Provable quantum computational advantage with the cyclic cluster state

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We propose two Bell-type nonlocal games that can be used to prove quantum computational advantage in an objective and hardware-agnostic manner. In these games, the circuit depth needed to prepare a cyclic cluster state and measure a subset of its Pauli stabilizers on a quantum computer is compared to that of classical Boolean circuits with the same gate connectivity. Using a circuit-based trapped-ion quantum computer, we prepare and measure a six-qubit cyclic cluster state with an overall fidelity of 60.6% and 66.4%, before and after correcting measurement-readout errors respectively. Our experimental results indicate that while this fidelity readily passes conventional (or depth-0) Bell bounds for local hidden-variable models, it is on the cusp of demonstrating quantum advantage against depth-1 classical circuits. Our games offer a practical and scalable set of quantitative benchmarks for quantum computers in the pre-fault-tolerant regime as the number of qubits available increases.

I. INTRODUCTION

There are many metrics to characterize the quality of noisy intermediate-scale quantum (NISQ) computers [1]. For example, qubit count, gate count, gate fidelities, and quantum volume [2] are common options. However, it is generally agreed upon that a more comprehensive picture is given by the device’s overall performance in executing a variety of computational tasks. To this end, we look toward computational tasks with two desirable properties in this work. First, they should have objective targets, beyond which one can prove that the NISQ computer has outperformed its classical counterpart. Note that the significance of demonstrating so-called quantum supremacy via sampling random quantum circuits [3, 4] or bosonic linear interferometers [5] relies on assumptions of the underlying problem’s computational hardness, making it a moving target based on current state-of-the-art classical hardware and algorithms. In contrast, we seek an unconditional demonstration of quantum computational advantage. Second, the computational task should be agnostic to the choice of hardware implementation, so that it allows a fair comparison of results among different architectures.

Natural examples with the first desideratum are Bell-type nonlocal games [6–8], since the violation of a Bell inequality sets an objective threshold the quantum computer must pass. Experimental demonstrations of Bell inequality violations have a long history (see Ref. [8, Sec. VII]). More recently, the nonlocal features of quantum mechanics implied by such a violation have been experimentally confirmed [9–12]. Nonlocal games have also been of recent interest for demonstrating classically verifiable quantum advantage in interactive proof systems [13–15]. The second desideratum is satisfied in cases where the violation is achievable via the preparation and measurement of multipartite entangled states with geometrically restricted, constant-depth formation circuits. A canonical example of such states are graph states [16, 17], which have a long history of study with respect to Bell nonlocality [18, 19], measurement-based quantum computation [20], and connections therein [21, 22]. Experimentally, nonlocal features of graph states have also been explored on a variety of hardware platforms [23–26]. It has also been shown that Bell-type scenarios utilizing Pauli measurements on graph states go beyond the traditional Bell’s theorem [27], as they are able to reject even classical theories assisted by a limited amount of communication, which are a particular class of nonlocal hidden-variable models. Much work has since gone into understanding how Bell’s theorem can be strengthened to more general causal scenarios [28–30]. By treating these communication-assisted classical strategies as classical circuits of limited depth, one can show that shallow-depth quantum circuits are more powerful than their classical counterparts [31]. In light of these considerations, the past few years have seen a number of novel works laying the theoretical foundation for unconditionally demonstrating a quantum advantage using only nearest-neighbor entangling gates [32–38]. This excludes the use of n-qubit Greenberger–Horne–Zeilinger (GHZ) states, which can be prepared in constant depth with an architecture that allows all-to-all connectivity of its entangling gates, but requires linear depth for architectures with only nearest-neighbor connectivity.

In this paper, we analyze two kinds of Bell-type nonlocal games and demonstrate proof-of-principle experimental implementations of them on a trapped-ion quantum computer [39]. Both games utilize the six-qubit cyclic
TABLE I. Summary of the nonlocal games studied in this work, along with our main experimental results using a trapped-ion quantum computer. The notation and definition of the games are provided in their respective sections. For the optimal quantum strategy, each game involves preparing the six-qubit cyclic cluster state $|C_6\rangle$ and then evaluating a number of stabilizers from global Pauli measurement settings, which may not necessarily be the same. As the success probability $\Pr_C[\text{win}]$ of a classical strategy may depend on the amount of communication allowed between parties, we report bounds for both depth-0 and depth-1 classical circuits. For the details of our quantum experiments, see Sec. V; the cubic Boolean function (CBF) results were obtained by a tomography experiment measuring the relevant stabilizers (data presented in Fig. 3). The stabilizer submeasurement (SS) results were obtained by a separate experiment wherein we faithfully played the nonlocal games (data presented in Table II). We report experimental success probabilities $\Pr_Q[\text{win}]$ estimated from both the raw output of our quantum device, and after state-preparation-and-measurement (SPAM) error correction. The experimental uncertainties, denoted by the values in parentheses, correspond to a 1σ standard error within the number of significant figures reported. For each input of a game, we took $N = 5000$ shots.

| Game                          | Number of (stabilizers, settings) | $\Pr_C[\text{win}]$ | $\Pr_Q[\text{win}]$  |
|-------------------------------|-----------------------------------|---------------------|-----------------------|
| CBF($C_6$, $(0, 1)^8$) (Sec. III A) | (63, 63)                          | Depth-0 bound = 23/32 = 71.875% | Depth-1 bound = 100% | Raw value = 80.30(8)% | 83.21(9)% |
| CBF($C_6$, $T_{\text{Mermin}}^{\text{II}}$) (Sec. III B) | (55, 55)                          | 37/55 ≈ 67.3%       | 100%                  | 79.51(9)% | 82.56(10)% |
| SS($C_6$, $T_{\text{HLF}}^6$) (Sec. IV A) | (5, 8)                            | 7/8 = 87.5%         | 87.5%                 | 81.35(59)% | 85.79(53)% |
| SS($C_6$, $T_{\text{HLF}}^6$) (Sec. IV B) | (5, 5)                            | 4/5 = 80%           | 80%                   | 79.42(25)% | 84.68(23)% |

cluster state (an example of a graph state) and are motivated as follows. The first game can be viewed as the computation of a particular nonlinear Boolean function via the measurement of the six-qubit cyclic cluster state’s stabilizer group, giving a new computational interpretation to the graph-state Bell inequalities introduced in Ref. [19]. Furthermore, it doubles as a method to determine the fidelity of the prepared state and is classically the hardest such game based on any six-qubit graph state.

The second game is based on the smallest instance of the so-called 2D hidden linear function problem introduced in Ref. [31], which is capable of demonstrating an unconditional separation between the power of a constant-depth quantum circuit and logarithmic-depth classical circuits. While prior works [32–38] focus on such quantum computational advantage in an asymptotic regime, we discuss in detail how quantum-advantageous strategies for this six-qubit game already demonstrate an unconditional separation from shallow-depth classical circuits with the same gate connectivity used on our quantum computer. Furthermore, we show that restricting the set of inputs for this game makes it harder classically, reducing the target success probability required to demonstrate quantum advantage from 87.5% to 80%.

These nonlocal games satisfy our desired criteria, as the corresponding quantum strategies require only nearest-neighbor entangling gates on a ring, which are readily available on most quantum-computing architectures. Furthermore the games scale well, leading to meaningful results for surpassing the same target success probability as the qubit count increases. For our six-qubit demonstration, we summarize the games implemented and our experimental results in Table I.

The paper is organized as follows. In Sec. II we review background material necessary to understand our key results and establish notation. In Sec. III we introduce the first type of nonlocal game, called cubic Boolean function games. In Sec. IV we discuss the second type of nonlocal game, called stabilizer submeasurement games. In Sec. V we present the results of our experimental implementation of the quantum strategies for these games. In Sec. VI we discuss the generalization of these two games to the $n$-qubit scenario. Finally, we conclude with a discussion and outlook in Sec. VII.

II. BACKGROUND

A. Graph States

Stabilizer states are a particular class of many-body quantum states that have a wide range of applications in quantum computing (e.g., for error correction, measurement-based quantum computing, and tests of Bell nonlocality). An $n$-qubit stabilizer state is defined as the joint $+1$ eigenstate of $2^n$ commuting Pauli operators, called the stabilizer group. Let $X$, $Y$, and $Z$ be the Pauli matrices. $I$ denotes the $2 \times 2$ identity matrix and $\mathbb{1}$ the identity operator acting on the whole system. Let $|0\rangle$ and $|1\rangle$ be the $+1$ and $-1$ eigenstates of $Z$, respectively. Denote by $X_j$ an $n$-qubit Pauli operator that acts as $X$ on the $j$th qubit and as the identity elsewhere (similarly for $Y_j$ and $Z_j$). Any $n$-qubit Pauli operator can be written (up to an overall phase) as $E(a, b) = \prod_{j=0}^{n-1} a_j b_j X_j^{a_j} Z_j^{b_j}$, where $a, b \in \{0, 1\}^n$. Denote the $j$th term in the product for $E(a, b)$ by $E_j(a_j, b_j)$.

It is convenient to study stabilizer states in the so-called graph-state formalism [17]. For each graph $G = (V, E)$, with vertex set $V$ and edge set $E$, the corresponding graph state $|G\rangle$ is prepared by initializing a qubit at each vertex $v \in V$ in the state $|+\rangle = (|0\rangle + |1\rangle)/\sqrt{2}$ and applying a two-qubit controlled-$Z$ gate, $CZ_{v,w} = (|+\rangle + Z_v + Z_w - Z_v Z_w)/2$, between each pair of vertices $(v, w) \in E$ that share an edge. That is, the graph state
FIG. 1. (Top): The quantum circuit that prepares the six-qubit cyclic cluster state $|C_6\rangle$ using CZ gates, followed by the measurement of the Pauli operator $E(a, b)$ for some $a, b \in \{0, 1\}^6$. Namely, the final conditional rotation changes the measurement basis of the $j$th qubit to $Z, X, \text{or } Y$ whenever the classical input $(a_j, b_j) \in \{0, 1\}^2$ is $(0, 1), (1, 0), \text{or } (1, 1)$, respectively. Input $(0, 0)$ denotes the presence of an identity in the Pauli operator, in which case the qubit is measured in the $Z$ basis. (Bottom): The same circuit, recompiled using $R_{XX}(\theta)$ gates which are native to our trapped-ion device, and optimized to reduce the number of single-qubit gates.

is defined as

$$|G\rangle = \prod_{(j,k) \in E} CZ_{j,k}|+\rangle^\otimes |V\rangle. \quad (1)$$

Any stabilizer state is equivalent to a graph state up to single-qubit Clifford gates (the set of gates generated by $H = (X + Z)/\sqrt{2}$ and $S = \sqrt{Z}$) [40–42].

