

Variational quantum algorithms 1



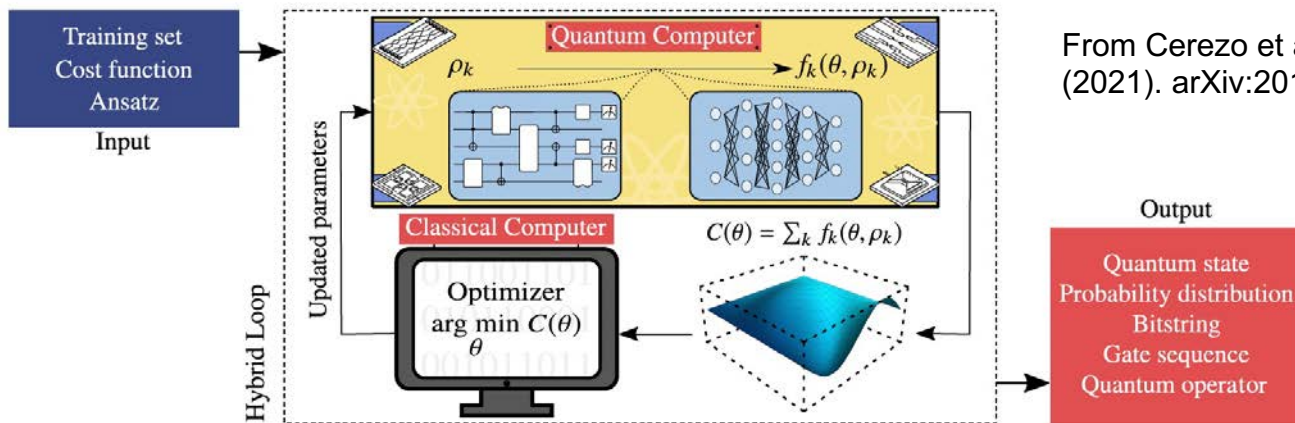
Peter P. Orth (Iowa State University & Ames Laboratory)



U.S. DEPARTMENT OF
ENERGY

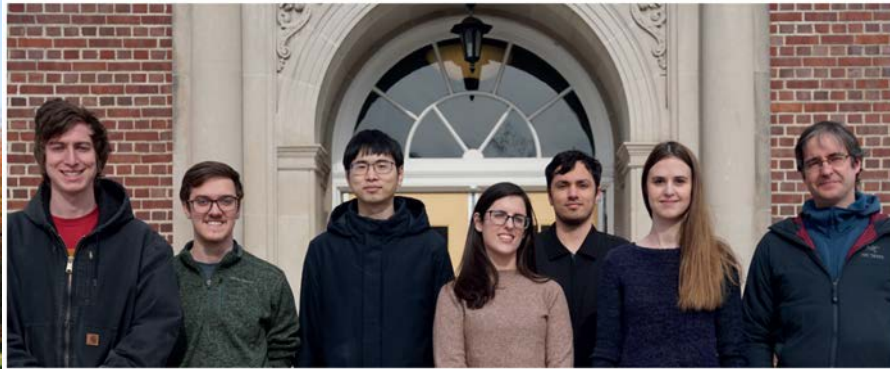
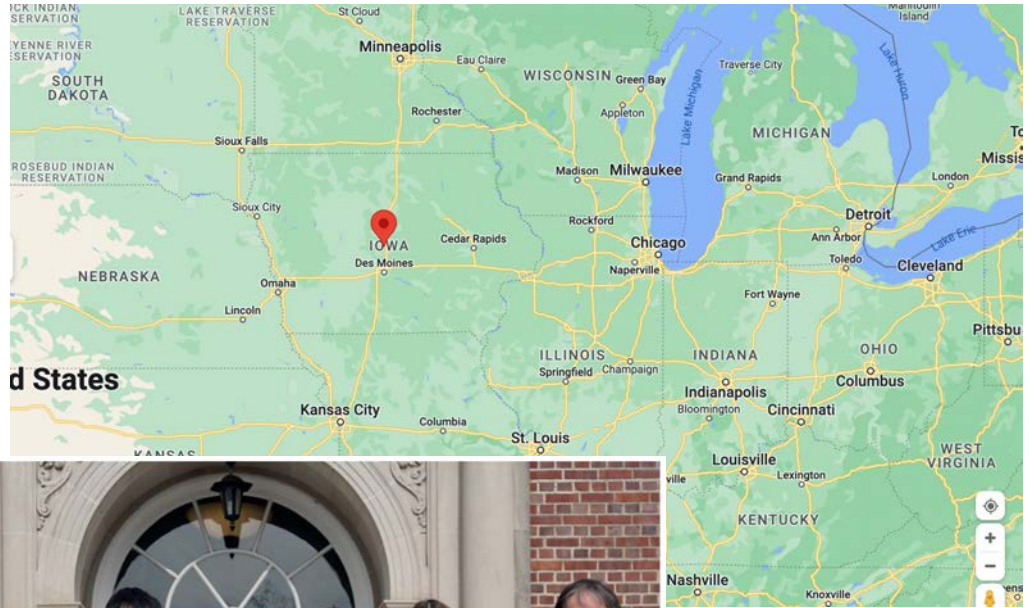
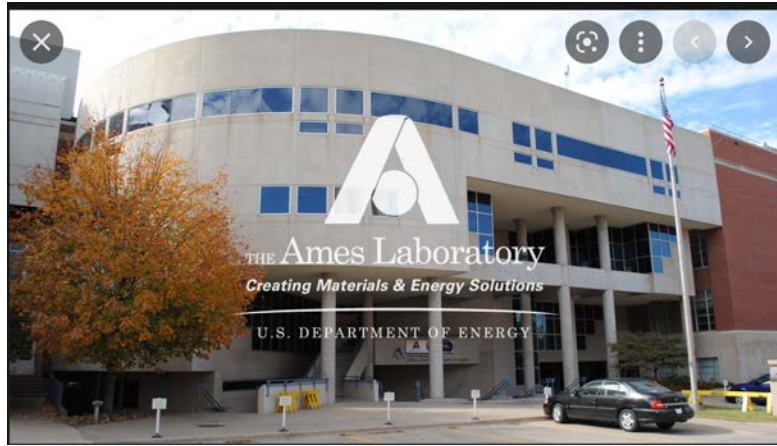
Office of
Science

