

QuFEM: Fast and Accurate Quantum Readout Calibration Using the Finite Element Method

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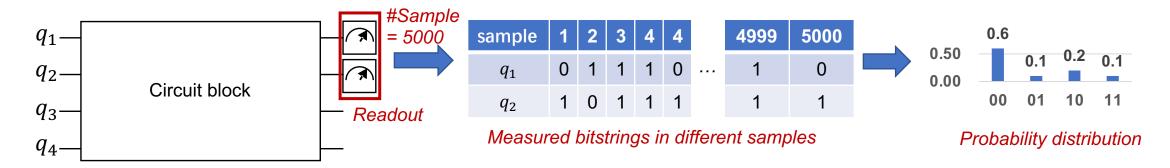
College of Computer Science and Technology, Zhejiang University

Background





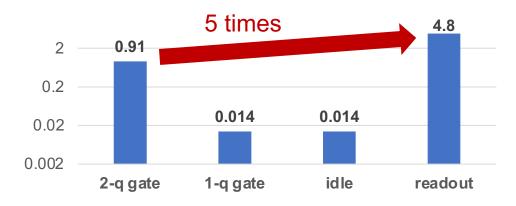
Quantum readout is an operation to read the information from quantum bits to classical bits.



Readout error is significant on current quantum hardware.



Noise on 127-qubit IBM Sherbrooke quantum device



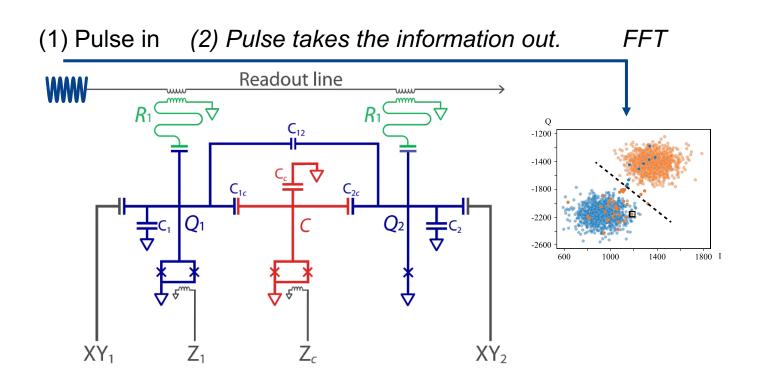
Noise on 10-qubit Tianmu quantum device

Background

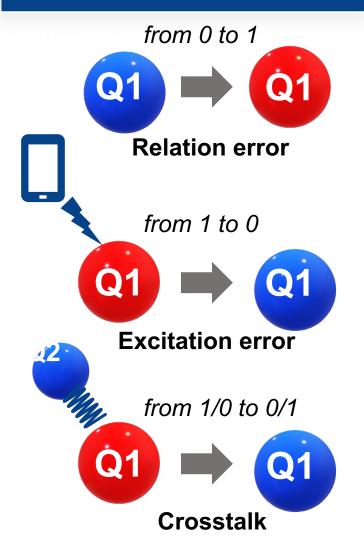




Implementation of readout on superconducting qubits



Source of readout error

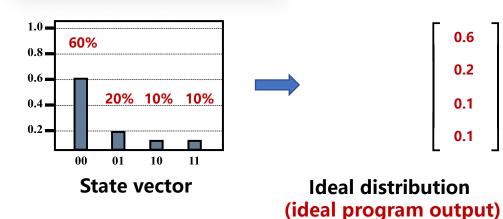


Background

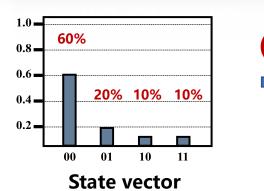




Ideal readout



Readout with noise

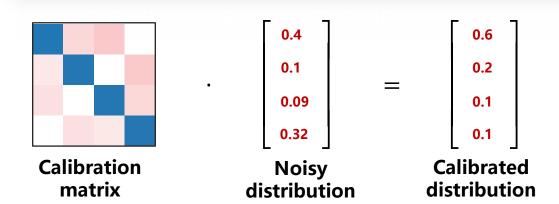


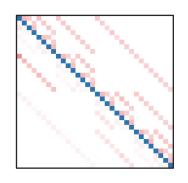


0.4 0.1 0.09 0.32

Noisy distribution (noisy program output)

Matrix-based readout error calibration





The size exponentially increases!

