## Probing finite-temperature observables in quantum simulators of spin systems with short-time dynamics

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Preparing finite-temperature states in quantum simulators of spin systems, such as trapped ions or Rydberg atoms in optical tweezers, is challenging due to their almost perfect isolation from the environment. Here, we show how finite-temperature observables can be obtained with an algorithm motivated from the Jarzynski equality and equivalent to the one in Lu *et al.*, PRX Quantum **2**, 020321 (2021). It consists of classical importance sampling of initial states and a measurement of the Loschmidt echo with a quantum simulator. We use the method as a quantum-inspired classical algorithm and simulate the protocol with matrix product states to analyze the requirements on a quantum simulator. This way, we show that a finite-temperature phase transition in the long-range transverse-field Ising model can be characterized in trapped ion quantum simulators. We propose a concrete measurement protocol for the Loschmidt echo and discuss the influence of measurement noise, dephasing, as well as state preparation and measurement errors. We argue that the algorithm is robust against those imperfections under realistic conditions.

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The excellent tunability of quantum simulators has enabled new insights into entire classes of many-body models. Recently, tremendous progress has been achieved in simulating unconventional nonequilibrium dynamics of quantum spin models with a large number of controlled degrees of freedom on different experimental platforms, including ultracold atoms in optical lattices [1–5], Rydberg atoms [6–8], and trapped ions [9–11]. A key question motivated from condensed matter physics is to study finite-temperature states of quantum spin models, which can host phases with symmetry breaking or even topological order, thermal phase transitions, and quantum criticality [12]. Preparing states at finite and, in particular, low temperatures, required to study these phenomena, is however a formidable challenge for quantum simulators because of their almost perfect isolation from the environment.

Here, we discuss a concrete approach to measure finitetemperature observables in quantum simulation experiments based on an algorithm recently derived by Lu *et al.* [13]. We motivate this algorithm from a different vantage point based on the Jaryznski equality [14], which provides a link between the nonequilibrium dynamics of a quantum system and its thermal properties. The key result of this algorithm is to obtain thermal observables from quantum simulators without preparing a thermal state directly, but to use a short real-time evolution instead. While this algorithm can be immediately applied to current quantum technology, even for large systems, other algorithms proposed for preparing finitetemperature states, including sampling methods [15–19], imaginary time evolution [20,21], variational methods [22], kernel based methods [23], and direct state preparation using fluctuation theorems [24] are challenging to implement on devices without fault tolerance due to significant resource overheads. Contrarily, the algorithm from Ref. [13] can be implemented on current devices due to its low requirement on evolution time and its error resilience.

We apply this algorithm to the detection of the thermal phase transition in the one-dimensional long-range transversefield Ising model (LTFIM)

$$\hat{H} = -J \sum_{i < j} \frac{1}{|i - j|^{\alpha}} \hat{\sigma}_i^z \hat{\sigma}_j^z - g \sum_i \hat{\sigma}_i^x, \qquad (1)$$

where g is the strength of the transverse field, J > 0 is the ferromagnetic coupling with long-range exponent  $\alpha$ , and  $\hat{\sigma}_i^a$  is the *a*th Pauli matrix on site *i*. This model can be implemented with trapped ions [25] [see Figs. 1(a) and 1(c)]. For  $\alpha \leq 2$ , the system exhibits a finite-temperature transition from a ferromagnet at low temperatures to a paramagnet at high temperatures, provided the transverse field is sufficiently weak [26,27] (cf. also Ref. [28] for an experiment with adiabatic state preparation).

