

Thermal state preparation via a variational quantum algorithm

Tony J.G. Apollaro, University of Malta

Overview:

- Noisy Intermediate-Scale Quantum (NISQ) Computers
- Structure of a Variational Quantum Algorithm (VQA)
- Gibbs state preparation via VQA
- Results on the quantum transverse field Ising model
- Conclusions and Perspectives



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ta' Malta**

THE NISQ ERA

John Preskill (2018) defines NISQ as:

- Noisy: imperfect control over qubits
- Intermediate-scale: 50-100 qubits
- Quantum: self-explanatory
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Factorisation of large integers

Experimental study of Shor's factoring algorithm using the IBM Q Experience

Mirko Amico, Zain H. Saleem, and Muir Kumph
Phys. Rev. A **100**, 012305 – Published 8 July 2019

✓ 15 ✓ 21 ✗ 35

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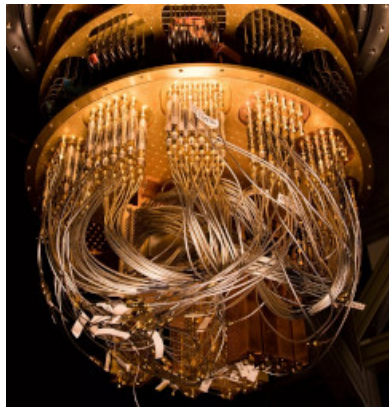
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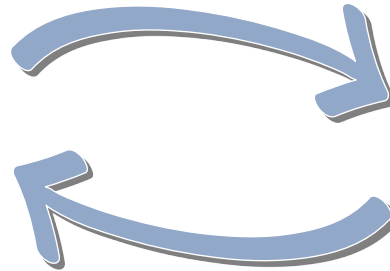
Take home message
Use a quantum computer as little as possible

Variational Quantum Algorithm (VQA)

Quantum Computer



Output: cost function

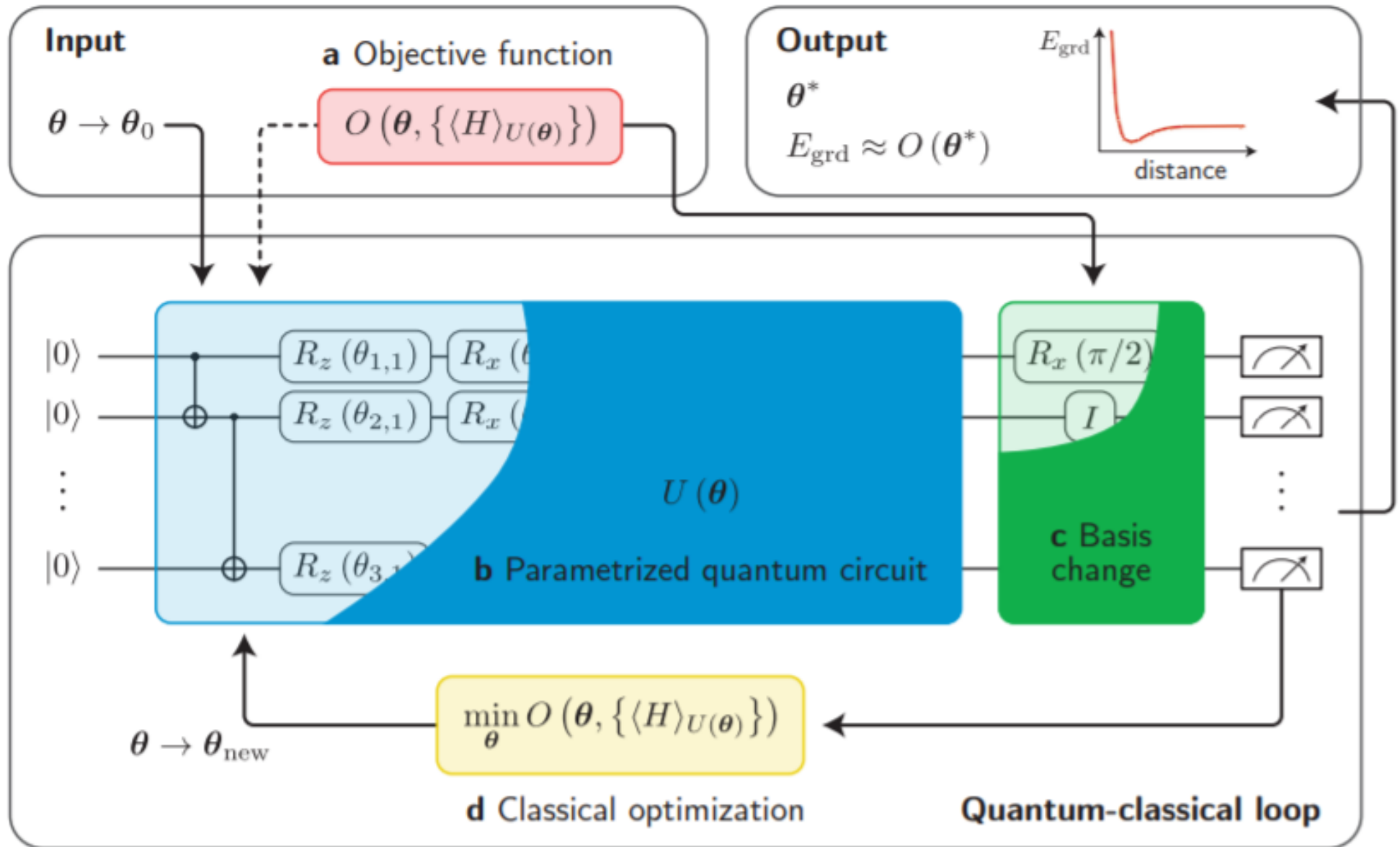


Classical optimizer



Update parameters
Call the QC

Variational Quantum Algorithm (VQA)



Bharti et al., *Noisy intermediate-scale quantum algorithms*, Rev. Mod. Phys. 94, 015004 (2022)

Structure of a VQA- A. Cost function

A1: An objective function can be defined as $\min_{\theta} O \left(\theta, \left\{ \langle \hat{H} \rangle_{U(\theta)} \right\} \right)$

A2: Given a parameterized unitary (PQC), one can expand it into

$$\langle \hat{H} \rangle_{U(\theta)} = \langle 0 |^{\otimes N} U^\dagger(\theta) \hat{H} U(\theta) | 0 \rangle^{\otimes N}$$

A3: The Hamiltonian is hence decomposed into Pauli strings

$$\hat{H} = \sum_{i=1}^N c_i \hat{P}_i, \quad \hat{P}_i = \prod_{j=1}^N \hat{\sigma}_j^{\alpha_j}$$

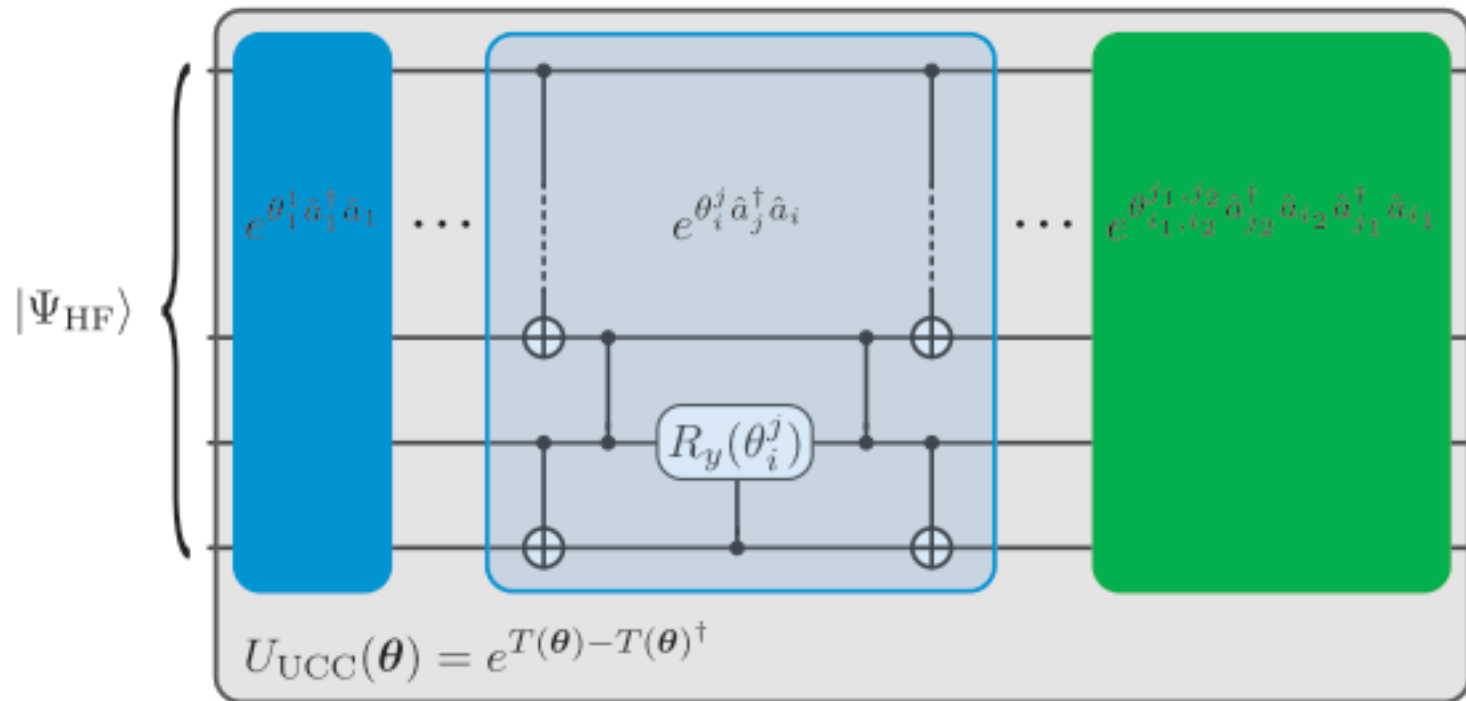
A4: The Hamiltonian expectation values is evaluated for each string

$$\langle \hat{H} \rangle_{U(\theta)} = \sum_{i=1}^N c_i \langle \hat{P}_i \rangle_{U(\theta)}$$