Calibration matrix of a 5-qubit readout 25 x 25

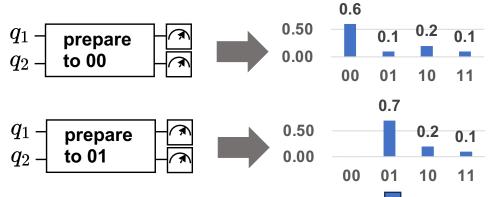
Basic Matrix-based readout calibration





Step 1. Matrix characterization

Prepares qubits to different basis states and apply measurement.



Fill in a noise matrix.

$$M = \begin{bmatrix} 0.6 & 0.1 & 0.2 & 0.1 \\ 0 & 0.7 & 0.5 & 0 \\ 0.2 & 0.1 & 0.1 & 0.8 \end{bmatrix}$$

Inverse the noise matrix

$$M^{-1} =$$

calibration matrix

Step 2. Calibration for any input

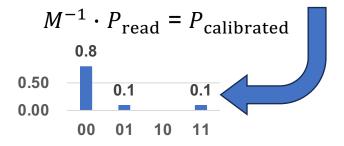
Represent the measured distribution as a vector.



数字算下

Apply matrix-vector multiplication.

$$\begin{bmatrix} 0.6 & 0.1 & 0.2 & 0.1 \\ 0 & 0.7 & 0.2 & 0.1 \\ 0.2 & 0.2 & 0.6 & 0 \\ 0 & 0.1 & 0.1 & 0.8 \end{bmatrix}^{-1} \cdot \begin{bmatrix} 0.6 \\ 0.1 \\ 0.2 \\ 0.1 \end{bmatrix} = \begin{bmatrix} 0.8 \\ 0.1 \\ 0 \\ 0.1 \end{bmatrix}$$



Basic Matrix-based readout calibration





Step 1. Matrix characterization

Prepares qubits to different basis states and apply measurement.

 2^N circuits are executed to measure qubits on all basis states.

Fill in a noise matrix.

The size of the noise matrix is $2^N \times 2^N$.

Inverse the noise matrix

Calcaute the inverse has $O(4^N)$ complexity.

Step 2. Calibration for any input

Represent the measured distribution as a vector.

Transformation has linear complexity.

Apply matrix-vector multiplication.

Multiplication has $\mathcal{O}(4^N)$ complexity.

8.8 TB and 2.3 min for a 18-qubit calibration

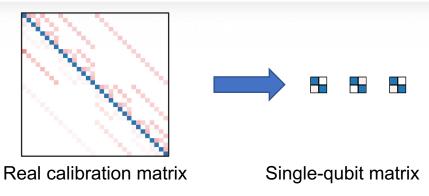


Limitations of Current Methods

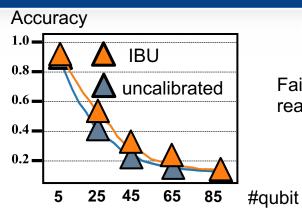




IBU (Google Science 2021) Realizing topologically ordered states on a quantum processor.



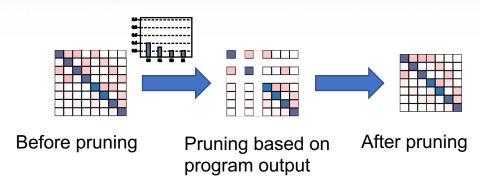
use tensor product of a series of single-qubit metamatrices



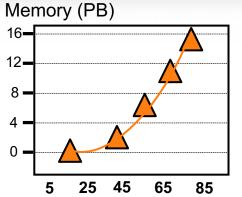
Fail to calibrate on 80-qubit readout output

Fast but not accurate: ignore the qubit interactions.