SQMS/GGI Summer School, Florence, Italy, 26 July 2022



From Cerezo et al., Nat. Phys. Reviews
(2021). arXiv:2012.09265

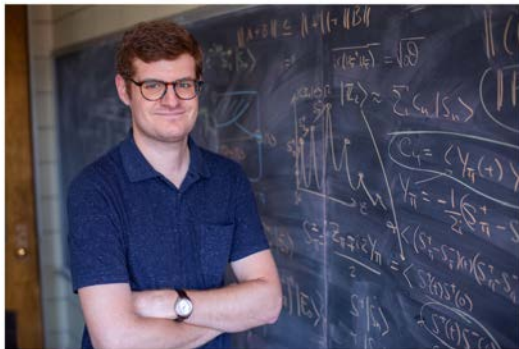
Ames National Lab & Iowa State University



(Part of) my group earlier this year

Check out:
www.physastro.iastate.edu

Quantum Computing at Ames National Laboratory & Iowa State



Thomas Iadecola studies how states of matter emerge from collections of atoms and subatomic particles and recently contributed theoretical work and data analysis to a paper published by the journal Nature. Larger photo. Photo by Christopher Giannon.



Srimoyee Sen



Yongxin Yao (materials simulations, quantum computing)

Joint group meeting of my group and the groups of Prof. Flint and Prof. Iadecola

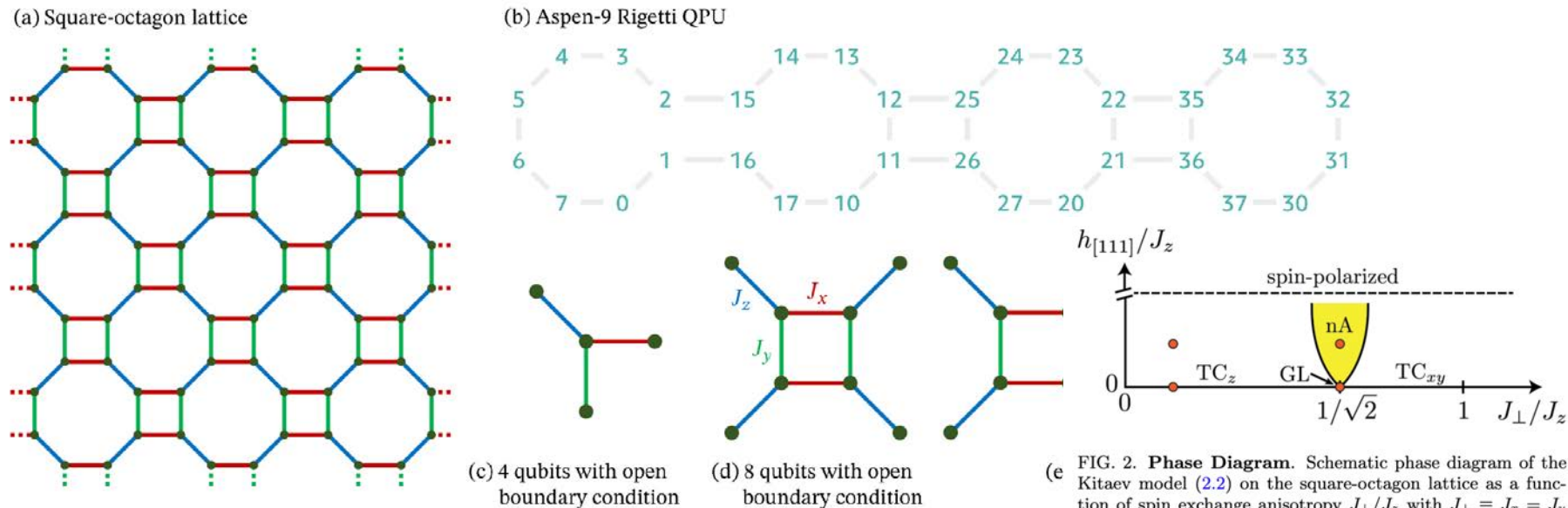
Tom Iadecola (Nonequilibrium dynamics, topological field theory, quantum computing) Srimoyee Sen (QCD, quantum field theory)

- Ames Lab participates in two national quantum initiative centers of the Department of Energy (SQMS & C2QA)
- Strong collaborations across fields and between theory and experiment at Ames National Lab



VQE for Kitaev spin model

From:
Li et al. (SQMS), arXiv (2021)



Kitaev model on square-octagon lattice matches Rigetti's QPU geometry. No SWAP gates needed as connectivities match.