Structure of a VQA- B. PQC

Problem inspired vs Hardware efficient Ansatz

B1: Problem-inspired Ansatz

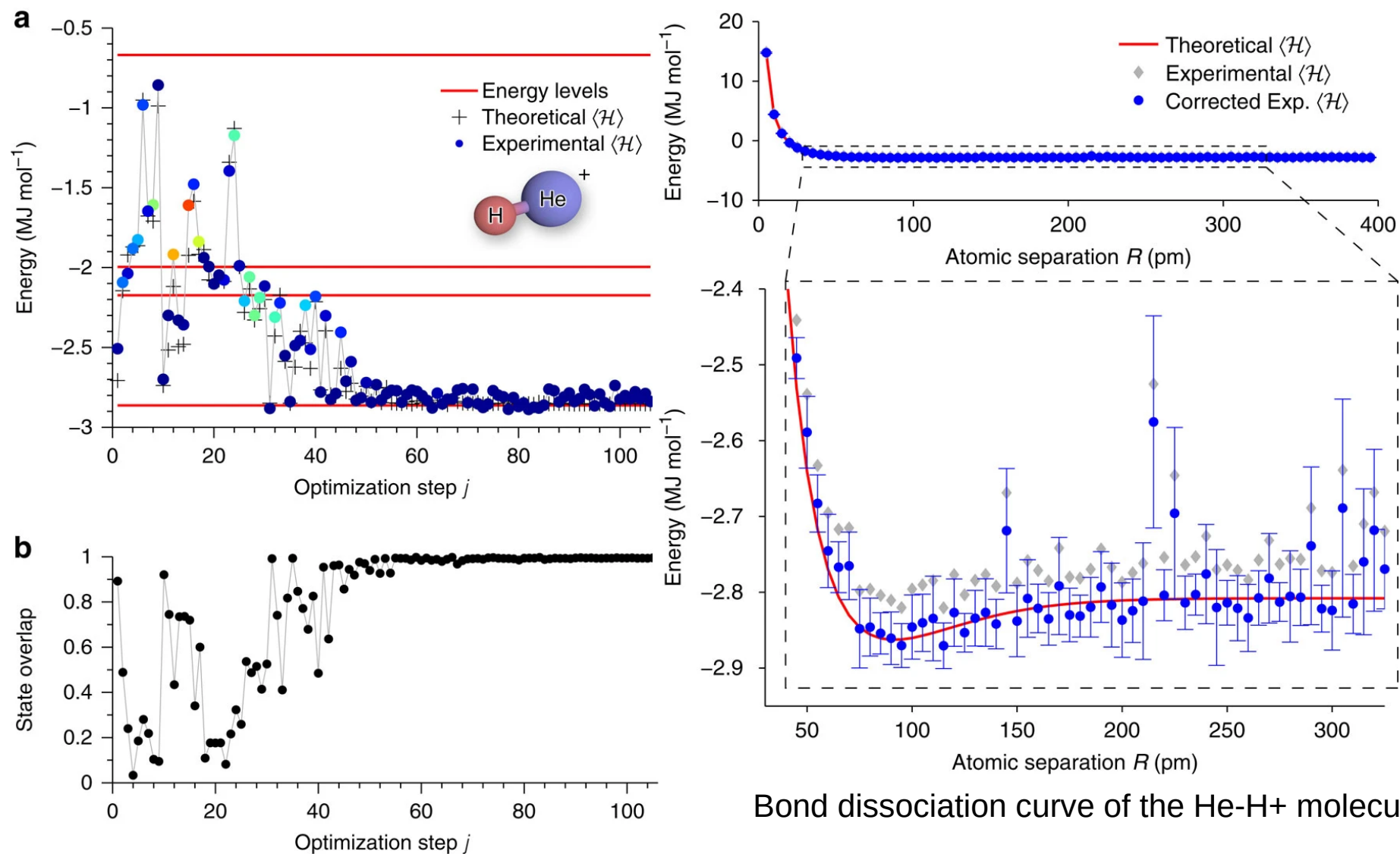


The Unitary Coupled Cluster (UCC): It adds quantum correlations to the Hartree-Fock (HF) approximation (a classically-inefficient procedure).

Other problem inspired Ansatz: QAOA, VQE, VHA,

Molecular quantum chemistry

Ground state of He-H⁺ for a specific molecular separation

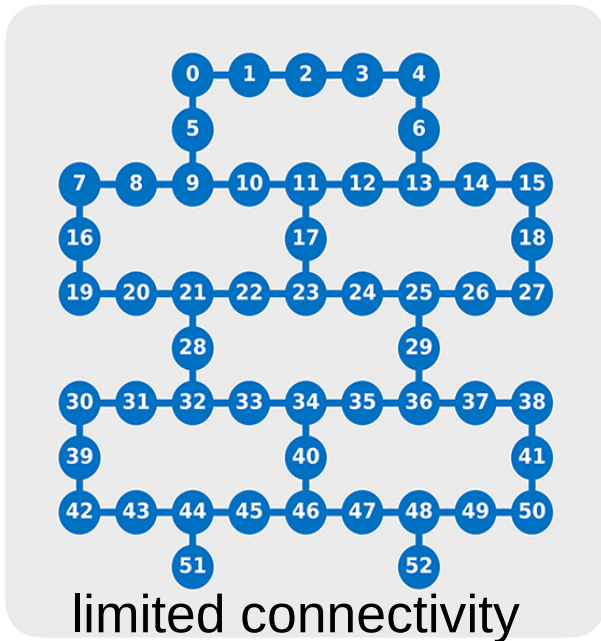


Bond dissociation curve of the He-H⁺ molecule.

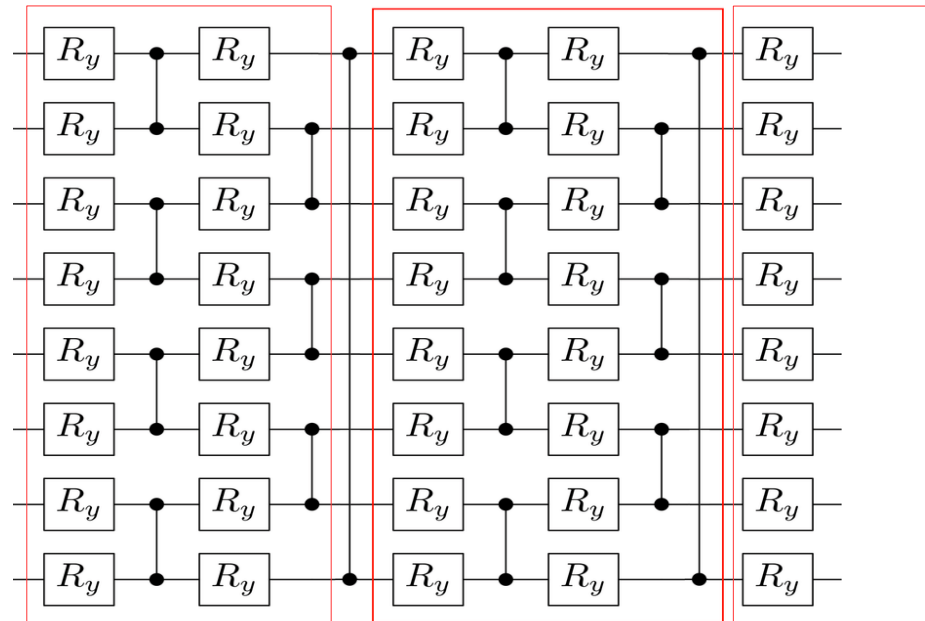
Peruzzo et al. *A variational eigenvalue solver on a photonic quantum processor.*
Nat Commun 5, 4213 (2014). <https://doi.org/10.1038/ncomms5213>

B2: Hardware-efficient Ansatz

53 Qubit Rochester Device



limited set of basis gates



1st layer

2nd layer

...

gates errors and decoherence

Qubit	Energy relaxation time T1 (μ s)	Dephasing time T2 (μ s)	Frequency (GHz)	Readout error rate	Single-qubit U2 error rate	CNOT error rate
Q0	98.33551265	46.82808058	4.641200919	5.50000e-2	3.19722e-4	cx0_1: 9.422e-3 cx1_0: 9.422e-3 cx1_2: 1.594e-2 cx1_3: 1.747e-2
Q1	66.80352570	84.08716319	4.719992563	6.05e-2	5.76695e-4	cx2_1: 1.594e-2 cx3_1: 1.747e-2
Q2	77.96055113	88.27547132	4.761962111	4.04999e-2	6.51329e-4	cx3_4: 1.208e-2
Q3	93.81392649	86.04122557	4.687003753	5.04999e-2	5.11595e-4	cx4_3: 1.208e-2
Q4	57.44885729	40.31605568	4.923978085	4.00000e-2	6.05968e-4	
Ave.	78.87247466	69.10959927	4.746827487	4.93e-2	5.33062e-4	1.37e-2

Limited Quantum Volume
8192
Quantinuum
(20 qubits)

Structure of a VQA

C. Measurements & D. Optimisation

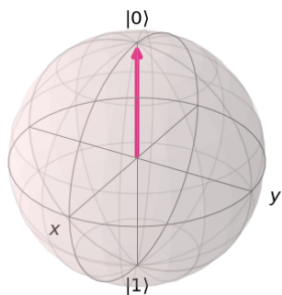
On a QC measures are performed in the computational basis

$$\langle \hat{\sigma}^z \rangle = p_0 - p_1$$

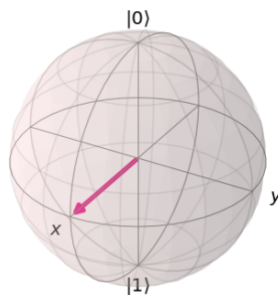
Measurements on a different basis are preceded by a rotation

$$\langle \hat{\sigma}^x \rangle = \langle R_y^\dagger \left(\frac{\pi}{2} \right) \hat{\sigma}^z R_y \left(\frac{\pi}{2} \right) \rangle$$

Before Applying RY Gate



After Applying RY Gate



Pauli strings measured accordingly

$$\langle \hat{P} \rangle = \left\langle \prod_{j=1}^N \hat{\sigma}_j^z \right\rangle R_j^\alpha$$

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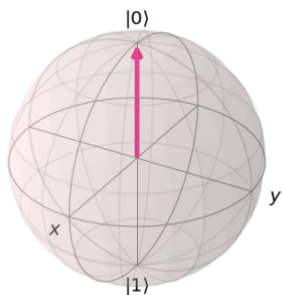
On a CC the cost function is minimized

$$\min_{\theta} O \left(\theta, \left\{ \langle \hat{H} \rangle_{U(\theta)} \right\} \right)$$

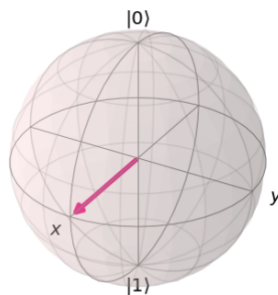
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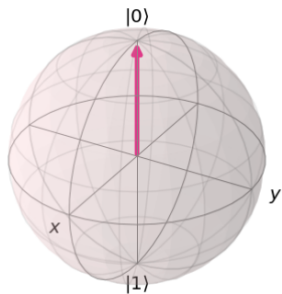
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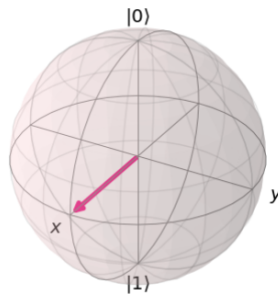
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- Gradient-based approaches (L-BFGS, QNG, SGD,...): Local optimisers that find efficiently a local minimum.
- Gradient-free approaches (Bayesian optimisation, Nelder-Mead method, genetic algorithms,...): Global optimisers that find efficiently a global minimum, but very costly.