The Jarzynski-inspired algorithm uses Monte Carlo (MC) importance sampling of product states, where the state probabilities are classically calculated from the Loschmidt echo  $G_{\psi}(t) = \langle \psi | e^{-i\hat{H}t} | \psi \rangle$  that is measured on a quantum simulator [Fig. 1(b)]. We develop a scheme for measuring  $G_{\psi}(t)$  based on Ramsey spectroscopy that involves a shelving state [Fig. 1(a)]. The Loschmidt echo  $G_{\psi}(t)$  only needs to be



FIG. 1. Setup and phase diagram. (a) Trapped ions in internal states  $|0\rangle$ ,  $|1\rangle$  interact with power-law couplings with exponent  $\alpha$ . Ions in the shelving state  $|s\rangle$  do not interact. (b) Thermal observables are obtained from Monte Carlo importance sampling of product states (see Ref. [13]). The probability of proposed states is classically evaluated from the Loschmidt echos  $G_{\psi}(t)$  measured on the quantum simulator (blue shading), in which the shelving state  $|s\rangle$  is used for obtaining the phase of  $G_{\psi}(t)$  with Ramsey experiments. The Loschmidt echo only needs to be measured to short times Jt = O(1) [13]. (c) The one-dimensional long-range transverse-field Ising model exhibits a finite-temperature phase transition from a ferromagnet to a paramagnet for  $\alpha \leq 2$  (and transverse field  $g < g_c$ ; here g = J). At  $\alpha = 2$ , the transition is in the Berenzinskii-Kosterlitz-Thouless universality class. In this work, we focus on  $\alpha = 1.5$  (gray dashed line).

evaluated to short times, as proven in Ref. [13]. To benchmark the algorithm, we interpret it as a quantum-inspired classical algorithm by evaluating the Loschmidt echoes  $G_{\psi}(t)$  with matrix product states (MPS). We find that the finite-temperature phase transition of the LTFIM can be efficiently characterized with this algorithm even for large systems. To assess the feasibility of the algorithm in a realistic quantum simulator, we study its robustness to a finite number of measurements and discuss dephasing noise as well as state preparation and measurement (SPAM) errors, demonstrating the immediate applicability of the algorithm in current experimental technology.

Thermal properties from the Jarzynski equality. The algorithm of Ref. [13] can be motivated from the Jarzynski equality [14], which is based on the following thought experiment [29,30]: A system is prepared in thermal equilibrium at temperature T with respect to a Hamiltonian  $\hat{H}_0$ . A measurement of  $\hat{H}_0$  is then performed, which projects the system with probability  $\frac{1}{Z_0}e^{-E_n^0/T}$  into the energy eigenstate  $|n^0\rangle$  with energy  $E_n^0$ , where  $Z_0$  is the partition sum of  $\hat{H}^0$ . Then, a second measurement in the eigenbasis of  $\hat{H}$ is performed, which yields the result  $E_m$  with probability  $|\langle n^0 | m \rangle|^2$ . In this process, the energy of the system changed and therefore work  $\omega = E_m - E_n^0$  has been performed. Repeating this experiment many times, we can measure the probability distribution of work  $\omega$ . It is given by  $p(\omega) =$  $\frac{1}{Z_0} \sum_n e^{-\tilde{E}_n^0/T} \sum_m |\langle n^0 | m \rangle|^2 \delta[\omega - (E_m - E_n^0)].$  By multiplying the work distribution with  $e^{-\omega/T}$  and integrating, we find the Jarzynski equality

$$Z = Z_0 \int d\omega \, e^{-\omega/T} p(\omega), \qquad (2)$$

which relates an equilibrium quantity—the thermal partition sum  $Z = \text{Tr}(e^{-\hat{H}/T})$ —to a nonequilibrium quantity—the work distribution. While in principle, all properties of a system can be obtained from Z, in practice it is hard to evaluate. Here we focus on the evaluation of finite-temperature observables  $\langle \hat{O} \rangle_T = \frac{1}{Z} \text{Tr}(\hat{O}e^{-\hat{H}/T})$ . In particular, by formally choosing  $\hat{H}_0 \propto \mathbb{1}$ , we can use the Jarzynski equality even without preparing a thermal state of  $\hat{H}_0$ , since for this choice, the dependence on  $\hat{H}_0$  becomes trivial.