Conversely, any graph state is uniquely defined in terms of its stabilizer group. From Eq. (1) it follows that $|G\rangle$ is the $+1$ eigenstate of all the elements of

$$S_G = \left\{ X_v \prod_{l \in N(v)} Z_l \mid \forall v \in V \right\}, \quad (2)$$

where $l \in N(v)$ if and only if $(l, v) \in E$. The notation $\langle \cdot \rangle$ indicates the set of all possible products generated by the operators contained in the brackets. The operators in the brackets are referred to as stabilizer generators. We may associate the generator $S_v = X_v \prod_{l \in N(v)} Z_l$ with the vertex $v$.

In this work we focus on the six-qubit graph state on the cycle graph $|C_6\rangle$, also known as the six-qubit cyclic cluster state, where $C_6$ denotes the six-vertex cycle graph. Following Eq. (1), the preparation of $|C_6\rangle$, followed by a measurement of an arbitrary Pauli operator $E(a, b)$, is implemented by the quantum circuit in Fig. 1 (top). For the experimental implementation using trapped ions, we recompile this circuit in terms of the native two-qubit Mølmer–Sørensen gates, defined as $R_{XX}(\theta) = \exp(-i\theta X \otimes X)$, yielding the circuit in Fig. 1 (bottom). (Further details of the experimental setup are discussed in Sec. V A.)

B. State fidelity and detection of entanglement

The fidelity between a target pure state $|\psi\rangle$ and a prepared (possibly mixed) state $\rho$ is given by

$$F(\rho, |\psi\rangle) = |\langle \psi \rangle_{\rho} | = \text{tr}(|\psi\rangle\langle \psi| \rho). \quad (3)$$

Accordingly, the fidelity between $|\psi\rangle$ and $\rho$ can be thought of as the expectation value of the projector onto $|\psi\rangle$ with respect to $\rho$. In particular, the projector onto any $n$-qubit stabilizer state, such as a graph state $|G\rangle$, can be expressed in terms of the $2^n$ elements of its stabilizer group $S_G$,

$$|G\rangle\langle G| = \frac{1}{2^n} \sum_{S \in S_G} S. \quad (4)$$

The fidelity between a graph state $|G\rangle$ and a state $\rho$ is therefore determined by the expectation values of the $2^n$ elements of the stabilizer group,

$$F(\rho, |G\rangle) = \frac{1}{2^n} \sum_{S \in S_G} \text{tr}(\rho S). \quad (5)$$
Another important property of the prepared state is whether or not it possesses genuine multipartite entanglement. A quantum state is said to have genuine \( n \)-partite entanglement if it cannot be expressed as a convex sum of biseparable states (i.e., states that are separable with respect to some bipartition of the \( n \)-qubit system). Since the maximum fidelity any biseparable state can have with a connected graph state is \( 1/2 \), the operator
\[
W = \frac{1}{2} - |G\rangle\langle G|
\] 
acts as a witness for genuine \( n \)-qubit entanglement [43]. Namely, \( \text{tr}(\rho W) < 0 \) implies the \( \rho \) has genuine multipartite entanglement. Thus the presence of multipartite entanglement may be detected directly from the fidelity as
\[
\text{tr}(\rho W) = \frac{1}{2} - \mathcal{F}(\rho, |G\rangle).
\] 

\( \text{C. Bell-type nonlocal games} \)

In this section we review the concept of nonlocal games and draw an analogy between classical circuits and classical strategies with communication. A nonlocal game [7, 8] is a computational scenario that takes place over several rounds. In each round, multiple parties are each provided a piece of information about a global input string \( x \in \text{Input} \subset \{0,1\}^* \) (the notation \( \{0,1\}^* \) denotes the set of arbitrary length binary strings, i.e., \( \{0,1\}^* = \{0,1,00,01,\ldots\} \)). The parties must then respond to a referee, each with one bit, to produce an output string \( y \in \{0,1\}^* \) that has certain properties depending on the input. The referee then checks that the output possesses the desired properties and accordingly tallies the round as a win or loss. For a particular game, denote the set of valid outputs \( y \) for a given input \( x \) as \( \text{Win}(x) \).

A strategy for a nonlocal game is a scheme by which the parties produce their outputs. Suppose each party receives \( l \) bits of information about the input \( x \). In a quantum strategy, each party holds one qubit from a multipartite entangled quantum state \( \rho \), measures their qubit in some basis depending on their received bits, and outputs the measurement outcome. We will compare quantum strategies implemented on the device to classical strategies with the same connectivity. In such a classical strategy, each party may communicate with neighboring parties to obtain a total of \( k \leq K \) bits of information about the input. Each party then manipulates their information via some Boolean function \( f : \{0,1\}^k \rightarrow \{0,1\}^* \) and repeats this process for a total of \( D \) rounds of communication; in the last round, each output is produced via some \( y : \{0,1\}^k \rightarrow \{0,1\} \). Each classical strategy can therefore be associated to a particular depth-\( D \) classical circuit, consisting of gates depending on no more than \( K \) outputs of its neighboring gates in the previous layer (i.e., each gate is geometrically restricted with respect to some bipartition of the \( n \)-qubit system). Since the maximum fidelity any biseparable state can have with a connected graph state is \( 1/2 \), the operator
\[
W = \frac{1}{2} - |G\rangle\langle G|
\] 
acts as a witness for genuine \( n \)-qubit entanglement [43]. Namely, \( \text{tr}(\rho W) < 0 \) implies the \( \rho \) has genuine multipartite entanglement. Thus the presence of multipartite entanglement may be detected directly from the fidelity as
\[
\text{tr}(\rho W) = \frac{1}{2} - \mathcal{F}(\rho, |G\rangle).
\] 

FIG. 2. Representation of a generic depth-1, fan-in \( \leq 3 \) classical circuit that is geometrically restricted with respect to the graph \( C_6 \), as a directed acyclic graph. This particular circuit is constructed for problems where each party receives \( l = 1 \) bit of information about the input. Each node labeled \( y_j \) denotes an output of an arbitrary Boolean function of the three adjacent bits. Periodic boundary conditions are imposed on the top and bottom edges on the directed acyclic graph.
inequality, which in turn demonstrates a quantum computational advantage.

III. $C_6$ CUBIC BOOLEAN FUNCTION GAMES

Here we describe a nonlocal game adapted from the graph-state Bell inequalities presented in Ref. [19]. In this game, parties are given partial information about an input string $x \in \{0,1\}^6$ and are tasked to produce outputs such that the joint parity is equal to a particular cubic Boolean function associated to a graph $G = (V,E)$ evaluated on the input. For a particular graph $G = (V,E)$ and input set $I \subseteq \{0,1\}^6$ we denote the game as CBF($G,I$).

For concreteness, we define the game over the six-vertex cycle graph.

$C_6$ cubic Boolean function games (CBF($C_6,I$)).

Consider a game where six parties corresponding to the vertices of the cycle graph $C_6$ are each given a two-bit input $s_j = (a_j,b_j) \in \{0,1\}^2$, $j = 0,\ldots,5$. The inputs $s_j$ are drawn as follows: from a global input $x = (x_0,\ldots,x_5) \in I \subseteq \{0,1\}^6$, each party $j$ is given the bit $a_j = x_j$ and the parity of their neighbors’ bits, $b_j = x_{j+1} + x_{j-1} \mod 2$. Each party $j$ then produces a one-bit output $y_j$. The parties win the game whenever

$$\sum_{j \in \text{supp}(s)} y_j = 5 \sum_{j=0}^{5} x_{j-1} x_j x_{j+1} \mod 2, \quad (10)$$

where $\text{supp}(s) = \{0 \leq j \leq 5 \mid s_j \neq (0,0)\}$ and all subscripts are taken mod 6.

In a perfect quantum strategy for this game, each party holds one qubit from the state $|C_6\rangle$ and measures the Pauli operator $E_j(a_j,b_j) = i^{a_j b_j} X_j^{a_j} Z_j^{b_j}$. This strategy is perfect because, collectively, the parties measure the Pauli part of the stabilizer $S_x = \prod_{j=0}^{5} S_j^{x_j}$, where $S_j$ denotes the stabilizer generator corresponding to vertex $j$ of $C_6$. The parity of the measurement outcomes for qubits $j \in \text{supp}(s)$ is then deterministically equal to the phase in front of $S_x$, which is $\exp\left(i\pi \sum_{j=0}^{5} x_{j-1} x_j x_{j+1}\right)$. See Fig. 1 and Appendix C for more details. On the other hand, classical strategies perform with varying success depending on the particular input set $I$. We now consider two particular input sets that give to different classical bounds on the depth-0 success probability.

A. Full-input cubic Boolean function game (fidelity game)

Setting the input set $I$ to be all possible six-bit strings, we have the game CBF($C_6,\{0,1\}^6$). In the perfect quantum strategy presented above, each of the $2^6$ inputs corresponds to a measurement of one element of the stabilizer group for the graph state $|C_6\rangle$. Playing many rounds of the game can be interpreted as a partial tomography experiment to determine the fidelity of the prepared state $|C_6\rangle$. The success probability $\Pr_Q[\text{win}]$ for the quantum strategy using $\rho$ is related to the fidelity of $\rho$ with respect to $|C_6\rangle$, via

$$\mathcal{F}(\rho,|C_6\rangle) = 2\Pr_Q[\text{win}] - 1. \quad (11)$$

For this specific input set, the depth-0 classical circuits cannot win this game with probability greater than $\Pr_C[\text{win}] \leq 23/32$. This comes from the local hidden-variable theory upper bound presented in Ref. [19] for the corresponding graph-state Bell inequality. However, depth-1 circuits can implement perfect strategies. This is achieved by each party individually computing one cubic term in the function of Eq. (10) as $y_j = a_j - a_j(b_j + a_{j-1}) \mod 2$, which satisfies the win conditions of the game. Hence,

$$\Pr_C[\text{win}] \leq \text{Depth-0} \ \frac{23}{32} \leq \text{Depth-1} \ 1 \leq \text{Depth-2} \ 1. \quad (12)$$

B. Restricted-input cubic Boolean function game

Consider the restriction to the 55-element set

$$x_{\text{Mermin}}^{(55)} = \left\{ x \in \{0,1\}^6 \mid |x| \neq 0,1, \quad x \neq 010101, 101010 \right\}, \quad (13)$$

where $|x| = \sum_{j=0}^{5} x_j$ denotes the Hamming weight of $x$ (i.e., the number of ones in the binary string). This input set yields the game CBF($C_6,x_{\text{Mermin}}^{(55)}$), first studied as a Bell scenario in Ref. [47], where it was shown to have the smallest depth-0 classical success probability for the state $|C_6\rangle$.

