## Solving Systems of Linear Equations with a Superconducting Quantum Processor

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Superconducting quantum circuits are a promising candidate for building scalable quantum computers. Here, we use a four-qubit superconducting quantum processor to solve a two-dimensional system of linear equations based on a quantum algorithm proposed by Harrow, Hassidim, and Lloyd [Phys. Rev. Lett. 103, 150502 (2009)], which promises an exponential speedup over classical algorithms under certain circumstances. We benchmark the solver with quantum inputs and outputs, and characterize it by nontrace-preserving quantum process tomography, which yields a process fidelity of  $0.837 \pm 0.006$ . Our results highlight the potential of superconducting quantum circuits for applications in solving large-scale linear systems, a ubiquitous task in science and engineering.

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Linear systems lie at the heart of many areas of science and engineering. To solve a system of linear equations with N variables, the best known classical algorithm requires a time of O(N). Harrow, Hassidim, and Lloyd (HHL) [1] showed that, in principle, quantum computers can solve linear systems exponentially faster by calculating the expectation value of an operator associated with the solution, which may lead to many practical applications of quantum computation other than those previously known [2–4]. For an s-sparse system matrix of size  $N \times N$  and condition number  $\kappa$ , the HHL algorithm can reach a desired computational accuracy  $\epsilon$  within a running time of  $O(\log(N)s^2\kappa^2/\epsilon)$ under certain circumstances [5], comparing to the best known classical algorithm of  $O(Ns\kappa/\log(\epsilon))$ . Such an exponential speedup promises widespread applications that address large-scale systems. Indeed, several applications based on the HHL algorithm, such as data processing [6], numerical calculation [7], and artificial intelligence [8,9], have been proposed in recent years.

A compiled version of the HHL algorithm was previously only demonstrated with parametric down-converted single photons [10,11] and liquid nuclear magnetic resonance [12], both of which are considered not easily scalable to a large number of qubits. For example, the optical demonstration was limited by the probabilistic photon generation and twophoton gate operation. For a deterministic and more scalable implementation, here, we turn to a solid-state system, i.e., a superconducting quantum circuit in this experiment, which has attracted significant attention due to a number of merits, including the much-improved coherence [13–15], the excellent scalability [16–18], and the remarkable high fidelity and fast control [19-21]. In addition, compiled versions of various quantum algorithms, such as the Deutsch-Jozsa algorithm [22,23], the Grover's algorithm [22], and the Shor's algorithm [24], have been successfully tested on this solid-state platform on a small scale.

In this Letter, we demonstrate a nontrivial instance of a quantum linear solver, based on the HHL algorithm for a  $2 \times 2$  system, with a superconducting circuit consisting of four transmon qubits of the Xmon variety [25]. We test the solver with 18 distinct quantum-state inputs that uniformly distribute on the Bloch sphere, from which the nontracepreserving quantum process tomography (QPT) can be reliably determined [26]. For various quantum inputs, our quantum solver can return the desired solutions with reasonably high precision, yielding an averaged QPT fidelity of  $0.837 \pm 0.006$ . As such, our experiment represents the first demonstration of the quantum algorithm for solving systems of linear equations on a solid-state platform.

The device was fabricated on a sapphire substrate in three steps following the procedure outlined previously [13]: (1) deposit the aluminum film on the degassed substrate; (2) define circuit wirings using wet etch; (3) double-angle evaporate the  $Al/AlO_x/Al$  Josephson junctions. Figure 1 shows the optical micrograph of the device, with the four Xmon qubits labeled from  $Q_1$  to  $Q_4$ . Each qubit has its own frequency-control Z line for

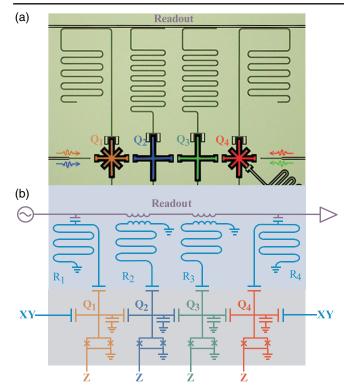


FIG. 1. (a) False color photomicrograph and (b) simplified circuit schematic of the superconducting quantum circuit for solving  $2 \times 2$  linear equations. Shown are the four Xmon qubits, marked from  $Q_1$  to  $Q_4$ , and their corresponding readout resonators, marked from  $R_1$  to  $R_4$ .