Structure of a VQA

C. Measurements & D. Optimisation

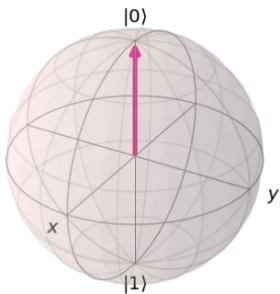
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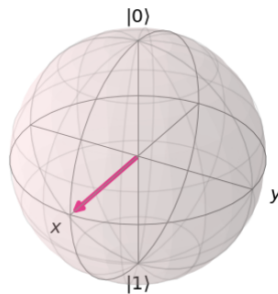
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However, on a QC:

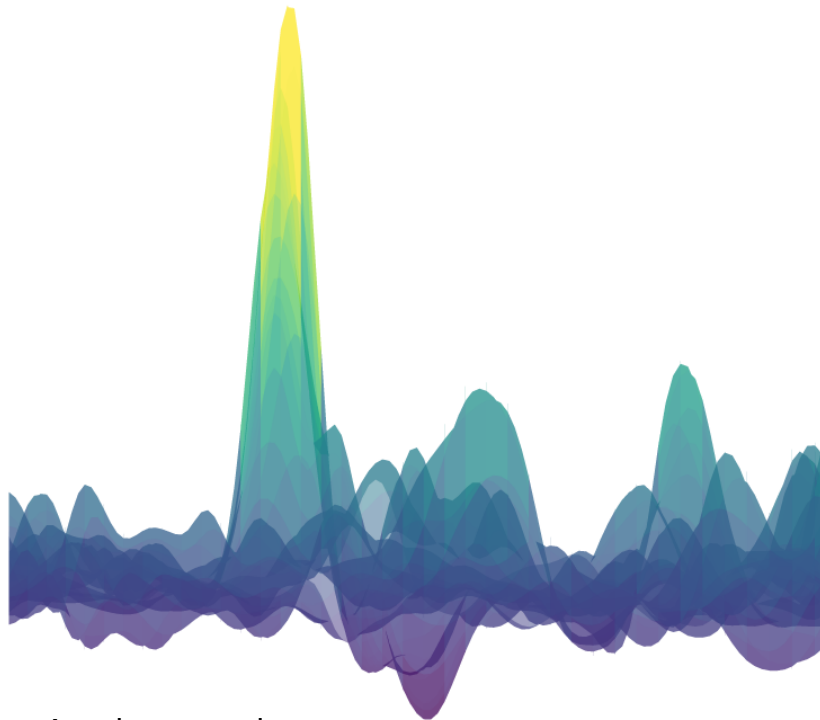
- Sampling noise and gate noise on NISQ devices disturb the landscape of the objective function
- The precision of the measured expectation value is limited by the sample shot number

Structure of a VQA

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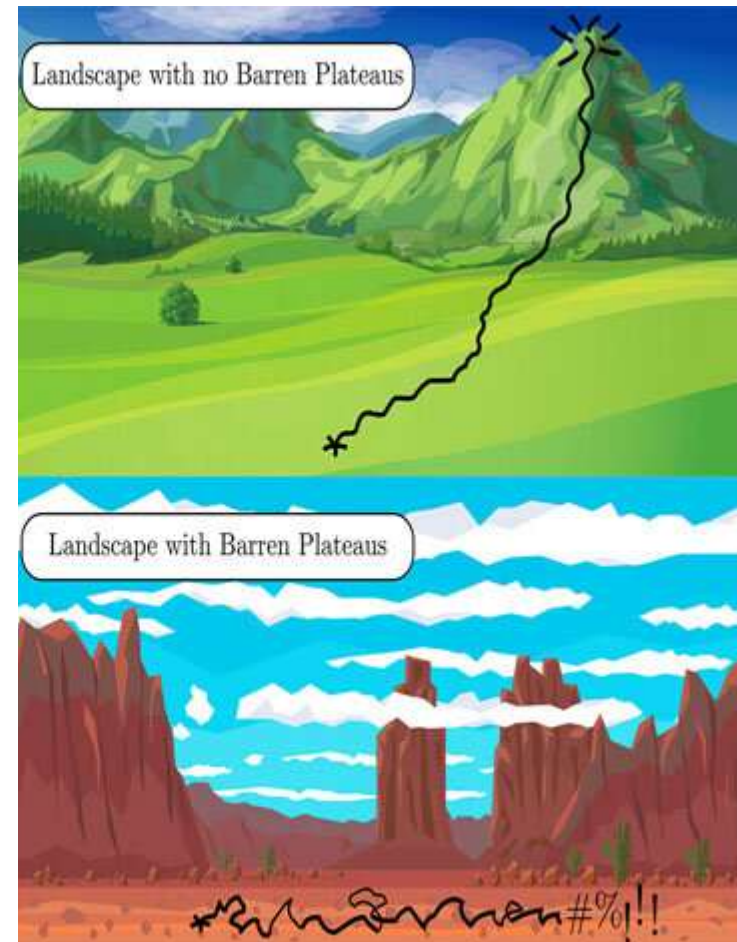
Local minima



Credits: Anschuetz et al.,
Quantum variational algorithms are swamped with traps.
Nat Commun 13, 7760 (2022).
<https://doi.org/10.1038/s41467-022-35364-5>

Arrasmith et al., *Effect of barren plateaus on gradient-free optimization*, Quantum 2021, 5, 558

Barren Plateaus



Credits: Los Alamos National Laboratory

Gibbs state on a QC

motivations

Definition of Gibbs states

$$\rho_\beta = \frac{e^{-\beta H}}{Z_\beta}, \quad \beta = \frac{1}{k_B T}$$

$$Z_\beta = \text{Tr} \{ e^{-\beta H} \} = \sum_{i=1}^d e^{-\beta E_i}$$

Usefulness of Gibbs states

quantum simulation
quantum machine learning
quantum optimization
open quantum systems

Sampling from Gibbs states used in:
combinatorial optimization problems
semidefinite programming
training quantum Boltzmann machines

Properties of Gibbs states

- Minimizes the Helmholtz free energy

$$F_\beta(\rho) = \text{Tr} [H\rho] - \beta^{-1} S(\rho)$$

where the von Neumann entropy is

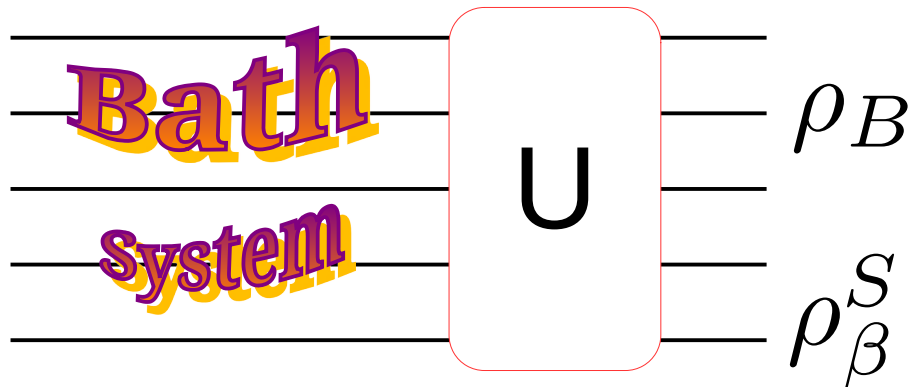
$$S(\rho) = -\text{Tr} [\rho \ln \rho] = -\sum_{i=1}^d p_i \ln p_i$$

- It is an equilibrium state
- It is unique (for quantum spin models on finite lattices)
- It is a completely passive state

Matsui, *Purification and Uniqueness of Quantum Gibbs States*, Commun. Math. Phys. 162, 321-332 (1994)

Gibbs state on a QC a set of approaches

Mimick the process of thermalisation



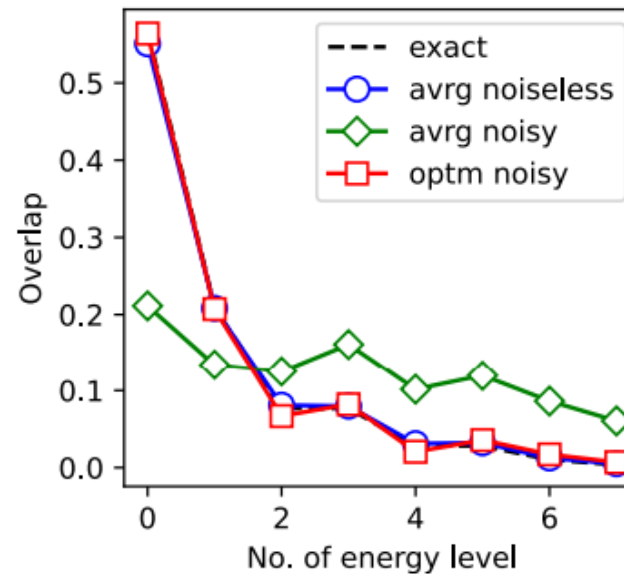
Drawbacks: thermalisation time, reservoir qubits,...

Riera, A.; Gogolin, C.; Eisert, J.
Thermalization in Nature and on a Quantum Computer.
Phys. Rev. Lett. 2012, 108, 080402–080406

Implement effective imaginary-time evolution

$$\mathbb{E}_{\psi} e^{-\beta H/2} |\psi\rangle \langle \psi| e^{-\beta H/2} \propto \exp(-\beta H)$$

$|\psi\rangle$ is a random initial state



Heisenberg
Hamiltonian
with 3 system
qubits and 2
ancilla qubits

Drawbacks: ITE requires non-unitary dynamics

Shtanko et al., *Algorithms for Gibbs state preparation on noiseless and noisy random quantum circuits*, arXiv:2112.14688

Motta et al., *Determining eigenstates and thermal states on a quantum computer using quantum imaginary time evolution*. Nat. Phys. 16, 205–210 (2020). <https://doi.org/10.1038/s41567-019-0704-4>

1. Ansatz selection

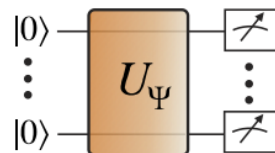
$$|\psi_i\rangle \in \mathcal{S}$$

initial pure state/
hybrid density matrix

$$|\phi(\alpha(t))\rangle = \sum_i \alpha_i(t) |\psi_i\rangle$$

$$\rho(\beta(t)) = \sum_{i,j} \beta_{i,j}(t) |\psi_i\rangle \langle \psi_j|$$

2. Quantum computer



measure overlaps

$$\mathcal{E}_{i,j} = \langle \psi_i | \psi_j \rangle$$

$$\mathcal{D}_{i,j} = \langle \psi_i | H | \psi_j \rangle$$

⋮

3. Classical evolution



evolve differential
equation involving
overlaps such as

$$\mathcal{E} \frac{\partial \alpha(t)}{\partial t} = -t \mathcal{D} \alpha(t)$$

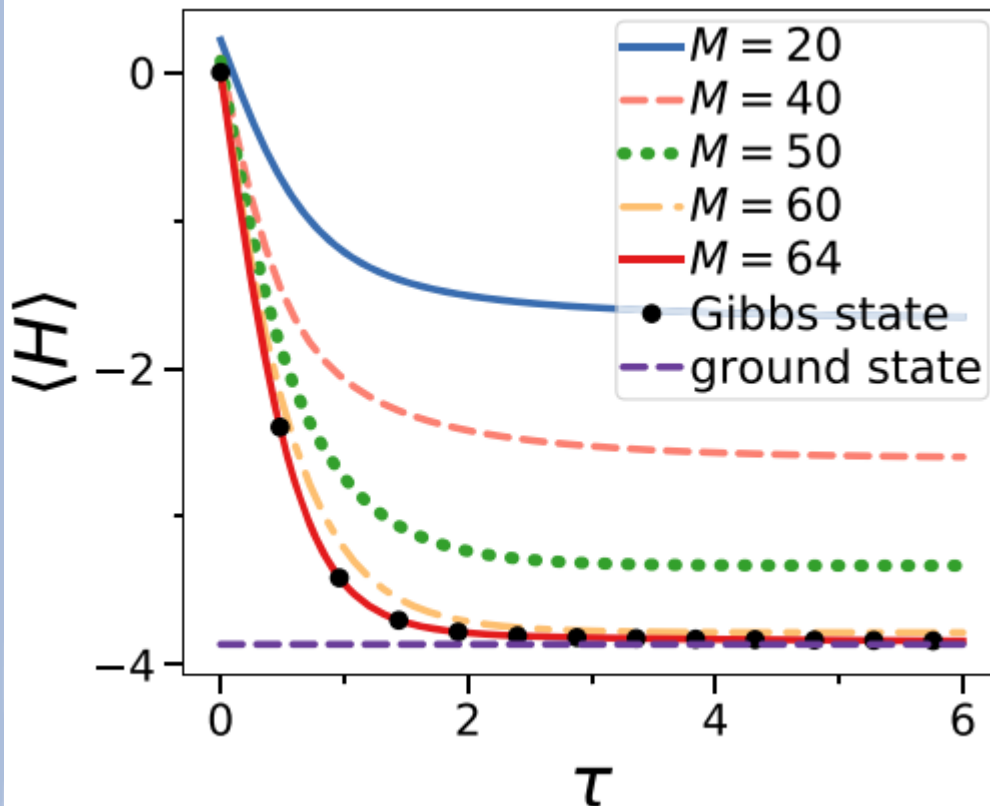
Haug and Bharti, *Generalized quantum assisted simulator* 2022
Quantum Sci. Technol. 7 045019

Gibbs state on a QC

a set of approaches

Mimic the process of thermalisation

Bath

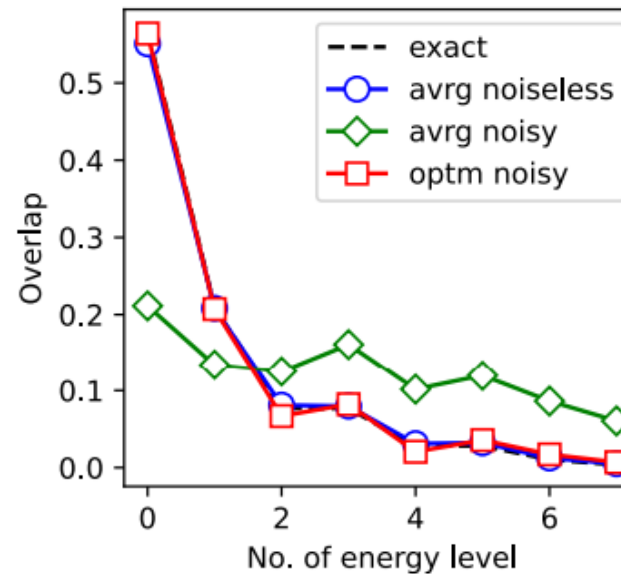



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Gibbs state on a QC

the variational approach

The free energy is a faithful cost function

$$F_\beta(\rho) = \text{Tr} \{ H \rho \} - \beta^{-1} S(\rho)$$

$\text{Tr}\{H\rho\}$ can be “easily” measured

$S(\rho)$ however, is not an observable!