The same quantum strategy presented for CBF($C_6,\{0,1\}^6$) is also perfect for this game. For the classical bounds, depth-0 classical circuits cannot win with probability greater than $37/55$, although the same depth-1 circuit presented in Sec. IIIA remains perfect. Hence,

$$\Pr_C[\text{win}] \leq \text{Depth-0} \ \frac{37}{55} \leq \text{Depth-1} \ 1 \leq \text{Depth-2} \ 1. \quad (14)$$

IV. $C_6$ STABILIZER SUBMEASUREMENT GAMES

We now move on to another class of games called stabilizer submeasurement games. Unlike the CBF games, the best depth-1 classical strategies are not perfect here. These games are defined as follows.

$C_6$ stabilizer submeasurement games (SS($C_6,I$)).

Consider a game where six parties corresponding to the vertices of the cycle graph $C_6$ are each given a one-bit input $x_j, j = 0,\ldots,5$, from a global input $x = \ldots,
$(x_0, \ldots, x_5) \in I \subseteq \{0,1\}^6$, drawn at random from a uniform distribution. With each input we can associate a global Pauli operator $E(1, x) = \prod_{j=0}^5 x_j^2 X_j^2$, where $1$ denotes the all-ones string. Each party then produces an output $y_j \in \{0,1\}$, which collectively forms a string $y \in \{0,1\}^6$. The parties are said to win the game whenever

$$\forall p \subseteq E(1, x) \text{ s.t. } P \in S_{C_6} \sum_{j \in \text{supp}(P)} y_j = 0,$$

$$\forall p \subseteq E(1, x) \text{ s.t. } P \in -S_{C_6} \sum_{j \in \text{supp}(P)} y_j = 1,$$

where $p \subseteq E(1, x)$ means that $P$ is a Pauli operator obtained by replacing some nonidentity tensor factors in $E(1, x)$ with the identity. Furthermore, $\text{supp}(P) = \text{supp}(a, b)$ where $P = E(a, b)$ (i.e., $\text{supp}(P)$ indexes the qubits where $P$ acts nontrivially) and all arithmetic is performed modulo 2.

The $SS(C_6, I)$ games always have a perfect quantum strategy wherein each party $j$ holds one qubit from the state $|C_6\rangle$, measures it in the basis of $X_j$ or $Y_j$ if $x_j = 0$ or 1, respectively, and then outputs the measurement outcome. For more details on the measurement outcomes expected when a stabilizer element exists as a submeasurement of a global Pauli measurement, see Appendix A. On the other hand, classical strategies perform with varying success depending on the particular input set $I$.

### A. 2D hidden linear function (HLF) game

Fixing the input set to be

$$\mathcal{T}^{(8)}_{\text{HLF}} = \left\{ x \in \{0,1\}^6 \mid (x_0, x_2, x_4) \in \{0,1\}^3, x_j = 0 \text{ for all other } j \right\},$$

we have the game $SS(C_6, \mathcal{T}^{(8)}_{\text{HLF}})$. This game is equivalent to the smallest instance of the 2D hidden linear function game introduced in Ref. [31] that was used to demonstrate an unconditional exponential separation between the power of classical and quantum circuits.

As shown in Appendix D2, all geometrically restricted depth-0 and depth-1 fan-in $\leq 3$ circuits cannot win this game on more than $7/8$ inputs; however, there is a depth-2 circuit that wins on all inputs. Hence,

$$\Pr_C[\text{win}] \leq_{\text{Depth-0}} \frac{7}{8} \leq_{\text{Depth-1}} \frac{7}{8} \leq_{\text{Depth-2}} 1.$$  

### B. Restricted-input 2D hidden linear function (HLF) game

By restricting the input set to be

$$\mathcal{T}^{(5)}_{\text{HLF}} = \left\{ x \in \{0,1\}^6 \mid (x_0, x_2, x_4) \in \mathcal{V}, x_j = 0 \text{ for all other } j \right\},$$

where the set $\mathcal{V} \subseteq \{0,1\}^3$ is defined as

$$\mathcal{V} = \{(0,0,0), (0,1,1), (1,0,1), (1,1,0), (1,1,1)\},$$

we obtain the game $SS(C_6, \mathcal{T}^{(5)}_{\text{HLF}})$. Analogously to how a restricted input set makes the game $CBF(C_6, \mathcal{T}^{(5)}_{\text{Mermin}})$ classically harder than $CBF(C_6, \{0,1\}^6)$ (i.e., the depth-0 lower bound on $\Pr_C[\text{win}]$ decreases), the same behavior occurs when restricting from $\mathcal{T}^{(8)}_{\text{HLF}}$ to $\mathcal{T}^{(5)}_{\text{HLF}}$. As shown in Appendix D2, the depth-0 and depth-1 classical bounds are reduced to

$$\Pr_C[\text{win}] \leq_{\text{Depth-0}} \frac{4}{5} \leq_{\text{Depth-1}} \frac{4}{5} \leq_{\text{Depth-2}} 1.$$
by applying a (noisy) $R_{XX}$ gate to the computational basis state $|00\rangle$ and measuring the odd-parity-state population $A = \langle 01|\rho|01\rangle + \langle 10|\rho|10\rangle$ of the resulting two-qubit state $\rho$ [53].

The implementation of a circuit consists of sequences of gate operations between qubits, compiled into the native single- and two-qubit gate operations. After executing the quantum circuit, the state of each ion is read out in parallel using state-dependent fluorescence detection [48]. State-preparation-and-measurement (SPAM) errors are characterized and corrected by applying the inverse of an independently measured state-to-state error matrix [54]. Detailed performance of the system has been described elsewhere [39, 55].

### B. Full-input cubic Boolean function game (fidelity game)

As discussed in Sec. IIIA, the fidelity $F(\rho, |C_6\rangle)$ indicates the probability of winning the game $\text{CBF}(C_6, \{0, 1\}^6)$, using $\rho$ as the quantum resource. We therefore directly estimate this fidelity by performing tomography over all $2^6 - 1 = 63$ nontrivial stabilizers. A straightforward implementation is to measure the 63 stabilizers one-by-one, with an equal number of samples each. Indeed, this would be essentially equivalent to known probabilistic protocols for estimating state fidelities, which in this context use uniformly randomized measurement settings over all stabilizers [45, 46]. In order to obtain an estimate of the fidelity within precision $\pm \epsilon$, with success probability $\geq 1 - \delta$, such methods require up to $[8 \ln(4/\delta)/\epsilon^2]$ measurements. This is notably independent of system size, despite the fact that the total number of stabilizers grows exponentially with $n$. However, since our six-qubit system is fairly small, this randomization turns out to incur a larger constant overhead compared to a more direct approach.

Therefore we instead use a greedy graph-coloring heuristic to find locally commuting subsets of the stabilizers [56, 57]. Two Pauli operators $\bigotimes_{j=0}^{n-1} P_j, \bigotimes_{j=0}^{n-1} Q_j \in \{I, X, Y, Z\}^\otimes n$ are said to locally commute if $[P_j, Q_j] = 0$ for all $j$. This allows us to parallelize the stabilizer measurements into fewer measurement settings, reducing the overall sampling runtime. For this experiment, we used a total of 37 global Pauli measurement settings to estimate all 63 nontrivial stabilizer expectation values (see Table III in Appendix F for details of this partitioning). Each measurement setting was sampled 5000 times to obtain adequate statistics.

The stabilizer expectation values are visualized in Fig. 3, with $1\sigma$ uncertainty bars associated with the sampling statistics. Explicit numerical values are provided in Table III in the Appendix. We include the values from both the raw experimental data, as well as after applying SPAM correction to the outcome distributions. From these results we estimate the fidelity of our prepared state...
\[ \rho \text{ to be} \]
\[ \hat{F}(\rho, \ket{C_6}) = \begin{cases} 0.6061(17) & \text{from raw data} \\ 0.6639(19) & \text{SPAM-corrected} \end{cases} \] (21)

The statistical uncertainty of this fidelity estimate is computed by propagating uncertainties from the stabilizer expectation values, paying special care to handle the covariances between stabilizers that are simultaneously evaluated within the same Pauli measurement setting. Details of this error analysis are discussed in Appendix E.

To put this result in context, we briefly review prior experimental work on graph states over six qubits. In Ref. [24], an H-graph state was prepared with fidelity $0.593(25)$ on a photonic device. More recently, Ref. [58] prepared a six-qubit path-graph state with fidelity just under 0.6 on a superconducting architecture; using the error-mitigation techniques also developed in that work, they then demonstrated an error-mitigated fidelity above 0.8. In both works, the underlying graphs required only five two-qubit entangling gates, whereas our $\ket{C_6}$ state requires six such gates.

As discussed in Sec. II B, we can also use the fidelity of $\rho$ to compute an estimate for the genuine multipartite-entanglement witness $W$ via Eq. (7). From the data, we obtain the estimate
\[ \text{tr}(\hat{\rho}W) = \begin{cases} -0.1061(17) & \text{from raw data} \\ -0.1639(19) & \text{SPAM-corrected} \end{cases} \] (22)

The value of the witness is negative, which indicates the presence of genuine multipartite entanglement in our experimentally prepared state. In other words, we guarantee that $\rho$ is entangled across all possible bipartitions of our six-qubit system.

Finally, using the relation $F(\rho, \ket{C_6}) = 2 \Pr_Q[\text{win}] - 1$ for the CBF games, we obtain a raw success probability for winning this full-input game with our experimentally prepared state $\rho$ as
\[ \hat{\Pr}_Q[\text{win}] = 80.30(8)\% \] (23)

This value significantly surpasses the classical bound of $23/32 \approx 72\%$.

