produce “inverted snapshots” $\hat{\rho}_{U,b} = \mathcal{M}^{-1}(\hat{\sigma}_{U,b})$. These form a compact, approximate description of the quantum state ρ —its *classical shadow* [11]. From this description one can extract many properties of the state, which remarkably do not have to be specified in advance—the

general philosophy of the method is to “*measure first, ask questions later*” [25].

The usefulness of classical shadows depends on their *sample complexity*, i.e., the number of experimental samples needed in order to estimate a certain property of ρ within a given error. To learn an expectation value $\text{Tr}(\rho O)$, one builds estimators $\hat{o}_{U,b} = \text{Tr}(\hat{\rho}_{U,b} O)$ that yield the desired value in expectation ($\mathbb{E}_{U,b}[\hat{o}_{U,b}] = \text{Tr}(\rho O)$). The sample complexity is determined by the variance of \hat{o} , captured by the *shadow norm* $\|O\|_{\text{sh}}$, itself a function of the twirling ensemble. The freedom in choosing the twirling ensemble can thus be leveraged to optimize the learnability of certain properties of a quantum state. For instance, “local twirling” (where $U = \bigotimes_i u_i$ is a product of single-qubit random unitaries) gives $\|O\|_{\text{sh}}^2 = 3^k$ for Pauli operators, where k is the number of qubits on which O acts nontrivially; this is best suited to learning the value of few-body operators. On the opposite end, “global twirling” (where U is a random Clifford unitary on the whole Hilbert space) gives $\|O\|_{\text{sh}}^2 = \text{Tr}(O^\dagger O)$, which favors learning e.g. the fidelity with a pure many-body state $O = |\psi\rangle\langle\psi|$, but performs poorly on Pauli operators ($\|O\|_{\text{sh}}^2 = 2^N$) irrespective of locality [11].

Intermediate schemes, dubbed *shallow shadows*, have been recently proposed [26–28] and use twirling ensembles made of shallow quantum circuits, whose depth t can be tuned to interpolate between the local and global twirling limits. The finite depth t makes these easier to implement on quantum hardware, and enables efficient classical computation of $\hat{\sigma}$ and $\hat{\rho}$ via tensor-network methods [26, 27]. Surprisingly, these schemes were numerically observed to perform better than local twirling for estimating the expectation value of contiguous, multi-site Pauli operators (interesting examples of such operators include string order parameters for characterizing

topological phases [29, 30] and check operators of a quantum code [31]). The optimal depth $t^*(k)$ for a Pauli operator acting on k contiguous sites was observed numerically to scale as polylog(k) in one dimension [26], with a significant gain in sample complexity over the local twirling protocol. The physical mechanism behind this behavior has remained elusive thus far.

Here we analyze this problem analytically and find a mapping of the shadow norm to the dynamics of *Hamming weight* (the number of sites on which a Pauli operator acts nontrivially, henceforth just ‘weight’) under the twirling evolution. This mapping reveals that the optimal depth for the estimation of contiguous Pauli operators is determined by the competition of two processes under chaotic unitary dynamics, sketched in Fig. 1(b): *operator spreading* [32–37] and *operator relaxation*, to be defined below. Based on this picture, we prove that at depth $t^*(k) \sim \log(k)$, shallow shadows realize an exponential-in- k gain in sample complexity over local twirling in any finite spatial dimension. We further develop an analytical mean-field approximation for the shadow norm in one dimension, indicating that at depth $t^*(k)$ the sample complexity nearly saturates a lower bound ($\sim 2^k$, up to poly(k) corrections), as sketched in Fig. 1(c); the prediction shows excellent agreement with numerics on large Pauli operators (up to $k = 1000$) in infinite 1D systems.

Our results shed light on the inner workings of the classical shadows protocol and how it relates to fundamental aspects of quantum dynamics. At the same time, they give a practical, operational meaning to ideas about operator dynamics, and promise applications towards highly optimized classical shadows protocols for near-term quantum devices.