VQE ansatz

From:
Li et al. (SQMS), arXiv (2021)

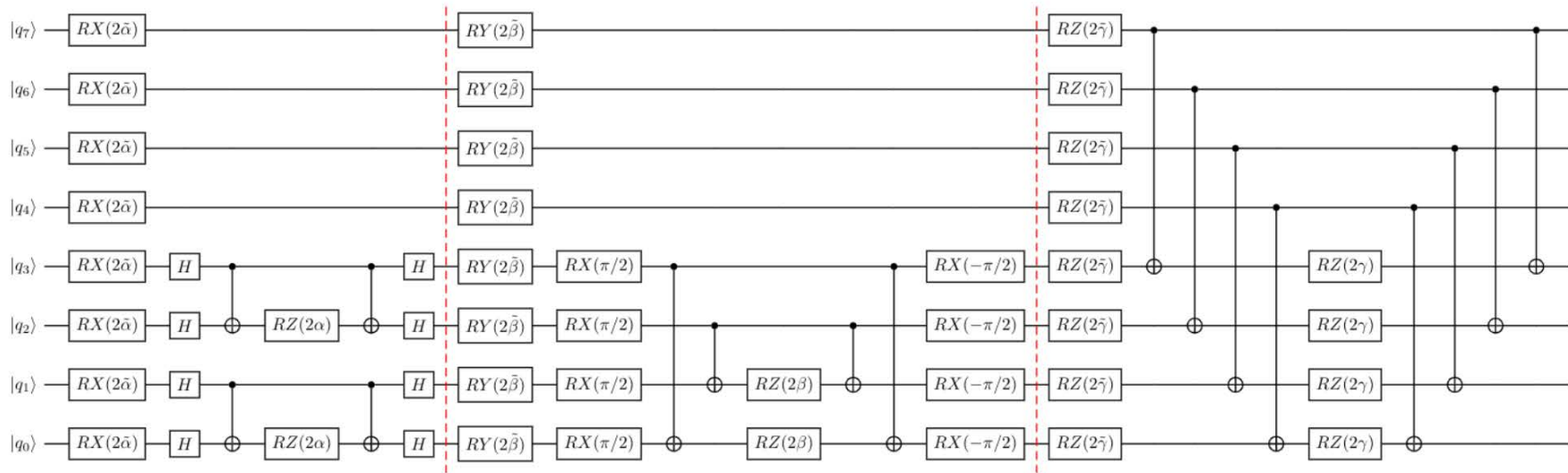
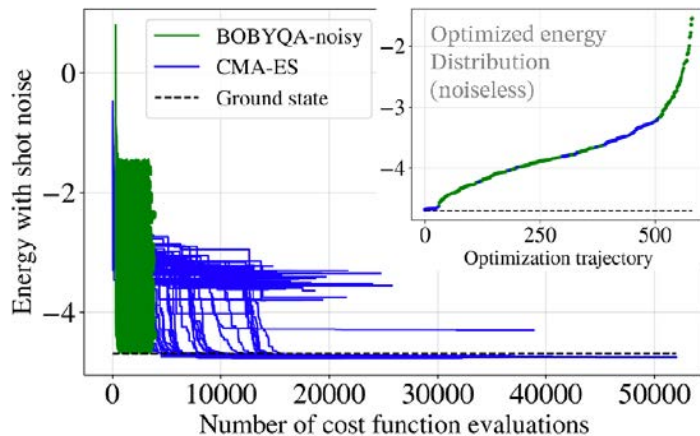
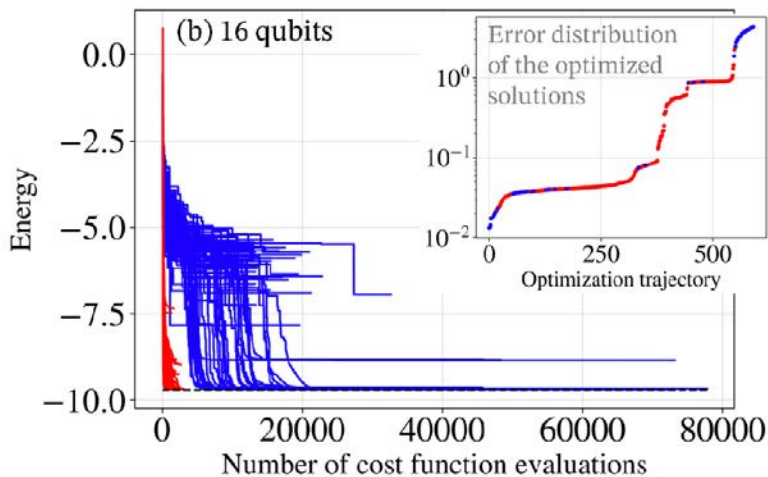


FIG. 3. **HVA with one layer on eight qubits.** The Hamiltonian Variational Ansatz (HVA) with one layer on eight qubits, split into commuting blocks. The first block corresponds to the operation $e^{-i\tilde{\alpha} \sum_q X_q} e^{-i\tilde{\alpha} \sum_{(i,j) \in X\text{-links}} X_i X_j}$, the second to $e^{-i\tilde{\beta} \sum_q Y_q} e^{-i\tilde{\beta} \sum_{(i,j) \in Y\text{-links}} Y_i Y_j}$, and the third to $e^{-i\tilde{\gamma} \sum_q Z_q} e^{-i\tilde{\gamma} \sum_{(i,j) \in Z\text{-links}} Z_i Z_j}$. For the circuit shown here, we used $X\text{-links} = \{(q_0, q_1), (q_2, q_3)\}$, $Y\text{-links} = \{(q_0, q_3), (q_1, q_2)\}$, and $Z\text{-links} = \{(q_0, q_4), (q_1, q_5), (q_2, q_6), (q_3, q_7)\}$.

