3-Dimensional Quenched $Z(2) \times Z(2)$ Gauge Theory

Seyong Kim Sejong University

In collaboration with M. Mueller (IQI, RWTH), M. Rispler (IQI, RWTH) and D. Vodola (BASF) Work in progress

Surface Code (Toric Code) Quantum Computing And

Motivation



Probing the QCD phase structure using event-by-event fluctuations - Nayak, Tapan K. - arXiv:2008.04643

Sequential suppression of Upsilon system

CMS, PRL109 (2012) 222301



FASTSUM, JHEP07 (2014) 097 (over 100 citations)



Lattice QCD in non-zero temperature faces challenges

• Non-equilibrium QCD • QCD in finite baryon density

1. Introduction to Quantum Computing 2. Quantum Error Correction, Threshold Probability and Statistical Mechanics Model

- Discussions

Content

3. Surface code/Toric code and Quenched $Z(2) \times Z(2)$



1. Introduction to Quantum Computing



International Journal of Theoretical Physics, Vol. 21, Nos. 6/7, 1982

Simulating Physics with Computers

Richard P. Feynman

Department of Physics, California Institute of Technology, Pasadena, California 91107

1. INTRODUCTION

On the program it says this is a keynote speech—and I don't know what a keynote speech is. I do not intend in any way to suggest what should be in this meeting as a keynote of the subjects or anything like that. I have my own things to say and to talk about and there's no implication that anybody needs to talk about the same thing or anything like it. So what I want to talk about is what Mike Dertouzos suggested that nobody would talk about. I want to talk about the problem of simulating physics with

Beginning

Received May 7, 1981



Google's Sycamore processor mounted in a cryostat, recently used to demonstrate quantum supremacy and the largest quantum chemistry simulation on a quantum computer. Credit: Rocco Ceselin





2025

KOOKABURRA 4,158 qubits



Communications

- Classical
- Qauntum

https://spectrum.ieee.org/ibm-condor







High-threshold and low-overhead fault-tolerant quantum memory

Sergey Bravyi¹, Andrew W. Cross¹, Jay M. Gambetta¹, Dmitri Maslov¹, Patrick Rall², and Theodore J. Yoder¹

¹IBM Quantum, IBM T.J. Watson Research Center, Yorktown Heights, NY 10598 (USA) ²IBM Quantum, MIT-IBM Watson AI Lab, Cambridge, MA 02142 (USA)

August 16, 2023

Quantum error correction becomes a practical possibility only if the physical error rate is below a threshold value that depends on a particular quantum code, syndrome measurement circuit, and a decoding algorithm. Here we present an end-to-end quantum error correction protocol that implements fault-tolerant memory based on a family of LDPC codes with a high encoding rate that achieves an error threshold of 0.8% for the standard circuit-based noise model. This is on par with the surface code which has remained an uncontested leader in terms of its high error threshold for nearly 20 years. The full syndrome measurement cycle for a length-n code in our family requires n ancillary qubits and a depth-7 circuit composed of nearest-neighbor CNOT gates. The required qubit connectivity is a degree-6 graph that consists of two edge-disjoint planar subgraphs. As a concrete example, we show that 12 logical qubits can be preserved for ten million syndrome cycles using 288 physical qubits in total, assuming the physical error rate of 0.1%. We argue that achieving the same level of error suppression on 12 logical qubits with the surface code would require more than 4000 physical qubits. Our findings bring demonstrations of a low-overhead fault-tolerant quantum memory within the reach of near-term quantum processors.

Abstract

15 Aug 2023

ant-ph]



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176KB HTML 문서





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Logical quantum processor based on reconfigurable atom arrays

Dolev Bluvstein¹, Simon J. Evered¹, Alexandra A. Geim¹, Sophie H. Li¹, Hengyun Zhou^{1,2}, Tom Manovitz¹, Sepehr Ebadi¹, Madelyn Cain¹, Marcin Kalinowski¹, Dominik Hangleiter³, J. Pablo Bonilla Ataides¹, Nishad Maskara¹, Iris Cong¹, Xun Gao¹, Pedro Sales Rodriguez², Thomas Karolyshyn², Giulia Semeghini⁴, Michael J. Gullans³, Markus Greiner¹, Vladan Vuletić⁵, and Mikhail D. Lukin^{1,†} ¹Department of Physics, Harvard University, Cambridge, MA 02138, USA ²QuEra Computing Inc., Boston, MA 02135, USA ³Joint Center for Quantum Information and Computer Science, NIST/University of Maryland, College Park, Maryland 20742, USA ⁴John A. Paulson School of Engineering and Applied Sciences, Harvard University, Cambridge, MA 02138, USA ⁵Department of Physics and Research Laboratory of Electronics, Massachusetts Institute of Technology, Cambridge, MA 02139, USA [†]Corresponding Author; E-mail: lukin@physics.harvard.edu