Approaches to estimate the von Neumann entropy

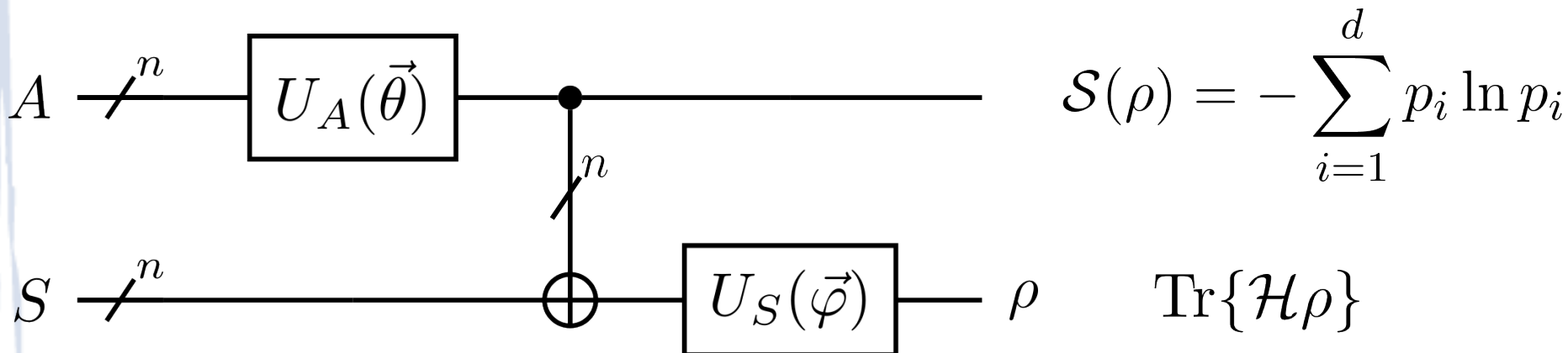
$$\tilde{S}(\rho) = \sum_{m=1}^M \left(b_m^{(1)} \text{Tr}(\rho \cos(pt_m)) + b_m^{(2)} \text{Tr}(\rho \sin(pt_m)) \right)$$

Chowdhury, A. N., G. H. Low, and N. Wiebe (2020), arXiv:2002.00055.

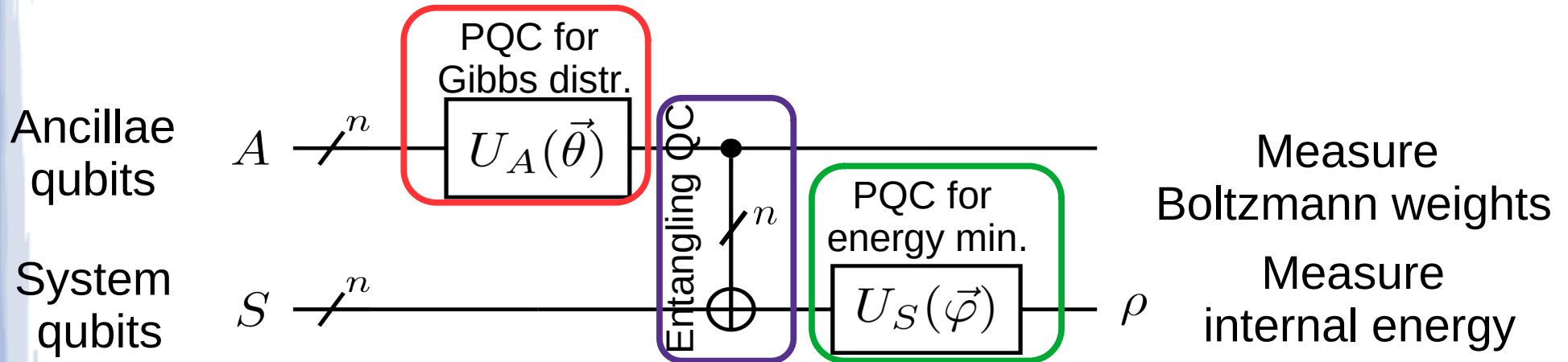
$$S_K(\rho) = \sum_{k=1}^K \frac{(-1)^k}{k} \text{tr}[(\rho - I)^k \rho] = \sum_{j=0}^K C_j \text{tr}(\rho^{j+1}).$$

Y. Wang, G. Li, and X. Wang, Phys. Rev. Applied 16,054035 (2021).

Our approach: generate a **diagonal** probability distribution function



Cost function $F_\beta(\rho) = \text{Tr} \{ H \rho \} - \beta^{-1} S(\rho)$



The unitary on the ancillae generates $(U_A \otimes I_S) |0\rangle_{AS}^{\otimes 2n} = |\psi\rangle_A \otimes |0\rangle_S^{\otimes n}$

A set of CNOT gates is applied between ancillae and system qubits

$$\text{CNOT}_{AS} \equiv \bigotimes_{i=1}^n \text{CNOT}_{A_i S_i}$$

The reduced density matrices of A and S are identical

$$\text{Tr}_S \left\{ \text{CNOT}_{AS} (|\psi\rangle\langle\psi|_A \otimes |0\rangle\langle 0|_S^{\otimes n}) \text{CNOT}_{AS}^\dagger \right\} = \text{diag} (p_1, p_2, \dots, p_d)$$

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The unitary on the system rotates from the computational basis to a different basis

$$\rho = U_S \text{diag}(p_1, p_2, \dots, p_d) U_S^\dagger = \sum_i p_i |\psi_i\rangle\langle\psi_i|$$

The result of the minimisation of the cost function

$$\rho_\beta = \arg \min_{\theta, \varphi} F_\beta(\rho(\theta, \varphi)) = \arg \min_{\theta, \varphi} (\text{Tr}\{H \rho_S(\theta, \varphi)\} - \beta^{-1} S(\rho_A(\theta)))$$

↳ $p_i = \frac{e^{-\beta E_i}}{Z}$ & $|\psi_i\rangle = |E_i\rangle$ **The Gibbs State**

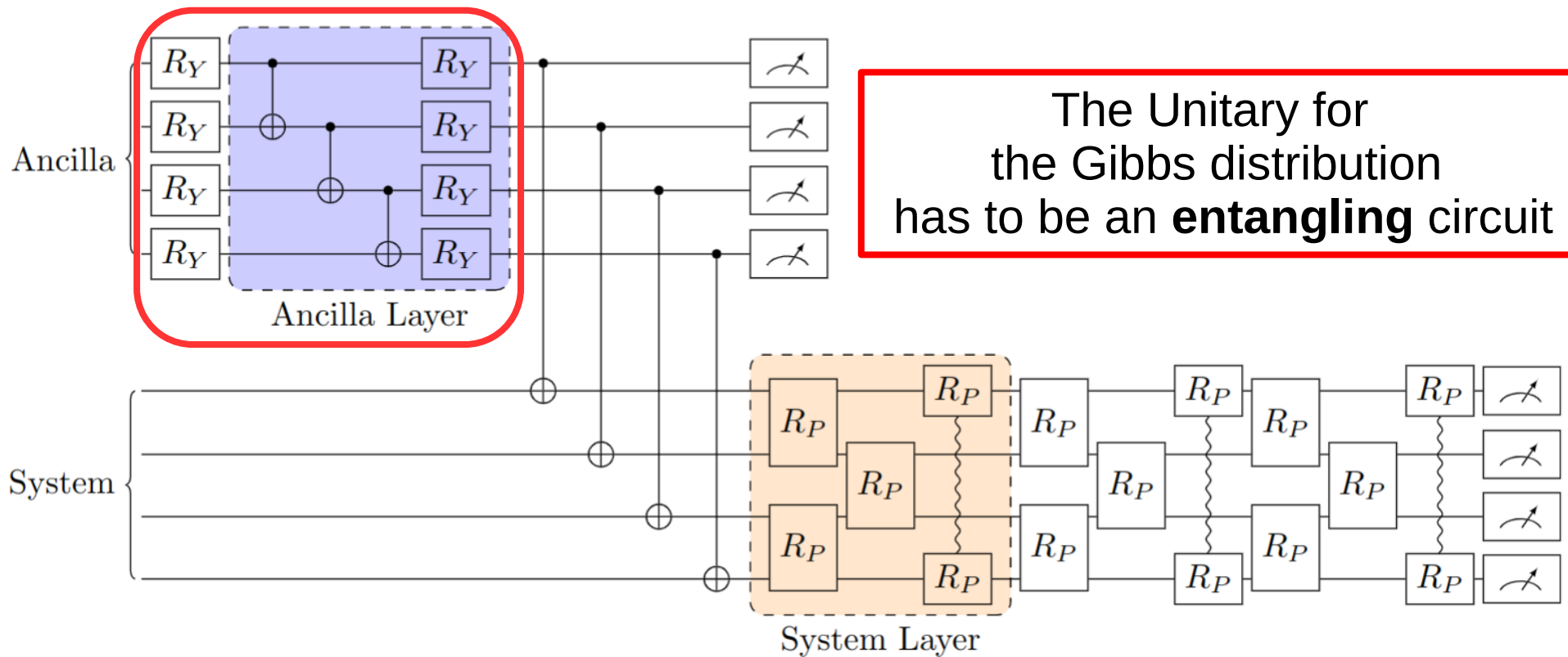
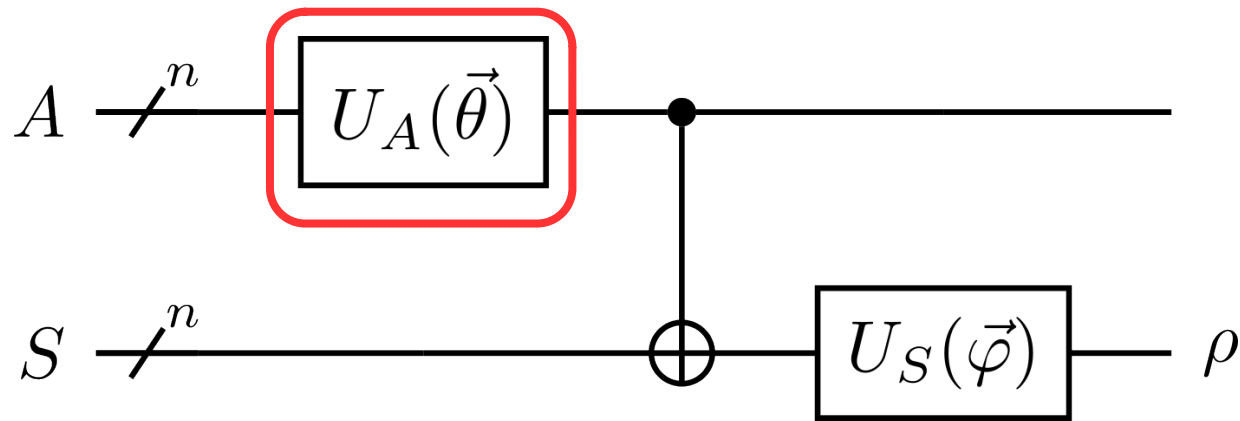
Generation of thermofield double states:
condensed matter physics, black holes and traversable wormholes



$$|\text{TFD}(\beta)\rangle = \frac{1}{\sqrt{Z}} \sum_i e^{-\frac{\beta E_i}{2}} |i\rangle_A |i\rangle_S$$

