

Solving differential equations on a superconducting quantum processor

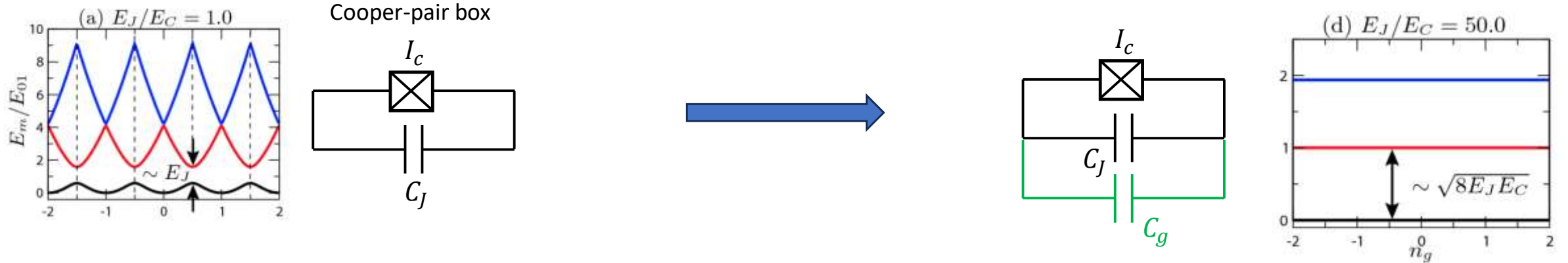
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Development of quantum computational methods

2008 : first unsensitive to background charge noise superconducting qubit is suggested ([J. Koch et. al., 2008](#))



2019-2023... - Noisy intermediate scale quantum (NISQ) devices epoch

- Quantum supremacy has been demonstrated in large size matrix multiplication ([Google team](#))
- First evidence of quantum simulation of spin dynamics has been demonstrated at a finite noise level of quantum gates ([IBM team](#))

Article

Quantum supremacy using a programmable superconducting processor

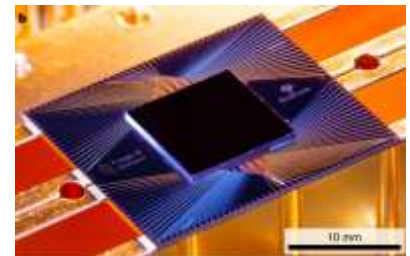
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Article

Evidence for the utility of quantum computing before fault tolerance

<https://doi.org/10.1038/s41586-023-06096-3>

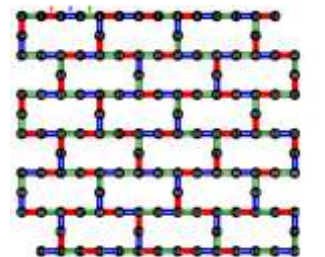
Received: 24 February 2023

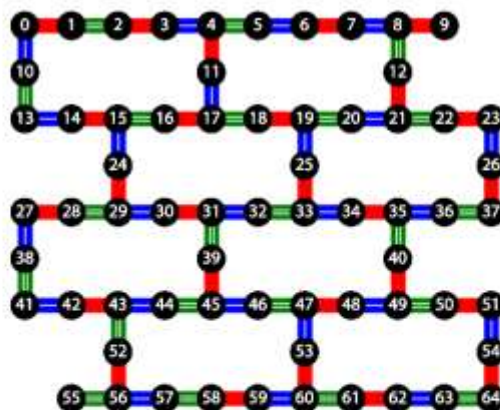
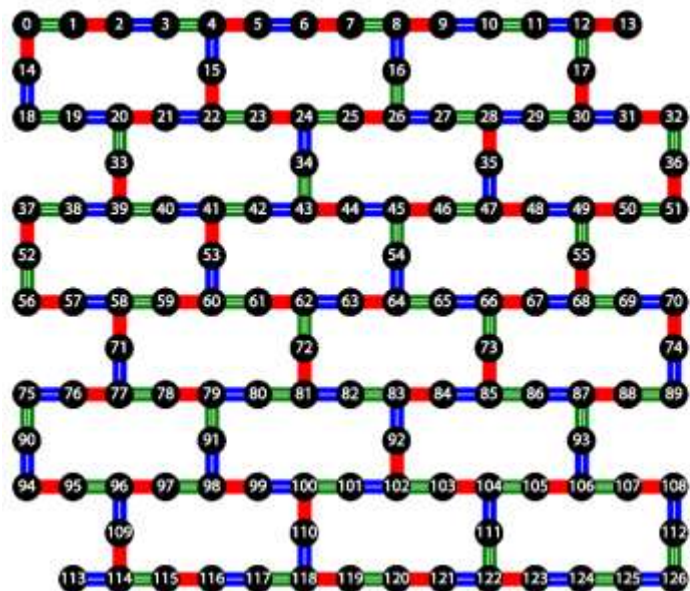
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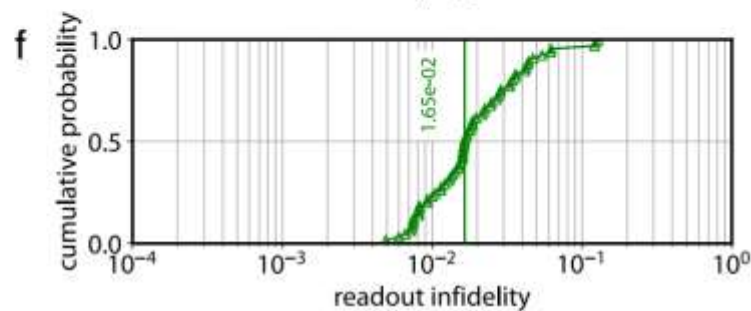
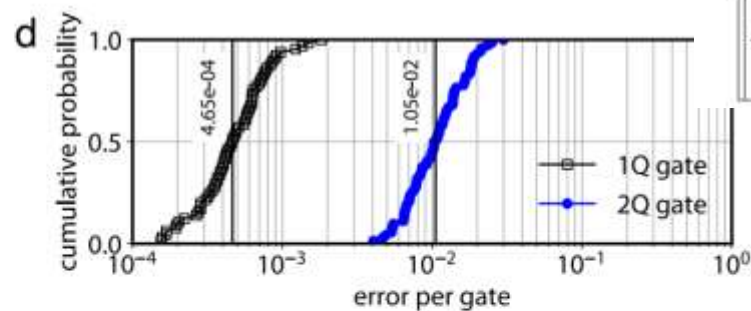
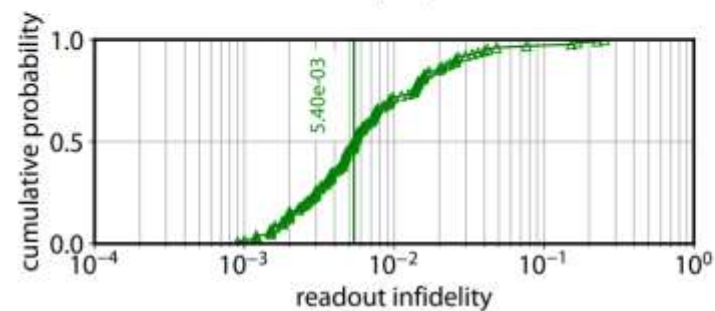
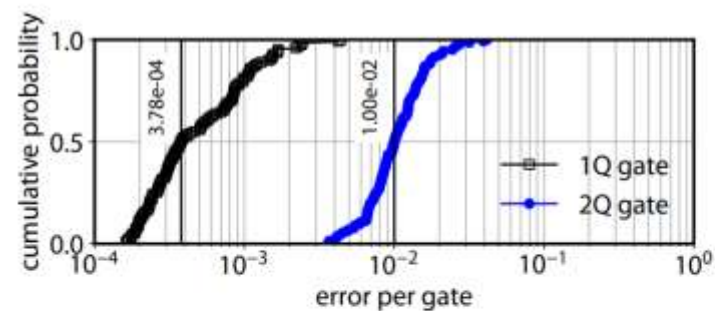
Youngseok Kim^{1,2,3,4}, Andrew Eddins^{2,3,4,5}, Soojan Anand³, Ken Xuan Wei¹, Ewout van den Berg¹, Sami Rosenblatt¹, Hasan Nayfeh¹, Yantao Wu^{1,4}, Michael Zaletel^{1,5}, Kristan Temme¹ & Abhinav Kandala^{1,6}