M3 (IBM PRA 2021): Scalable mitigation of measurement errors on quantum computers



use a sparsity-aware method prune the matrix under a threshold of Hamming distance



Require 16PB to calibrate a 85-qubit result.

(4 times the Fugaku supercomputer)

qubit

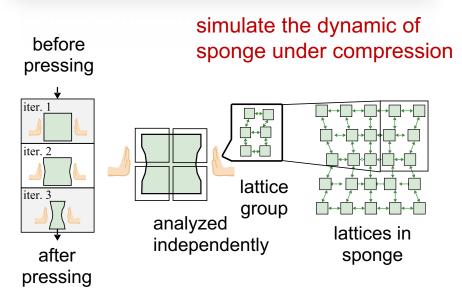
Accurate but not fast: many matrix elements cannot be ignored

Calibration based on Finite Element method





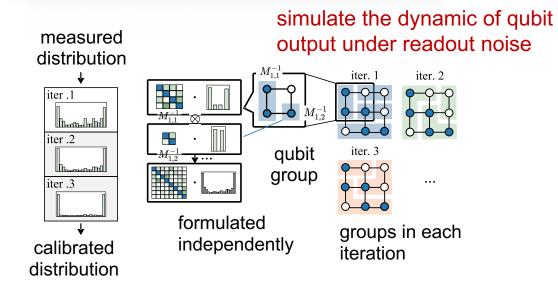
Classical Finite Element Method



- 1 partitions the sponge into lattices
- ② analyzes the state of each lattice independently
- 3 simulate the interaction
- 4 update the state of sponge



Quantum Finite Element Method



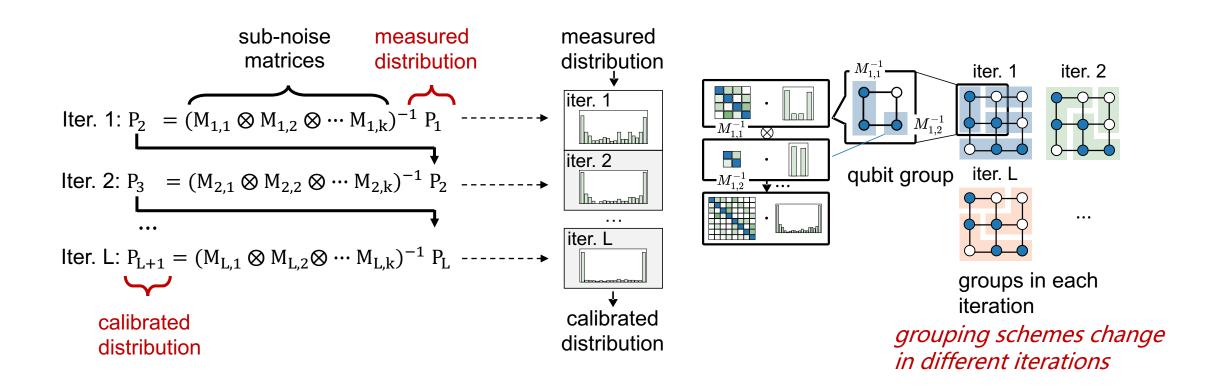
- ① partitions qubits into groups
- 2 analyze the noise in each group independently
- 3 simulate the interaction
- 4 update the calibration result of qubits

A divide-and-conquer strategy to calibrate measured distribution

Calibration formulation



QuFEM reformulates the calibration as an iterative process with a series of sub-noise matrices.