Wu and Hsieh, *Variational Thermal Quantum Simulation via Thermofield Double States*, Phys. Rev. Lett. 123, 220502 (2019)
Sagastizabal et al, *Variational preparation of finite-temperature states on a quantum computer*, npj Quantum Information 7, 1 (2021),

Generating the Gibbs distribution



With only local gates (non-entangling gates) only specific PDFs can be searched

$$\bigotimes_{i=0}^{n-1} R_Y(\theta_i) |0\rangle_i = \bigotimes_{i=0}^{n-1} (b_i |0\rangle_i + a_i |1\rangle_i) = \sum_{i=0}^{d-1} \prod_{j \in S_{i=0}} a_j \prod_{k \in S_{i=1}} b_k |i\rangle = \sum_{i=0}^{d-1} p_i |i\rangle$$

As each p_i is the Boltzmann weight of the Gibbs distribution

$$p_i = \prod_{j \in S_{i=0}} \cos\left(\frac{\theta_j}{2}\right) \prod_{k \in S_{i=1}} \sin\left(\frac{\theta_k}{2}\right) = \frac{e^{-\beta E_i}}{Z}$$

Only PDFs satisfying very restrictive criteria can be searched
(the non-entangling Ansatz has little expressibility)

$$\frac{p_0}{p_1} = \frac{a_0}{b_0}$$

$$\frac{p_1}{p_2} = \frac{a_1}{b_1}$$

$$\frac{p_0}{p_2} = \frac{a_0 a_1}{b_0 b_1}$$

$$\frac{p_0}{p_3} = \frac{a_0 a_1}{b_0 b_1}$$

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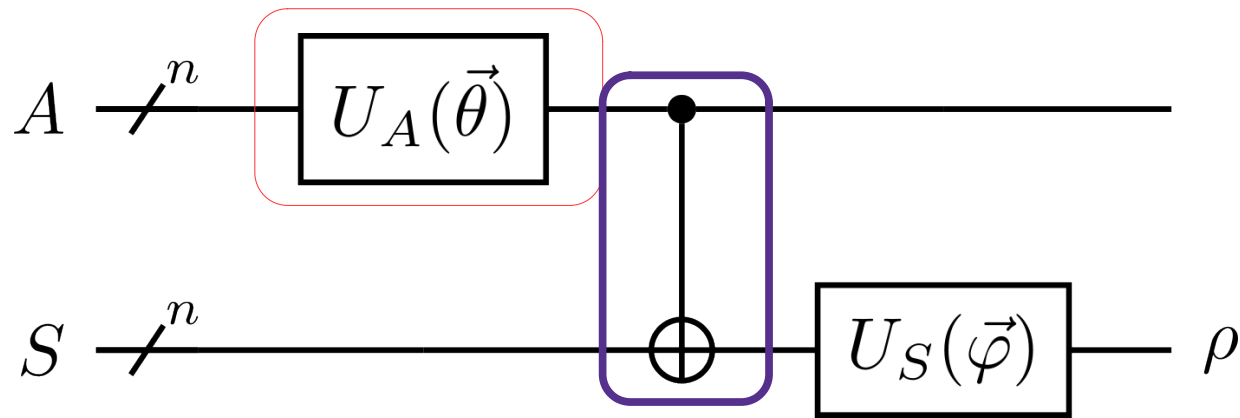
$$\frac{p_0}{p_3} = \frac{a_0 a_1}{b_0 b_1}$$

$$p_0 p_3 = p_1 p_2$$

$$e^{-\beta E_0} e^{-\beta E_3} = e^{-\beta E_1} e^{-\beta E_2}$$

$$E_0 + E_3 = E_1 + E_2$$

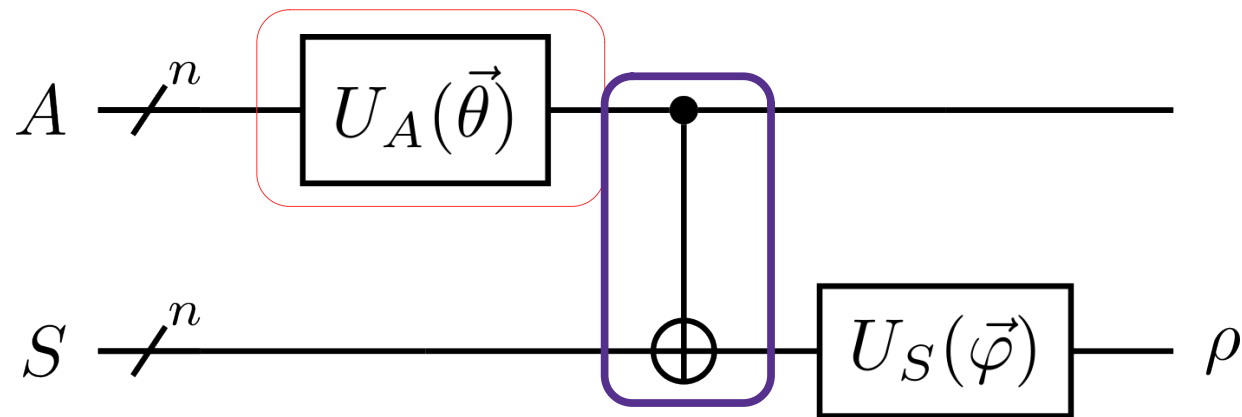
Mapping the PDF to the system



$$U_A = \begin{pmatrix} u_{11} & u_{12} & \cdots & u_{1d} \\ u_{21} & u_{22} & \cdots & u_{2d} \\ \vdots & \vdots & \ddots & \vdots \\ u_{d1} & u_{d2} & \cdots & u_{dd} \end{pmatrix}$$

$$|\psi_{AS}\rangle = \sum_{i=1}^d u_{i1} |\mathbf{i}\rangle_A |\mathbf{0}\rangle_S$$

Mapping the PDF to the system

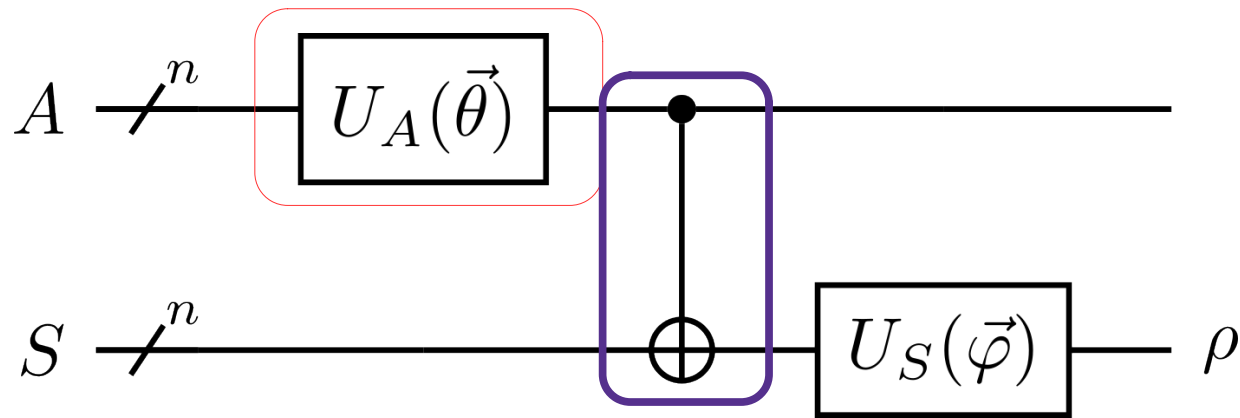


$$U_A = \begin{pmatrix} u_{11} & u_{12} & \cdots & u_{1d} \\ u_{21} & u_{22} & \cdots & u_{2d} \\ \vdots & \vdots & \ddots & \vdots \\ u_{d1} & u_{d2} & \cdots & u_{dd} \end{pmatrix}$$

$$|\psi_{AS}\rangle = \sum_{i=1}^d u_{i1} |\mathbf{i}\rangle_A |\mathbf{0}\rangle_S$$

$$|\psi_{AS}\rangle = \sum_{i=1}^d u_{i1} |\mathbf{i}\rangle_A |\mathbf{i}\rangle_S \implies \rho_S = \text{Tr}_A[\rho_{AS}] = \sum_i |u_{i1}|^2 |\mathbf{i}\rangle\langle\mathbf{i}|$$