rotations of the qubit state around the Z axis on the Bloch sphere.  $Q_1$  and  $Q_2$  ( $Q_3$  and  $Q_4$ ) share the microwave XY line on the left (right) that is closer to  $Q_1$  ( $Q_4$ ) for singlequbit rotations around X and Y axes. The microwave pulses transmitted through each XY line have two-frequency components, and the drive strength to  $Q_1$  ( $Q_4$ ) is more than that to  $Q_2$  ( $Q_3$ ) by a factor of 11.7 (6.7) as experimentally calibrated. Each qubit dispersively couples to its own readout resonator, and all readout resonators couple to a common transmission line, enabling simultaneous single-shot quantum nondemolition measurement on multiple qubits using frequency-domain multiplexing. The signal-to-noise ratio is further improved by a quantumlimited parametric amplifier, similar to that described previously [27].

The circuit Hamiltonian under the rotating wave approximation is

$$H = -\sum_{j=1}^{4} \omega_j(t) \sigma_j^z / 2 + \sum_{j=1}^{3} g_{j,j+1} (\sigma_j^+ \sigma_{j+1}^- + \sigma_j^- \sigma_{j+1}^+), \quad (1)$$

where  $\omega_j(t)$  is the resonant frequency of the *j*th qubit that can be tuned over time,  $\sigma_j^z$  is the Pauli operator,  $\sigma_j^{\pm}$  are the raising and lowering operators, and  $g_{j,j+1}$  is the

nearest-neighbor coupling strength. The coupling strengths of  $g_{j,j+1}/2\pi$  for j = 1 to 3 are measured to be around 13.0, 9.8, and 14.1 MHz, respectively. These Xmon qubits typically have a maximum frequency around 5.1 GHz and an anharmonicity around 250 MHz. In this experiment, qubit idle frequencies  $\omega_j(t=0)/2\pi$  for j=1 to 4 are arranged in a zigzag pattern at 5.073, 4.074, 4.948, and 4.547 GHz, respectively, which ensures that the nearest-neighbor and next-nearest-neighbor couplings along the qubit chain are effectively turned off when idling. At the above-listed frequencies, the qubit lifetimes  $T_1$ s are measured to be around 15.9, 7.4, 7.8, and 14.1  $\mu$ s, and the Gaussian dephasing times [28]  $T_2^*$ s are around 8.7, 2.3, 5.2, and 3.4  $\mu$ s, respectively (see [29] for more information on the device).

The HHL algorithm aims to solve a system of linear equations  $A\vec{x} = \vec{b}$  for  $\vec{x}$ , given the  $N \times N$  Hermitian matrix A and the input vector  $\vec{b}$ . The process involves three subsets of qubits: a single ancilla qubit, a register of k qubits used to store the eigenvalues of A to a binary precision of k bits, and a memory of  $O(\log(N))$  qubits used to load  $\vec{b}$  and also store the output  $\vec{x}$ . For simplicity, we assume that  $\vec{b}$  is a unit vector, whose entries  $\{b_i\}$  can be encoded in the memory formatted to a quantum state  $|b\rangle = \sum b_i |i\rangle$ , where  $|i\rangle$ denotes the computational basis of the  $O(\log(N))$  qubits. Next is the core of the HHL algorithm responsible for the exponential speedup: with carefully designed quantum logic gates, including mapping the Hermitian matrix A to the system Hamiltonian, the quantum state  $|x\rangle$  representing the desired solution  $\vec{x}$  can be synthesized in the memory conditional upon the state of the ancilla qubit. Afterward, one can either map the quantum state  $|x\rangle$  to recover all entries of the vector  $\vec{x}$ , or, more efficiently, perform the quantum measurement corresponding to an operator M that one is interested in to extract its expectation value  $\langle x|M|x\rangle$ .

With the system initialized in the state  $|0\rangle_a|0\rangle_r|b\rangle_m$ , where the subscripts *a*, *r*, and *m* index, respectively, the subsets of qubits in the ancilla, the register, and the memory (here and below, we keep the subscripts in wave functions only when the states of two or three subsets are quoted simultaneously), a general description of the HHL core is as follows: (1) with quantum phase estimation [1,32], using the controlled unitary transformations in the form  $e^{-iAt}$  for a variable time *t*, decompose  $|b\rangle$  in the eigenbasis of *A*, i.e.,  $|b\rangle = \sum_j \beta_j |u_j\rangle$ , and map the corresponding eigenvalues  $\lambda_j$ into the register in a binary form to transform the system to  $\sum_j \beta_j |0\rangle_a |\lambda_j\rangle_r |u_j\rangle_m$ ; (2) perform controlled rotation  $R(\lambda^{-1})$ on the ancilla, according to  $\lambda_j$  stored in the register, which transforms the system to