Noiseless and noisy (shots) VQE optimization

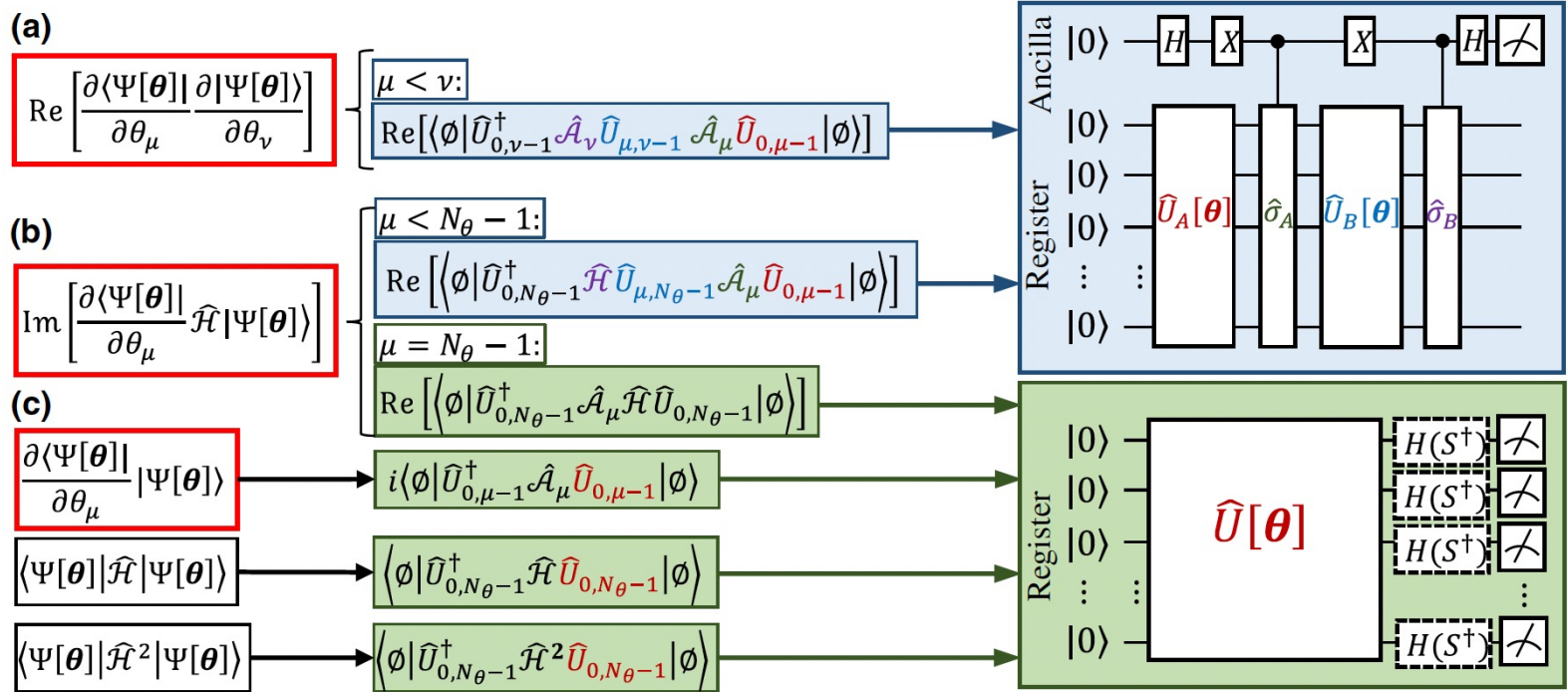
From:
Li et al. (SQMS), arXiv (2021)



Noiseless: 16 qubits
Noisy: 8 qubits (8000 shots)

Optimizer	Error (noiseless)	Measured deviation	Cost function evaluations
BFGS, 501 initial values	0.45069	0.42052	mean: 747, max: 1994
BOBYQA, 501 initial values	0.27485	0.21843	mean: 471, max: 610
BOBYQA-noisy, 501 initial values	0.07989	-0.00453	mean: 3532, max: 4004
CMA-ES	0.02416	-0.06462	37570
CMA-ES, 80 initial values	0.01610	-0.07125	mean: 21042, max: 52000
Dual annealing	0.04534	-0.01631	60101
SPSA	0.00612	0.00879	100000 (cutoff)

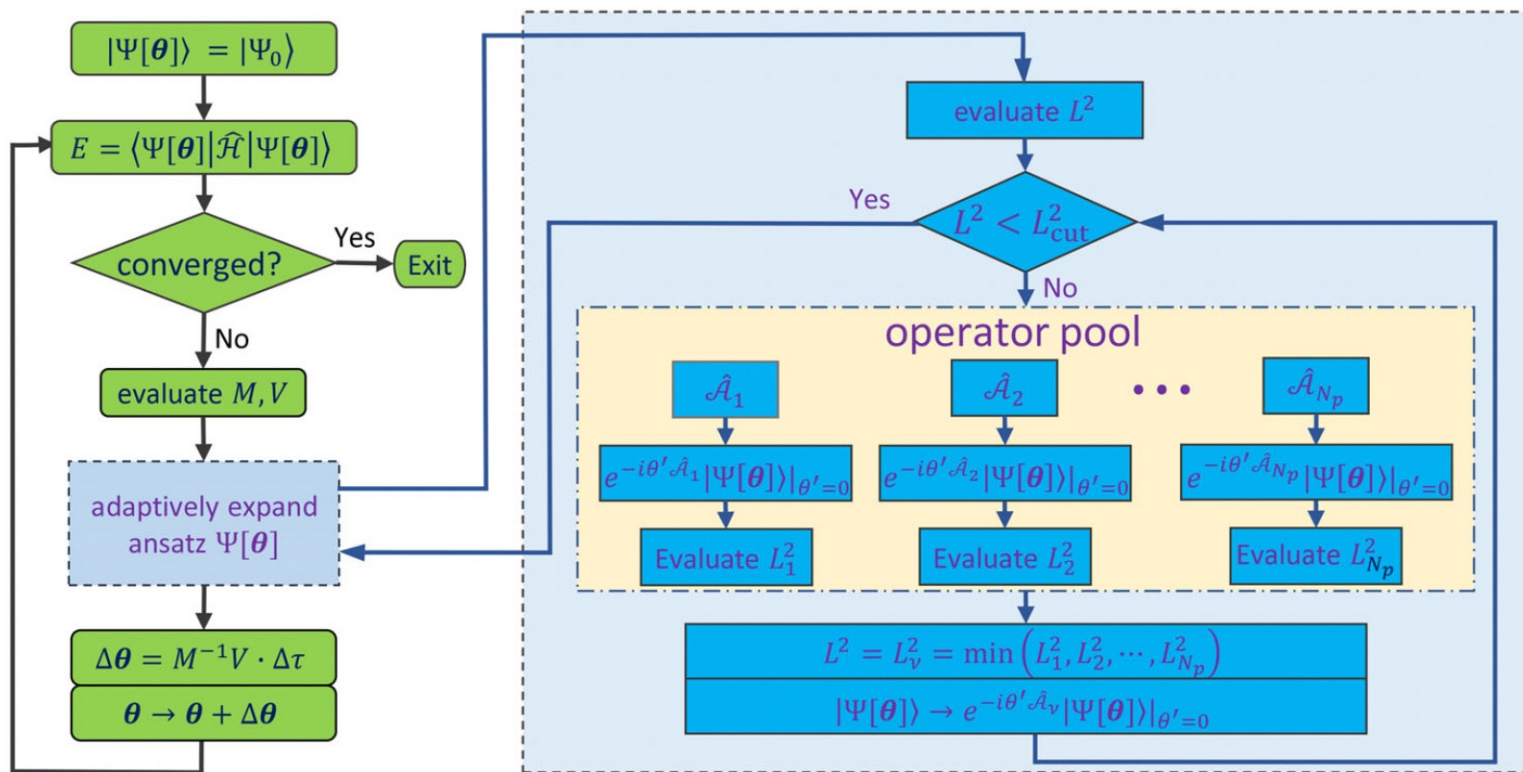
Quantum circuits for measurement of $\text{Re}[G]$ and energy gradient