In order to relate the work distribution function to the Loschmidt echo, we interpret work as the Fourier conjugate to time by writing the delta function as  $\delta(\omega) = \int \frac{dt}{2\pi} e^{i\omega t}$ . Then, the Jarzynski equality becomes  $Z = \int d\omega e^{-\omega/T} \int \frac{dt}{2\pi} e^{i\omega t} \operatorname{Tr}(e^{-i\hat{H}t})$ . We expand the trace in a basis of product states  $\mathbb{1} = \sum_{\psi} |\psi\rangle \langle \psi|$  to write  $Z = \sum_{\psi} \int d\omega e^{-\omega/T} p_{\psi}(\omega)$  with

$$p_{\psi}(\omega) = \int \frac{dt}{2\pi} e^{i\omega t} \langle \psi | e^{-i\hat{H}t} | \psi \rangle.$$
 (3)

This way, we have reduced the evaluation of Z via the Jarzynski equality to a measurement of the Loschmidt echo  $G_{\psi}(t) = \langle \psi | e^{-i\hat{H}t} | \psi \rangle$  with respect to product states  $| \psi \rangle$  without requiring one to prepare a thermal state in the quantum simulator.

In order to obtain observables  $\hat{O}$  from Z, we shift  $\hat{H} \rightarrow \hat{H} + h\hat{O}$  and evaluate

$$\langle \hat{O} \rangle_T = -\frac{T}{Z} \frac{dZ}{dh} \bigg|_{h=0}.$$
 (4)

While in principle any observable can be evaluated this way (see Supplemental Material [31]), the simplest algorithm can be derived for observables satisfying  $\hat{O} |\psi\rangle = O_{\psi} |\psi\rangle$ . Inserting the Jarzynski equality, Eq. (2), we find in this case

$$\langle \hat{O} \rangle_T = \frac{\sum_{\psi} p_{\psi}(T) O_{\psi}}{\sum_{\psi} p_{\psi}(T)},\tag{5}$$

where

$$p_{\psi}(T) = \int p_{\psi}(\omega) e^{-\omega/T} d\omega.$$
 (6)

Due to the exponential size of the Hilbert space, we cannot evaluate Eq. (5) exactly. However, because  $p_{\psi}(T) / \sum_{\psi} p_{\psi}(T)$  is a probability distribution (even for frustrated and fermionic models), we can use a classical Monte Carlo importance sampling algorithm to select product initial states:

- (1) Start with some product state  $|\psi\rangle$ .
- (2) Repeat  $N_{\rm MC}$  times for a given temperature T:
  - (a) Propose a new state  $|\psi'\rangle$ .
  - (b) Measure  $G_{\psi'}(t)$  on a quantum simulator.
  - (c) Evaluate  $p_{\psi'}(\omega)$  from Eq. (3).
  - (d) Evaluate  $p_{\psi'}(T)$  from Eq. (6).
  - (e) Accept  $|\psi'\rangle$  with probability  $p_{\psi'}(T)/p_{\psi}(T)$ .
  - (f) Evaluate  $O_{\psi}$ .
- (3) Obtain thermal expectation value by averaging  $O_{\psi}$ .



FIG. 2. Data processing with classical resources. (a) Loschmidt echo shown for two different initial states: a completely polarized state and a state with vanishing magnetization. This data can be measured directly on a quantum simulator or can be obtained from numerical simulations for the quantum-inspired classical algorithm. (b) The corresponding work distributions  $p_{\psi}(\omega)$  are centered around the energies of the initial states, with the width given by the energy fluctuations. (c) Magnetization evaluated from Monte Carlo importance sampling. (d) The absolute error of the squared magnetization  $\langle (S^z)^2 \rangle$  at temperature T/J = 7 calculated by a jackknife estimate. Parameters: L = 16,  $\delta/J = 4$ ,  $Jt_{max} = 1$ ,  $\Delta t = 0.1$ .