Quantum computing promises to offer substantial speed-ups over its classical





— layer 0
— layer 1
— layer 2



2008 isolated high-coherence qubits



5-qubit public quantum computer IBM (2017)

Coherence time T_2^* : 41.0, 43.5, 39.4 μs

2-qubit error rate: 2.79%, 2.46%, 1.68%



2023

	ibm_kyiv (127Q)				ibm_ithaca (65Q)			
	median	mean	min	max	median	mean	min	max
f_{01} (GHz)	4.61	4.62 ± 0.11	4.34	4.96	4.73	4.74 ± 0.10	4.54	4.93
$f_{01} - f_{12}$ (MHz)	311.07	310.77 ± 11.48	280.56	356.92	333.36	333.99 ± 5.61	316.04	355.56
T_1 (μs)	287.87	293.39 ± 84.78	85.30	567.55	183.54	180.52 ± 43.44	77.92	278.08
T_2 (μs)	127.49	156.73 ± 109.47	16.18	456.02	183.70	182.88 ± 91.05	19.27	407.15



What is next? 202? – 203?

Quantum error correction codes:

- Several **physical qubits** comprise a **logical qubit**
- The required **qubit number** is about **10000**
- The required qubit **error rate** **< 0.001%**

Evidence for the utility of quantum computing before fault tolerance

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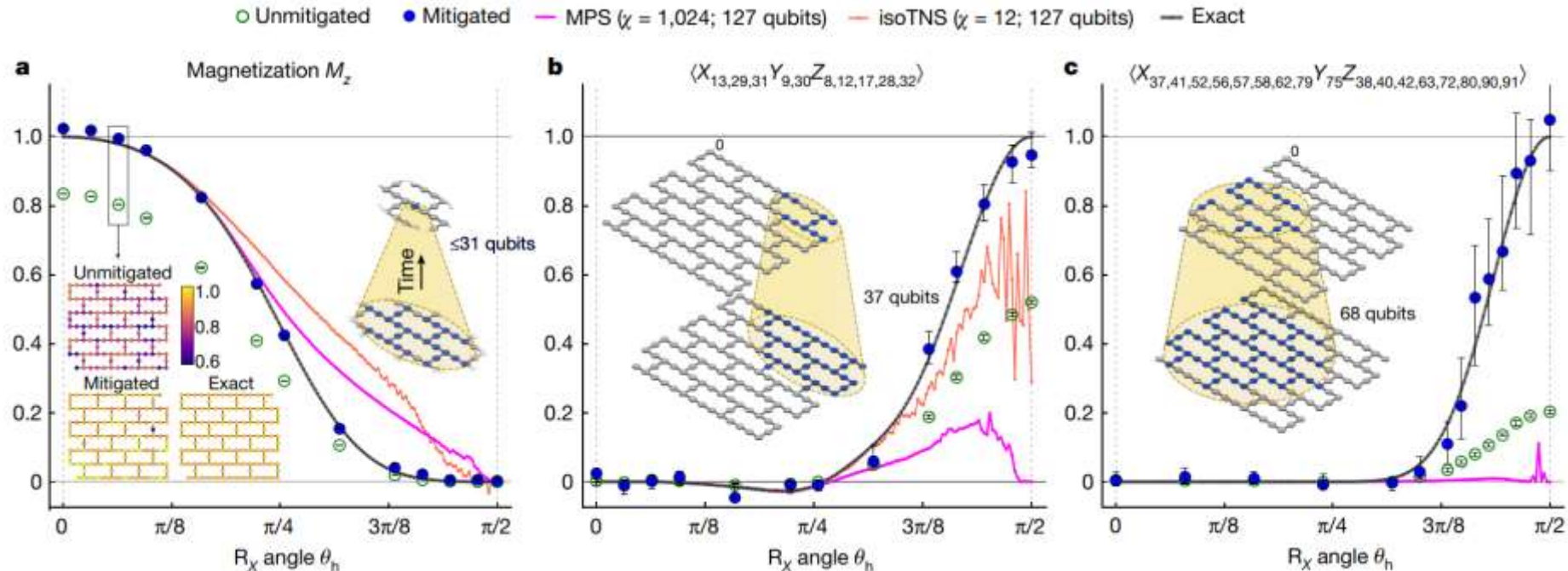
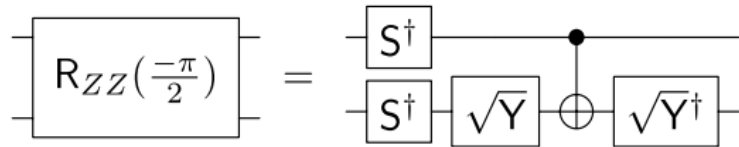
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Quantum computing promises to offer substantial speed-ups over its classical