From:
Yao, ...,
PPO, PRX
Quantum
(2021).

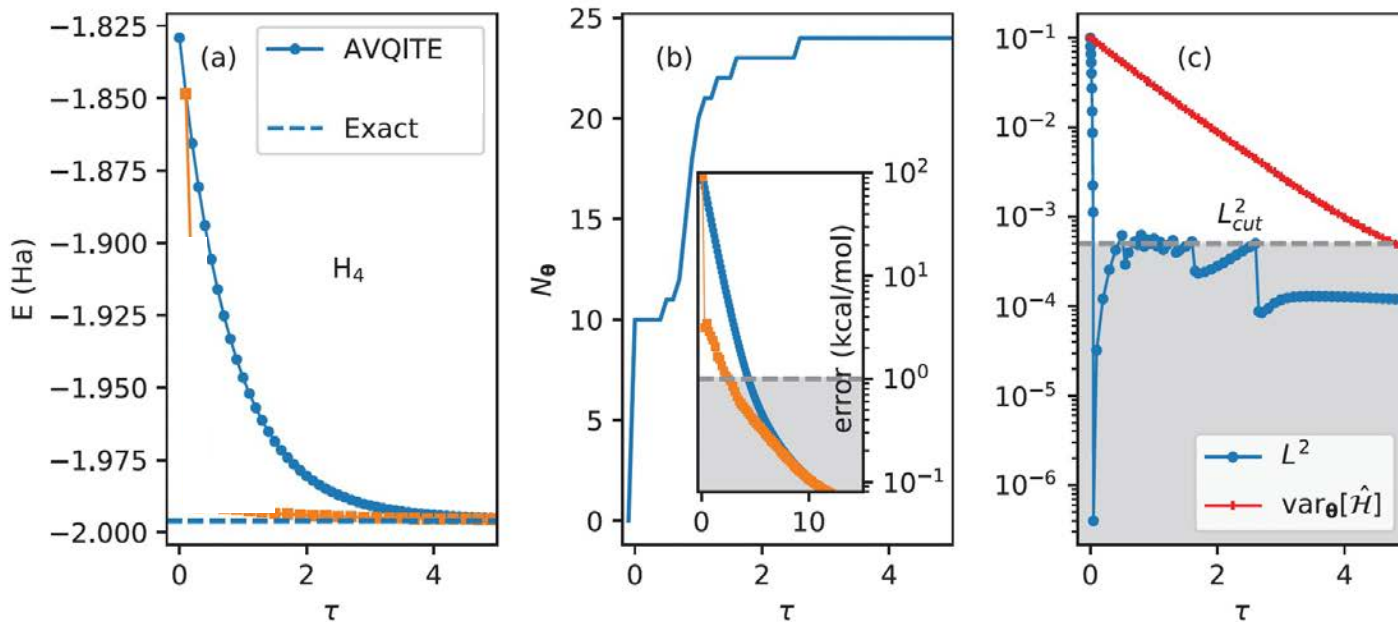
These circuits are for real-time evolution. For imaginary time evolution one applies an additional S^\dagger gate after the first Hadamard gate (and removes the X gates) in order to measure the imaginary part in the Hadamard test.