This algorithm is the same as the one introduced in Ref. [13] based on energy filtering. Here, we provide an interpretation of this algorithm based on the Jarzynski equality.

In the following, we discuss each step in more detail and apply it to the LTFIM. For step (1), a basis needs to be chosen. We use *z*-product states for which  $\hat{\sigma}_i^z |\psi\rangle = \sigma_i^z |\psi\rangle$ . As a state proposal in step (2a) we flip a single, randomly chosen spin, which fulfills ergodicity and, together with the acceptance step (2e), detailed balance [32].

In step (2b), we measure  $G_{\psi}(t)$  on the quantum simulator in a time interval  $[0, t_{\text{max}}]$  with equal time steps  $\Delta t$  [Fig. 2(a)]. The evaluation of  $G_{\psi}(t)$  is the only step performed on the quantum simulator; all other steps use classical resources and take negligible computation time. We will discuss below how to measure this quantity in trapped ion simulators.

From  $G_{\psi}(t)$ , we then evaluate  $p_{\psi}(\omega)$  from Eq. (3) by a discrete Fourier transform in step (2c),

$$p_{\psi}(\omega) = \frac{\Delta t}{2\pi} \sum_{n=-N}^{n=N} e^{i\omega n \Delta t} G_{\psi}(t) e^{-(t\delta)^2/2}, \qquad (7)$$

where  $\omega = 2\pi n/[\Delta t(2N + 1)]$ ,  $N = t_{max}/\Delta t$ . Above, we have introduced a Gaussian filter with standard deviation  $1/\delta$  in order to suppress artifacts due to the finite  $t_{max}$  [Fig. 2(b)]. Because  $p_{\psi}(\omega)$  in Eq. (3) is the density of states weighted by the overlap of  $|\psi\rangle$  with the eigenstates, its width is given by the energy fluctuations  $g\sqrt{L}$  (see Supplemental Material [31]). Hence, in order for the frequency range of the discrete Fourier transform to cover  $p(\omega)$ , the time step needs to be scaled as

 $\Delta t \propto 1/\sqrt{L}$ . The width of the Gaussian filter  $\delta$  can be chosen independently of *L* and, hence,  $t_{\text{max}} \sim 1/\delta$ . In fact, even a scaling  $\delta \propto \sqrt{L}$  leads to convergence (with  $t_{\text{max}} \sim 1/\sqrt{L}$ ) leaving the number of time-step evaluations constant [13,33].

From  $p_{\psi}(\omega)$ , we classically evaluate  $p_{\psi}(T)$  according to Eq. (6) in step (2e). In this evaluation, small errors introduced by experimental errors or numerical imprecision are exponentially amplified at low temperatures for large negative  $\omega$  and low T. To mitigate this problem, we set  $p_{\psi}(\omega) =$ 0 when  $p_{\psi}(\omega) < p_{\text{cut}}$ . Moreover,  $p_{\psi}(T)$  is centered around  $E_{\psi} = \langle \psi | \hat{H} | \psi \rangle$ , which can be problematic as  $E_{\psi} \propto L$  if  $\psi$ has a large overlap with states on the edges of the spectrum. This is, for example, the case for the totally polarized state in Fig. 2(a). In order to resolve the fast oscillation,  $\Delta t \propto 1/L$ would have to be chosen. However, we can circumvent this problem by shifting the zero of the frequency by  $E_{\psi}$  when evaluating Eqs. (7) and (6). This guarantees that  $\Delta t \propto 1/\sqrt{L}$ .