What has been done: $\delta t = \frac{\pi}{4J}$



The dynamics of a spin lattice with nearest neighbor interaction has been simulated

$$H = -J \sum_{\langle i,j \rangle} Z_i Z_j + h \sum_i X_i,$$

The exact simulation requires Trotter decomposition of evolution operator

$$U(t) = [e^{-iH_{ZZ} \delta t} e^{-iH_X \delta t}]^N$$

$$e^{-iH_{ZZ} \delta t} = \prod_{\langle i,j \rangle} e^{i\delta t Z_i Z_j} = \prod_{\langle i,j \rangle} R_{Z_i Z_j}(-2J\delta t)$$

$$e^{-iH_X \delta t} = \prod_i e^{-i\delta t X_i} = \prod_i R_{X_i}(2h\delta t),$$

$$2J\delta t \ll 1$$

The roadmap for practical quantum computation

Formulate a set of industry motivated computational problems:

- Continuum dynamics
- Flow dynamics



Develop and analyse the quantum counterpart algorithm:

- Quantum scheme size (number of qubits, of quantum gates)
- Noise sensitivity (fidelity)
- Computational gain at a given noise level



Quantum hardware requirements :

- Required accuracy of quantum gates to reach quantum supremacy
- Number of qubits and scheme connectivity
- Required coherence time

These problems are hard to solve on a classical computer !

Quantum Poisson Solver (QPS): direct finite difference method

$$-\sum_{i=1}^n \frac{\partial^2 f}{\partial x_i^2} = q(\vec{x})$$

Solution is encoded into **amplitudes** of a quantum state of a **qubit register**:

$$|f\rangle = \sum_{x,y} f(x,y) |x\rangle |y\rangle$$

$$|x_k\rangle = |b_{d-1} b_{d-2} \dots b_0\rangle$$

bit string $b_{k-1} b_{k-2} \dots b_0$ represents index k of a discrete spatial lattice

Poisson equation is solved on a **discrete spatial lattice** of $N \times N \times \dots$ nodes.

Number of nodes quantify **complexity of the problem**

1D case

$$-\frac{f_{n+1} + f_{n-1} - 2f_n}{h^2} = q_n$$

$$x \in \{x_0, \dots, x_N\}$$

$$h = x_{n+1} - x_n$$



$$\begin{bmatrix} -2 & 1 & 0 & 0 & \dots & 0 & 0 & 0 \\ 1 & -2 & 1 & 0 & \dots & 0 & 0 & 0 \\ 0 & 1 & -2 & 1 & \dots & 0 & 0 & 0 \\ & \dots & \dots & & \dots & & \dots & \\ & \dots & \dots & & \dots & & \dots & \\ 0 & 0 & 0 & 0 & \dots & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 & \dots & -2 & 1 & 0 \\ 0 & 0 & 0 & 0 & \dots & 1 & -2 & 1 \end{bmatrix} \begin{bmatrix} f_1 \\ f_2 \\ f_3 \\ \vdots \\ \vdots \\ f_{N-3} \\ f_{N-2} \\ f_{N-1} \end{bmatrix} = \begin{bmatrix} h^2 q_1 + f_0 \\ h^2 q_2 \\ h^2 q_3 \\ \vdots \\ \vdots \\ h^2 q_{N-3} \\ h^2 q_{N-2} \\ h^2 q_{N-1} + f_N \end{bmatrix}$$

The QPS reduces to the solution of a large sparse linear system

$$N = 2^d$$

$$H \vec{f} = \vec{b}$$

Quantum Algorithm for Linear Systems of Equations

Aram W. Harrow, Avinatan Hassidim, and Seth Lloyd
Phys. Rev. Lett. **103**, 150502 – Published 7 October 2009

Physics See Synopsis: [The quantum shortcut to a solution](#)

$$A \vec{x} = \vec{b}$$

$$|x\rangle = A^{-1}|b\rangle$$

The main idea of the HHL algorithm to find inverse matrix through its spectral decomposition

$$A \xrightarrow{\text{phase estimation algorithm}} A = \sum_{\lambda} \lambda |a_{\lambda}\rangle\langle a_{\lambda}| \xrightarrow{\text{approximate reciprocal function calculation}} A^{-1} = \sum_{\lambda} \frac{1}{\lambda} |a_{\lambda}\rangle\langle a_{\lambda}|$$

Quantum linear solver provides an exponential speed up for sparse linear systems $N \Rightarrow \log N$

Hidden requirements of HHL algorithm: **efficient** unitary operation generated by matrix A

$$A \Rightarrow U_A(m) = \exp(-2\pi i m A), \quad m = 1, 2, \dots$$

What does it mean **efficient**?

$$U_A = \prod_{i=1}^{K_A} G_i \quad K_A \sim (\log N)^p$$

1 or 2-qubit quantum gate polylogarithmic complexity

$$\begin{bmatrix} -2 & 1 & 0 & \dots & 0 \\ 1 & -2 & 1 & \dots & 0 \\ 0 & 1 & -2 & \ddots & \vdots \\ & \vdots & & \ddots & \vdots \\ 0 & 0 & \dots & \dots & 0 \end{bmatrix}$$