Adaptive variational quantum imaginary time evolution



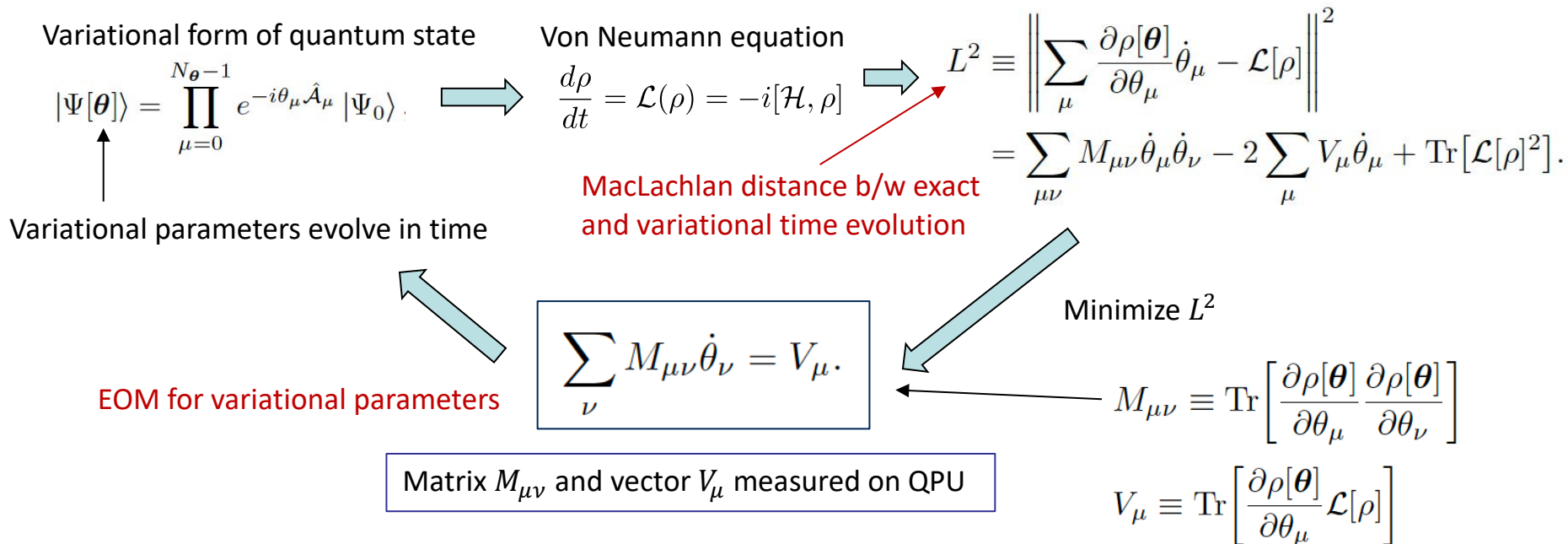
From:
Gomes,
..., PPO,
Yao, Adv.
Qu. Tech
(2021).

Adaptive variational quantum imaginary time evolution



From:
Gomes,
..., PPO,
Yao, Adv.
Qu. Tech
(2021).

Time-dependent variational quantum algorithms



Only so good as the ansatz can follow the dynamics

☞ How to select an efficient yet flexible variational ansatz?

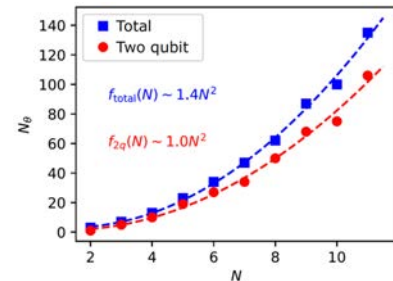
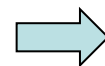
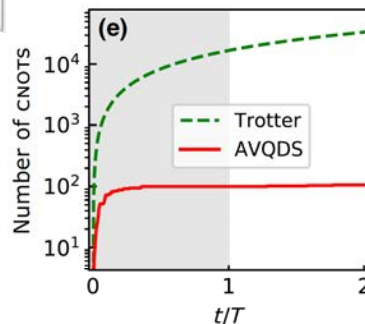
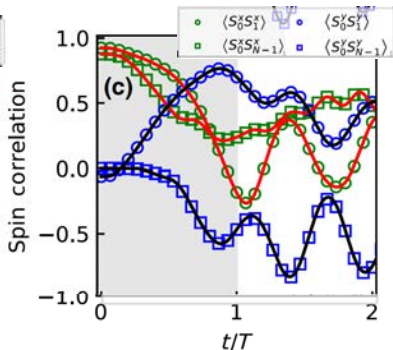
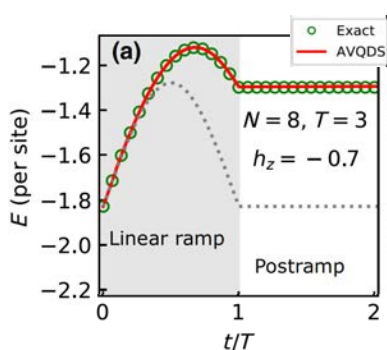
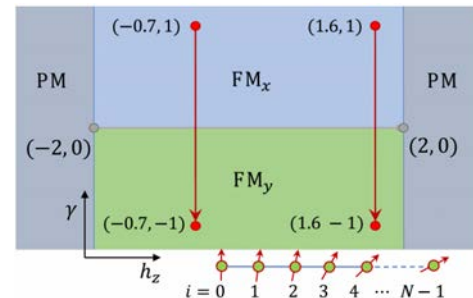
[1] Li, Benjamin, Endo, Yuan (2019).

Application I: continuous quench in integrable spin chain

- > Linear quench of anisotropic XY chain in transverse magnetic field

$$\hat{H} = -J \sum_{i=0}^{N-2} \left[(1 + \gamma) \hat{X}_i \hat{X}_{i+1} + (1 - \gamma) \hat{Y}_i \hat{Y}_{i+1} \right] + h_z \sum_{i=0}^{N-1} \hat{Z}_i \quad \text{with} \quad \gamma(t) = 1 - \frac{2t}{T}$$

- > AVQDS follows exact solution during and after quench, shown for $N = 8$
- > Circuit depth saturates at 100 CNOTs \ll Trotter circuit depth 10^4 CNOTs
- > Can simulate system with gate depth independent of time t \Rightarrow can simulate to arbitrary times!