Shadow norm and operator weight. We begin by deriving a relationship between the shadow norm and operator dynamics valid if the twirling ensemble is *locally scrambled* [38, 39], i.e., such that measure dU over the ensemble is invariant under $U \mapsto VU$ and $U \mapsto UV$ for all product Clifford unitaries [40] $V = \bigotimes_i v_i$, $v_i \in \text{Cliff}(q)$ (this holds for local and global twirling, as well as for shallow shadows [26–28]).

We will consider a system of q -state qudits arranged on a d -dimensional lattice consisting of N qudits. For qudits with $q > 2$, we use ‘generalized Pauli operators’ defined by products of clock and shift unitary operators [41]. The measurement channel reads

$$\mathcal{M}(\rho) = \sum_b \int dU \overbrace{\langle b | U \rho U^\dagger | b \rangle}^{\text{Prob}(b|\rho,U)} \overbrace{U^\dagger | b \rangle \langle b | U}^{\text{snapshot } \hat{\sigma}_{U,b}}, \quad (1)$$

where b ranges over all $D = q^N$ computational basis states.

All Pauli operators are eigenmodes of the channel [20, 26, 27], and the eigenvalue depends solely on the twirling ensemble and on the support A of the Pauli operator: $\mathcal{M}[O_A] = \lambda_A O_A$, where O_A denotes a Pauli operator

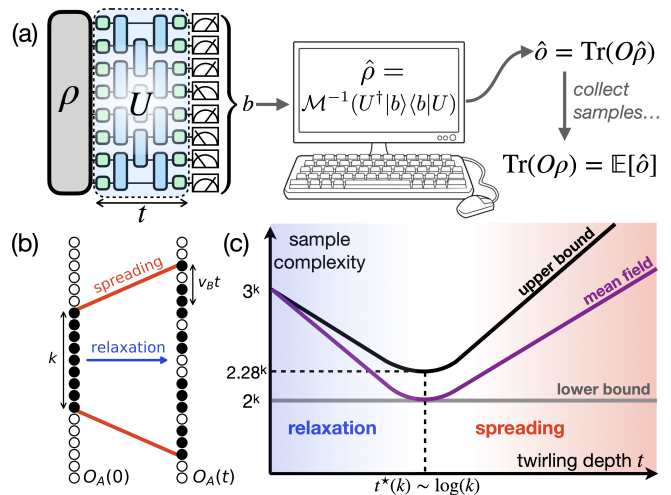


FIG. 1. (a) Schematic of classical shadows via shallow circuits: a state ρ is randomized by a ‘twirling’ circuit U of depth t , then measured; data is classically processed to estimate Pauli expectation values. (b) Operator spreading and relaxation under chaotic dynamics. \circ/\bullet denote identity and traceless Pauli matrices, respectively. (c) Summary of main results of this work. The competition between operator spreading and relaxation determines the optimal sample complexity of learning Pauli expectation values.

supported in region A . The eigenvalues can be expressed as [42]

$$\lambda_A = \sum_{w=1}^N \pi_{A,t}(w) (q+1)^{-w}, \quad (2)$$

where $\pi_{A,t}(w)$ is the averaged *weight distribution* [43] of the twirled operator $O_A(t) \equiv U O_A U^\dagger$:

$$\pi_{A,t}(w) = \sum_{P: |P|=w} \mathbb{E}_U |D^{-1} \text{Tr}(P O_A(t))|^2. \quad (3)$$

The sum runs over Pauli operators P , and $|P|$ is the weight of P .

With this result, we can exactly compute the shadow norm: $\|O_A\|_{\text{sh}}^2 = \text{Tr}(O_A^\dagger \mathcal{M}^{-1}[O_A])/D = \lambda_A^{-1}$ [26, 27, 42]. Combined with Eq. (2), this yields an exact relationship between the shadow norm and the weight distribution of a twirled operator,

$$\|O_A\|_{\text{sh}}^2 = \left[\overline{(q+1)^{-w}} \right]^{-1} \quad (4)$$

where the overline denotes averaging over w according to $\pi_{A,t}(w)$. Eq. (4) constitutes one of the main results of our work.