$$\sum_{j} \beta_{j} \left( \sqrt{1 - \frac{C^{2}}{\lambda_{j}^{2}}} |0\rangle_{a} + \frac{C}{\lambda_{j}} |1\rangle_{a} \right) |\lambda_{j}\rangle_{r} |u_{j}\rangle_{m}, \qquad (2)$$

where  $C (\leq 1)$  is a constant that can be selected as any real number to make the controlled rotation physical [1]; (3) reverse the procedure in (1) to disentangle and clear the register, and the system state evolves to

$$\sum_{j} \beta_{j} \left( \sqrt{1 - \frac{C^{2}}{\lambda_{j}^{2}}} |0\rangle_{a} + \frac{C}{\lambda_{j}} |1\rangle_{a} \right) |0\rangle_{r} |u_{j}\rangle_{m}.$$
(3)

A postselection of the  $|1\rangle$ -state outcome of the ancilla will yield the desired output in the memory  $|x\rangle \sim \sum_{j} C(\beta_{j}/\lambda_{j})|u_{j}\rangle$ , with a success probability of  $\sum_{i} (C\beta_{i}/\lambda_{i})^{2}$ .

As argued elsewhere [5], conversions between the classical vectors and their quantum counterparts, i.e.,  $\vec{b} \leftrightarrow |b\rangle$  and  $\vec{x} \leftrightarrow |x\rangle$ , take extra time and may eventually kill the exponential speedup gained during execution of the HHL core. Nevertheless, the HHL algorithm provides a general template and represents a real advance in the theory of quantum algorithms. Our immediate goal in this experiment is to implement a *purely* quantum version of the HHL algorithm; i.e., we aim to test the above-mentioned HHL core with quantum inputs and outputs.

In our demonstration, the four-qubit solver is set to run a nontrivial instance, where the system matrix *A* is chosen as  $A = \begin{pmatrix} 1.5 & 0.5 \\ 0.5 & 1.5 \end{pmatrix}$ . The two quantum states corresponding to the eigenvectors of *A* are  $|u_1\rangle = (|0\rangle - |1\rangle)/\sqrt{2}$  and  $|u_2\rangle = (|0\rangle + |1\rangle)/\sqrt{2}$ , with eigenvalues of  $\lambda_1 = 1$  and

 $\lambda_2 = 2$ , respectively. Accordingly, our four qubits are distributed into three subsets: the ancilla  $(Q_4)$ , the register  $(Q_2Q_3)$ , and the memory  $(Q_1)$ . The binary representations of " $Q_2Q_3$ ", "01" and "10", record eigenvalues  $\lambda_1$  and  $\lambda_2$ , respectively. The input  $|b\rangle$  is prepared in  $Q_1$  by single-qubit rotations.

The prechosen  $2 \times 2$  Hermitian matrix A allows us to optimize the circuit that consists of three subroutines as shown in Fig. 2(a), where all two-qubit gates fit to our system Hamiltonian [Eq. (1)]. Subroutine 1: the phase estimation subroutine is precompiled with a controlledphase (cz) gate, two  $\sqrt{i}$ SWAP gates, and two single-qubit gates acting on the memory  $(Q_1)$  and the register  $(Q_2Q_3)$ , which can be described as follows: first, prepare  $|b\rangle =$  $\beta_1 |u_1\rangle + \beta_2 |u_2\rangle$  in  $Q_1$ , following which, the Hadamard gate transforms  $Q_1$  to  $|b\rangle = \beta_1 |1\rangle + \beta_2 |0\rangle$ ; meanwhile a  $\pi$ rotation on  $Q_3$  yields  $|01\rangle$  in the  $|Q_2Q_3\rangle$  register; next, the sandwiched  $\sqrt{iSWAP}$ -CZ- $\sqrt{iSWAP}$  gate combination fulfills a controlled-iSWAP gate, which swaps the states between  $Q_2$  and  $Q_3$  up to a phase factor of -i only if  $Q_1$  is in  $|0\rangle$ . At the end of this subroutine, the state of  $|Q_2Q_3\rangle_r |Q_1\rangle_m$  goes to  $\beta_1|01\rangle_r |1\rangle_m - i\beta_2|10\rangle_r |0\rangle_m$ , which correlates the binary representations of the eigenvalues in the register (ignoring the phase factor as we omit up to a single-qubit phase gate here) with the eigenstates in the memory since  $|1\rangle$  and  $|0\rangle$  correspond to  $|u_1\rangle$  and  $|u_2\rangle$ , respectively, up to a Hadamard gate. Subroutine 2: the controlled rotation  $R(\lambda^{-1})$  subroutine acts on the ancilla  $(Q_4)$ , depending on the state of the register  $|Q_2Q_3\rangle$ , whose