Having evaluated  $p_{\psi}(T)$ , we can now accept the state with probability  $p_{\psi'}(T)/p_{\psi}(T)$  (step 2e) and store the value of  $O_{\psi}$  of the state after the acceptance step [Fig. 2(c)]. After repeating the importance sampling iteration for  $N_{\rm MC}$ times, we evaluate thermal observables by averaging the  $O_{\psi}$ [step (3)]. For the LTFIM, we evaluate the squared magnetization  $(\hat{S}^z/L)^2$  as well as the Binder cumulant (3/2) –  $\langle (\hat{S}^z)^4 \rangle / (2 \langle (\hat{S}^z)^2 \rangle^2)$  by calculating a power of the magnetization  $\hat{S}^z = \sum_i \hat{\sigma}_i^z$  in the importance sampling. The Binder cumulant is a standard observable for the detection of Ising phase transitions as the Binder cumulant approaches 1 (0) in the ferromagnetic (paramagnetic) phase [34]. Because importance sampling creates correlated samples, we use a jackknife binning analysis to determine error bars. In Fig. 2(d) we show that these errors scale as  $1/\sqrt{N_{\rm MC}}$  as expected from the central limit theorem.

Quantum-inspired classical algorithm. The algorithm described above can also be used as a purely classical method, requiring an exact method for calculating the Loschmidt echoes  $G_{\psi}(t)$  up to short times. We use the time-dependent variational principle (TDVP) for MPS [35,36], which can be readily applied to systems with long-range interactions. In order to reach the required times, a relatively small bond dimension  $\chi = 15$  is sufficient (see Supplemental Material for convergence [31]).

In Fig. 3 we show the observables obtained from the algorithm for system sizes L = 8-64 and compare them to exact results from matrix-product operator based imaginary time evolution [37] of a purified MPS [38]. We find excellent agreement even in the vicinity of the phase transition. To simulate the algorithm, we used a time step  $J\Delta t = 0.1-0.05$ , approximately following the scaling  $\Delta t \propto 1/\sqrt{L}$  noted above. The maximum time  $Jt_{\text{max}} = 2$  (1) for L = 8-32 (64) is small. We chose a filter width  $\delta/J = 2$  (4) for L = 8-32 (64), a cutoff  $p_{\text{cut}} = 10^{-6}$ , and averaged over 10–19 independent MC runs with  $N_{\text{MC}} = 6000$ , 10 000, 2000, 4000 iterations each for L = 8, 16, 32, 64, where the first 1000 iterations were discarded as burn-in of the Markov chain.

Measurement of Loschmidt echo with trapped ions. While we have now shown that the algorithm can detect the phase transition in the LTFIM once the Loschmidt echoes  $G_{\psi}(t)$  are known, we have yet to show how  $G_{\psi}(t)$  can be measured



FIG. 3. Detecting the phase transition in the LTFIM. Data points for (a) the squared magnetization and (b) the Binder cumulant are obtained by the quantum-inspired classical algorithm discussed in the main text. In this algorithm the Loschmidt echoes  $G_{\psi}(t)$  are numerically simulated and could alternatively be obtained from a quantum simulator [cf. blue box in Fig. 1(c)]. Error bars are estimated from a jackknife binning analysis.

in a trapped ion quantum simulator [step (2b) in the algorithm above]. To this end, consider the polar decomposition of  $G_{\psi}(t) = re^{i\varphi}$ . The absolute value r is given by the probability of measuring  $|\psi\rangle$  after time evolving  $|\psi\rangle$  for a time t, i.e.,  $r^2 = |\langle \psi | e^{-i\hat{H}t} | \psi \rangle|^2$ , which has previously been measured in trapped ions [39]. The phase  $\varphi$  can be obtained from a Ramsey-type experiment by interfering a state which evolves under  $\hat{H}$  with one that does not. To engineer such a state, we introduce a shelving state  $|s\rangle$  which does not couple to the qubit levels under  $\hat{H}$ , such as one of the  $D_{5/2}$  Zeeman sublevels in  ${}^{40}Ca^+$  [40] or the  ${}^{2}D_{5/2}$  state in  ${}^{171}Yb^+$  [41]. This allows us to obtain the phase  $\varphi$  by rotating into a superposition between  $|\psi\rangle$  and  $|s \cdots s\rangle$ , evolving it in time, rotating back and measuring the return probability to  $|\psi\rangle$  (see Supplemental Material [31] and Ref. [13]). However, this superposition is a Greenberger-Horne-Zeilinger (GHZ) type state, which is in general difficult to prepare.