Suppressing errors is the central challenge for useful quantum computing [1], requiring quantum error correction [2–6] for large-scale processing. However, the overhead in the realization of error-corrected "logical" qubits, where information is encoded across many physical qubits for redundancy [2–4], poses significant challenges to large-scale logical quantum computing. Here we report the realization of a programmable quantum processor based on encoded logical qubits operating with up to 280 physical qubits. Utilizing logical-level control and a zoned architecture in reconfigurable neutral atom arrays [7], our system combines high two-qubit gate fidelities [8], arbitrary connectivity [7, 9], as well as fully programmable single-qubit rotations and mid-circuit readout [10-15]. Operating this logical processor with various types of encodings, we demonstrate improvement of a two-qubit logic gate by scaling surface code [6] distance from d = 3 to d = 7, preparation of color code qubits with break-even fidelities [5], fault-tolerant creation of logical GHZ states and feedforward entanglement teleportation, as well as operation of 40 color code qubits. Finally, using three-dimensional [[8,3,2]] code blocks [16, 17], we realize computationally complex sampling circuits [18] with up to 48 logical qubits entangled with hypercube connectivity [19] with 228 logical two-qubit gates and 48 logical CCZ gates [20]. We find that this logical encoding substantially improves algorithmic performance with error detection, outperforming physical qubit fidelities at both cross-entropy benchmarking and quantum simulations of fast scrambling [21, 22]. These results herald the advent of early error-corrected quantum computation and chart a path toward large-scale



Example of QC: Shor's Algorithm

- RSA algorithm —> integer factoring problem
- change integer factoring problem into order-finding problem using a digital computer
- solve order-finding problem using a Quantum Computer
- P. Shor, proceedings of 35th annual symposium on the foundations of computer science, 1994, p124-134

- How to factor N = pq, a product of two primes https://www.youtube.com/watch?v=-UrdExQW0cs&t=712s back to step 1
- 1. Make a guess, g < N that shares no factors with N 2. Find r such that $g^r = mN + 1$ 3. If r is even, calculate $(q^{r/2} + 1)$ and $(q^{r/2} - 1)$. If r is odd, go
- 4. Use Euclid's algorithm to find the greatest common divisor of $g^{r/2} + 1$ and N



How to factor N, a product of two primes (example)

https://www.youtube.com/watch?v=-UrdExQW0cs&t=712s

- N = 77 = pq
- 1. g = 8
- 2. Find *r* such that $g^r = mN + 1 \rightarrow r = 10$
- 3. $(g^{r/2} + 1) = 32,769, (g^{r/2} 1) = 32,767, (g^{r/2} + 1)(g^{r/2} 1) = mN$
- 4. Use Euclid's algorithm to find the greatest common divisor of $g^{r/2} + 1$ and N: 32769/77 = 425R44 - 77/44 = 1R33 - 44/33 = 1R11 - 33/11 = 3R0
- 5. 11 is a prime factor of 77!



How to factor N, a product of two primes (example)

https://www.youtube.com/watch?v=-UrdExQW0cs&t=712s

- QC is better at step 3, finding r of $g^r = mN + 1$, because $g, g^1, g^2, \dots, g^{r-1}, g^r, g^{r+1}, \dots, g^{2r}, \dots, g^{3r}, \dots,$ are periodic
- Fast Fourier transform in DC and quantum Fourier transform in QC!