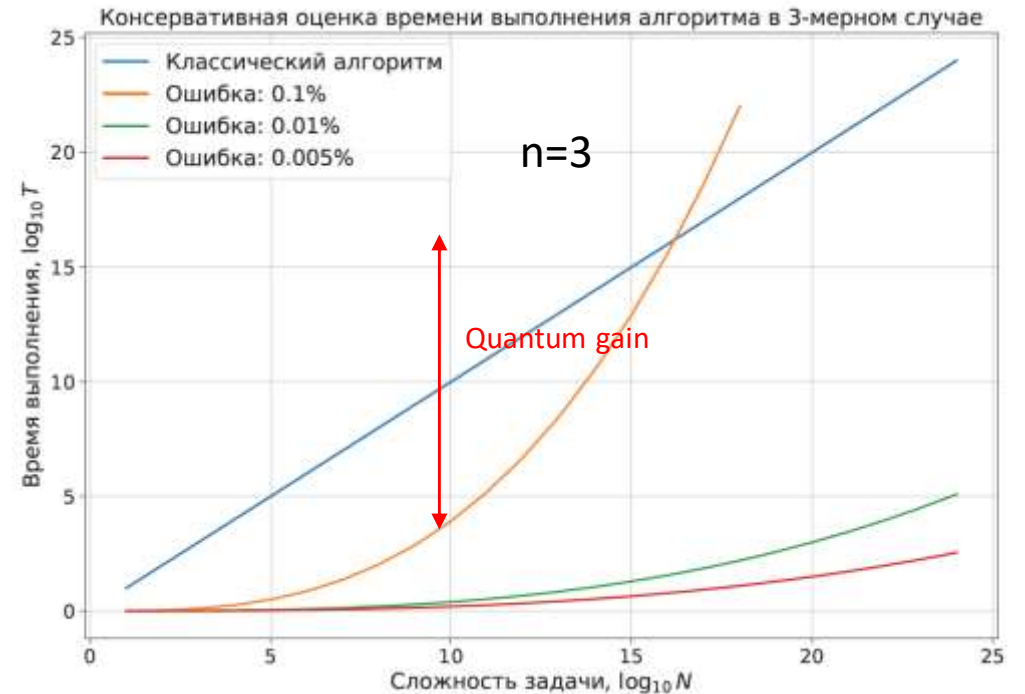
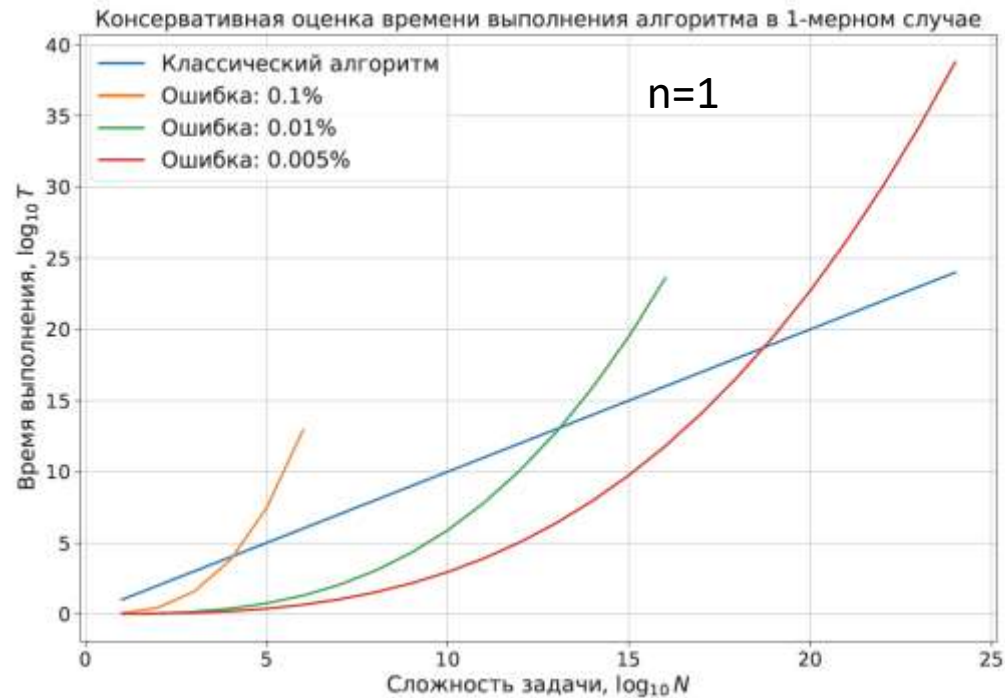
$$U_A = U_{QFT} \left(\prod_{i,j}^{d-1} e^{-i\phi_{ij} Z_i Z_j - i a_i Z_i} \right) U_{QFT}^+$$

$$K_A = \frac{(\log_2 N)^2}{2} + \frac{\log_2 N}{2}$$

QPS algorithm characteristics

(for exaflop $N \sim 10^{18}$ size of computational task)

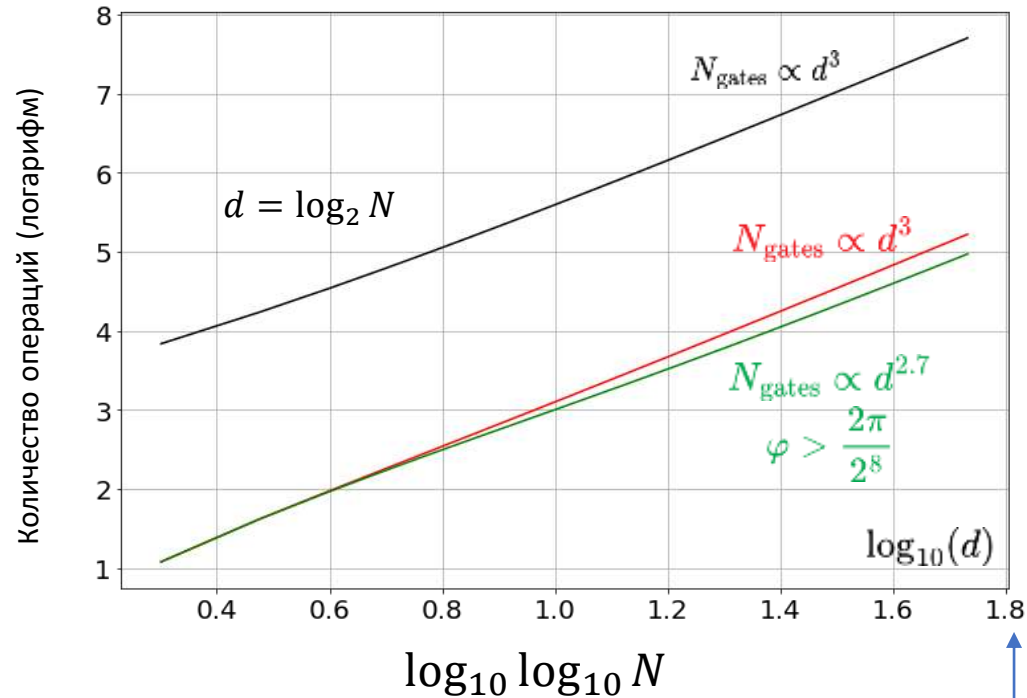
problem dimensionality	n=1	n=2	n=3
qubits number	295	355	415
number of quantum operations	775000 (2600 per qubit)	133000 (375 per qubit)	51000 (123 per qubit)
computational gain	20	10^{12}	10^{15}
required gate accuracy	99.995%	99.99%	99.98%



Comparison with others algorithms

Quantum fast Poisson solver:
the algorithm and complete and modular
circuit design,
Quantum Information Processing
19:170 (2020)