Here, we have reported only the probability estimated from the raw data; we argue that any conclusions drawn from SPAM-corrected data in this context can be criticized. This is because the nonlocal game is evaluated on a shot-to-shot basis: a given output either does or does not satisfy the requisite win conditions. The final success rate is then merely an average over many individual shots (rounds of the game). SPAM correction, on the other hand, is a postprocessing technique that is necessarily applied to the observed distribution of outputs, rather than to each output individually. In contrast, we show corrected values to quantify the contribution of SPAM errors in our results and illustrate the quality of the entangled state created.

\begin{table}[h]
\centering
\begin{tabular}{c|ccc}
\hline
Input & Raw value & SPAM-corrected & Classical bound (%) \\
\hline
000000 & 80.42(56) & 86.45(48) & - \\
000100 & 84.76(51) & 87.85(46) & - \\
001000 & 83.92(52) & 86.94(48) & - \\
001010 & 75.58(61) & 81.12(55) & - \\
100000 & 85.04(50) & 88.11(46) & - \\
100100 & 78.12(58) & 83.98(52) & - \\
101000 & 78.46(58) & 84.28(51) & - \\
101010 & 84.52(51) & 87.56(47) & - \\
\hline
\hline
$\mathcal{T}^{(8)}_{\text{HLF}}$ & 81.35(59) & 85.79(53) & 87.5 \\
$\mathcal{T}^{(5)}_{\text{HLF}}$ & 79.42(25) & 84.68(23) & 80 \\
\hline
\end{tabular}
\caption{Experimental results for the stabilizer submeasurement games, SS($C_6, \mathcal{T}^{(8)}_{\text{HLF}}$) and SS($C_6, \mathcal{T}^{(5)}_{\text{HLF}}$). The rules and winning conditions of the game are described in Sec. IV. For each input, 5000 rounds of the game were played, from which we estimate a win rate and a 1σ statistical uncertainty. For completeness we include the SPAM-corrected probabilities, although as argued in the main text, the proper figure of merit for claiming quantum computational advantage here (or lack thereof) is the raw value.}
\end{table}

\section{C. Restricted-input cubic Boolean function game}

Using the same experimental data, we can also estimate our device’s average success probability for the game $\mathcal{CBF}(C_6, \mathcal{T}^{(5)}_{\text{Mermin}})$ discussed in Sec. III B. This game was also recently implemented in Ref. [59] using IonQ’s cloud-accessible quantum computer, where a winning probability of 87(1)% was reported. Using the data, we estimate
\[ \hat{\Pr}_Q[\text{win}] = 79.51(9)\% \] (24)

far surpassing the classical bound of $37/55 \approx 67\%$. We remind the reader that this result corresponds to the raw output of our quantum device.

\section{D. 2D hidden linear function game}

As another benchmark of our device, we performed the game SS($C_6, \mathcal{T}^{(8)}_{\text{HLF}}$), described in Sec. IV A, which is the smallest example of the so-called 2D hidden linear function game of Ref. [31]. This game was also recently implemented in Ref. [59] on IBM, IonQ, and Honeywell’s cloud-accessible quantum computers. Their highest reported success probability was 85(1)%; using Honeywell’s H0 trapped-ion device, which does not surpass the depth-1 classical bound of 87.5%.

In this experiment, we played 5000 rounds of the game per input to determine the average success probability for our device. The results for each input are presented in Table II. The estimated average success probability is
\[ \hat{\Pr}_Q[\text{win}] = 81.35(59)\% \] (25)

which is below the depth-1 classical bound.
E. Restricted-input 2D hidden linear function game

Using the same experimental data from Table II, we also estimate our device’s average success probability for the game \( SS(C_6, I_{HLF}^{(5)}) \), described in Sec. IV B. The estimated average success probability in this case is

\[
\widehat{p}_Q[\text{win}] = 79.42(25)\% \quad \text{[SS(C_6, I_{HLF}^{(5)}) game]},
\]

which is very close to meeting the depth-1 classical bound of 80%.

F. Characterization of device noise

In this section we present a method for characterizing the various types of noise in our trapped-ion device based on the SPAM-corrected stabilizer information (cf. Fig. 3) and classical simulations. Since the experiments we implemented involved only six qubits, a single instance of the experiment with a particular set of parameters can be simulated on a laptop in a few minutes, making the following method tractable for small system sizes. We simulate noisy quantum circuits with a physically motivated error model over many different noise rates and determine the values which best fit the experimental data.

1. Error model

The error model we chose consists of three noise channels with variable error rates, and two channels with fixed error rates that were determined from device calibration beforehand. The variable noise channels consist of a single-qubit depolarizing error after every single-qubit gate, and both a two-qubit depolarizing error and a stochastic joint bit-flip error after every two-qubit gate. The corresponding error channels (acting on qubits \( j,k \)) are defined as

\[
\mathcal{E}_{1d}^{(j)}(\rho) = (1 - p_{1d}^j)\rho + \frac{p_{1d}^j}{3}(X_j\rho X_j + Y_j\rho Y_j + Z_j\rho Z_j),
\]

\[
\mathcal{E}_{2d}^{(j,k)}(\rho) = (1 - p_{2d}^j)\rho + \frac{p_{2d}^j}{15}\sum_{(P,Q)\in K_{\text{dep}}} P_j Q_k\rho Q_k P_j,
\]

\[
\mathcal{E}_{XX}^{(j,k)}(\rho) = (1 - p_{XX}^j)\rho + p_{XX}^j X_j\rho X_j + Y_k\rho Y_k + Z_j\rho Z_j,
\]

where \( K_{\text{dep}} = \{I, X, Y, Z\}^2 \setminus \{(I, I)\} \) and \( p_{1d}^j, p_{2d}^j, \) and \( p_{XX}^j \) are the characteristic noise parameters, respectively. The use of two types of two-qubit gate errors is motivated by the physics of the ion-trap device. The \( XX \) error channel can be interpreted as the ensemble average of random, normally distributed overrotations of the \( R_{XX} \) gate (with variance related to \( p_{XX}^j \)) [60]. However this type of error preserves two-qubit parity, so we additionally include the depolarizing error channel as an effective model of more general noise features.

For the fixed error channels, we assume that a single-qubit dephasing error occurs whenever a qubit idles in the circuit, and a two-qubit crosstalk error occurs between ion pairs that are unintentionally coupled when we apply an \( R_{XX} \) gate [60]. Idling errors are modeled by the channel

\[
\mathcal{E}_{\text{idle}}^{(j)}(\rho) = (1 - p_{\text{idle}})\rho + p_{\text{idle}} Z_j Z_j,
\]

where the error rate depends on the gate time \( t \) as \( p_{\text{idle}} = (1 - e^{t/T_2})/2 \). Based on prior calibration of our device, we estimate \( T_2 = 200\, \text{ms} \). The single- and two-qubit gate times are \( t_1 = 10\, \mu\text{s} \) and \( t_2 = 350\, \mu\text{s} \), respectively. Crosstalk errors are modeled with \( XX \) errors [see Eq. (29)] between each pair of unintentionally coupled ions with error rate \( p_c = 0.06\% \), as determined by the crosstalk Rabi ratio \( \Omega_c/\Omega_B = 3\% \) of our device. Measurement noise is set to zero, as we compare the outcome of the classical simulation to the SPAM-corrected data, which ideally corrects for all measurement errors. We use Qiskit [61] to simulate the noisy circuit using the quantum-trajectories method, i.e., by stochastically applying Kraus operators with their corresponding probabilities at the appropriate steps in the circuit. We take \( 10^4 \) shots per measurement setting for each simulation point, so that sampling errors are negligible.

2. Identifying noise parameters of best fit

Given the noise model, we classically simulate the noisy circuit for a range of noise parameters \( (p_{1d}, p_{XX}^j, p_{2d}^j) \) and find the values that most closely approximate the experiment by analyzing two figures of merit. The first is the absolute difference between the fidelity value obtained from the experiment, \( \hat{F}_\text{exp} \), and the fidelity value obtained from classical simulation, \( F_\text{sim} \),

\[
\Delta F = |\hat{F}_\text{exp} - F_\text{sim}| = \frac{1}{64} \sum_{S \in S_{\text{Ca}}} |\langle S\rangle_\text{exp} - \langle S\rangle_\text{sim}|.
\]

Here, \( S_{\text{Ca}} \) is the stabilizer group of \( |C_6\rangle \), and \( \langle S\rangle_\text{exp} \) and \( \langle S\rangle_\text{sim} \) denote the expectation value of stabilizer \( S \) calculated from the experimental and simulated data, respectively. The second metric is the average absolute difference of stabilizer values,

\[
\Delta S = \frac{1}{64} \sum_{S \in S_{\text{Ca}}} |\langle S\rangle_\text{exp} - \langle S\rangle_\text{sim}|.
\]

If our simulation exactly mirrors the experiment, then \( \Delta F = \Delta S = 0 \). For our task of fitting the noise parameters, we employ both metrics. The weaker \( \Delta F = 0 \) condition works
nels [Eqs. (27–29)] are the gate infidelities [62, Sec. 9.3] of our three error channels. From these figures, we estimate the best-fit parameters to be

\[
\Delta S = (1.2\%, 3.5\%, 3.5\%)
\]

corresponding to the best-fit parameters for this noise model of our device.

VI. n-QUBIT GENERALIZATION

While the work presented so far concerned only a six-qubit system, we present a generalization of our games to the n-qubit scenario. Such games offer a scalable demonstration of the advantage that constant-depth quantum circuits have over geometrically restricted classical circuits of depth linearly growing with n.

A. C_n cubic Boolean function games

The C_6 cubic Boolean function games introduced in Sec. III generalize to the n-qubit scenario as follows.