• Finding r is related to Fourier transform and QC Fourier transform is faster

For digital computer,

the best Fourier transform algorithm is Cooley-Tukey algorithm and the number of operations is

 $\sim N \times \log_2 N$, $N = 2^n$



Cooley-Tukey Algorithm

 $ex_{0}, x_{1}, x_{2}, x_{3}, x_{4}, x_{5}, x_{6}, x_{7}, N = 2^{3}, m = 3$

For $k = 1, \cdots, N$,

$$X_{k} = \sum_{n=0}^{n=7} x_{n}(\omega_{N})^{-kn}, \quad \omega_{N} = e^{\frac{2\pi i}{N}} = e^{\frac{2\pi i}{8}} = e^{\frac{2\pi i}{8}}$$



Note that $X_{k+N/2} = X_{\ell}^e + \omega_N^{-k} X_{\ell}^o$

$$X_{k}^{e} = X_{k}^{ee} + X_{k}^{eo} \ \omega_{\frac{N}{2}}^{-k} = X_{k}^{ee} + X_{k}^{eo} \ \omega_{N}^{-2k} \text{ since } \omega_{\frac{N}{2}} = e^{\frac{2\pi i}{N}} = e^{\frac{4\pi i}{N}} = \omega_{N}^{2}$$

Similarly, $X_k^o = X_k^{oe} + X_k^{oo} \omega_{\frac{N}{2}}^{-k} = X_k^{oe} + X_k^{oo} \omega_N^{-2k}$

 $e^{\frac{i\pi}{4}}$

Cooley-Tukey Algorithm

$$ex)x_{0}, x_{1}, x_{2}, x_{3}, x_{4}, x_{5}, x_{6}, x_{7}, \quad N = 2^{3}, m =$$

For $k = 1, \dots, N$,
$$X_{k} = x_{0}\omega_{N}^{0} + x_{1}\omega_{N}^{-k} + x_{2}\omega_{N}^{-2k} \dots + x_{7}\alpha$$
$$= X_{k}^{e} + X_{k}^{o} \omega_{N}^{-k}$$
$$= (X_{k}^{ee} + X_{k}^{eo} \omega_{N}^{-2k}) + \omega_{N}^{-k}(X_{k}^{oe} + X_{k}^{oe})$$

Note that since

$$X_{k}^{ee} = x_{0} + x_{4}\omega_{N}^{-4k}, X_{k}^{eo} = x_{2} + x_{6}\omega_{N}^{-4k}$$
$$X_{k} = (x_{0} + x_{4}\omega_{N}^{-4k}) + \omega_{N}^{-2k}(x_{2} + x_{6}\omega_{N}^{-4k})$$
$$= x_{0} + x_{4}\omega_{N}^{-4k} + x_{2}\omega_{N}^{-2k} + x_{6}\omega_{N}^{-6k} + x_{6}\omega_{N}^{-6k}$$

3,

 v_N^{-7k}

 $X_k^{oo} \omega_N^{-2k}$

^{4k}, $X_k^{oe} = x_1 + x_5 \omega_N^{-4k}$, $X_k^{oo} = x_3 + x_7 \omega_N^{-4k}$ $(x_1^{-4k}) + \omega_N^{-k}[(x_1 + x_5\omega_N^{-4k}) + \omega_N^{-2k}(x_3 + x_7\omega_N^{-4k})]$ $+ x_1 \omega_N^{-k} + x_5 \omega_N^{-5k} + x_3 \omega_N^{-3k} + x_7 \omega_N^{-7k}$

Cooley-Tukey Algorithm

Bit-ordering —) bit-reverse, swap so that the data x_i ordered as bit-reversed $n = 0, (0,0,0) \rightarrow (0,0,0), n = 0$ $n = 1, (0,0,1) \rightarrow (1,0,0), n = 4$ $n = 2, (0,1,0) \rightarrow (0,1,0), n = 2$ $n = 3, (0,1,1) \rightarrow (1,1,0), n = 6$ $n = 4, (1,0,0) \rightarrow (0,0,1), n = 1$ $n = 5, (1,0,1) \rightarrow (1,0,1), n = 5$ $n = 6, (1,1,0) \rightarrow (0,1,1), n = 3$ $n = 7, (1,1,1) \rightarrow (1,1,1), n = 7$ $x_0, x_1, x_2, x_3, x_4, x_5, x_6, x_7 \rightarrow x_0, x_4, x_2, x_6, x_1, x_5, x_3, x_7$

Cooley-Tukey Algorithm Bit-ordering —) bit-reverse, swap so that the data x_i ordered as

- 1. bit-reversed
- 2. Then, form the first loop

 $(x_0 + x_4\omega_N^{-4k}), (x_2 + x_6\omega_N^{-4k}), (x_1 + x_5\omega_N^{-4k}), (x_3 + x_7\omega_N^{-4k})$