Saturated # of parameters

Application II: sudden quench in nonintegrable spin chain

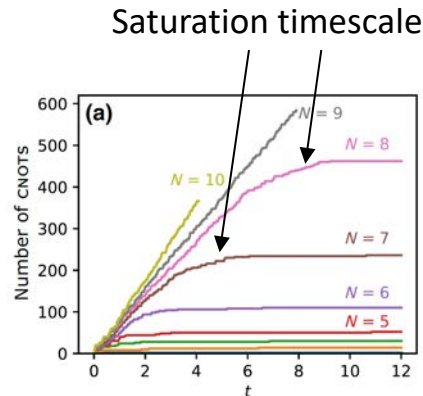
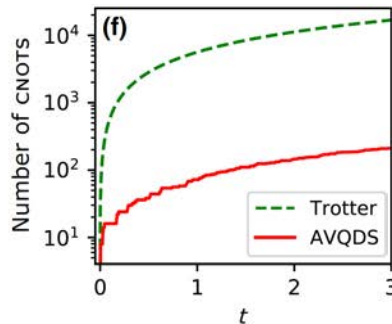
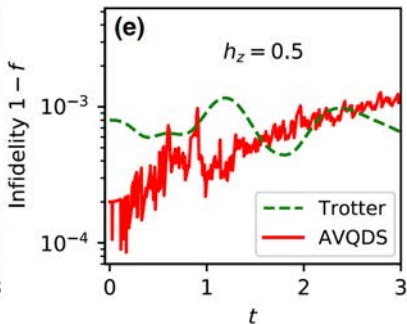
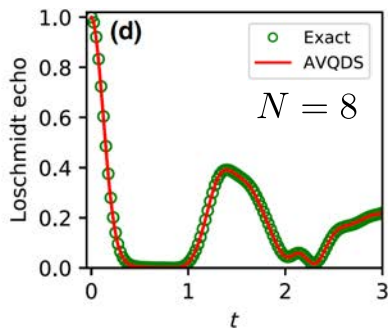
- Sudden quench in mixed-field Ising model

$$\hat{\mathcal{H}} = -J \sum_{i=0}^{N-1} \hat{Z}_i \hat{Z}_{i+1} + \sum_{i=0}^{N-1} (h_x \hat{X}_i + h_z \hat{Z}_i)$$

Loschmidt echo: $\mathcal{L}(t) = \left| \langle \Psi_0 | e^{-i\hat{\mathcal{H}}_f t} | \Psi_0 \rangle \right|^2$

Initial state: $|\Psi_0\rangle = |\uparrow \cdots \uparrow\rangle$

- AVQDS follows exact solution, shown for $N \leq 10$
- Circuit depth two orders of magnitude smaller than Trotter circuit depth
- AVQDS circuit depth saturates at system size N dependent time scale



Application II: sudden quench in nonintegrable spin chain

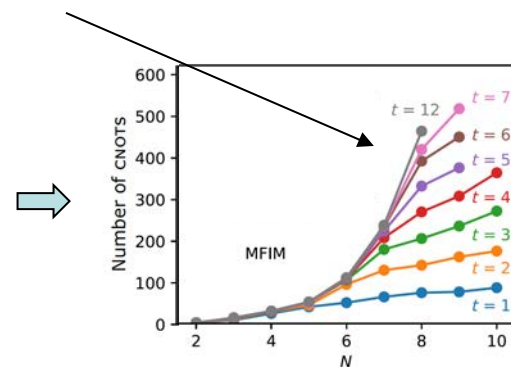
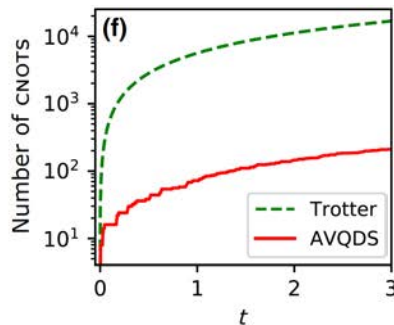
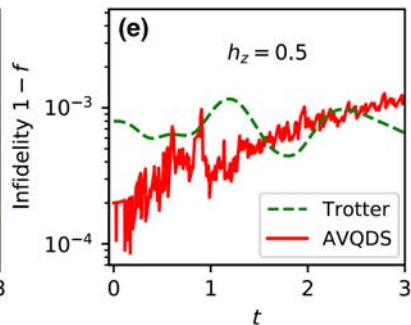
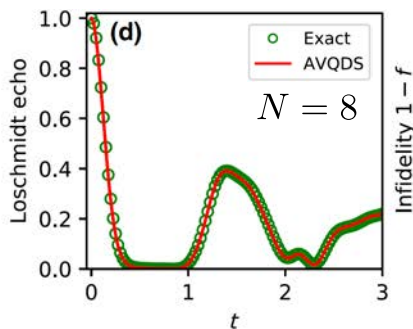
- > Sudden quench in mixed-field Ising model

$$\hat{\mathcal{H}} = -J \sum_{i=0}^{N-1} \hat{Z}_i \hat{Z}_{i+1} + \sum_{i=0}^{N-1} (h_x \hat{X}_i + h_z \hat{Z}_i)$$

Loschmidt echo: $\mathcal{L}(t) = \left| \langle \Psi_0 | e^{-i\hat{\mathcal{H}}_f t} | \Psi_0 \rangle \right|^2$

Initial state: $|\Psi_0\rangle = |\uparrow \cdots \uparrow\rangle$

- > Circuit depth two orders of magnitude smaller than Trotter circuit depth
- > Saturated AVQDS circuit depth scales exponentially with system size N
- > # measurements is main bottleneck of algorithm



VQE-X: variational quantum eigensolver for highly excited states

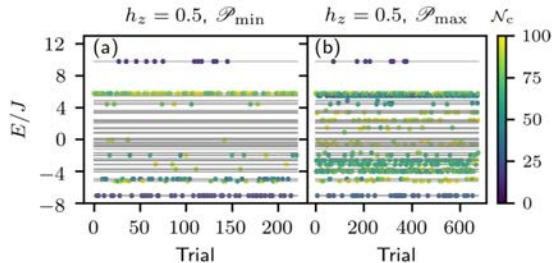
Adaptive variational quantum eigensolver for excited states