C_n cubic Boolean function games (CBF(C_n, I)). Consider a game where n parties corresponding to the vertices of the cycle graph C_n are each given a two-bit input \( s_j = (a_j, b_j) \in \{0,1\}^2 \), \( j = 0,\ldots, n-1 \). The inputs \( s_j \) are drawn as follows: from a global input \( \mathbf{x} = (x_0,\ldots,x_{n-1}) \in \mathcal{I} \subseteq \{0,1\}^n \), each party \( j \) is given the bit \( a_j = x_j \) and the parity of their neighbors’ bits, \( b_j = x_{j+1} + x_{j-1} \mod 2 \). Each party \( j \) then produces a one-bit output \( y_j \). The parties win the game whenever

\[
\sum_{j \in \text{supp}(s)} y_j = \sum_{j=0}^{n-1} x_{j-1}x_jx_{j+1} \mod 2,
\]

where \( \text{supp}(s) = \{0 \leq j \leq n-1 \mid s_j \neq (0,0)\} \) and all subscripts are taken \( \mod n \).

As shown in Appendix C, the structure of the signs appearing in the stabilizer group of the n-qubit cycle graph state \( |C_n\rangle \) always has the form of a cubic Boolean
function \( g_n(x) = \sum_{j=0}^{n-1} x_{j-1}x_jx_{j+1} \). Thus the quantum strategy where each party \( j \) measures the Pauli observable \( E_j(a_j, b_j) \) is always perfect. As for classical strategies, the game \( \operatorname{CBF}(C_n, \{0,1\}^n) \) for \( n \) parties always exhibits a depth-0 bound \( \beta_0 < 1 \) (cf. Theorem 1 of Ref. [19]). However, the exact value of \( \beta_0 \) for \( n > 1 \) is unknown and cannot be obtained efficiently numerically. A general discussion of cubic Boolean function games defined on arbitrary graphs will be presented in future work. On the other hand, this game exhibits a trivial depth-1 bound \( \beta_1 = 1 \), as the strategy whereby each party outputs \( y_j = a_{j-1}a_j(b_j + a_{j-1}) = x_{j-1}x_jx_{j+1} \mod 2 \) is perfect.

**B. \( C_n \) stabilizer submeasurement games**

The \( C_6 \) stabilizer submeasurement games introduced in Sec. IV generalize to the \( n \)-qubit scenario as follows.

**\( C_n \) stabilizer submeasurement games (SS\((C_n, I)\)).** Consider a game where \( n \) parties corresponding to the vertices of the cycle graph \( C_n \) are each given a one-bit input \( x_j, \ j = 0, \ldots, n-1 \), from a global input \( x = (x_0, \ldots, x_{n-1}) \in I \subseteq \{0,1\}^n \) drawn at random from the uniform distribution. To each input we can associate a global Pauli operator \( E(1, x) = \prod_{j=0}^{n-1} (x_jX_jZ_j^2) \), where \( 1 \) denotes the all-ones string. Each party then produces an output \( y_j \in \{0,1\} \), which collectively forms a string \( y \in \{0,1\}^n \). The parties are said to win the game whenever

\[
\forall P \subseteq E(1, x) \text{ s.t. } P \in S_{C_n}, \quad \sum_{j \in \text{supp}(P)} y_j = 0, \quad (34a)
\]

\[
\forall P \subseteq E(1, x) \text{ s.t. } P \in -S_{C_n}, \quad \sum_{j \in \text{supp}(P)} y_j = 1, \quad (34b)
\]

where \( P \subseteq E(1, x) \) means that \( P \) is a Pauli operator obtained by replacing some nonidentity tensor factors in \( E(1, x) \) with the identity. Furthermore, \( \text{supp}(P) = \text{supp}(a, b) \) where \( P = E(a, b) \) (i.e., \( \text{supp}(P) \) indexes the qubits where \( P \) acts nontrivially) and all arithmetic is performed modulo 2.

By increasing the number of qubits used in the our stabilizer submeasurement games, we can demonstrate a separation against circuits of greater odd depth \( D \) and fan-in \( \leq K \) that are geometrically restricted with respect to the cycle graph \( C_n \). Let

\[
(5)
\]

\[
\forall P \subseteq E(1, x) \text{ s.t. } P \in S_{C_n}, \quad \sum_{j \in \text{supp}(P)} y_j = 0, \quad (34a)
\]

\[
\forall P \subseteq E(1, x) \text{ s.t. } P \in -S_{C_n}, \quad \sum_{j \in \text{supp}(P)} y_j = 1, \quad (34b)
\]

where the set \( V \subseteq \{0,1\}^3 \) is defined as

\[
V = \{(0,0,0), (0,1,1), (1,0,1), (1,1,0), (1,1,1)\}. \quad (36)
\]

The following theorem then holds.

**Theorem 1.** Let \( D \) be odd and let \( n = 6D \). No classical Boolean circuit with depth \( D \) and fan-in \( \leq 3 \) that is geometrically restricted with respect to the cycle graph \( C_n \) can win \( \operatorname{SS}(C_n, \mathcal{T}_{\text{HLF}, n}^{(5)}) \) with average success probability greater than \( \beta_D = 4/5 \).

We relegate the proof of this statement to Appendix D.

**Remark 2.** For every even number \( 6(2m - 1) \leq n \leq 6(2m + 1) \) the game \( \operatorname{SS}(C_n, \mathcal{T}_{\text{HLF}, n}^{(5)}) \) satisfies the bounds \( \beta_{2m-1} = 4/5 \) and \( \beta_{2m} = 1 \). Hence, for each odd \( D \), \( n = 6D \) is the smallest example giving the next nontrivial bound of \( \beta_D = 4/5 \).

It is important to note that the above theorem demonstrates an unconditional separation between the power of constant-depth quantum circuits and linear-depth classical circuits with the same cyclic geometry. In order to demonstrate advantage against the more powerful class of classical circuits without geometric restriction (namely those in Nick’s class \( NC^0 \)), it is imperative either to embed the stabilizer submeasurement game into a 2D grid [31], or to change the game to a more complex one, such as the magic square game [36], that is hard even on a 1D geometry. Note that small instances of the 1D magic square game incur a constant overhead in the number of qubits and entangling gates required, and they have a higher classical bound of \( 8/9 \), in contrast to the games proposed here.

**VII. CONCLUSION**

In this work, we have proposed two types of nonlocal games, which can be interpreted as computational tasks that unconditionally prove quantum advantage against a particular class of geometrically restricted, shallow-depth classical circuits. We implemented proof-of-principle experimental demonstrations using six qubits on a gate-based trapped-ion quantum computer. As summarized in Table I, our NISQ device surpasses the conventional depth-0 bounds for the cubic Boolean function games by a significant margin. These games also provide fine-grained details like state fidelity, which is useful for the characterization of noise in a quantum device. On the other hand, our results suggest that state-of-the-art devices are now at the level of challenging the more difficult depth-1 bounds of the stabilizer submeasurement games.

As discussed in Sec. VI, both games generalize nicely to the \( n \)-qubit scenario. However, demonstration of an advantageous quantum strategy for the \( C_n \) stabilizer submeasurement games gives a particularly meaningful separation against classical circuits with larger circuit depth, as we scale \( n \) according to Theorem 1. For example, the stabilizer submeasurement game on an 18-qubit device provides a clear future target, in that surpassing the corresponding bounds would demonstrate an advantage against depth-3 classical circuits. Indeed, as this
classical depth is greater than the number of layers of two-qubit CZ gates without overlap, it would make more rigorous the comparison between the required quantum and classical circuits. Furthermore, these nonlocal games are friendly to a number of experimental platforms as their advantageous quantum strategies could generalize to condensed-matter platforms realizing 1D symmetry-protected topological order [63].

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Appendix A: Additional details on stabilizer submeasurements and expected populations

In this appendix we elaborate on the notion of stabilizer submeasurements used in the main text. We discuss the measurement outcomes expected when a stabilizer exists as a subset of the observables measured in a global Pauli string, which is indicative of the quantum correlations between those outcomes.

The following lemma regarding the stabilizer formalism is well known [62, Sec. 10.5.3].

**Lemma 3.** The expectation value of any $n$-qubit Pauli operator over an $n$-qubit stabilizer state $|\psi\rangle$ with stabilizer group $S$ is

\[
\langle \psi | P | \psi \rangle = +1 \quad \text{iff } P \in S \quad \text{(A1a)}
\]

\[
\langle \psi | P | \psi \rangle = -1 \quad \text{iff } P \in -S \quad \text{(A1b)}
\]

\[
\langle \psi | P | \psi \rangle = 0 \quad \text{iff } P \not\in S. \quad \text{(A1c)}
\]

The following is an extension of this lemma for the case when $P$ contains stabilizers within it.

**Lemma 4.** Consider a global $n$-qubit Pauli operator $P = \prod_{j=1}^{n} P_j$ for some choice of $P_j \in \{X_j, Y_j, Z_j\}$ (i.e., a Pauli string $P$ that has nontrivial support on every qubit). Let $P \cap S \subseteq S$ denote stabilizer elements that are contained in $P$. Denote each such $S \in P \cap S$ with a binary string $f(S) \in \{0, 1\}^n$ whereby $S = (-1)^{\lambda_S} \prod_{j=1}^{n} P_j^{f(S)}$ with $\lambda_S \in \{0, 1\}$: If a measurement of $P$ is performed by locally measuring each Pauli operator in the string, then the outcomes $y \in \{0, 1\}^n$ occur with probability

\[
\Pr[y] = \begin{cases} 
|P \cap S|/2^n & \text{if } f(S) \cdot y = \lambda_S \text{ for all } S \in P \cap S \\
0 & \text{otherwise.}
\end{cases} \quad \text{(A2)}
\]

**Proof.** Suppose that $U$ is the local Clifford unitary that diagonalizes $P$ in the sense that $U P U^\dagger = \prod_{j=1}^{n} Z_j$. Then for each $S \in P \cap S$, $U S U^\dagger = (-1)^{\lambda_S} \prod_{j=1}^{n} Z_j^{f(S)}$. First note that

\[
\langle y | U | \psi \rangle = \langle y | U S U^\dagger | \psi \rangle = \langle y | U S U^\dagger U | \psi \rangle = \langle y | (-1)^{\lambda_S} \prod_{j=1}^{n} Z_j^{f(S)} | \psi \rangle = (-1)^{f(S) \cdot y + \lambda_S} \langle y | U | \psi \rangle,
\]

which implies that either $\lambda_S = f(S) \cdot y$ or $|\langle y | U | \psi \rangle|^2 = \Pr[y] = 0$.