3. Then, form the 2nd loop

 $(x_0 + x_4\omega_N^{-4k}) + \omega_N^{-2k}(x_2 + x_6\omega_N^{-4k}), (x_1 + x_5\omega_N^{-4k}) + \omega_N^{-2k}(x_3 + x_7\omega_N^{-4k})$

4. Then, form the 3rd loop

 $[(x_0 + x_4\omega_N^{-4k}) + \omega_N^{-2k}(x_2 + x_6\omega_N^{-4k})] + \omega_N^{-k}[(x_1 + x_5\omega_N^{-4k}) + \omega_N^{-2k}(x_3 + x_7\omega_N^{-4k})]$



Cooley-Tukey Algorithm $k = 3, \left[(x_0 + x_4 \omega_N^{-12}) + \omega_N^{-6} (x_2 + x_6 \omega_N^{-12}) \right] + \omega_N^{-3} \left[(x_1 + x_5 \omega_N^{-12}) + \omega_N^{-6} (x_3 + x_7 \omega_N^{-12}) \right]$ $k = 4, \left[(x_0 + x_4 \omega_N^{-16}) + \omega_N^{-8} (x_2 + x_6 \omega_N^{-16}) \right] + \omega_N^{-4} \left[(x_1 + x_5 \omega_N^{-16}) + \omega_N^{-8} (x_3 + x_7 \omega_N^{-16}) \right]$ $k = 5, \left[(x_0 + x_4 \omega_N^{-20}) + \omega_N^{-10} (x_2 + x_6 \omega_N^{-20}) \right] + \omega_N^{-5} \left[(x_1 + x_5 \omega_N^{-20}) + \omega_N^{-10} (x_3 + x_7 \omega_N^{-20}) \right]$ $k = 6, \left[(x_0 + x_4 \omega_N^{-24}) + \omega_N^{-12} (x_2 + x_6 \omega_N^{-24}) \right] + \omega_N^{-6} \left[(x_1 + x_5 \omega_N^{-24}) + \omega_N^{-12} (x_3 + x_7 \omega_N^{-24}) \right]$ $k = 7, \left[(x_0 + x_4 \omega_N^{-28}) + \omega_N^{-14} (x_2 + x_6 \omega_N^{-28}) \right] + \omega_N^{-7} \left[(x_1 + x_5 \omega_N^{-28}) + \omega_N^{-14} (x_3 + x_7 \omega_N^{-28}) \right]$

Quantum Fourier transform

 $[(x_0 + x_4\omega_N^{-4k}) + \omega_N^{-2k}(x_2 + x_6\omega_N^{-4k})] + \omega_N^{-k}[(x_1 + x_5\omega_N^{-4k}) + \omega^{-2k}(x_3 + x_7\omega_N^{-4k})], N = 2^3 = 8$



Figure 5.4: $QFT_{M/2}$ and a Hadamard gate correspond to $FFT_M/2$ on the odd and even terms