- > Long-time dynamics of observable determined by diagonal ensemble of excited states

$$\langle \mathcal{O}(t) \rangle = \sum_{n,m} c_n c_m^* e^{i(E_m - E_n)t} \langle m | \mathcal{O} | n \rangle \xrightarrow{t \rightarrow \infty} \langle \mathcal{O}(t) \rangle \approx \sum_n |c_n|^2 \langle n | \mathcal{O} | n \rangle$$

- > Adaptive variational quantum eigensolver to prepare highly excited states (VQE-X)

- > Minimize energy variance (instead of energy): $\mathcal{C}(|\psi(\theta)\rangle) = \langle \psi(\theta) | H^2 | \psi(\theta) \rangle - \langle \psi(\theta) | H | \psi(\theta) \rangle^2$



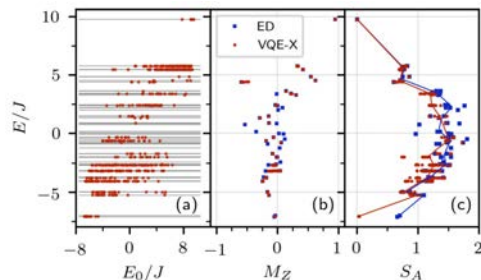
Nontrivial pool dependence

$$\mathcal{P}_{\min} = \{Y_i\}_{i=1}^N \cup \{Y_i Z_{i+1}\}_{i=1}^N$$

$$\mathcal{P}_{\max} = \{Y_i\}_{i=1}^N \cup \{Y_i Z_j\}_{i,j=1}^N \cup \{Y_i X_j\}_{i,j=1}^N$$

Full coverage of energy spectrum for operator pool with long-range Pauli strings

[1] Zhang, Gomes, Yao, PPO, Iadecola, PRB **104**, 075159 (2021).



Can investigate properties of volume law highly excited states

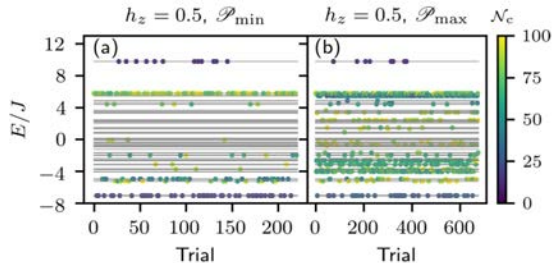
Adaptive variational quantum eigensolver for excited states

- Long-time dynamics of observable determined by diagonal ensemble of excited states

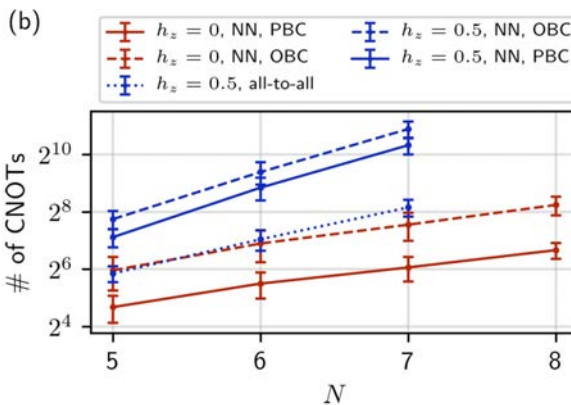
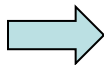
$$\langle \mathcal{O}(t) \rangle = \sum_{n,m} c_n c_m^* e^{i(E_m - E_n)t} \langle m | \mathcal{O} | n \rangle \xrightarrow{t \rightarrow \infty} \langle \mathcal{O}(t) \rangle \approx \sum_n |c_n|^2 \langle n | \mathcal{O} | n \rangle$$

- Adaptive variational quantum eigensolver to prepare highly excited states (VQE-X)

- Minimize energy variance (instead of energy): $\mathcal{C}(|\psi(\theta)\rangle) = \langle \psi(\theta) | H^2 | \psi(\theta) \rangle - \langle \psi(\theta) | H | \psi(\theta) \rangle^2$



Full coverage of energy spectrum for operator pool with long-range Pauli strings

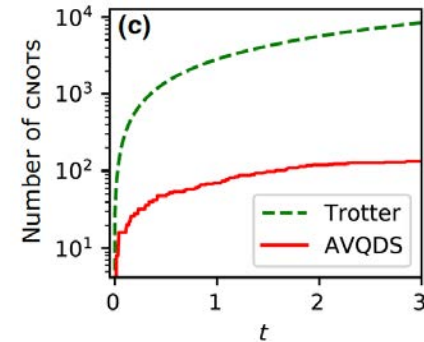


Exponential scaling of # CNOTs with system size
 ☞ relax convergence condition to represent microcanonical averages instead [1]

[1] Banuls, Huse, Cirac (2020)

Summary and outlook

- > Adaptive variational quantum dynamics simulation (AVQDS) framework
 - > Orders of magnitude shallower circuits than Trotter simulations
 - > Linear growth of # CNOTs with time initially, then saturation at e^N number of gates
 - > Allows simulation out to arbitrary times for small systems
- > Adaptive variational quantum eigensolver for highly excited states (adaptive VQE-X)
 - > Minimize energy variance allows preparing arbitrary excited states
 - > Generate volume-law states requires exponentially many parameters
 - > May be alleviated in simulations of microcanonical averages



References:

- N. Gomes *et al.*, arXiv:2102.01544 (2021)
- Y. Yao *et al.*, PRX Quantum **2**, 030307 (2021)
- F. Zhang *et al.*, PRB **104**, 075159 (2021)

Thank you for your attention!