To avoid the creation of a GHZ state, we can instead use a sequence of Ramsey experiments, akin to the sequential protocol proposed in Ref. [13]: For each  $j \in [0, L-1]$ , we consider the state  $|\psi_i\rangle$ , in which the *j* leftmost ions are in the qubit state corresponding to  $|\psi\rangle$  and the rest are in the shelving state  $|s\rangle$  (cf. Fig. 4). The phase difference  $\Delta \varphi_{j+1} =$  $\varphi_{i+1} - \varphi_i$  of the Loschmidt echoes of two states is then obtained through the following Ramsey experiments:

(1) For a set of phases  $\theta$ :

(a) Prepare the state  $|\psi_i\rangle$ .

- (b) Act with a single ion operation  $\hat{V}_{j+1}(\theta)$  on ion j+1:  $\hat{V}_{j+1}(\theta) |\psi_j\rangle = \frac{1}{\sqrt{2}}(|\psi_j\rangle + e^{i\theta} |\psi_{j+1}\rangle).$
- (c) Evolve with  $\hat{H}$  for time *t*. (d) Act with  $\hat{V}_{j+1}^{\dagger}(0)$  on ion j + 1.



FIG. 4. Ramsey protocol and robustness. (a) Pseudocircuit diagram of the Ramsey protocol for measuring the phase. For j = 0, j = 1, etc., there are L - 1, L - 2, etc., additional ions in the  $|s\rangle$ state which we do not display for simplicity as the time evolution acts trivially as an identity on these states. (b) Results of a full simulation of the Ramsey protocol for a finite number of measurement repetitions. Inset: work distribution of a totally polarized state, showing the influence of the noise. Gray dashed line illustrates  $p_{cut}$ . We used 10<sup>4</sup> MC iterations,  $\delta = 1$ , L = 10,  $Jt_{max} = 4$ ,  $J\Delta t = 0.1$ , cut  $p_{\text{cut}} = 5 \times 10^{-2}$  (100 shots),  $8 \times 10^{-4}$  (100 000 shots).

(e) Measure the return probability  $M(\theta)$  to state  $|\psi_i\rangle$ :

$$M(\theta) = \frac{1}{4} \left( r_j^2 + r_{j+1}^2 + 2r_{j+1}r_j \cos(\theta + \Delta\varphi_{j+1}) \right).$$
(8)

(2) Fit the measured  $M(\theta)$  to Eq. (8), to obtain the phase difference  $\Delta \varphi_{j+1}$ . We found this fit to perform best if  $r_j$  and  $r_{i+1}$  are also measured, such that  $\Delta \varphi_{i+1}$  is the only unknown.

(3) Add up all L phase differences to find  $\varphi$ . To see this, we use that  $\varphi_0 = 0$  because  $G_{|s \dots s\rangle} = 1$ , which gives

$$\varphi = \sum_{j=0}^{L-1} \Delta \varphi_{j+1}.$$
(9)

We have therefore found an algorithm which determines the phase of  $G_{\psi}(t)$  using O(L) single-ion Ramsey experiments, that can be implemented directly on current experiments.