On the other hand, we have that

\[
|\langle y | U | \psi \rangle|^2 = \langle y | \frac{1}{2^n} \sum_{S \in \mathcal{S}} U S U^\dagger | y \rangle
\]

\[
= \frac{1}{2^n} \sum_{S \in \mathcal{S}} (-1)^{\lambda_S} \langle y | \prod_{j=1}^{n} Z_j^{f(S)} | y \rangle
\]

\[
= \frac{1}{2^n} \sum_{S \in \mathcal{S}} (-1)^{\lambda_S} (-1)^{f(S) \cdot y}
\]

\[
= \frac{1}{2^n} \sum_{S \in \mathcal{S}} (-1)^{f(S) \cdot y}
\]

\[
= \frac{|P \cap S|}{2^n}.
\]

Therefore, the measurement outcomes $y \in \{0, 1\}^n$ that occur with nonzero probability are those satisfying $f(S) \cdot y = \lambda_S \forall S \in P \cap S$. Furthermore, each such outcome occurs with uniform probability $|P \cap S|/2^n$. \qed

Appendix B: Bell operators for stabilizer submeasurement games

Here we prove a useful expression for computing the average success probability for the stabilizer submeasurement games in terms of the average output parities for each stabilizer submeasurement $P \subseteq E(1, x)$, which we shall denote by $\langle P \rangle$. This result applies for both quantum and classical strategies: for a quantum strategy using the state $\rho$,

\[
\langle P \rangle = \text{tr}(\rho P), \quad \text{(B1)}
\]
while for a classical deterministic strategy assigning outputs $y \in \{0, 1\}^n$ to each observable in $P$ depending on the input,

$$\langle P \rangle = (-1)^{\sum_{j \in \text{supp}(P)} y_j}. \quad \text{(B2)}$$

**Lemma 5.** Consider a strategy for any stabilizer sub-measurement game $SS(C_n, \mathcal{I})$. The average success probability can be written in terms of expectation values of the stabilizing operators,

$$\text{Pr}[\text{win}] = \frac{1}{|\mathcal{I}|} \sum_{x \in \mathcal{I}} \frac{1}{|P_x|} \sum_{P \in P_x} \text{sgn}(P) \langle P \rangle,$$  \quad \text{(B3)}$$

where $P_x = \{P \leq E(1, x) \mid P \in \pm \mathcal{S}_n \}$ and $\text{sgn}(P) = \pm 1$ whenever $P \in \pm \mathcal{S}_n$

**Proof.** Recall from Eq. (8) that the average success probability can be computed as

$$\text{Pr}[\text{win}] = \frac{1}{|\mathcal{I}|} \sum_{x \in \mathcal{I}} \sum_{y \in \text{Win}(x)} \text{Pr}[y | x],$$  \quad \text{(B4)}$$

where for the game $SS(C_n, \mathcal{I})$ we have

$$\text{Win}(x) = \left\{ y \in \{0, 1\}^n \mid \forall P \in P_x, \ (-1)^{\sum_{j \in \text{supp}(P)} y_j} = \text{sgn}(P) \right\}. \quad \text{(B5)}$$

Define the single-qubit Weyl operators as $E(a, b) = i^a X^a Z^b$ for $a, b \in \{0, 1\}$. The quantity $\text{Pr}[y | x]$ can be computed as

$$\text{Pr}[y | x] = \left\langle \bigotimes_{j=0}^{n-1} \Pi_{y_j}^{E(1, x_j)} \right\rangle,$$  \quad \text{(B6)}$$

where $\Pi_{y_j}^{E(1, x_j)}$ denotes the projector onto the eigenspace corresponding to eigenvalue $(-1)^{y_j}$ of the operator $E(1, x_j)$. Since $E(1, x_j)$ is a single-qubit Pauli observable, we can write

$$\Pi_{y_j}^{E(1, x_j)} = \frac{I + (-1)^{y_j} E(1, x_j)}{2}, \quad \text{(B7)}$$

hence

$$\text{Pr}[\text{win}] = \frac{1}{|\mathcal{I}|} \sum_{x \in \mathcal{I}} \sum_{y \in \text{Win}(x)} \left\langle \bigotimes_{j=0}^{n-1} \frac{I + (-1)^{y_j} E(1, x_j)}{2} \right\rangle.$$  \quad \text{(B8)}$$

Upon binomial-expanding the product in the expectation value, we get

$$\text{Pr}[\text{win}] = \frac{1}{2^n |\mathcal{I}|} \sum_{x \in \mathcal{I}} \sum_{y \in \text{Win}(x)} \sum_{q \in \{0, 1\}^n} \left\langle \bigotimes_{j=0}^{n-1} (-1)^{q_j y_j} E(1, x_j)^y \right\rangle,$$  \quad \text{(B9)}$$

$$= \frac{1}{2^n |\mathcal{I}|} \sum_{x \in \mathcal{I}} \sum_{y \in \text{Win}(x)} \sum_{q \in \{0, 1\}^n} (-1)^{q \cdot y} \left\langle E(q, q \odot x) \right\rangle,$$  \quad \text{(B10)}$$

where $q \odot x = (q_0 x_0, \ldots, q_{n-1} x_{n-1})$ denotes elementwise multiplication of vectors. Notice that the $j$th term in the $n$-qubit Pauli operator $E(q, q \odot x)$ is $I$ if $q_j = 0$, and $E(1, x_j)$ if $q_j = 1$. Therefore, the sum over $q$ is the same as a sum over all Pauli operators $P$ such that $P \subseteq E(1, x)$. Hence,

$$\text{Pr}[\text{win}] = \frac{1}{2^n |\mathcal{I}|} \sum_{x \in \mathcal{I}} \sum_{y \in \text{Win}(x)} \sum_{P \subseteq E(1, x)} \left\langle \bigotimes_{j \in \text{supp}(P)} (-1)^{y_j} \right\rangle \langle P \rangle,$$  \quad \text{(B11)}$$

$$= \frac{1}{2^n |\mathcal{I}|} \sum_{x \in \mathcal{I}} \sum_{y \in \text{Win}(x)} \sum_{P \subseteq E(1, x)} \sum_{y \in \text{Win}(x)} \left\langle \bigotimes_{j \in \text{supp}(P)} (-1)^{y_j} \right\rangle \langle P \rangle.$$  \quad \text{(B12)}$$

If $P \in P_x$, then $\sum_{y \in \text{Win}(x)} (-1)^{\sum_{j \in \text{supp}(P)} y_j} = |\text{Win}(x)| \text{sgn}(P)$ by definition of $\text{Win}(x)$. On the other hand, if $P \notin P_x$, then $\sum_{y \in \text{Win}(x)} (-1)^{\sum_{j \in \text{supp}(P)} y_j} = 0$. Intuitively, this can be understood from the fact that if $P \notin \pm S$, then
\(\sum_{j\in\text{supp}(P)} y_j\) is not constrained by a win condition. Hence, this quantity will be unconstrained and the corresponding sum over \(y \in \text{Win}(x)\) will vanish.

More rigorously, this occurs because for each \(x \in \mathcal{I}\), \(\text{Win}(x)\) corresponds to the solution set of an inhomogeneous system of linear equations \(Ay = b\) over \(\mathbb{F}_2\), where \(A\) is a matrix where each row \(\mu\) corresponds to a binary vector \(a_\mu\) such that \(E(a_\mu, a_\mu \otimes x) \in \pm S\) and \((-1)^{b_\mu} = \text{sgn}(E(a_\mu, a_\mu \otimes x))\). Denote the row space of this matrix as \(\text{Row}(A)\). It is then clear that \(\text{rank}(A) = \log_2 |P_x|\), hence \(|\text{Win}(x)| = 2^{|P_x|} / |\text{Row}(A)|\) by the rank-nullity theorem. Furthermore, each \(y \in \text{Win}(x)\) can be written as \(y = y_0 + \sum_{j=1}^{n-\text{rank}(A)} z_j \eta_j\), where \(Ay_0 = b\), \(z_j \in \mathbb{F}_2\), and \(\{\eta_j\}_{j=1}^{n-\text{rank}(A)}\) spans the null space of \(A\). It then follows that for any \(P = E(q, q \otimes x) \not\in \pm S\),

\[
\sum_{y \in \text{Win}(x)} (-1)\sum_{j \in \text{supp}(P)} y_j = (-1)^q y_0 \prod_{j=1}^{n-\text{rank}(A)} \sum_{z_j = 0}^{1} (-1)^{z_j} q \eta_j. \quad (B13)
\]

The only way for the right-hand side to be nonzero is for \(q \cdot \eta_j = 0\) for each \(j = 1, \ldots, n - \text{rank}(A)\); however, this can only be true if \(q \in \text{Row}(A)\), which does not hold by virtue of the fact that \(P \not\in \pm S\). Therefore

\[
\Pr[\text{win}] = \frac{1}{|\mathcal{I}|} \sum_{x \in \mathcal{I}} \frac{1}{|P_x|} \sum_{P \in P_x} \text{sgn}(P), \quad (B14)
\]

which completes the proof.

Using the above expression, for the game \(\text{SS}(C_6, \mathcal{I}^{(5)}_{\text{HLF}})\) we obtain

\[
\Pr[\text{win}] = \frac{1}{32} (12 + 12 \langle X_1 X_3 X_5 \rangle + \langle X_0 X_2 X_4 \rangle + \langle X_0 X_1 X_2 X_3 X_4 X_5 \rangle - \langle X_0 Y_2 X_3 Y_4 - \langle X_0 X_1 Y_2 Y_4 X_5 \rangle - \langle Y_0 X_2 Y_4 X_5 \rangle - \langle Y_0 X_1 X_2 X_3 Y_4 \rangle - \langle Y_0 X_1 Y_2 X_4 \rangle - \langle Y_0 Y_2 X_3 X_4 X_5 \rangle). \quad (B15)
\]

Similarly, for the game \(\text{SS}(C_6, \mathcal{I}^{(5)}_{\text{HLF}})\) we obtain

\[
\Pr[\text{win}] = \frac{1}{20} (6 + 6 \langle X_1 X_3 X_5 \rangle + \langle X_0 X_2 X_4 \rangle + \langle X_0 X_1 X_2 X_3 X_4 X_5 \rangle - \langle X_0 Y_2 X_3 Y_4 - \langle X_0 X_1 Y_2 Y_4 X_5 \rangle - \langle Y_0 X_2 Y_4 X_5 \rangle - \langle Y_0 X_1 X_2 X_3 Y_4 \rangle - \langle Y_0 X_1 Y_2 X_4 \rangle - \langle Y_0 Y_2 X_3 X_4 X_5 \rangle). \quad (B16)
\]

**Appendix C: Quantum strategy for the \(C_6\) cubic Boolean function game**

To demonstrate that the quantum strategy for the \(C_6\) cubic Boolean function game is perfect, we prove a lemma regarding the structure of the signs appearing in the stabilizer group of the \(n\)-qubit cyclic cluster state. Define the single-qubit Weyl operators as \(E(a, b) = j^{ab} X^a Z^b\) for \(a, b \in \{0, 1\}\). Furthermore, denote the generators of the \(n\)-qubit cyclic cluster state stabilizer group as \(S_j = Z_{j-1} X_j Z_{j+1}\), where the subscripts are taken mod \(n\). We then have the following lemma.