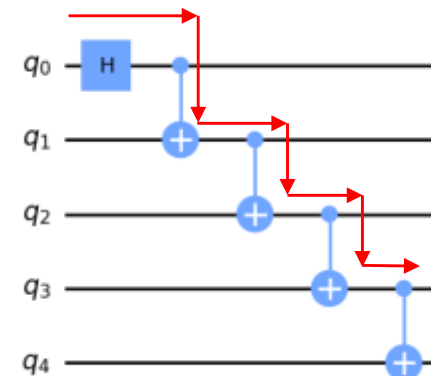
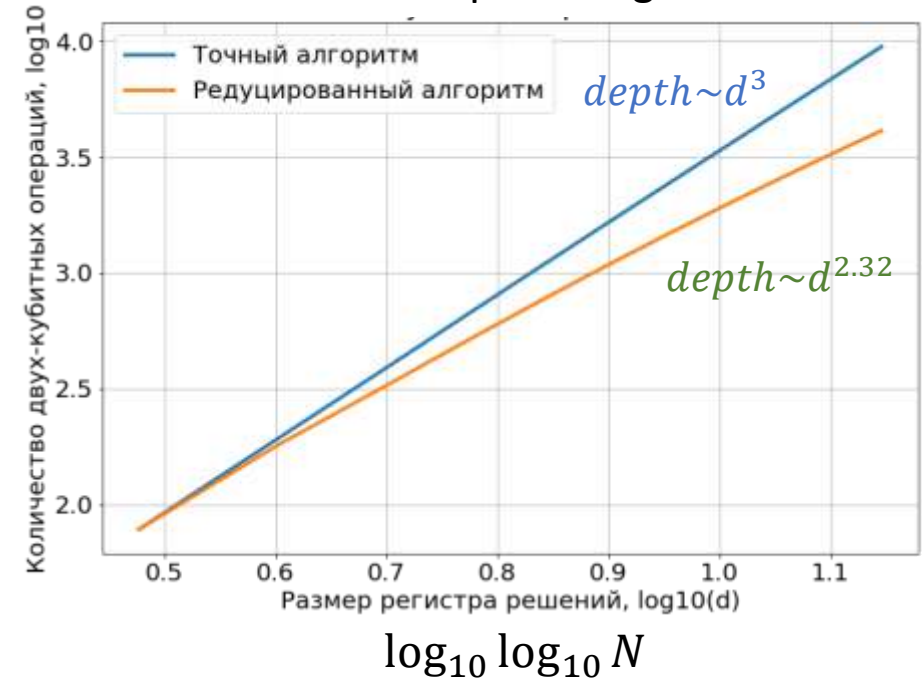
Number of gates scaling



Exceeds number of atoms
in the Universe

GHZ state preparation $(|00000\rangle + |11111\rangle)/\sqrt{2}$

Quantum depth of algorithm



Quantum depth: number of gates
along the longest path in the circuit

depth = 5

Main disadvantages of the direct method:

- Still requires a big qubit number $(4 + n) \log_2 N - 5$
- Number of quantum gates (quantum depth) is about 100000: **execution time exceeds** $1 - 10 \text{ msec}$
- Algorithm is **difficult to generalize** for time dynamics (for example diffusion or heat equation).
- We have **no classical information** about solution and need to repeat algorithm execution many times to infer it.

Are there any alternatives for HHL algorithm to solve linear systems?

Quantum variational solvers - QVS (main idea) $A\vec{x} = \vec{b}$

1. QVS prepares **candidates for solution** using a shallow sequence of quantum gates.
The solution is prepared in the form of **parametrized gate sequence** acting on a reference state.

variational ansatz

$$|x\rangle = U_K(\theta_K)U_{K-1}(\theta_{K-1}) \dots U_j(\theta_j) \dots U_1(\theta_1) |ref\ state\rangle$$

2. Measurement is performed on a candidate to evaluate its quality in terms of a **loss function**:

$$L(|x\rangle) \Rightarrow number$$

3. Optimization loop updates variational parameters to minimize the loss function with a classical optimization methods e.g. gradient descend

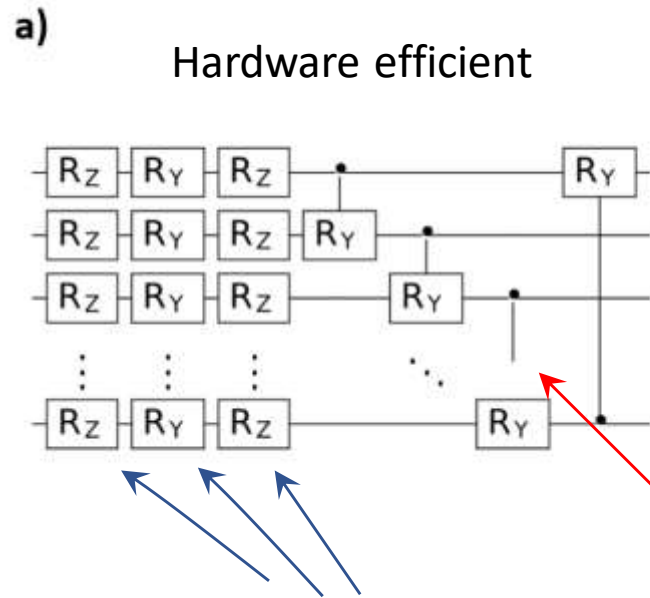
$$\min_{\vec{\theta}} L(|x(\vec{\theta})\rangle)$$

What are the possible loss functions?

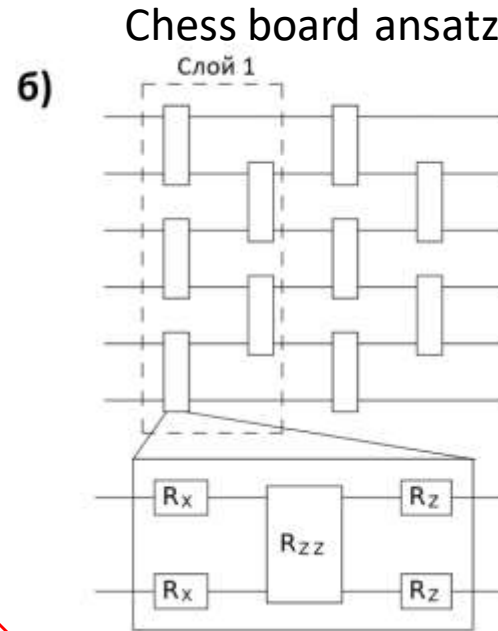
$$L_R(\vec{x}) = \|A\vec{x} - \vec{b}\|^2$$

$$L_T(\vec{x}) = \|A\vec{x} - \vec{b}\|^2 + \|\vec{x}\|^2$$

First class of VQS: **Agnostic variational ansatz** (uses no information about linear system)