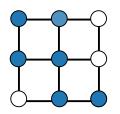
An example



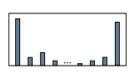


Input:

measured qubits



measured distribution



换成6比特 的分布

直接写成 拆开的形 式

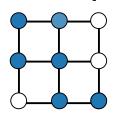
An example





Input:

measured qubits

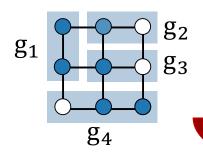


measured distribution

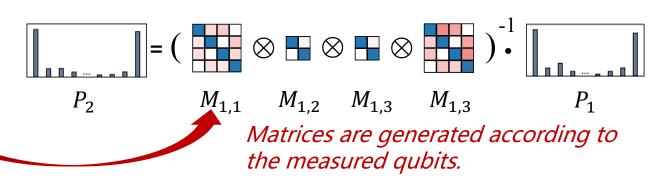


Iteration 1:

grouping scheme



computation formulation



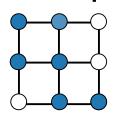
An example



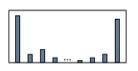


Input:

measured qubits

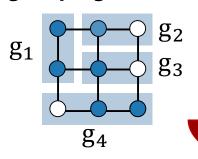


measured distribution

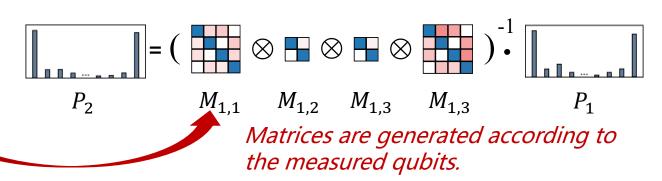


Iteration 1:

grouping scheme

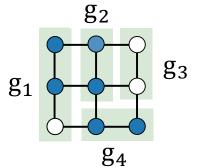


computation formulation

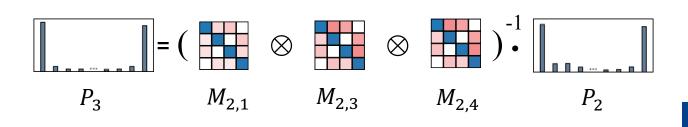


Iteration 2:

grouping scheme



computation formulation







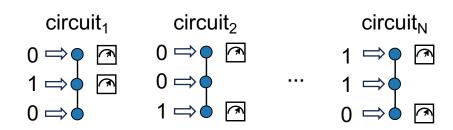
Characterize readout error in one iteration

Goals:

- 1. Collect the readout data under different qubit states.
- 2. Determine the grouping scheme in each iteration.

1. Data collection

Run benchmarking circuits.



Prepare qubits to different basis states and measure a subset of them.

2. Determine the grouping scheme

Possible states of a qubit in a benchmarking circuit:

- 1: qubit is set 0 and measured
- 2: qubit is set 1 and measured
- 3: qubit is set 0 or 1 and not measured

Characterize the **readout error** of a qubit under different states:

$$P(q_i.error = 1 | q_i.state = x, q_j.state = y)$$
 q_i is error q_j is set x q_j is set y

$$y \in \{1,2,3\}$$
, $x \in \{1,2,3\}$



要不直接简写,然后 说key insight





Characterize readout error in one iteration

2. Determine the grouping scheme

Possible states of a qubit in a benchmarking circuit:

- 0: qubit is set 0 and measured
- 1: qubit is set 1 and measured
- 2: qubit is set 0 or 1 and not measured

Characterize the **readout error** of a qubit under different states:

$$P(q_i.error = 1 \mid q_i.state = x, q_j.state = y)$$

$$q_i \text{ is error} \qquad q_j \text{ is set } x \qquad q_j \text{ is set } y$$

$$y \in \{0,1,2\} \ , x \in \{0,1\}$$

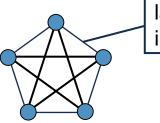
给个例子

Characterize the **interaction** from one qubit to another qubit under different states:

interact(q_i. state =
$$x \rightarrow q_j$$
. state = x)
$$= P(q_j. error = 1 | C1, C2) - P(q_j. error = 1 | C2)$$
error rate of q_j under C1, C2 average error rate of qz_j

$$C1: q_i. state = x, \qquad C2: q_j. state = y$$

Construct a weighted qubit graph:



labeled by the sum of interactions under all states





Characterizing readout error in one iteration

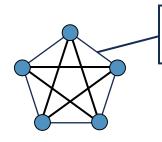
2. Determine the grouping scheme

Characterize the **interaction** from one qubit to another qubit under different states:

interact(q_i. state =
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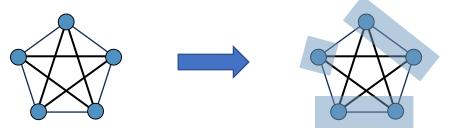
$$C1: q_i. state = x, \qquad C2: q_i. state = y$$

Construct a weighted qubit graph:



labeled by the sum of interactions under all states

Partitions with a **MAX-CUT solver**:

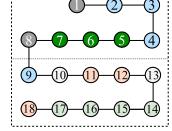


An Example

Prior knowledge of hardware helps grouping

Readout resonator 1

Readout resonator 2



18-qubit topology

group 1: (4) (5) (6) (7)
same readout resonator
group 2: (2) (3) (4) (9)
similar readout frequency

group 3: ① ⑧

overlapping frequency shift region





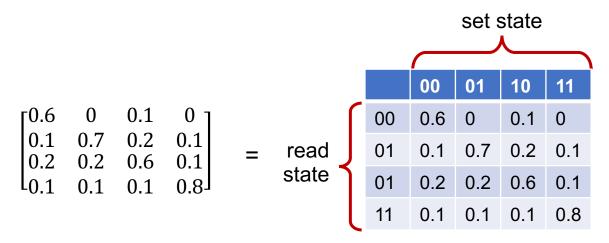
Calibration in one iteration

Perform matrix-vector multiplication

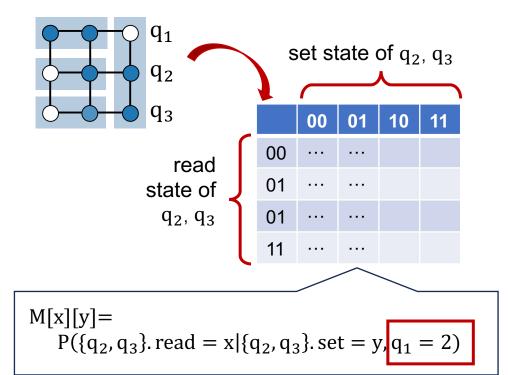
Iter. i:
$$P_{i+1} = (M_{i,1} \otimes M_{i,2} \otimes \cdots M_{i,k})^{-1} P_i$$

Matrix generation

Noise matrix formulates the transformation probability from the ideal state to measured state.



Sub-noise matrices of QuFEM formulates the transformation probability of states inside the qubit groups.



Transformation probability when q_1 is not measured

Put all together

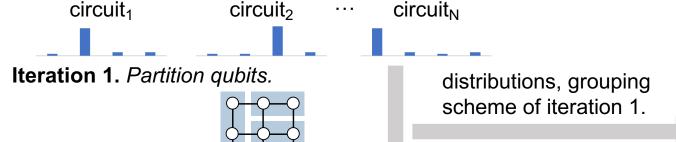




Characterization

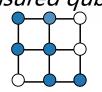
Iteration 1. Calibrate.

Iteration 1. Run benchmarking circuits.



Calibration

Input. measured qubits measured distribution





Iteration 1. Generate sub-noise matrices.









Iteration 1. Calibrate.





distributions, grouping scheme of iteration 2.

Iteration 2. Generate sub-noise matrices.

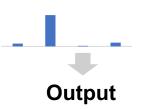






Iteration 2. Calibrate.