Eq. (4) reproduces the well-known results for local and global twirling of qubits (3^k and 2^N respectively [11]) in the $t = 0$ and $t \rightarrow \infty$ limits [42]. However, our result allows us to understand the behavior of the shadow norm away from these well-known limits, by leveraging the connection to the dynamics of operator weight under chaotic

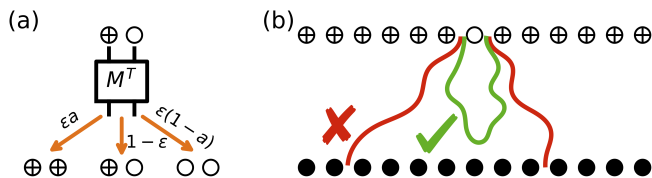


FIG. 2. (a) Update rules for a domain wall between \oplus and \circ states. (b) Random-walk calculation for the average density of holes $\bar{h}_i(t)$: if the two walkers fail to annihilate within t steps, the diagram vanishes.

evolution (i.e. the twirling ensemble U) as a function of time (i.e. the variable depth t).

Relaxation of operator weight. We focus on Pauli operators whose support A is a spatially-contiguous region (though our results also have implications for more general, non-contiguous Pauli operators [42]). We consider twirling ensembles of *diluted* random brickwork circuits, i.e. circuits where each gate is Haar-random [44] with probability ϵ and is the identity otherwise. These include conventional random circuits ($\epsilon = 1$), but allow us to slow down the twirling dynamics and discretize time more finely. To study the dynamics of operator weight during twirling, we introduce “occupation” variables n_i ($n_i = 0$ if a Pauli operator is the identity at site i , $n_i = 1$ otherwise). Before twirling, we have a *fully-packed* Pauli operator in region A : $n_i = 1$ iff $i \in A$. As the twirling depth t increases, two things happen: (i) *Operator spreading*—the boundary of the operator moves outwards, so that $\bar{n}_i(t) > 0$ also on sites $i \notin A$ that were initially empty, leading to an *increase* in weight; and (ii) *Operator relaxation*—the bulk of the operator relaxes from its fully-packed initial state ($n_i = 1 \forall i \in A$) towards an equilibrium density $\bar{n}_i(t) \rightarrow 1 - q^{-2}$ (when all q^2 Pauli operators are equally likely), leading to a *decrease* in weight.

As the latter is a bulk effect, it always dominates (at early times) for a sufficiently large region A . Thus the shadow norm must initially decrease from its $t = 0$ value (local twirling), before eventually becoming dominated by operator spreading and increasing again towards its $t \rightarrow \infty$ value (global twirling), implying a minimum at some finite optimal depth t^* .

To characterize the relaxation process, we focus on an infinite, fully-packed Pauli operator, and consider the average occupation of a site $\bar{n}_i(t)$ as a function of twirling depth t . For the twirling ensembles under consideration this problem can be addressed analytically in one spatial dimension. We leverage the fact that the vector of occupation probabilities $p_{\mathbf{n}}$ ($\mathbf{n} \in \{0, 1\}^N$ labels occupation configurations) evolves under the circuit-averaged dynamics via a Markov process, $p'_{\mathbf{n}} = \sum_{\mathbf{m}} \mathbb{M}_{\mathbf{n}, \mathbf{m}} p_{\mathbf{m}}$ with \mathbb{M} a stochastic matrix ($\sum_{\mathbf{m}} \mathbb{M}_{\mathbf{m}, \mathbf{n}} = 1 \forall \mathbf{n}$), to solve for the local occupation number analytically [34, 36]. We focus on the “density of holes” \bar{h}_i ($h_i \equiv 1 - n_i$) and introduce vectors in the binary space of (identity, traceless Pauli): $|\circ\rangle = (1, 0)^T$, $|\bullet\rangle = (0, 1)^T$, and $|\oplus\rangle = (1, 1)^T$. The

fully-packed initial state $p_{\mathbf{n}}^{\text{init}} = \prod_i \delta_{n_{i,1}}$ evolves under the averaged circuit into a final state $p_{\mathbf{n}}^{\text{final}}$, and we have $\bar{h}_i = \sum_{\mathbf{n}} p_{\mathbf{n}}^{\text{final}} \delta_{n_i, 0}$, corresponding to a matrix element $(\cdots \oplus \oplus \circ \oplus \oplus \cdots | \mathbb{M}_t | \cdots \bullet \bullet \bullet \cdots)$ where \mathbb{M}_t is the transition matrix for the averaged depth- t twirling circuit.