**Lemma 6.** Every element of the stabilizer group of the \(n\)-qubit cyclic cluster state \(|C_n\rangle\), when denoted as a product of generators \(S_x = \prod_{j=0}^{n-1} S_j^{g_j(x)}\) for some \(x \in \{0, 1\}^n\), can be expressed as a Pauli string times a phase via

\[
S_x = (-1)^{g_n(x)} \prod_{j=0}^{n-1} E(x_j, x_{j-1} + x_{j+1}), \quad (C1)
\]

where \(g_n(x) = \sum_{j=0}^{n-1} x_{j-1} x_j x_{j+1}\) and all arithmetic is performed mod 2.

**Proof.** For the \(n\)-qubit cycle graph state,

\[
S_x = \prod_{j=0}^{n-1} (Z_{j-1} X_j Z_{j+1})^{x_j} = \prod_{j=0}^{n-1} Z^{x_j} X^{x_{j-1}} Z^{x_{j+1}} \quad (C2)
\]

Each local Pauli operator can be expressed as a Weyl operator times a phase. In terms of the Weyl operators, this expression becomes

\[
S_x = \prod_{j=0}^{n-1} (-1)^{x_j x_{j-1} x_j + x_{j+1}}
\]

\[
= (-1)^{x_j x_{j+1}} \prod_{j=0}^{n-1} i^{x_j (x_{j-1} + x_{j+1})} E(x_j, x_{j-1} + x_{j+1})
\]

\[
= (-1)^{\sum_{j=0}^{n-1} x_j x_{j+1}} \prod_{j=0}^{n-1} i^{x_j (x_{j-1} + x_{j+1})}
\]

\[
\times \prod_{k=0}^{n-1} E(x_k, x_{k-1} + x_{k+1}) \quad (C3)
\]

Here we have used \(\oplus\) to denote the XOR operation (i.e., sum mod 2) in the exponent of \(i\) to remind the reader that the exponent must be computed mod 2. The phase appearing in the last line then has the rather unnatural
form of $i$ raised to the power of a Boolean function. A simple analysis shows that the total number of factors of $i$ accumulated is $\sum_{j=0}^{n-1} x_j(x_{j-1} \oplus x_{j+1})$, which is simply twice the number of connected strings of 1’s in $x$ with length greater than one. We thus obtain an overall phase of $\pm 1$ whenever this number of connected strings equals 0 or 2 mod 4, respectively. This phase is also expressed by the function $\sum_{j=0}^{n-1} (x_{j-1}+1)x_{j+1}$, but using a base of $-1$ instead of $i$. Thus we have

$$S_x = (-1)^{g_n(x)} \prod_{k=0}^{n-1} E(x_k, x_{k-1} + x_{k+1}),$$

which is what we sought to prove.

It is clear that since $S_x|C_n = |C_n$ we have that,

$$\langle C_n | \prod_{j=0}^{n-1} E(x_j, x_{j-1} + x_{j+1}) | C_n \rangle = (-1)^{g_n(x)}.$$

Therefore, locally measuring each Pauli observable in the string $\prod_{j=0}^{n-1} E(x_j, x_{j-1} + x_{j+1})$ gives measurement outcomes whose parity is equal to $g_n(x) \mod 2$. Hence the strategy presented in the text for the $C_6$ cubic Boolean function game is perfect.

Appendix D: Classical strategies for stabilizer submeasurement games

Here we derive the classical bounds for the stabilizer submeasurement games, assuming the classical strategy is restricted to depth-0, depth-1, or depth-$D$ circuits. In particular, we show that for the particular input set $I_{\text{HLF},n}^{(8)}$ (defined in Eq. (35) in the main text), the classical bounds are 4/5 in all cases considered. This improves upon the previously known bound of 7/8, which corresponds to the input set $I_{\text{HLF},n}^{(8)}$ [31].

1. Depth-0 bounds for stabilizer submeasurement games

Lemma 7. The depth-0 bound for any of the $\text{SS}(C_n, I)$ games can be calculated using the expression in Lemma 5. It follows that no depth-0 circuit can win the game $\text{SS}(C_6, I_{\text{HLF}}^{(8)})$ with probability greater than 7/8, nor can they win the game $\text{SS}(C_6, I_{\text{HLF}}^{(8)})$ with probability greater than 4/5.

Proof. Our proof gives a general technique to calculate this bound for any of the $\text{SS}(C_n, I)$ games. As a classical circuit, depth-0 strategies for these games correspond to circuits with one layer of single-input, single-output gates each computing an affine Boolean function $y_j(x_j)$. The average success probability can then be calculated from the Bell expression given in Lemma 5 as

$$2 \Pr[\text{win}] - 1 = \frac{1}{|I|} \sum_{x \in I} \left| \sum_{y \in \text{supp}(P)} (-1)^{\sum_{j \in \text{supp}(P)} y_j(x_j) \text{sgn}(P)} \right|$$

The floor function takes care of the fact that if one or more on the parity constraints imposed by the win conditions is not met, then the round is counted as a failure [cf. Eq. (34)]. By maximizing over all possible local functions $y_j : \{0, 1\} \to \{0, 1\}$ we can numerically find these bounds for $\text{SS}(C_6, I_{\text{HLF}}^{(8)})$ and $\text{SS}(C_6, I_{\text{HLF}}^{(5)})$ via brute-force search.

Note that Lemma 7 could also be proven by the proof technique in the following Lemma 8.

2. Depth-1 bounds for $\text{SS}(C_6, I_{\text{HLF}}^{(8)})$ and $\text{SS}(C_6, I_{\text{HLF}}^{(5)})$

Lemma 8. Depth-1, fan-in $\leq 3$ classical circuits that are geometrically restricted with respect to the geometry of $C_6$ cannot win the game $\text{SS}(C_6, I_{\text{HLF}}^{(8)})$ on more than 7/8 of the inputs, nor can they win the game $\text{SS}(C_6, I_{\text{HLF}}^{(5)})$ on more than 4/5 of the inputs.

Proof. In the following proof all arithmetic is performed modulo 2. Consider a classical circuit of depth-1 and fan-in $\leq 3$ that is geometrically restricted with respect to the geometry of $C_6$. When given an input from $I_{\text{HLF}}^{(8)}$ or $I_{\text{HLF}}^{(5)}$, as depicted in Fig. 5, the circuit produces outputs that
can be parameterized as follows:

\[
\begin{align*}
y_0 &= \alpha_0 + \beta_0 x_0 & (D2a) \\
y_1 &= \alpha_1 + \beta_1 x_0 + \gamma_1 x_2 + \delta_1 x_0 x_2 & (D2b) \\
y_2 &= \alpha_2 + \beta_2 x_2 & (D2c) \\
y_3 &= \alpha_3 + \beta_3 x_2 + \gamma_3 x_4 + \delta_3 x_2 x_4 & (D2d) \\
y_4 &= \alpha_4 + \beta_4 x_4 & (D2e) \\
y_5 &= \alpha_5 + \beta_5 x_4 + \gamma_5 x_0 + \delta_5 x_0 x_4, & (D2f)
\end{align*}
\]

where \(\alpha_j, \beta_j, \gamma_j, \delta_j \in \{0, 1\}\) for \(j = 0, \ldots, 5\).

If this strategy is to win the game \(SS(C_6, I_{\text{HLF}})\), it must hold that \(y_1 + y_3 + y_5 = 0\) for every input. In particular, for the game \(SS(C_6, I_{\text{HLF}})\) this must hold for the input \(x = 101010\). By inserting Eqs. (D2b), (D2d), and (D2f) into this expression for \(x = 101010\), we obtain a constraint on \(\alpha_j, \beta_j, \gamma_j, \text{ and } \delta_j\) for each \(j \in \{1, 3, 5\}\). Namely,

\[
\sum_{j \in \{1, 3, 5\}} \alpha_j + \beta_j + \gamma_j + \delta_j = 0. \quad (D3)
\]

Let \(e(x) = y_0 + y_2 + y_4 = \alpha + \beta_0 x_0 + \beta_2 x_2 + \beta_4 x_4\). Here we have defined \(\alpha = \alpha_0 + \alpha_2 + \alpha_4\). The win conditions for inputs \(x \in I_{\text{HLF}}^{(5)} \setminus \{101010\}\) imply that the following linear constraints must hold:

\[
\begin{align*}
e(000000) &= 0 & (D4a) \\
e(101000) + y_1(101000) &= 1 & (D4b) \\
e(100010) + y_3(100010) &= 1 & (D4c) \\
e(001010) + y_5(001010) &= 1. & (D4d)
\end{align*}
\]

Inserting the aforementioned Boolean functions into these expressions gives

\[
\begin{align*}
\alpha &= 0 & (D5a) \\
\alpha + \beta_0 + \beta_2 + \alpha_1 + \beta_1 + \gamma_1 + \delta_1 &= 1 & (D5b) \\
\alpha + \beta_2 + \alpha_3 + \beta_3 + \gamma_3 + \delta_3 &= 1 & (D5c) \\
\alpha + \beta_0 + \alpha_5 + \beta_5 + \gamma_5 + \delta_5 &= 1. & (D5d)
\end{align*}
\]

Summing Eq. (D3) and Eqs. (D5a)–(D5d) modulo 2 gives the contradictory statement that \(0 = 1\). Therefore, this class of circuits wins the game \(SS(C_6, I_{\text{HLF}})\) on no more than \(7/8\) of the inputs, and the game \(SS(C_6, I_{\text{HLF}})\) on no more than \(4/5\) of the inputs. \(\square\)

### 3. Depth-D bounds for \(SS(C_6,D, I_{\text{HLF,6D}})\)

**Theorem 1** (Restated from main text.). Let \(D\) be odd and let \(n = 6D\). No classical circuit with depth-\(D\) and fan-in \(\leq 3\) that is geometrically restricted with respect to the cycle graph \(C_n\) can win \(SS(C_6, I_{\text{HLF,6D}})\) with average success probability greater than \(\beta_D = 4/5\).