Robustness. In order to demonstrate the robustness of the algorithm, we consider several sources of error in the measurement of  $G_{\psi}(t)$  in the following. In an experiment,  $M(\theta)$  and r can only be determined up to a precision  $\sim 1/\sqrt{N_s}$  due to quantum projection noise [42] introduced by measuring  $N_s$  times. In order to show that this algorithm works even for a finite  $N_s$ , we simulated the above Ramsey protocol and show the results in Fig. 4(b). We used  $N_s$  repetitions for  $r_i$  as well as  $N_s/4$  repetitions for  $M(\theta)$  for four equally spaced values of  $\theta \in [0, \pi]$ , such that  $2N_sL$  measurements are performed per time point. The algorithm therefore

requires  $N_{\rm MC}(t_{\rm max}/\Delta t)2LN_s$  measurements, which reduces to  $N_{\rm MC}(t_{\rm max}/\Delta t)N_s$  when using the GHZ protocol. The noise leads to errors when  $p_{\psi}(\omega)$  is small, which we remove by using a cut  $p_{\rm cut} \propto 1/\sqrt{N_s}$  [cf. Fig. 4(b) inset]. We find good results even for a small number of measurements. At low temperatures, more measurements are needed as small values of  $p_{\psi}(\omega)$  for large negative  $\omega$  become more important due to the factor  $e^{-\omega/T}$  in Eq. (6).

Another source of imprecision is errors in SPAM. The leading contribution is given by uncorrelated single-qubit state assignment errors with probability p. The return probability  $|G_{\psi}(t)|^2$  then has an error of  $1 - (1 - p)^L$ . Current trappedion devices have  $p \approx 10^{-3}$  (see, e.g., Ref. [11]), such that the error on  $|G_{\psi}(t)|^2$  is 5% for L = 50. Uncorrelated SPAM errors can be corrected by multiplying the measured probability distribution with the tensor product of the inverse single-qubit measurement error matrices (see, e.g., [43,44]).

The effect of dephasing noise may be modeled by an exponential decay of the Loschmidt amplitude  $G_{\psi}(t) \rightarrow G_{\psi}(t) \exp(-\gamma Lt)$ . The resulting  $1/\omega^2$  tails in the work distribution lead to errors at large positive or negative frequencies. Similar to measurement noise, this error can in principle be mitigated by using a suitably small cutoff. If  $\gamma$  is known, multiplying the measured signal with  $\exp(\gamma Lt)$  might lead to an even more efficient removal of decoherence effects. However, we expect decoherence effects not to be very strong as the timescales  $Jt \sim 1-2$  are small compared to those routinely employed in trapped-ion quantum simulators.

*Discussion and outlook.* To summarize, we have applied a protocol to probe finite-temperature observables in analog quantum simulators of spin systems. We benchmarked the protocol by studying a thermal phase transition in the transverse-field Ising mode with long-range interactions as realized with trapped ions. This algorithm is well suited to current noisy devices as only short times are needed, independent of system size. Due to the importance sampling, the main bottleneck is the number of measurements. Current trappedion simulators employing <sup>40</sup>Ca<sup>+</sup> (<sup>171</sup>Yb<sup>+</sup>) have a short time on the order of 100 ms (10 ms), such that  $N_{MC} = 1000$  samples seem to be realistically achievable.

The algorithm can be directly employed in other platforms. Rydberg atoms in tweezers have access to efficient GHZ state preparation [45], shelving states as well as a high measurement repetition rate. They could probe two-dimensional or three-dimensional interacting Ising models in frustrated lat-

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tices, for which classical algorithms are plagued by the sign problem, thus providing a route to obtain quantum advantage [46]. In these cases, finding more efficient update rules for the importance sampling, e.g., by using the quantum simulator for the proposal step [47], might be advantageous. Moreover, the free energy of the system can in principle be evaluated via  $F = -T \ln Z$  using, for example, multicanonical sampling strategies [48]. The range of models accessible in analog simulators can be extended by applying prethermalization (e.g., the XY model in trapped ions [49]), or Floquet engineering (e.g., Heisenberg models [50,51]).

While our primary motivation was the application to quantum simulators, this algorithm can also be used to study previously inaccessible regimes with MPS methods due to the low requirement on the maximum time and hence the entanglement.

*Note added.* During the completion of this manuscript, a work appeared proposing this algorithm as a quantum-inspired classical algorithm for MPS [33].

Data analysis and simulation codes are available on Zenodo upon reasonable request [52].

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