It is advantageous to consider the *backward* evolution \mathbb{M}_t^T acting on the state $|\cdots \oplus \oplus \oplus \oplus \oplus \oplus \cdots\rangle$: we have $M^T |\oplus \oplus\rangle = \epsilon a |\oplus \oplus\rangle + (1 - \epsilon) |\oplus \oplus\rangle + \epsilon(1 - a) |\circ \circ\rangle$ (Fig. 2(a)), where M is the transition matrix for a single two-qudit gate, $a = 1/(q^2 + 1)$, and ϵ is the dilution parameter (see [42]). Moreover we have $M^T |\circ \circ\rangle = |\circ \circ\rangle$ (unitary invariance of the identity operator) and $M^T |\oplus \oplus\rangle = |\oplus \oplus\rangle$ (conservation of total probability under the Markov process [45]). Thus the structure of a domain of \circ in a background of \oplus is preserved under \mathbb{M}_t^T , and domain walls undergo a random walk with a bias that tends to expand the \circ domain. When the domain walls are adjacent, they may annihilate, leading to an all- \oplus state which is invariant under M^T and yields a contribution $(\oplus |\bullet\rangle)^N = 1$; if the domain of \circ survives all the way to $t = 0$, the result vanishes as it involves at least one overlap $(\circ |\bullet\rangle) = 0$ (Fig. 2(b)).

In all, the average density of holes $\bar{h}_i(t)$ equals the probability that the two random walkers annihilate in t steps or less; conversely, $\bar{n}_i(t)$ equals their *survival probability*, which can be computed analytically: at large t ,

$$\bar{n}_i(t) = 1 - q^{-2} + ct^{-3/2} e^{-\gamma t} + \dots \quad (5)$$

for any site i in the bulk of the operator, with $c > 0$ a constant and \dots denoting subleading corrections in t [42]. The relaxation rate γ is related to the circuit’s *entanglement velocity* v_E (which sets the decay of half-system purity as $\sim q^{-v_E t}$) [46] via $\gamma = 2 \ln(q) v_E$, see [42]; the $t^{-3/2}$ is a universal correction related to the first return of a random walker in one dimension [47]. We conjecture that the convergence to equilibrium is exponential in any finite spatial dimension, and numerically verify it in two dimensions [42].

Scaling of the optimal depth. With these key results in hand, we return to the question of the optimal depth. From Eq. (4) and Jensen’s inequality, we have $\|O_A\|_{\text{sh}}^2 \leq (q + 1)^{\bar{w}}$; in one dimension, the average weight obeys $\bar{w}(t) = \sum_i \bar{n}_i(t) \simeq \bar{\ell}(t) \bar{n}_{\text{bulk}}(t)$, with $\bar{\ell}(t) = k + 2v_B t$ the average spatial length of the twirled operator, which spreads with butterfly velocity v_B [34, 36, 48], and $\bar{n}_{\text{bulk}}(t)$ the bulk density of traceless Paulis, Eq. (5) (the structure of the operator’s fronts can be neglected at large k). The bound is minimized at depth

$$t^*(k) = \gamma^{-1} \left(\ln(k) - \frac{3}{2} \ln \ln(k) + o(\ln \ln(k)) \right) \quad (6)$$