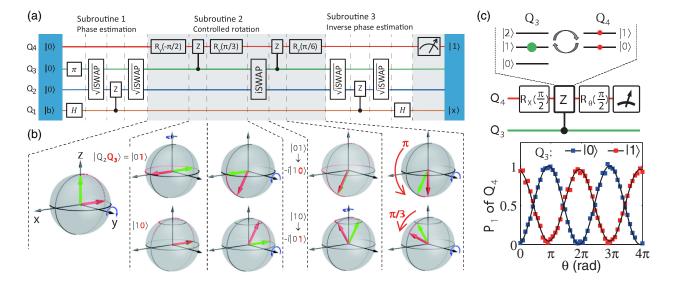


FIG. 2. (a) Compiled quantum circuits for solving  $2 \times 2$  linear equations with four qubits. There are three subroutines and more than 15 gates as indicated. (b) The Bloch-sphere illustration of the controlled rotation subroutine for C = 1, where the two rotation angles of  $\pi$  and  $\pi/3$  are achieved for eigenvalues  $\lambda_1 = 1$  ( $|01\rangle$ ) and  $\lambda_2 = 2$  ( $|10\rangle$ ), respectively. (c) The  $Q_3-Q_4$  cz gate. Top: qubit energy level arrangement showing that the  $|0\rangle \leftrightarrow |1\rangle$  transition of the target  $Q_4$  is on resonance with the  $|1\rangle \leftrightarrow |2\rangle$  transition of the control  $Q_3$ . Only when  $Q_3$  is in  $|1\rangle$ , the state in  $Q_4$  will make a full cycle and gain an additional phase of  $\pi$ . Middle: the quantum circuits for calibrating the cz gate sandwiched in between two  $\pi/2$  rotations, where the second  $\pi/2$  rotation axis has an angle  $\theta$  from the X axis in the XY plane of the Bloch sphere. Bottom: the calibrated Ramsey interference curves of  $Q_4$  when  $Q_3$  is in  $|0\rangle$  (blue) and  $|1\rangle$  (red), which differ by a phase of  $\pi$ . No single-qubit phase gates were used during this measurement to cancel the dynamical phase due to the change of qubit frequency.

effect can be best visualized using the Bloch sphere representation as illustrated Fig. 2(b). Here, we choose C = 1, and according to Eq. (2), the rotation angles around the Y axis for  $Q_4$  should be  $\pi$  for  $\lambda_1 = 1$  and  $\pi/3$  for  $\lambda_2 = 2$ . Subroutine 3: this one is exactly the reverse of Subroutine 1, in which the register  $|Q_2Q_3\rangle$  is disentangled and cleared, and  $|1\rangle$  and  $|0\rangle$  are transformed back to the eigenstates  $|u_1\rangle$ and  $|u_2\rangle$ , respectively, with the final Hadamard gate on  $Q_1$ .

The compiled HHL circuits in Fig. 2(a) consist of more than 15 one- and two-qubit gates, excluding phase compensation and tomography gates that are not shown. Execution of these gates with reasonably high precisions are therefore important, though phase errors may not be critical at certain steps, e.g., the process of mapping eigenvalues to the register. Our single-qubit rotations have been calibrated by randomized benchmarking to be of reasonably high fidelity as limited by qubit coherence [33].  $Q_3$ 's  $\pi$  gate is 300 ns in length, and qubit coherence would limit the gate fidelity to be just under 0.98. For the  $Q_2-Q_3$  $\sqrt{iSWAP}$  and iSWAP gates, we tune the qubits on resonance for periods of  $\pi/(4g_{2,3})$  and  $\pi/(2g_{2,3})$ , respectively, similar to those demonstrated using the qubit-resonator architecture [34]. However, here, due to the crosstalk issue, which likely results from insufficient crossover wirings to tie the grounds together [35], our  $\sqrt{iSWAP}$  and iSWAP gates show slightly lower performances, with fidelity values estimated to be slightly above 0.98 and 0.97, respectively. For the cz gate, we include the qubit's second excited state  $|2\rangle$  and implement similarly as that demonstrated using the qubitresonator architecture [36]. The fidelity of our CZ gate is estimated to be about 0.95, using the calibrated Ramsey interference data shown in Fig. 2(c). We note that the CZ gate can also be implemented, with high fidelity and without exchange of excitations, by optimizing one qubit's frequency trajectory to mediate the two-qubit  $|11\rangle$  state close to the avoided level crossing with the  $|02\rangle$  state [19], which would be hard to be implemented in our device due to the crosstalk issue.