Figure 5.6: First Iteration



Figure 5.5: QFT_M is reduced to $QFT_{M/2}$ and M additional gates



Figure 5.7: Second Iteration. Recall that $H = QFT_1$

https://courses.edx.org/c4x/BerkeleyX/CS191x/asset/chap5.pdf







$$\frac{1}{\sqrt{2}} \left(|0\rangle + e^{2\pi i [0.x_1...x_n]} |1\rangle - \frac{1}{\sqrt{2}} \left(|0\rangle + e^{2\pi i [0.x_2...x_n]} |1\rangle - \frac{1}{\sqrt{2}} \left(|0\rangle + e^{2\pi i [0.x_3...x_n]} |1\rangle - \frac{1}{\sqrt{2}} \left(|0\rangle + e^{2\pi i [0.x_3...x_n]} |1\rangle - \frac{1}{\sqrt{2}} \left(|0\rangle + e^{2\pi i [0.x_3...x_n]} |1\rangle - \frac{1}{\sqrt{2}} \left(|0\rangle + e^{2\pi i [0.x_3...x_n]} |1\rangle - \frac{1}{\sqrt{2}} \left(|0\rangle + e^{2\pi i [0.x_3...x_n]} |1\rangle - \frac{1}{\sqrt{2}} \left(|0\rangle + e^{2\pi i [0.x_3...x_n]} |1\rangle - \frac{1}{\sqrt{2}} \left(|0\rangle + e^{2\pi i [0.x_3...x_n]} |1\rangle - \frac{1}{\sqrt{2}} \left(|0\rangle + e^{2\pi i [0.x_3...x_n]} |1\rangle - \frac{1}{\sqrt{2}} \left(|0\rangle + e^{2\pi i [0.x_3...x_n]} |1\rangle - \frac{1}{\sqrt{2}} \left(|0\rangle + e^{2\pi i [0.x_3...x_n]} |1\rangle - \frac{1}{\sqrt{2}} \left(|0\rangle + e^{2\pi i [0.x_3...x_n]} |1\rangle - \frac{1}{\sqrt{2}} \left(|0\rangle + e^{2\pi i [0.x_3...x_n]} |1\rangle - \frac{1}{\sqrt{2}} \left(|0\rangle + e^{2\pi i [0.x_3...x_n]} |1\rangle - \frac{1}{\sqrt{2}} \left(|0\rangle + e^{2\pi i [0.x_3...x_n]} |1\rangle - \frac{1}{\sqrt{2}} \left(|0\rangle + e^{2\pi i [0.x_3...x_n]} |1\rangle - \frac{1}{\sqrt{2}} \left(|0\rangle + e^{2\pi i [0.x_3...x_n]} |1\rangle - \frac{1}{\sqrt{2}} \left(|0\rangle + e^{2\pi i [0.x_3...x_n]} |1\rangle - \frac{1}{\sqrt{2}} \left(|0\rangle + e^{2\pi i [0.x_3...x_n]} |1\rangle - \frac{1}{\sqrt{2}} \left(|0\rangle + e^{2\pi i [0.x_3...x_n]} |1\rangle - \frac{1}{\sqrt{2}} \left(|0\rangle + e^{2\pi i [0.x_3...x_n]} |1\rangle - \frac{1}{\sqrt{2}} \left(|0\rangle + e^{2\pi i [0.x_3...x_n]} |1\rangle - \frac{1}{\sqrt{2}} \left(|0\rangle + e^{2\pi i [0.x_3...x_n]} |1\rangle - \frac{1}{\sqrt{2}} \left(|0\rangle + e^{2\pi i [0.x_3...x_n]} |1\rangle - \frac{1}{\sqrt{2}} \left(|0\rangle + e^{2\pi i [0.x_3...x_n]} |1\rangle - \frac{1}{\sqrt{2}} \left(|0\rangle + e^{2\pi i [0.x_3...x_n]} |1\rangle - \frac{1}{\sqrt{2}} \left(|0\rangle + e^{2\pi i [0.x_3...x_n]} \right) \right)$$

$$----H - R_{2} - \frac{1}{\sqrt{2}} \left(|0\rangle + e^{2\pi i [0.x_{n-1}x_{n}]} |1\rangle \right)$$
$$----H - \frac{1}{\sqrt{2}} \left(|0\rangle + e^{2\pi i [0.x_{n}]} |1\rangle \right)$$

https://en.wikipedia.org/wiki/Quantum_Fourier_transform



 $|0>|0> \rightarrow \frac{1}{\sqrt{N}} \sum_{x=0}^{N-1} |x>|0> \text{ under } QFT_N$

 $\frac{1}{\sqrt{N}} \sum_{x=0}^{N-1} |x| > |0| > \to \frac{1}{\sqrt{N}} \sum_{x=0}^{N-1} |x| > |f(x)| > \text{ under } U_f$

 $\frac{1}{\sqrt{N}} \sum_{x=0}^{N-1} |x| |f(x)| > \to \sqrt{\frac{r}{N}} \sum_{x=0}^{N/r-1} |ir + x_0| > |f(x_0)| >$

under measurement on $|f(x)\rangle$

QFT for Shor's Algorithm

Example: QFT for Shor's algorithm

 $|0>|0> \rightarrow \frac{1}{\sqrt{8}} \sum_{r=0}^{\prime} |x>|0>$ under QFT_8

 $\frac{1}{\sqrt{8}} \sum_{x=0}^{7} |x| > |0| > \rightarrow \frac{1}{\sqrt{8}} \sum_{x=0}^{7} |x| > |f(x)| > \text{ under } U_f, f(x) = x \pmod{2}$