Mapping the PDF to the system



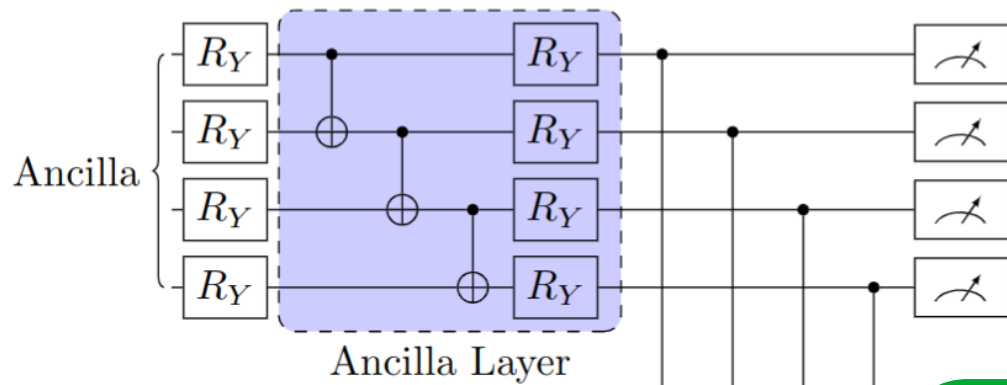
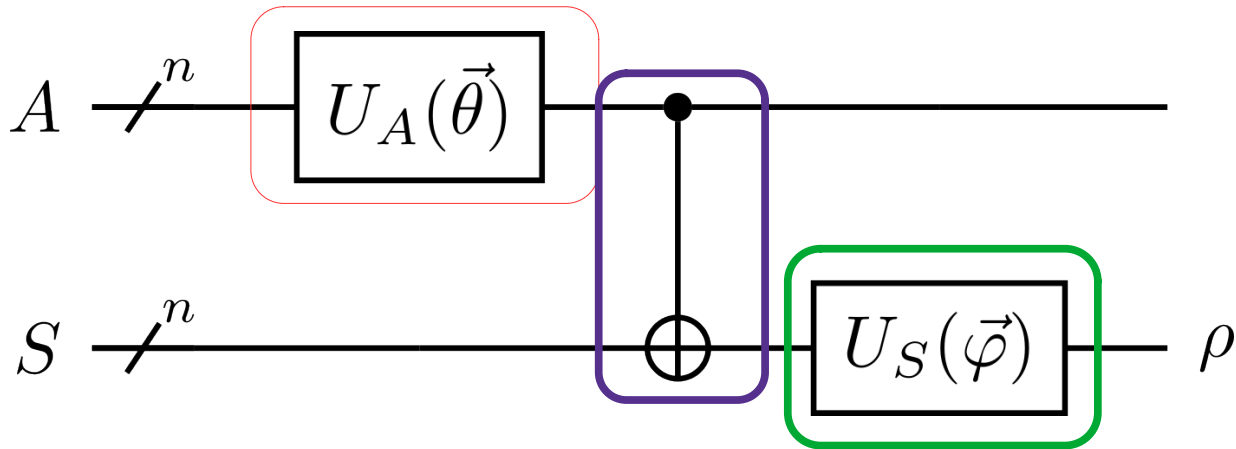
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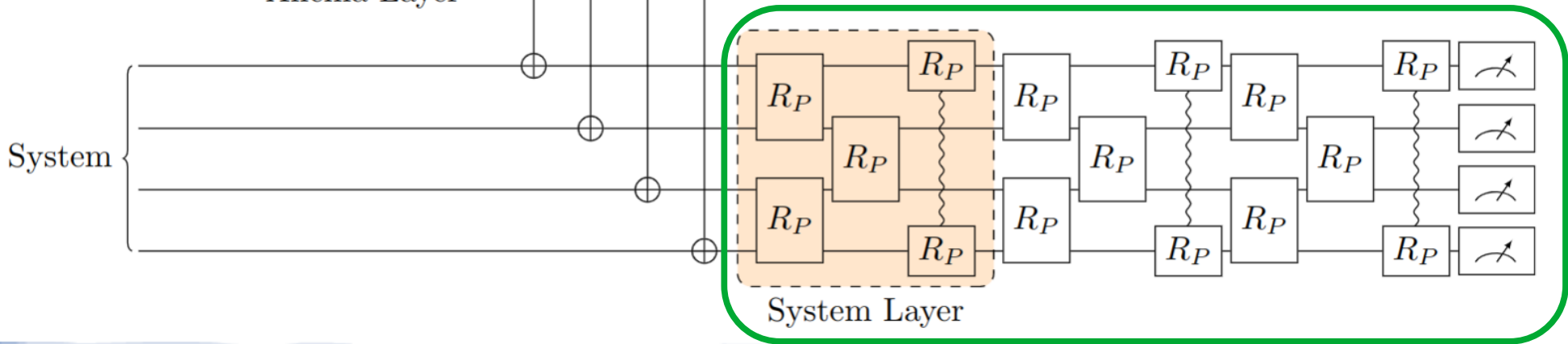
Only the 1st column
of U_A matters

$$|\psi_{AS}\rangle = \sum_{i=1}^d u_{i1} |\mathbf{i}\rangle_A |\mathbf{i}\rangle_S \implies \rho_S = \text{Tr}_A[\rho_{AS}] = \sum_i |u_{i1}|^2 |\mathbf{i}\rangle\langle\mathbf{i}|$$

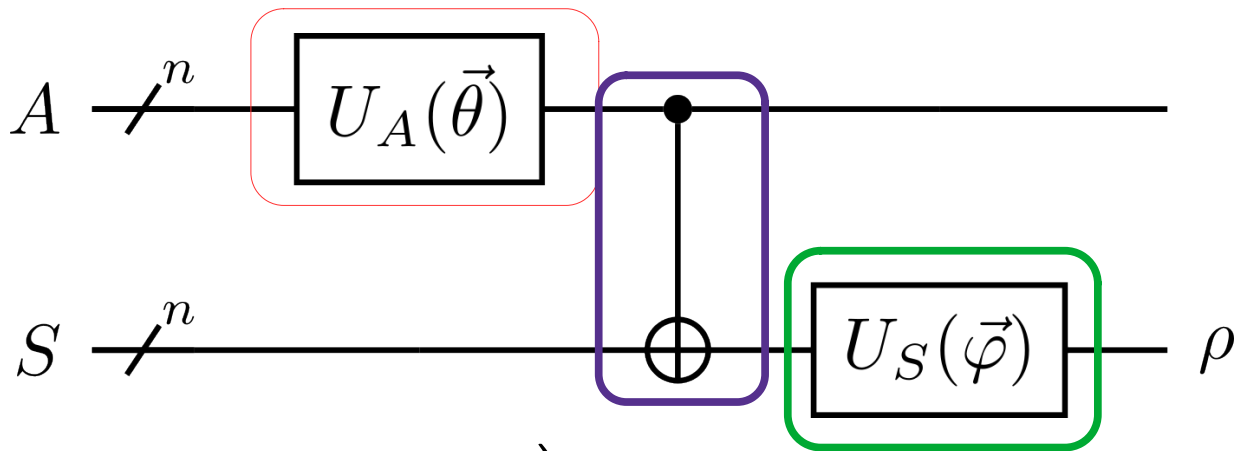
Minimising the energy



The Unitary for the minimisation of the energy is (in general) model-dependent



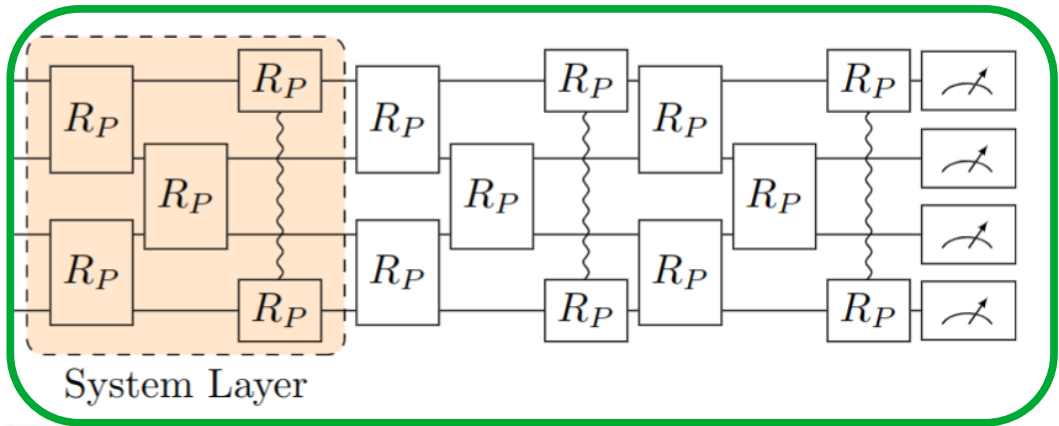
Minimising the energy



$$U_S = \begin{pmatrix} u_{11} & u_{12} & \cdots & u_{1d} \\ u_{21} & u_{22} & \cdots & u_{2d} \\ \vdots & \vdots & \ddots & \vdots \\ u_{d1} & u_{d2} & \cdots & u_{dd} \end{pmatrix}$$

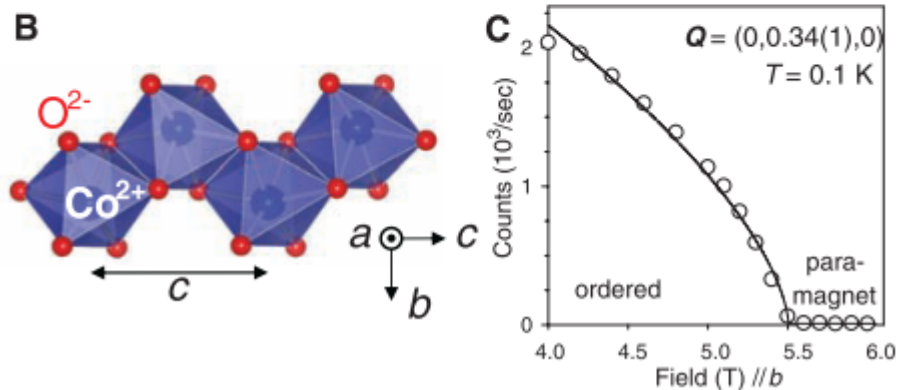
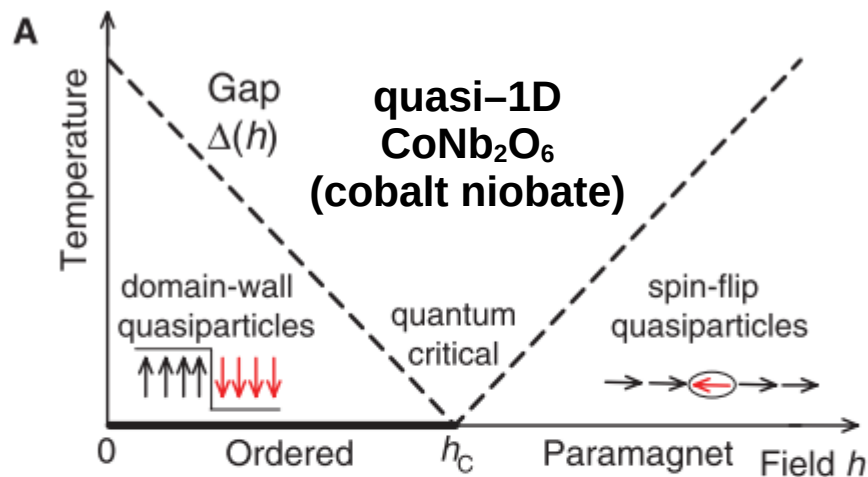
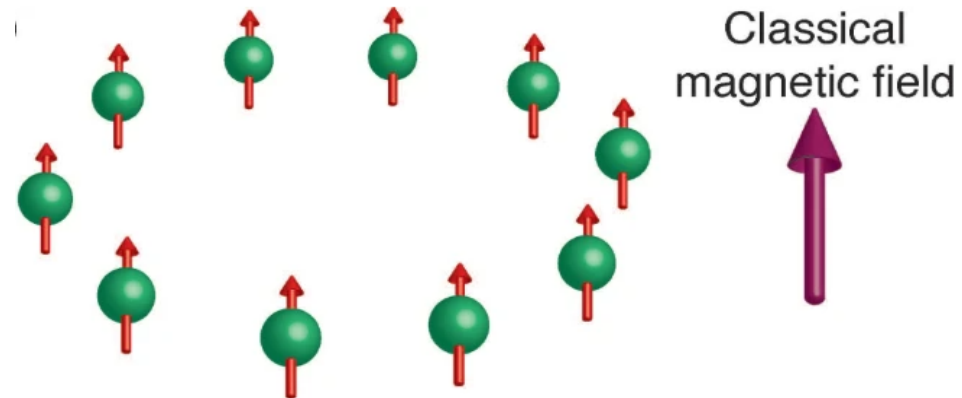
The Unitary for the minimisation of the energy is (in general) model-dependent

U_S rotates from the computational basis to the energy basis

$$U_S |i\rangle = |E_i\rangle$$


The quantum Ising Hamiltonian

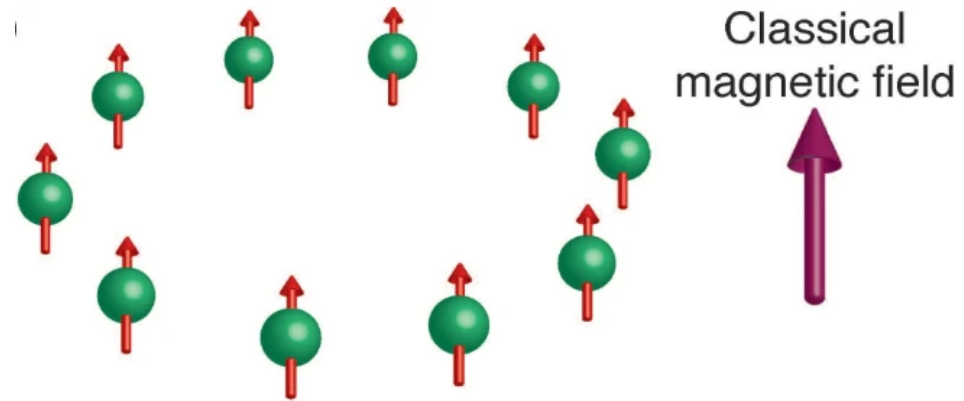
$$\hat{H} = -J \sum_n \hat{\sigma}_n^x \hat{\sigma}_{n+1}^x - h \hat{\sigma}_n^z$$