We prepare 18 different input states  $|b_i\rangle$  for j = 1 to 18 on the Bloch sphere as shown in Fig. 3(a). With the output state stored in  $Q_1$ , we can measure the expectation value of a certain operator that we are interested in [see, e.g., the case of  $|b_1\rangle = |0\rangle$  in Fig. 3(b)]. We can also characterize the input and output states of the above instance by quantum state tomography (QST): qubit polarization along the Z axis of the Bloch sphere can be directly measured; for polarizations along the X and Y axes, the qubit is rotated around the Y and X axes, respectively, before measurement. The corresponding output  $|x_i\rangle$  in  $Q_1$  and the state of  $Q_4$  are simultaneously measured [29], which allows us to reconstruct  $|x_i\rangle$  using only the data corresponding to the  $|1\rangle$ -state outcome of  $Q_4$ . As shown in Fig. 3(c), fidelity values of the 18 output states by QST are reasonably high, ranging from  $0.840 \pm 0.006$  to  $0.923 \pm 0.008$ .

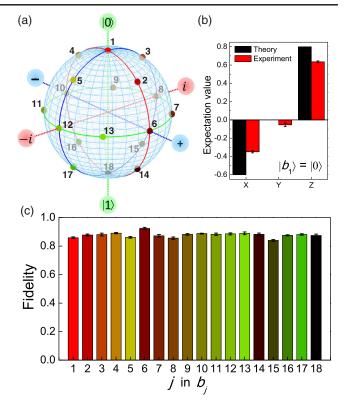


FIG. 3. (a) 18 input quantum states indexed by the number *j* on the Bloch sphere that are used to test the quantum linear solver. (b) Expectation values of the three operators, where  $\{X, Y, Z\}$  are the Pauli operators  $\{\sigma_x, \sigma_y, \sigma_z\}$  for the output state  $|x\rangle$  when the input state  $|b_1\rangle = |0\rangle$ . (c) Fidelity values of the output states corresponding to the 18 input states  $|b_j\rangle$  that are labeled on the Bloch sphere in (a). Statistical errors are shown by error bars, defined as  $\pm 1$  s.d., using the repeated sets of the QST measurement.

With all the input and output states being measured, we are able to characterize the solver more exactly by QPT, where a nontrace-preserving process matrix in the Pauli basis is used to linearly map  $|b_j\rangle$  to  $|x_j\rangle$ . The experimental  $\chi_{exp}$  matrix, inferred reliably only with all 18 input states

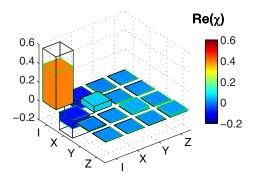


FIG. 4. Nontrace-preserving QPT characterizing the quantum linear solver. Shown are the real parts of the experimental  $\chi_{exp}$  matrix (bars with color) and the ideal  $\chi_{id}$  matrix (black frames), where *I* is the identity and  $\{X, Y, Z\}$  are the Pauli operators  $\{\sigma_x, \sigma_y, \sigma_z\}$ . All imaginary components (data not shown) of  $\chi_{exp}$  are measured to be no higher than 0.015 in magnitude.

being used, and the ideal matrix  $\chi_{id}$  are shown together in Fig. 4 for comparison, which yields a process fidelity [37] of  $F = \text{Tr}(\chi_{id}\chi_{exp})/[\text{Tr}(\chi_{id})\text{Tr}(\chi_{exp})] = 0.837 \pm 0.006$ . Here, the error bar is estimated using simulated random normal distributed noise associated with the QST data [38].  $\chi_{exp}$  has a trace of 0.497, which indicates the averaged success probability of our solver for the 18 input states, while ideally, the trace is 0.625 for C = 1 [29].

In summary, we have demonstrated a superconducting quantum linear solver for a  $2 \times 2$  system with output state fidelities ranging from  $0.840 \pm 0.006$  to  $0.923 \pm 0.008$  and quantum process fidelity of  $0.837 \pm 0.006$ . The achieved fidelities are limited by decoherence and by errors in our implementation of the two-qubit gates, the latter of which is related to the insufficient grounding that can be fixed by adding an extra lithography layer in the device fabrication [17,19]. With future improvements on superconducting quantum circuits could be used to implement more intricate quantum algorithms on a larger scale and ultimately reach quantum computational speedup.

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