(see [42]). At $t = t^*(k)$, the shadow norm is bounded above by $(q + 1)^{(1 - q^{-2})k} \times \text{poly}(k)$, *exponentially smaller* than the $t = 0$ (local twirling) value of $(q + 1)^k$; e.g., for qubits ($q = 2$) the scaling is $3^{\frac{3}{2}k} \simeq 2.28^k$ vs 3^k . The scaling $\log(k)$ (as opposed to more general polylog(k)) [26])

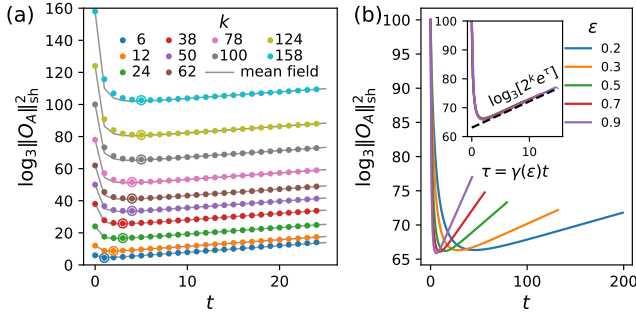


FIG. 3. (a) Shadow norm of a weight- k Pauli string O_A in an infinite 1D system of qubits ($q = 2$), under twirling by depth- t brickwork circuits of Haar-random gates (no gate dilution, $\epsilon = 1$). Data from iMPS simulations with bond dimension $\chi = 2048$. Circled dots indicate the optimal depth. (b) Same quantity for fixed $k = 100$ and variable gate dilution ϵ . Inset: same data as a function of “effective depth” $\tau = \gamma(\epsilon)t$, compared to $q^k e^{\gamma t}$ (dashed line).

is especially important as it ensures an MPO representation for \mathcal{M}^{-1} with poly(k) bond dimension, key to the classical computational cost of the method [26, 27].

We conjecture that $t = t^*(k)$ minimizes not just the upper bound $(q+1)^{\overline{w}}$, but the shadow norm itself, and that the achievable scaling of the latter is poly(k) \times q^k —nearly saturating the q^k lower bound obtained by full relaxation with no spreading. This is supported by an analytical calculation within a mean-field approximation, where we neglect correlations between occupations n_i, n_j at different sites, see [42]. We find that $\|O_A\|_{\text{sh}}$ is dominated by Pauli operators of size $k + 2v_B^{\text{sp}}t$, with a renormalized “saddle-point butterfly velocity” v_B^{sp} smaller than the original v_B , and equal to the entanglement velocity $v_E = \gamma/\ln(q^2)$. This predicts the late-time behavior $\|O_A\|_{\text{sh}}^2 \sim q^{k+2v_B^{\text{sp}}t} = q^k e^{\gamma t}$. Minimizing the mean-field shadow norm over t yields the same $t^*(k)$ as in Eq. (6), and thus the optimal shadow norm $\sim kq^k$.

It follows also that shallow shadows can be advantageous over local twirling not just for operators with contiguous support, but also for various types of non-contiguous operators, notably including typical random Pauli strings on a finite segment [42].

Numerical simulations. To check the validity of the above results, we perform numerical simulations of the averaged twirling dynamics with infinite matrix product states (iMPS) [49] (see [42]). Fig. 3(a) shows the shadow norm for contiguous operators in a 1D chain of qubits ($q = 2$), as a function of depth t . Three regimes are clearly visible: the $t = 0$ (local-twirling) value of 3^k , a minimum at $t \sim \log(k)$, and finally exponential growth due to continued operator spreading after relaxation. In un-diluted circuits ($\epsilon = 1$) the optimal depth $t^*(k)$ takes very small integer values, severely limiting the resolution on its scaling [26]. This issue is greatly alleviated by gate dilution: the shadow norm approximately behaves as a smooth function of an “effective depth” $\tau = \gamma(\epsilon)t$