R. Coldea et al., *Quantum Criticality in an Ising Chain: Experimental Evidence for Emergent E_8 Symmetry*, Science 327, 177 (2010);

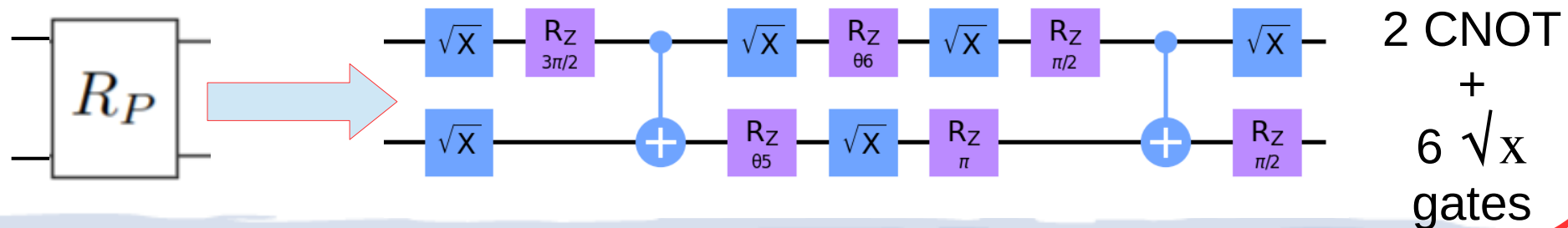
The quantum Ising Hamiltonian

$$\hat{H} = -J \sum_n \hat{\sigma}_n^x \hat{\sigma}_{n+1}^x - h \hat{\sigma}_n^z$$



Z_2 symmetry \rightarrow Parity-Preserving Ansatz

$$R_P(\varphi_i, \varphi_j) = R_{YX}(\varphi_j) \cdot R_{XY}(\varphi_i) = \begin{pmatrix} \cos\left(\frac{\varphi_i + \varphi_j}{2}\right) & 0 & 0 & \sin\left(\frac{\varphi_i + \varphi_j}{2}\right) \\ 0 & \cos\left(\frac{\varphi_i - \varphi_j}{2}\right) & -\sin\left(\frac{\varphi_i - \varphi_j}{2}\right) & 0 \\ 0 & \sin\left(\frac{\varphi_i - \varphi_j}{2}\right) & \cos\left(\frac{\varphi_i - \varphi_j}{2}\right) & 0 \\ -\sin\left(\frac{\varphi_i + \varphi_j}{2}\right) & 0 & 0 & \cos\left(\frac{\varphi_i + \varphi_j}{2}\right) \end{pmatrix}$$



Results (state vector)

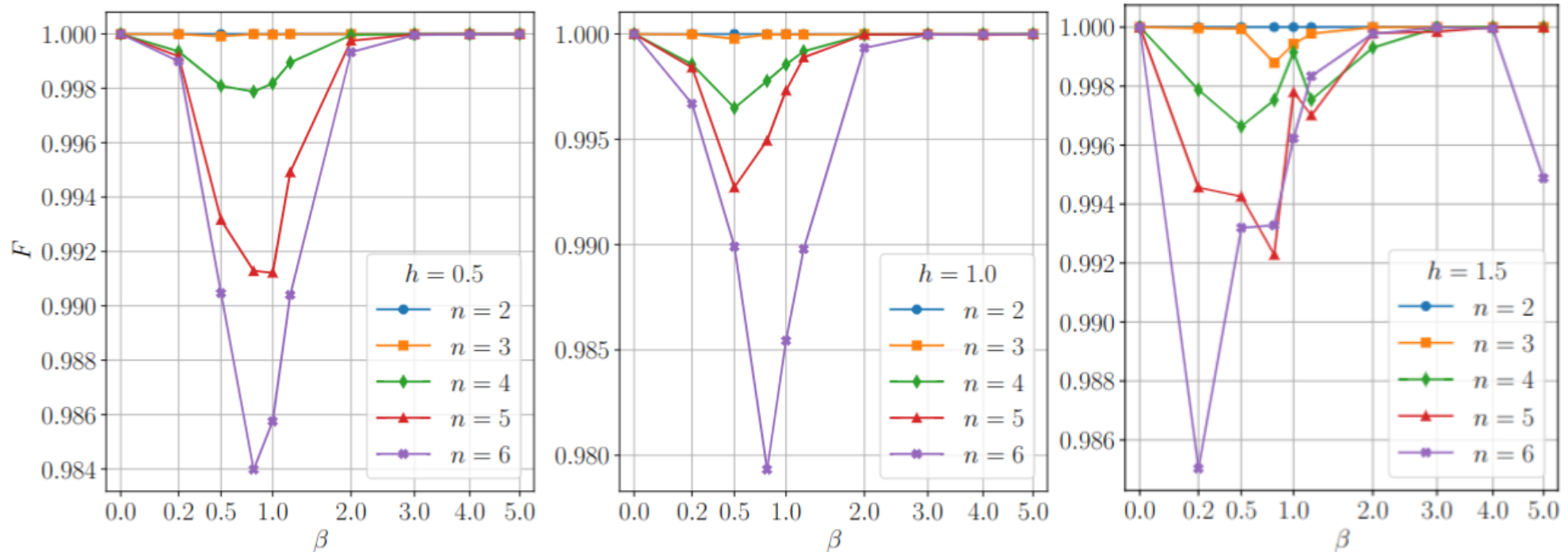
Measures for determining the quality of the prepared Gibbs state

$$D(\rho_1, \rho_2) = \frac{1}{2} \|\rho_1 - \rho_2\|_1$$

$$F(\rho_1, \rho_2) = \sqrt{\sqrt{\rho_2} \rho_1 \sqrt{\rho_2}}$$

$$\left(\|A\|_1 := \text{Tr} \left\{ \sqrt{A^\dagger A} \right\} \right)$$

$$1 - F(\rho_1, \rho_2) \leq D(\rho_1, \rho_2) \leq \sqrt{1 - F^2(\rho_1, \rho_2)}$$



Results (noisy-simulations & QC)

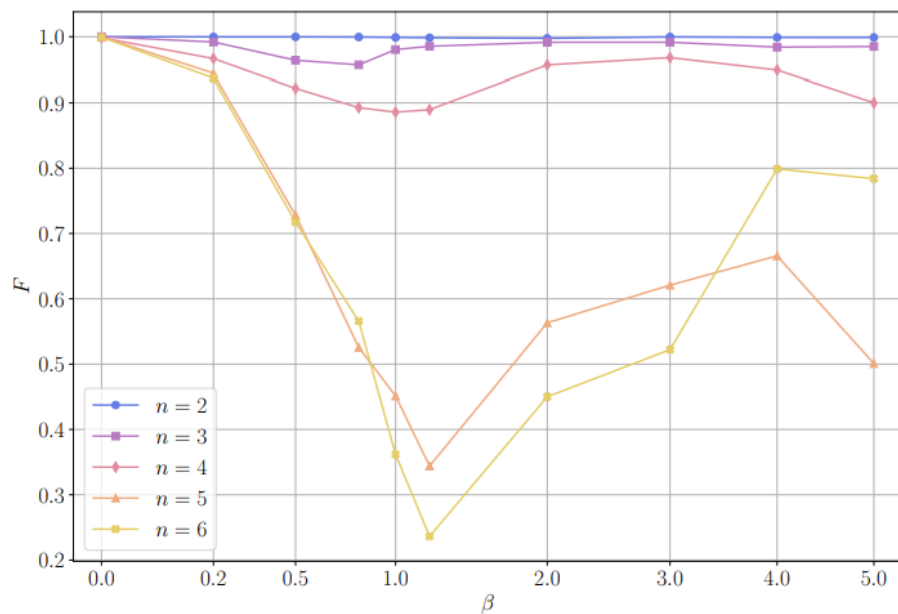


Figure 5: Fidelity F , of the obtained state via noisy simulations (using SPSA) of `ibmq_guadalupe` with the exact Gibbs state, vs inverse temperature β , for two to six qubits of the Ising model with $h = 0.5$. A total of ten runs are made for each point, with the optimal state taken to be the one that maximizes the fidelity.

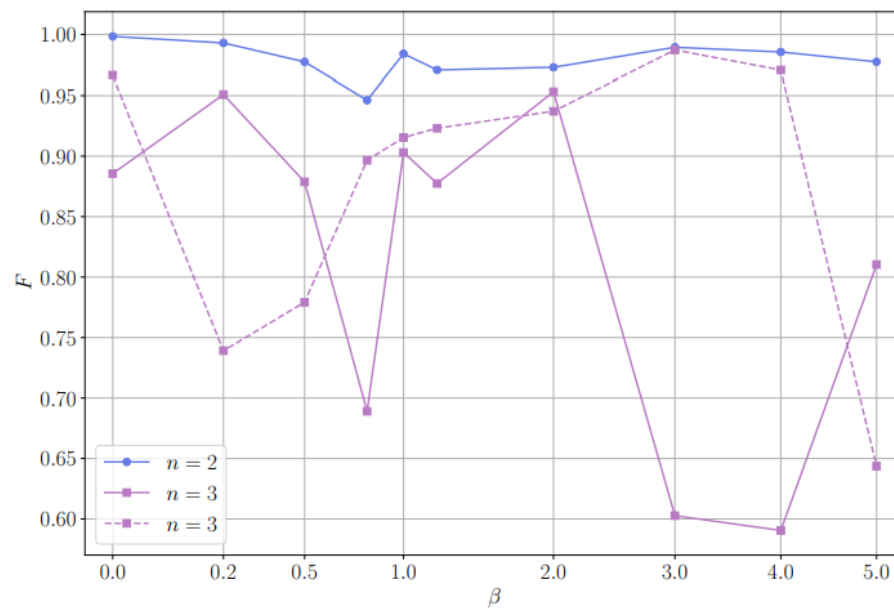


Figure 6: Fidelity F , of the obtained state (using SPSA) running directly on `ibmq_nairobi` with the exact Gibbs state, vs inverse temperature β , for two and three qubits of the Ising model with $h = 0.5$. The dashed line represents the run with no R_P gate between non-adjacent qubits in the system layers. One run is carried out for $n = 2$, and $n = 3$ for the dashed line, and two runs for $n = 3$ for the solid line.