Note that x(mod 2) = 0 or 1. Choose 1

$$\frac{1}{\sqrt{8}} \sum_{x=0}^{7} |x| > |1| > \rightarrow \frac{1}{2} (|1| > + |3| > + |5|)$$

 $\frac{1}{2}(|1 > + |3 > + |5 > + |7 >) \rightarrow |0 > - |4 > \text{ under } QFT_8$

5 > + |7 > |1 > under measurement on f

Example: QFT for Shor's algorithm



Figure 5.9: Circuit for factoring

Fourier transform of the same signal requires ~ $2^n \times \log_2 2^n$ number of operations for digital computer $\sim n^2$ number of operations tor quantum computer





Quantum Error Correction, Threshold Probability and Statistical Mechanics Model

• Fighting quantum decoherence with entanglement

• Quantum Error Correction (QEC)

cf. B.M Terhal, Rev. Mod. Phys. 87 (2015) 307

Quantum Computing in "noisy environment" or Fault-Tolerant QC

QEC: Shor's code (classical counter-part)

- Measure each bit and do majority choice
- Not applicable to quantum computer
- the proceedings of 37th FOCS, p55-65

• Smallest classical code, $|0\rangle_L = |000\rangle$, $|1\rangle_L = |111\rangle$

• cf. P. W. Shor, "Fault-tolerant quantum computation" in



QEC: Shor's code (9-bit concatenated code)

• $|0\rangle_L = |000\rangle$, $|1\rangle_L = |111\rangle$ (entangled qubits)

supperposition/entanglement, $|\psi\rangle_L =$

- Measuring any of qubits by Z_1, Z_2, Z_3 collapses entangled data qubits
- Measuring the ancilla qubit collapses entangled data qubits if 1 (or 2, or 3) data qubit is entangled with the ancilla qubit (i.e. measuring Z_1 , or (Z_2, Z_3))
- Measuring ancilla qubit doesn't collapse entangled data qubits if 1,2 (or 2, 3) data qubits are entangled with the ancilla qubit (i.e. measuring Z_1Z_2 , or Z_2Z_3)

• Not measure data qubits, but measure parity of data without collapsing qubits

$$= \cos\frac{\theta}{2} |0>_L + \sin\frac{\theta}{2} e^{\phi} |1>_L$$



Z₁Z₂ parity check

Before measurement

|000 > |0 >, |001 > |1 >, |010 > |0 >, |011 > |1 >, |100 > |0 >, |101 > |1 >, |110 > |0 >, |111 > |1 >

$c_1 |000 > + c_2 |111 >$ $|0>_{a}$

|000 > |0 >, |001 > |1 >, |010 > |1 >, |011 > |0 >, |100 > |0 >, |101 > |1 >, |110 > |1 >, |111 > |0 >



Before measurement





Quantum error and statistical model

- Modeling quantum error pattern
- Mapping quantum error pattern to statistical model
- cf. simple case: Dennis et al, J. Math. Phys. 43 (2002) 4452

• Specific quantum code with stabilizer formalism

Quantum Error Detection/Correction

- (quantum error detection)
- Correct quantum error

• Check whether error happens via the measurement of "ancilla" qubits: measurement result is called syndrome

• From the syndrome, guess quantum error probabilistically



Error rate and threshold probability

- If the quantum error rate is higher than the "threshold probability", QEC is not possible.
- Above the threshold probability, "probabilistic correction"
- is not possible.
- "Probabilistic interpretation model

• "Probabilistic interpretation" is related to some statistical



M. Rispler, D. Vodola, SK, M. Muller, "Fundamental Thresholds of Realistic Quantum Error Correction Circuits from Classical Spin Models", Quantum 6 (2022), 618

 realistic quantum circuit diagram for 1–D repetition code with phase flip error and mapping to a statistical model (quenched 2-D Ising model on a triangular lattice)

1-d repetition code and correlated error

- Protecting against phase flip error (Z-error)
- Realistic quantum circuit which implements the algorithm
- Analysis of the correlated quantum error from 1-qubit
- error, 1-qubit gate error, 2-qubit gate error and etc
- Mapping into 2-d random bond Ising model on triangular lattice



cf. J. Kelly et al (Google Quantum AI), "State preservation by repetitive error detection in a superconducting quantum circuit", Nature 519 (2015) 66