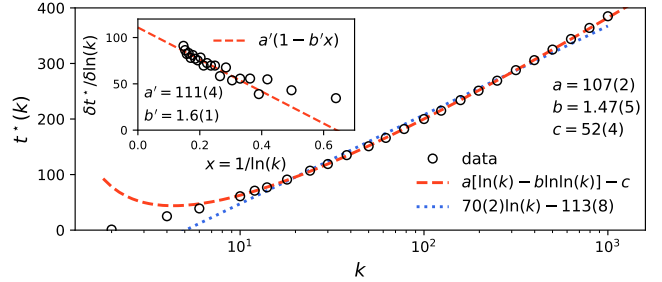


FIG. 4. Optimal depth $t^*(k)$ as a function of Pauli operator weight k , obtained from iMPS data as in Fig. 3, for k up to 1000. The gate dilution is $\epsilon = 0.05$ and bond dimension is $\chi = 2048$. Best fits to $t^*(k) = a'' \ln(k) - c''$ (dotted line) and $t^*(k) = a[\ln(k) - b \ln \ln(k)] - c$ (dashed line) are shown. The doubly-logarithmic correction is found to be $b = 1.47(5)$, consistent with the predicted $3/2$ in Eq. (6). Inset: discrete derivatives $\delta t^*(k)/\delta \ln(k)$, plotted vs $1/\ln(k)$, indicate a doubly-logarithmic correction $b = 1.6(1)$, also consistent with $3/2$.

(Fig. 3(b)), where γ is the Pauli density relaxation rate in Eq. (5)—smaller ϵ yields a finer sampling of τ . To finely resolve the scaling of $t^*(k)$, we set $\epsilon = 0.05$ obtaining the results in Fig. 4. The data show remarkable agreement with Eq. (6), including the subleading correction $\sim \ln \ln(k)$. The value of $3/2$ for the ratio of coefficients is universal (determined by the probability of first return of a random walk via Eq. (5)), which constitutes a non-trivial check of our analytical results.

Higher dimensions. While several details of the above discussion are special to one dimension, the general picture applies to systems in any finite spatial dimension. The leading-order result ($t^*(k) \sim \ln k$) depends only on the balancing of operator spreading and relaxation for operators whose boundary is much smaller than the bulk. In systems with all-to-all connectivity or on expander graphs, where a subsystem’s bulk and boundary generally have comparable sizes, the optimal twirling depth is expected to be zero, i.e., local twirling performs best. We test this expectation on a “Brownian circuit” model whose operator dynamics are described by simple, closed equations, and are amenable to exact treatment; we find the optimal depth is $t = 0$ unless the operator is supported on a sufficiently large fraction of the system ($k \gtrsim N/2$), see [42].

Discussion. We have studied how classical shadows based on shallow quantum circuits can be used to learn expectation values of Pauli operators. We have connected the sample complexity of classical shadows to the dynamics of operator weight, identifying two competing dynamical processes (operator spreading and relaxation) whose balance determines the optimal depth t^* of the twirling circuits. This picture elegantly explains previous numerical observations on one-dimensional systems [26, 27], and extends the result to systems in any finite dimension. Further, it shows that the optimal depth scales as

$t^* = O(\ln k)$ with the weight k of the learned operator, as opposed to a more general $t^* = \text{polylog}(k)$ scaling [26], ensuring a $\text{poly}(k)$ classical computational cost for the optimal protocol.

Our work opens up several directions for future research. It would be interesting to generalize our results to different settings for classical shadows, beyond shallow brickwork circuits on qudits. The recent proposals for classical shadows in analog simulators [50, 51] or on fermionic [17, 18] and bosonic [52] systems are interesting possible directions. The validity of our results in higher dimension also suggests interesting applications to e.g. topological or fracton codes and phases [53–58]. Further, it would be interesting to extend our analysis to measures of entanglement [23, 24], and to make contact with NISQ experiments [15, 59] by understanding the impact of noise on our results [13, 60].

Finally, the concept of operator relaxation may be of independent interest from the point of view of quantum dynamics. While operator spreading is central to the study of quantum chaos [32–37], operator relaxation and similar diagnostics of local equilibration in operator space [35, 37, 61] are comparatively under-explored, and may prove similarly useful in understanding signatures of

quantum-chaotic behavior [62].

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