换一下分布数 据,逐渐校准

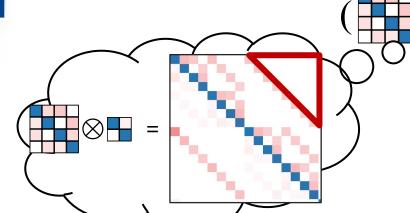


Sparse Tensor-Product Engine





Observation



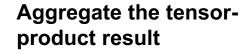
 \otimes \blacksquare \otimes \blacksquare \otimes

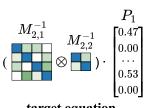
A large number of sparse intermediate vectors is generated in the tensor-product.

Implementation

Use a key-value table to store sparse vector

Calculate the tensor product





P_1	
47	
.00	
• •	
53	
വ	

target equation

	prob.	1		value
(000)	0.47	→	00	0.50
$_{1}(011)$	0.53		01	-0.02
			10	0.01
			11	0.01

	value	2	
00	0.50	\otimes	0
01	-0.02		1
10	0.01		
11	0.01		

•		vaine	
\otimes	0	0.99	• 0.47 =
	1	0.01	
			1 1 4 0

value

 $\Im |value| < \beta$

		value	4)	X	prob.
:	000	0.49	 → >	〕→	$P_2(000)$	0.48
	001	0.01	1	.	$P_2(001)$	6×10 ⁻³
	010	-0.01			$P_2(010)$	6×10 ⁻³
,	100	0.01			$P_2(011)$	0.50
	101	10-4-			$P_2(111)$	6×10 ⁻³
	110	10-4				

Prune values < threshold (10⁻⁵)

For each basis states

- ① calculate the matrix-vector multiplication
- ② calculate the tensor-product
- (3) prune intermediate values
- (4) sum intermediate values to obtain output.

Compute the tensor-product of other basis states

Experiment



Setup

Platform	#Qubits	1-q fidelity	2-q fidelity	Instructions
0	136	94.6±3.1%	94.6±3.0%	ID,RX,RY,RZ,H,CX
Quafu	18	95.9±1.3%	95.9±1.3%	ID,RX,RY,RZ,H,CX
Rigetti	79	99.5±1.1%	90.0±6.4%	CPHASE,XY
Self-developed	36	99.9±0.1%	98.7±0.8%	U3,CZ
IBMQ	7	99.9±0.1%	99.2±0.1%	CX,ID,RZ,SX,X

Evaluated hardware

IBU: KJ Satzinger, et al. Realizing topologically ordered states on a quantum processor. Science 2021

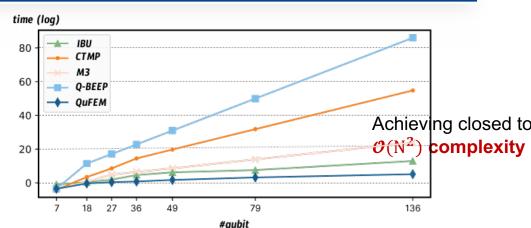
CTMP: Sergey, et al. Mitigating measurement errors in multiqubit experiments. PRA 2021.

M3: Paul D Nation, et al. Scalable mitigation of measurement errors on quantum computers. PRX Quantum 2021.

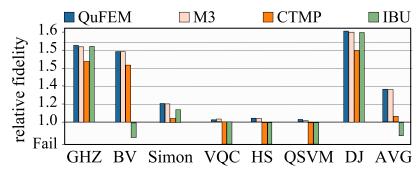
Q-BEEP: Nathan Wiebe, et al. Qbeep: Quantum Bayesian error mitigation employing Poisson modeling over the hamming spectrum. ISCA 2023.

Baselines

Result



QuFEM reduces the calibration time of the 136-qubit program output from 119.44 hours (IBU) to 169.65 seconds (119.44 \times reduction).



QuFEM shows an average improvement in relative fidelity of 1.003×, 1.2×, and 1.4× compared to M3, CTMP, and IBU, respectively.



Thanks for listening

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