Figure 3 | **Protecting the GHZ state from bit-flip errors. a**, Quantum circuit for generating the GHZ state and two cycles of the repetition code. CNOT gates are physically implemented with controlled-phase (CZ) and single qubit gates. **b**, Quantum state tomography on the input (top left 'Input', left of black dashed line), and after the repetition code conditional on the detection events (between black dashed lines): we input a GHZ state with a fidelity (*F*) of 82%, and find, for the case of no detection events (top right 'Output', above grey dashed line), a 78% fidelity GHZ state. For the detection event connecting both measurement qubits (bottom left 'Raw output', below grey dashed line), indicating a likely bit-flip error on the central data qubit, we find that through

cf. J. Kelly et al (Google Quantum AI), "Exponential suppression of bit or phase errors with cyclic error correction ", Nature 595 (2021) 383

Article



Fig. 1 | **Stabilizer circuits on Sycamore. a**, Layout of distance-11 repetition code and distance-2 surface code in the Sycamore processor. In the experiment, the two codes use overlapping sets of qubits, which are offset in the figure for clarity. **b**, Pauli error rates for single-qubit and CZ gates and identification error rates for measurement, each benchmarked in simultaneous operation. **c**, Circuit schematic for the phase-flip code. Data qubits are randomly initialized into $|+\rangle$ or $|-\rangle$, followed by repeated application of XX stabilizer measurements and finally X-basis measurements of the data qubits. Hadamard refers to the Hadamard gate, a quantum operation. **d**, Illustration of error detection events that occur when a measurement disagrees with the previous round. **e**, Fraction of measurements (out of 80,000) that detected an error versus measurement round for the d=11

3. Surface code/Toric code and Quenched $Z(2) \times Z(2)$ gauge theory

Surface code(Toric Code) circuit layout



V(i-1,j) $S^x(i,j)$ 1 - 1H(i, j - 1)3 V(i, j)



Qubit error interpretation



 $J_{y}(i,j,k) = J_{x}(i,j,k)J_{z}(i,j,k)$ $J_{y}(i,j,k) = J_{x}(i,j,k)J_{z}(i,j,k)$ 5_{x} (\tilde{n}, \tilde{j}) k $-\hat{g}(n,j,k), if x-error, y:(n,j,k), 1$ $(y-z,-) \geq;(n+1,j,k), 2$ $J_{z}(ijk)$ if y = error, z(ijk), 1(-z, -), z(ijk), 2 $\chi^{2}(x,y,-)$ $\chi^{2}(x,y,-)$ $\chi^{2}(x,y,-)$ $\chi^{2}(x,y,-)$ $\chi^{2}(x,y,-)$

Measurement error interpretation



Lattice hamiltonian for simulation Phase flip or Bit flip

$$\begin{aligned} H_{x,z}(i,j,k) &= -J_x^{-1}(i,j,k)\sigma_y(i,j) \\ &-J_x^{-2}(i,j,k)\sigma_t(i,j) \\ &-J_x^{-3}(i,j,k)\sigma_x(i,j) \\ &-J_z^{-1}(i,j,k)\tau_y(i,j) \\ &-J_z^{-2}(i,j,k)\tau_t(i,j) \\ &-J_z^{-3}(i,j,k)\tau_x(i,j) \end{aligned}$$

 $\sigma_t(i, j+1, k)\sigma_v(i, j, k+1)\sigma_t(i, j, k)$ $(i, k)\sigma_{x}(i, j, k+1)\sigma_{t}(i+1, j, k)\sigma_{x}(i, j, k)$ $\sigma_v(i+1,j,k)\sigma_x(i,j+1,k)\sigma_v(i,j,k)$ $(j,k)\tau_t(i,j+1,k)\tau_v(i,j,k+1)\tau_t(i,j,k)$ $(j, k)\tau_{x}(i, j, k+1)\tau_{t}(i+1, j, k)\tau_{x}(i, j, k)$ $(j,k)\tau_y(i+1,j,k)\tau_x(i,j+1,k)\tau_y(i,j,k)$