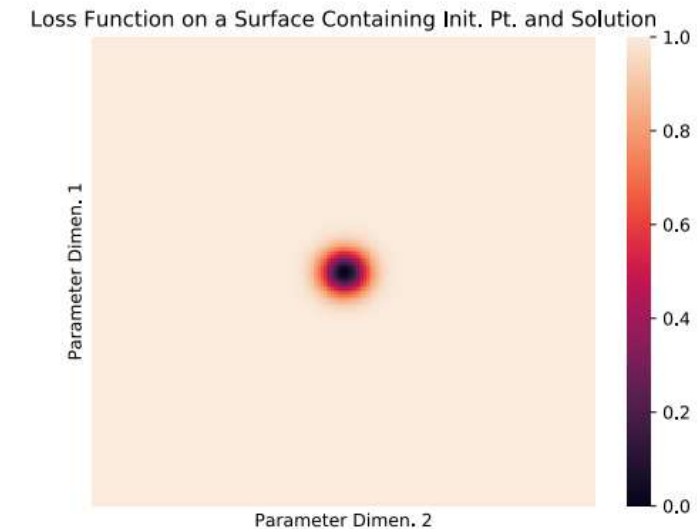
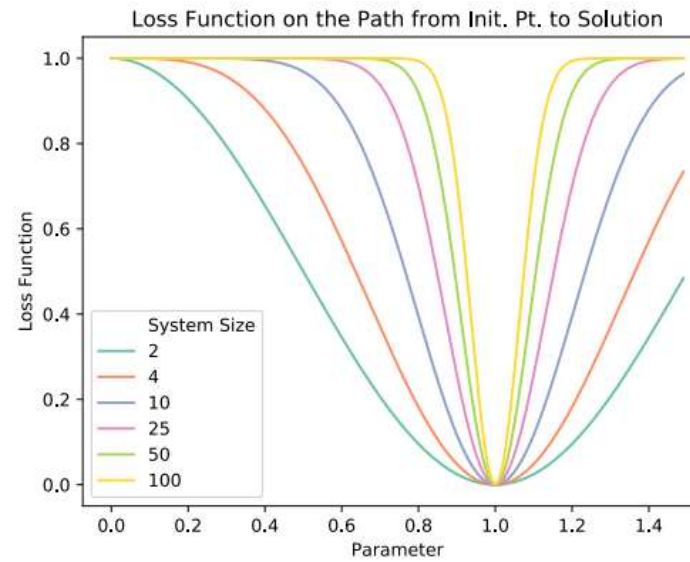
Variational parameters are encoded into single-qubit rotations at **fixed entangled gates**



The main problem of agnostic variational solvers is an **exponentially vanishing sensitivity** to the variational parameters: “**plateau effect**”


$$|\partial_{\theta} L| \leq \mathcal{O}(1/2^n)$$

The Hilbert space is huge and probe state hardly overlaps with the solution



Second class of VQS: Classical combination of variational quantum states

The main idea: different candidate states are **classically combined** to form a solution: $A\vec{x} = \vec{b}$

$$|x\rangle \in \text{Span}\{ |\psi_1\rangle, \dots, |\psi_K\rangle \}$$


candidate states


Search for the solution in the form:

$$|x\rangle = \alpha_1 |\psi_1\rangle + \alpha_2 |\psi_2\rangle + \dots + \alpha_K |\psi_K\rangle$$

Optimize variational parameters $\alpha_1, \alpha_2, \dots$ using classical convex optimization (**polynomially hard problem**)

Advantages of the method:

- The quantum state $|x\rangle$ **is never created** during algorithm execution, only simple quantum states $|\psi_k\rangle$ are generated.
- Algorithm requires the measurement of $K \times K$ simple averages:

$$L_R(|x\rangle) = \langle x|A^+A|x\rangle - 2\text{Re}\langle x|A|b\rangle + 1$$

$$\begin{aligned} &\langle \psi_i|A^+A|\psi_j\rangle \\ &\langle \psi_i|A|b\rangle \end{aligned}$$

Measured on a QC

- The loss function is calculated and optimized using classical computer

$$\sum_{i=1}^K \sum_{j=1}^K \langle \psi_i|A^+A|\psi_j\rangle$$

The main question: Where do we take good candidate states from?

Ansatz tree approach

(Huang H. Y., Bharti K., Rebentrost P. arXiv:1909.07344.)

$$A\vec{x} = \vec{b}$$

Every Hermitian matrix has a **unitary decomposition**: $A = \sum_i \beta_i U_i$

We make use of the form of the linear system!

Candidate states: $|\psi_{klm\dots}\rangle = \dots U_m U_l U_k |b\rangle$ are generated by unitary from decomposition of A

The variational state is built up step by step in the form of the tree:

$$|x\rangle = \alpha_0 |b\rangle + \alpha_1 U_1 |b\rangle + \alpha_2 U_2 U_1 |b\rangle + \dots$$



Tree root
(0 step)



1 gen. child
(1 step)



2 gen. child
(2 step)

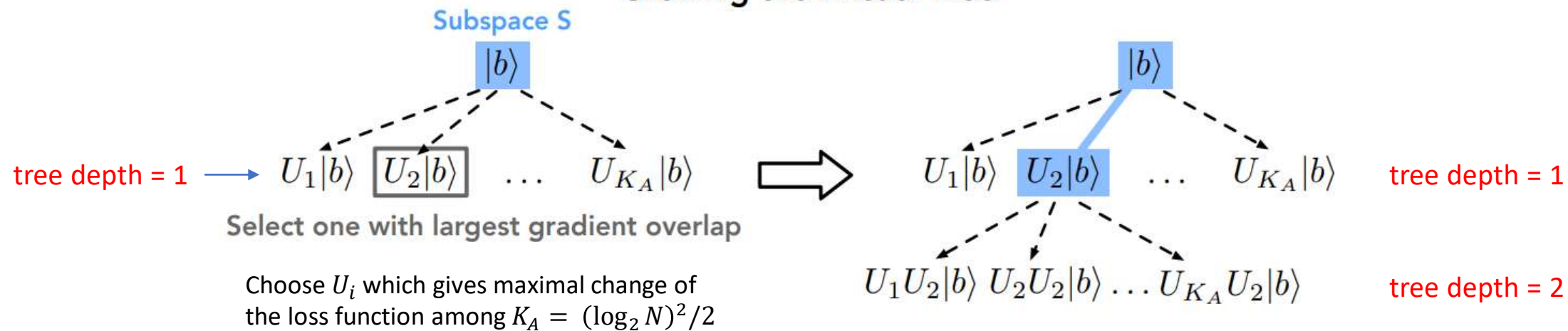
There is an efficient unitary decomposition:

The candidate states are generated by a Pauli Z operators followed by Fourier transform

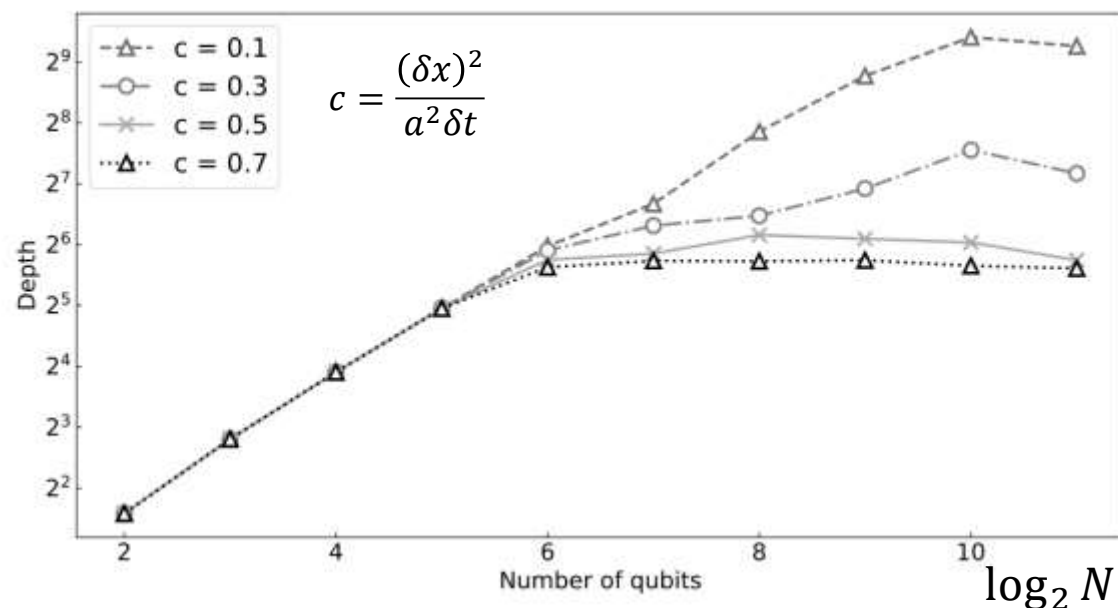
$$A = U_{QFT} \left(\sum_{k,m=0}^{d-1} d_{k,m} Z_k Z_m + \zeta_k Z_k \right) U_{QFT}^+$$

$$|\psi_k\rangle = Z_{i_1} Z_{i_2} \dots Z_{i_k} U_{QFT} |b\rangle$$

Growing the Ansatz Tree



The important question: How depth of the tree grows with the size of the problem?



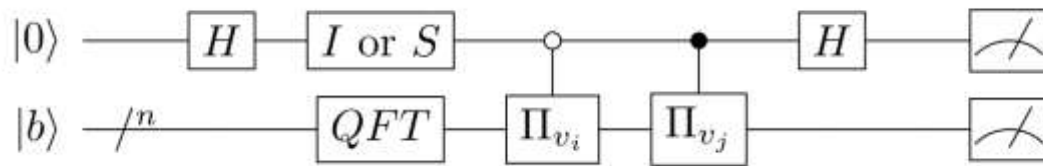
Comparison between direct and variational (Ansatz tree) methods for solving Poisson equation

1-dimensional case

	Direct solver	Ansatz tree
Qubit number	295	75
Number of gates	775000	25000 (needed to create solution)
required gate accuracy	99.995%	> 99.83%
Execution time	$(\log N)^3$	$(\log N)^{4+p}$

Main advantages: 1) during algorithm execution one uses short quantum circuits with low quantum depth

Circuit for measurement of $\langle \psi_i | A | b \rangle$



$$(\log_2 N)^2 / 2$$

$$\leq 4 \log_2 N$$

$$\Pi_{v_i} = Z_{i_1} Z_{i_2} \dots Z_{i_k}$$

Quantum depth
2026 for
 $N = 10^{18}$

2) Solution constructed in the form of a classical combination of states and thus known classically

Ansatz tree for solution of heat equation

$$\frac{\partial U}{\partial t} = a^2 \frac{\partial^2 U}{\partial x^2} + f(x, t)$$

implicit
finite difference
scheme



$$U_{n+1}^{\tau+1} + U_{n-1}^{\tau+1} - (2 + c)U_n^{\tau+1} = b_n^{\tau}$$

$$b_n^{\tau} = f_x^{\tau} \delta t + c U_n^{\tau}$$

lattice
parameter

$$c = \frac{(\delta x)^2}{a^2 \delta t}$$

On each time step algorithm solves Poisson-like equation with updated vector \vec{b} $A(c) \vec{x} = \vec{b}$