Results (noisy-simulations & QC)

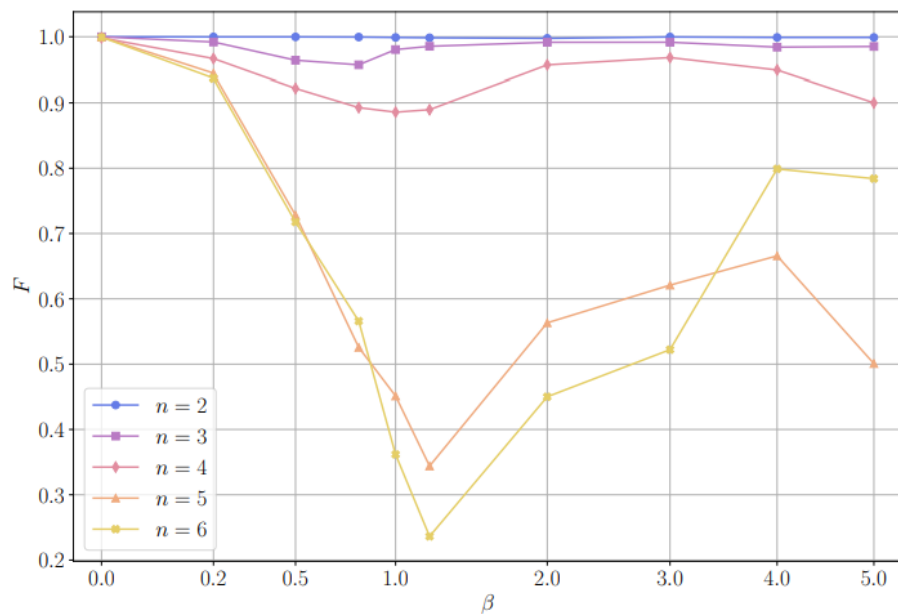


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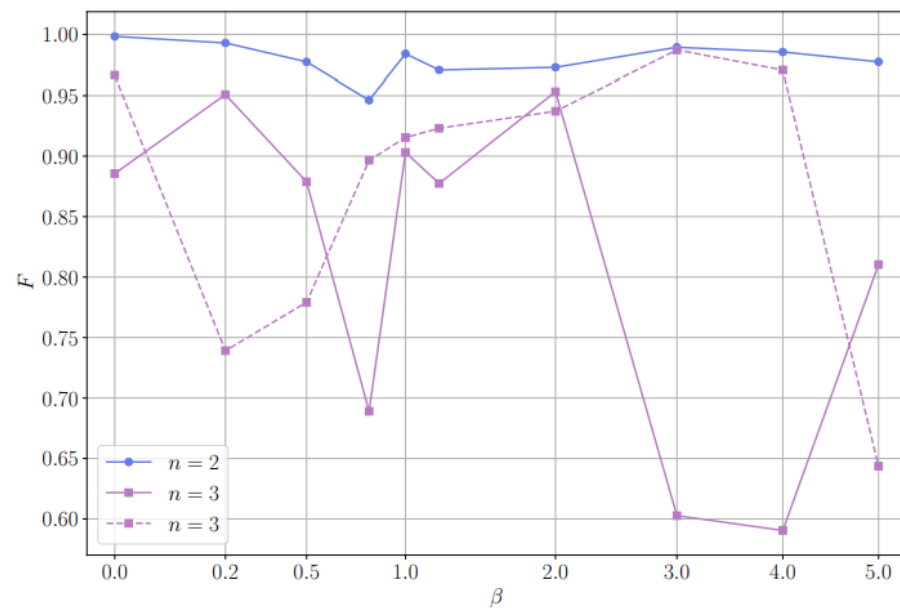


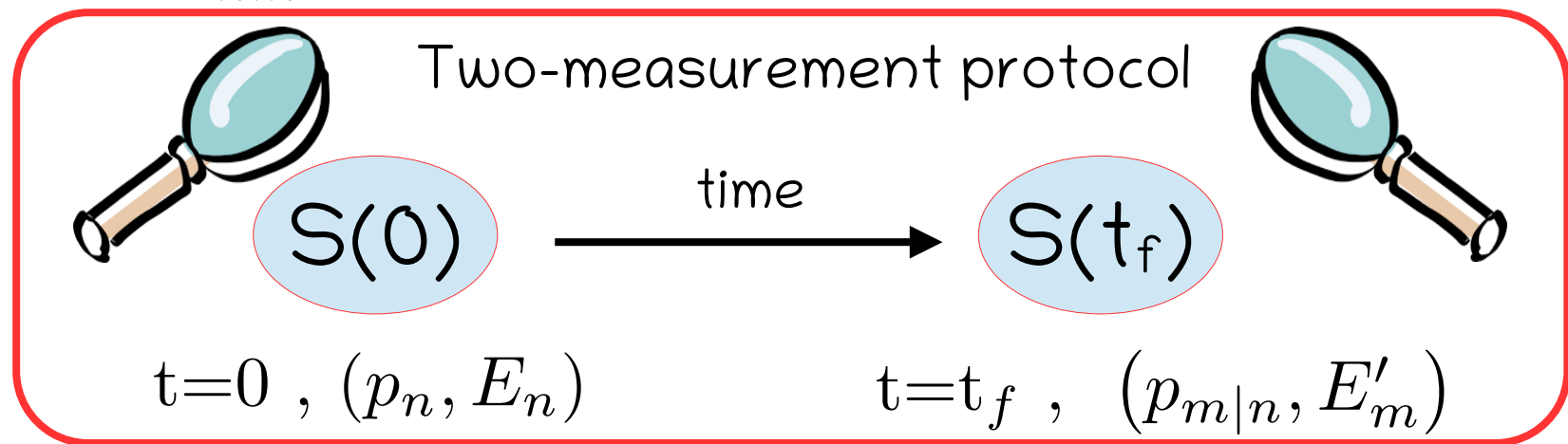
Figure 6: Fidelity F , of the obtained state (using SPSA) running directly on `ibmq_nairobi` with the exact Gibbs state, vs inverse temperature β , for two and three qubits of the Ising model with $h = 0.5$. The dashed line represents the run with no R_P gate between non-adjacent qubits in the system layers. One run is carried out for $n = 2$, and $n = 3$ for the dashed line, and two runs for $n = 3$ for the solid line.

Take home messages:

- The VQA works very well...on a classical computer;
- “Use the QC as little as possible” \rightarrow the algorithm works (well) within the given available quantum volume;
- Intermediate temperatures ($\beta \approx 1$) are more difficult to realise.

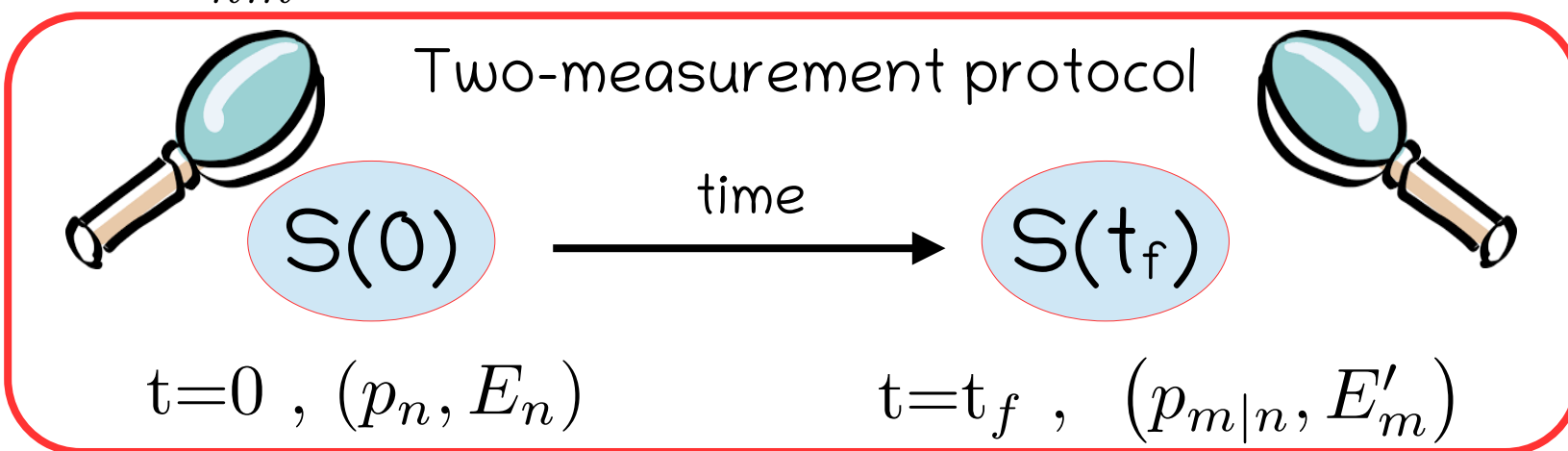
Applications in quantum thermodynamics: PDF of Work

$$p(w) = \sum_{nm} p_{m|n}(t_f) p_n(0) \delta(w - (E'_m(t_f) - E_n(0)))$$



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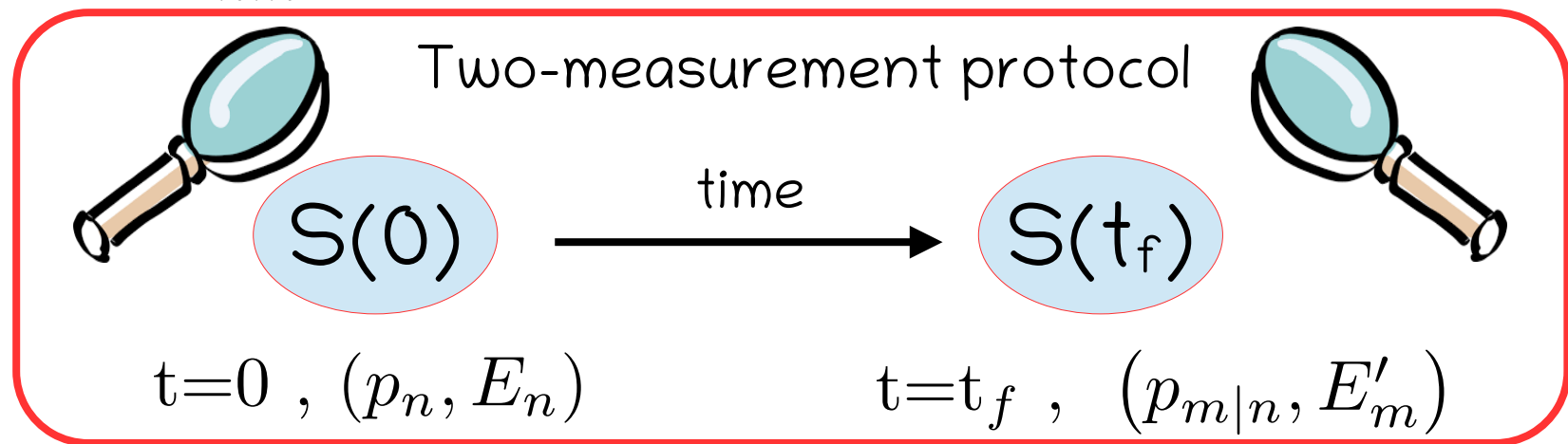
$$p_{m|n}(t_f) = |\langle E'_m | U(t_f) | E_n \rangle|^2 \xrightarrow{\text{sudden quench}} |\langle E'_m | E_n \rangle|^2$$

where U diagonalises H
 and
 V diagonalises H'

$$p_{m|n} = |\langle m' | V^\dagger U | n \rangle|^2$$

Applications in quantum thermodynamics: PDF of Work

$$p(w) = \sum_{nm} p_{m|n}(t_f) p_n(0) \delta(w - (E'_m(t_f) - E_n(0)))$$



Method for preparing Gibbs states in

Holmes et al.,

Quantum algorithms from fluctuation theorems:

Thermal-state preparation,

Quantum 6, 825 (2022)

Conclusions

- Preparation of a thermal state (of an arbitrary system) on a quantum computer;
- Variational Quantum Algorithm to generate the thermal state (with an exact method to evaluate the von Neumann entropy);
- Efficient generation of the Gibbs state for the quantum Ising model at arbitrary temperatures (state-vector simulation);
- Efficient generation on a real NISQ device only for few qubits.

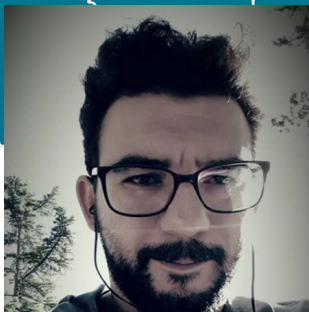
Outlooks

- Preparation of a thermal state of molecular systems (VQE works generally good);
- Application to the work PDF of an out-of-equilibrium protocol

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Francesco Plastina
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Jacopo Settino
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Andrea Giordano
ICAR-CNR

Variational Gibbs State Preparation on NISQ devices
<https://arxiv.org/abs/2303.11276>