Lattice hamiltonian for simulation Phase flip and Bit flip together

$$H_{y}(i,j,k) = -J_{y}^{1}(i,j,k)\sigma_{y}(i,j,k)$$

$$\times \tau_{t}(i+1,j,k)\tau_{x}(i+1,j,k)\tau_{x}(i+1,j,k)$$

$$-J_{y}^{2}(i,j,k)\sigma_{t}(i,j,k)$$

$$\times \tau_{y}(i,j+1,k)\tau_{t}(i,j,k)$$

$$\begin{split} \dot{i}, k)\sigma_t(i, j+1, k)\sigma_y(i, j, k+1)\sigma_t(i, j, k) \\ &+ 1, j, k+1)\tau_t(i+2, j, k)\tau_x(i+1, j, k) \\ \dot{i}, k)\sigma_x(i, j, k+1)\sigma_t(i+1, j, k)\sigma_x(i, j, k) \\ &+ j + 2, k)\tau_y(i, j+1, k+1)\tau_t(i, j+1, k) \end{split}$$

Monte Carlo Simulation

Finding the threshold probability for fault-tolerant surface code/toric code is used to express logical informations is equivalent to finding the maximum quenching probability above which there is no

- quantum computing with a depolarizing noise in which
- ferromagnetic phase in quenched $Z(2) \times Z(2)$ gauge theory



Monte Carlo Simulation (preliminary)

p = 0.00



p = 0.00



cf. R. Acharya et al (Google Quantum AI), "Suppressing quantum errors by scaling a surface code logical qubit", Nature 614 (2023) 676



Fig. 2 | Error detection in the surface code. a, Illustration of a surface code correction. **c**, Detection probability heatmap, averaging over t = 1 to 24. experiment, in perspective view with time progressing to the right. We begin d,e, Similar to b,c for four separate distance-3 experiments covering the four with an initial data qubit state that has known parities in one stabilizer basis quadrants of the distance-5 code. **f**,**g**, Similar to **b**,**c** using a simulation with Pauli (here, Z). We show example errors that manifest in detection pairs: a Z error errors plus leakage, crosstalk and stray interactions (Pauli+). h, Bar chart summarizing the detection correlation matrix p_{ii} , comparing the distance-5 (red) on a data qubit (spacelike pair), a measurement error (purple) on a measure qubit (timelike pair), an X error (blue) during the CZ gates (spacetimelike pair) experiment from **b** to the simulation in f (Pauli+) and a simpler simulation with and a measurement error (green) on a data qubit (detected in the final inferred only Pauli errors. We aggregate four groups of correlations: timelike pairs; spacelike pairs; spacetimelike pairs expected for Pauli noise; and spacetimelike Z parities). **b**, Detection probability for each stabilizer over a 25-cycle distance-5 experiment (50,000 repetitions). Darker lines: average over all stabilizers with pairs unexpected for Pauli noise (Unexp.), including correlations over two the same weight. There are fewer detections at timestep t = 0 because there is timesteps. Each bar shows a mean and standard deviation of correlations from a no preceding syndrome extraction, and at t = 25 because the final parities are 25-cycle, 50,000-repetition dataset. calculated from data qubit measurements directly. QEC, quantum error

cf. R. Acharya et al (Google Quantum AI), "Suppressing quantum errors by scaling a surfac code logical qubit", Nature 614 (2023) 676



Fig. 4 | Towards algorithmically relevant error rates. a, Estimated error budget for the surface code, based on component errors (see Fig. 1c) and Pauli+ simulations. $\Lambda_{3/5} = \varepsilon_3/\varepsilon_5$. CZ, contributions from CZ error (excluding leakage and stray interactions). CZ stray int., CZ error from unwanted interactions. DD, dynamical decoupling (data qubit idle error during measurement and reset). Measure, measurement and reset error. Leakage, leakage during CZs and due to heating. 1Q, single-qubit gate error. **b**, Logical error for repetition codes.

4. Discussions

- Model mapping
- model is quenched $Z(2) \ge Z(2)$ gauge theory in the 3-dimension
- is currently being performed

• "Threshold probability", the largest error rate below which fault-tolerant quantum computing is possible can be investigated through QEC - Statistical • We show that for the depolarizing noise and the measurement error from a realistic quantum circuits of QEC for the surface code, the mapped statistical

• To find the threshold probability of this model, Monte Carlo simulation of the chosen statistical model using parallel tempering and metropolis updates

Stay tuned!







Figure 1. Scaling of number field sieve (NFS) on classical computers and Shor's algorithm for factoring on a quantum computer, using BCDP modular exponentiation with various clock rates. Both horizontal and vertical axes are log scale. The horizontal axis is the size of the number being factored.

R. Van Meter et al., arXiv:quant-ph/0507023

Physical systems



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