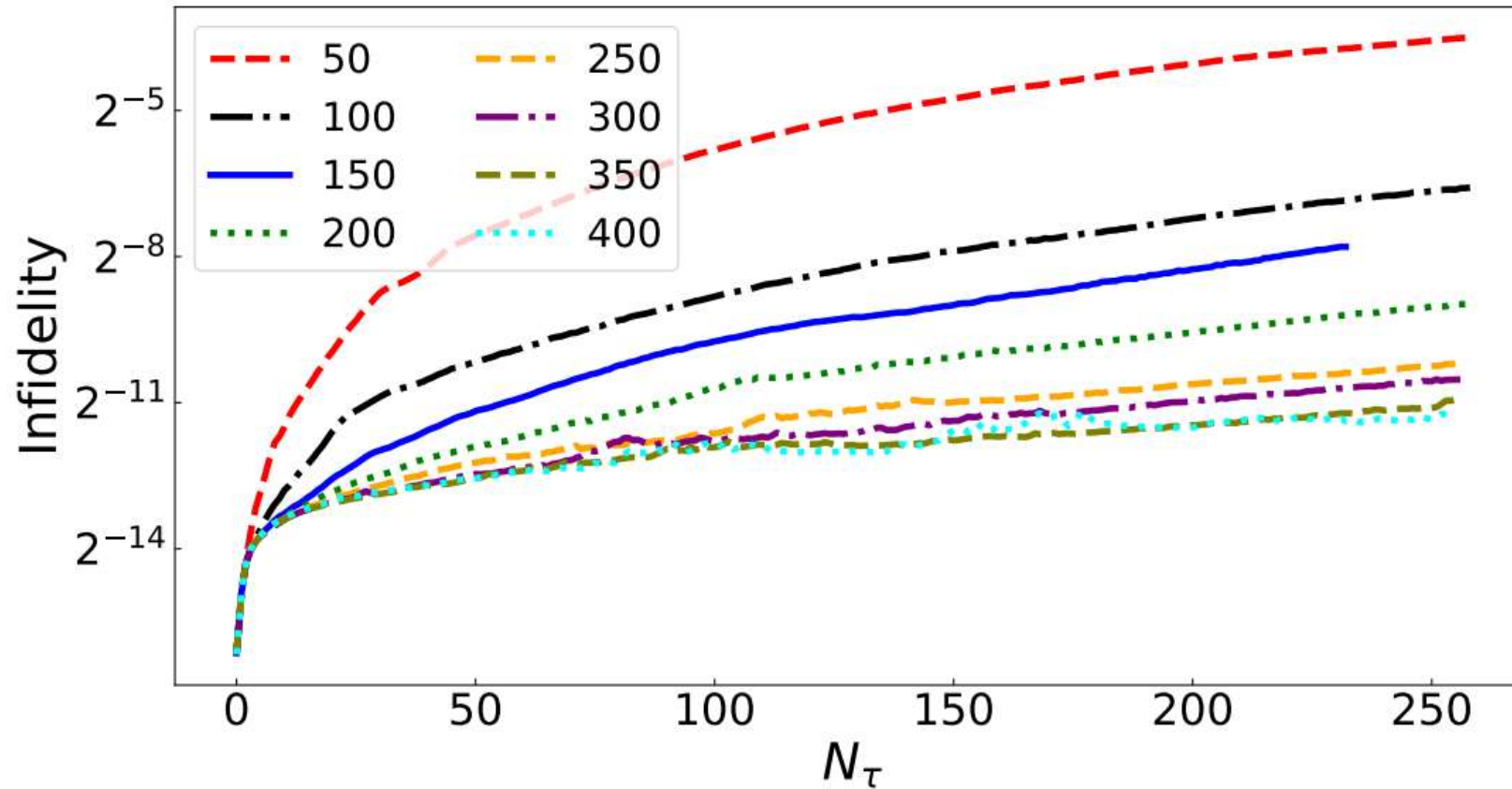
$$A(c) = \begin{pmatrix} -2-c & 1 & 0 & \cdots & 0 & 1 \\ 1 & -2-c & 1 & \cdots & 0 & 0 \\ 0 & 1 & -2-c & \cdots & 0 & 0 \\ \vdots & & \ddots & \ddots & & \vdots \\ 0 & 0 & \cdots & -2-c & 1 & 0 \\ 0 & 0 & \cdots & 1 & -2-c & 1 \\ 1 & 0 & \cdots & 0 & 1 & -2-c \end{pmatrix}$$

Depth analysis of variational quantum algorithms for the heat equation

N. M. Guseynov, A. A. Zhukov, W. V. Pogoso, and A. V. Lebedev
Phys. Rev. A **107**, 052422 – Published 26 May 2023

Full numerical simulation of a time dependent heat equation

$$N = 2^{11} = 2048$$



Variational quantum Ansatz tree approach for the heat equation

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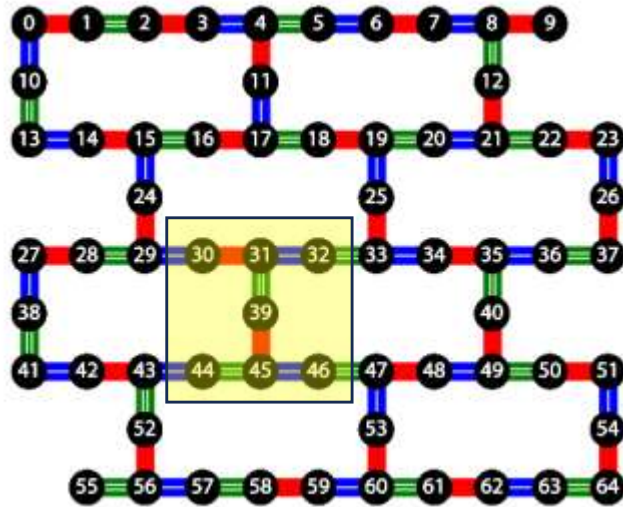
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Russian-Chinese International School "Superconducting functional materials for advanced quantum technologies"



Where are we from hardware point of view



$$T_2^* = 31\mu s$$

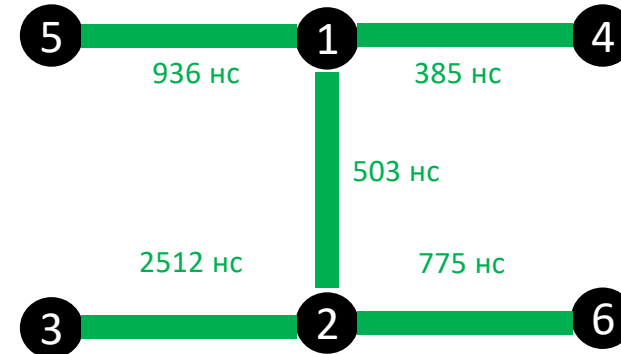
$$T_2 = 80\mu s$$

$$T_2^* = 32\mu s$$

$$T_2 = 38\mu s$$

$$T_2^* = 63\mu s$$

$$T_2 = 63\mu s$$



$$T_2 = 78\mu s$$

$$T_2^* = 40\mu s$$

$$T_2 = 46\mu s$$

$$T_2^* = 27\mu s$$

$$T_2 = 66\mu s$$

$$T_2^* = 42\mu s$$

Conclusions:

- We have developed two types of algorithms for solving linear differential equations.
- For the exaflops problem size we have found the necessary requirements for the quantum processor:

	495 qubits (direct method)	75 qubit (variational solver)
error rates:	< 0.02%	< 0.1%

- We are on the route to develop a large scale quantum processor (5-year lag from leading teams)
- 6-qubit quantum processor has been deployed

Thank you for attention!