**Proof.** The proof follows similarly to the six-qubit case. All arithmetic here is performed modulo 2. Let \(D\) be an odd positive integer. Let \(I_{\text{HLF,6D}}^{(5)} = \{x \in \{0,1\}^{6D} \mid x_0 x_2 D x_4 \in \{000, 011, 101, 110, 111\} \text{ and } x_2 = 0 \text{ otherwise}\}\). A general circuit with depth-\(D\) and fan-in \(\leq 3\) that is geometrically restricted with respect to the cycle graph \(C_n\) allows each output \(y_j\) to depend on the input bits \(\{x_{j+k} \mid k = 1, \ldots, D\}\). Thus, upon feeding an input from \(I_{\text{HLF,6D}}^{(5)}\) to such a circuit, the only outputs that depend on more than one of the bits \(x_0, x_2 D, \text{ or } x_4 D\) are \(y_D = y_D(x_0, x_2 D), y_3 D = y_3 D(x_2 D, x_4 D), \text{ and } y_5 D = y_5 D(x_0, x_4 D)\). Now let

\[
\begin{align*}
y^{(0,2D)}_0 &= \sum_{j=1}^{D} y_{2j-1} & (D6a) \\
y^{(2D,4D)}_2 &= \sum_{j=D+1}^{2D} y_{2j-1} & (D6b) \\
y^{(4D,0)}_4 &= \sum_{j=2D+1}^{3D} y_{2j-1} & (D6c) \\
y^{(\text{even})} &= \sum_{j=0}^{3D-1} y_{2j} & (D6d)
\end{align*}
\]

Then we can parameterize each such function as

\[
\begin{align*}
y^{(0,2D)}_0 &= \alpha_0 + \beta_0 x_0 + \gamma_0 x_2 D + \delta_0 x_0 x_2 D & (D7a) \\
y^{(2D,4D)}_2 &= \alpha_3 D + \beta_3 D x_2 D + \gamma_3 D x_4 D + \delta_3 D x_2 D x_4 D & (D7b) \\
y^{(4D,0)}_4 &= \alpha_5 D + \beta_5 D x_4 D + \gamma_5 D x_0 \delta_5 D x_0 x_4 D & (D7c) \\
y^{(\text{even})} &= \alpha_0 + \beta_2 D x_2 D + \beta_4 D x_4 D & (D7d)
\end{align*}
\]
The win conditions for \( SS(C_{6D}, T^{(5)}_{\mathcal{H}_{L,F,6D}}) \) state that

\[
\begin{align*}
    y^{(even)}(0, 0, 0) &= 0 \quad (D8a) \\
    y^{(even)}(1, 0, 1) + y^{(odd)}_{[0,2D]}(1, 1) &= 1 \quad (D8b) \\
    y^{(even)}(0, 1, 1) + y^{(odd)}_{[2D,4D]}(1, 1) &= 1 \quad (D8c) \\
    y^{(even)}(1, 0, 1) + y^{(odd)}_{[4D,0]}(1, 1) &= 1 \quad (D8d) \\
    y^{(odd)}_{[0,2D]}(1, 1) + y^{(odd)}_{[2D,4D]}(1, 1) + y^{(odd)}_{[4D,0]}(1, 1) &= 0. \quad (D8e)
\end{align*}
\]

However, substituting in the parameterization above and summing all the equations returns 0 = 1. Therefore, no circuit with depth-\( D \) and fan-in \( \leq 3 \) that is geometrically restricted with respect to the cycle graph \( C_n \) can win the game \( SS(C_{6D}, T^{(5)}_{\mathcal{H}_{L,F,6D}}) \) with probability greater than 4/5.

**Corollary 9.** Circuits in with depth-(\( D + 1 \)) and fan-in \( \leq 3 \) that are geometrically restricted with respect to the cycle graph \( C_n \) can win the game \( SS(C_{6D}, T^{(5)}_{\mathcal{H}_{L,F,6D}}) \) on all inputs.

**Proof.** In the following all arithmetic is performed modulo 2. Notice that in this case, \( y_{D+1} = y_{2D+1}(x_0, x_{2D}), \) \( y_{3D+1} = y_{3D+1}(x_{2D}, x_{4D}), \) and \( y_{5D+1} = y_{5D+1}(x_0, x_{4D}). \) Taking \( y_{D+1} = x_0 x_{2D}, \) \( y_{3D+1} = x_2 x_{4D}, \) \( y_{5D+1} = x_0 x_{4D}, \) and \( y_j = 0 \) otherwise gives \( y^{(even)} \equiv x_0 x_{2D} + x_{2D} x_{4D} + x_0 x_{4D} \) and \( y^{(odd)}_{[0,2D]} = y^{(odd)}_{[2D,4D]} = y^{(odd)}_{[4D,0]} = 0, \) which satisfies all the above constraints imposed by the win conditions.

### Appendix E: Uncertainty analysis

In this appendix we provide details regarding the statistical estimates of our experiments. In particular, we make explicit the uncertainty contributions due to covariances between simultaneously measured stabilizers, and we show how the SPAM-correction procedure affects these error bars.

#### 1. Raw-data estimates

In our tomography experiment, we prepare the \( n \)-qubit state \( \rho \) using the graph-state formation circuit, further apply a local Clifford unitary \( U \) to rotate into the appropriate Pauli basis, and then perform a projective measurement in the computational basis \( \{ |z\rangle | z \in \{0, 1\}^n \} \). Repeating this procedure \( N \) times yields a collection of \( N \) samples \( b_1, \ldots, b_N \in \{0, 1\}^n \).

Let \( S \in \{I, X, Y, Z\}^\otimes n \times \{\pm 1\} \) be a (signed) \( n \)-qubit Pauli observable diagonalized by \( U \), i.e.,

\[
USU^\dagger = \omega_S \prod_{j=1}^n Z^{f(S)}_j,
\]

where \( f(S)_j = 1 \) if \( S \) acts nontrivially on the \( j \)-th qubit, and 0 otherwise. Here, \( \omega_S \in \{\pm 1\} \) denotes the phase factor of \( S \). Then we can construct an estimator for \( \text{tr}(\rho S) \) as

\[
\hat{\mu}_S = \frac{1}{N} \sum_{k=1}^N \omega_S(-1)^{f(S)} b_k,
\]

where \( f(S) \equiv (f_1(S), \ldots, f_n(S)) \). Since \( S^2 = I \), an estimate for the variance is given by \( \delta^2_{\hat{\mu}_S} = 1 - \hat{\mu}_S^2 \), and in particular the standard error of the sample mean \( \hat{\mu}_S \) is

\[
\delta_{\hat{\mu}_S} = \frac{\delta_{\hat{\mu}_S}}{\sqrt{N}}. \quad (E3)
\]

This is the statistical uncertainty reported for each (non-SPAM-corrected) stabilizer expectation value in Table III.

To compute the standard error of our fidelity estimate, we need to address the covariances between stabilizer expectation values that are evaluated within the same measurement setting. Given a set \( C \) of locally commuting Pauli operators, referred to as a clique, there exists some local Clifford transformation \( U = U_C \) such that Eq. (E1) holds for all \( S \in C \). Then we can evaluate \( \hat{\mu}_S \) for each \( S \in C \), using the same samples \( b_k \), as prescribed above.

Now consider

\[
O_C = \sum_{S \in C} h_S \hat{S}_S, \quad h_S \in \mathbb{R}. \quad (E4)
\]

By linearity, \( \text{tr}(\rho O_C) \) is estimated via

\[
\hat{\mu}_{O_C} = \sum_{S \in C} h_S \hat{\mu}_S, \quad (E5)
\]

which has variance

\[
\text{Var}[\hat{\mu}_{O_C}] = \text{Cov}[\hat{\mu}_{O_C}, \hat{\mu}_{O_C}] = \sum_{S \neq T} h_S h_T \text{Cov}[\hat{\mu}_S, \hat{\mu}_T]. \quad (E6)
\]

We already have an estimate for the diagonal terms, \( \text{Cov}[\hat{\mu}_S, \hat{\mu}_S] = \text{Var}[\hat{\mu}_S], \) via \( \sigma^2_{\hat{\mu}_S} \). The off-diagonal covariances \( \text{Cov}[\hat{\mu}_S, \hat{\mu}_T] \) are estimated as

\[
\hat{\Sigma}_{S,T} = \hat{\mu}_{ST} - \hat{\mu}_S \hat{\mu}_T. \quad (E7)
\]

Note that this expression involves the Pauli operator \( ST \), even if \( ST \notin C \). Nonetheless, since \( S \) and \( T \) locally commute, the samples obtained from this measurement setting contain sufficient information to compute \( \hat{\mu}_{ST} \). Recognizing that \( \hat{\Sigma}_{S,T} = \hat{\sigma}^2_{\hat{\mu}_S} \), we have a compact expression for the variance of \( \hat{\mu}_{O_C} \),

\[
\delta^2_{\hat{\mu}_{O_C}} = \sum_{S \neq T} h_S h_T \hat{\Sigma}_{S,T}. \quad (E8)
\]

Finally, suppose we have disjoint cliques \( C_1, \ldots, C_L \), corresponding to the different measurement settings of
our experiment. Such cliques are constructed in order to estimate the observable $O = \sum_{\ell=1}^{L} O_{\ell}$. (For instance, our fidelity experiment featured $O = |C_{0}\rangle\langle C_{0}|$, $L = 37$, and $h_{S} = 2^{-6} \forall S \in S_{C_{0}}$.) Then we set

$$\hat{\mu}_{O} = \sum_{\ell=1}^{L} \sum_{S \in C_{\ell}} h_{S} \hat{\mu}_{S},$$

(E9)

which has standard error

$$\hat{\sigma}_{\hat{\mu}_{O}} = \sqrt{\frac{1}{N} \sum_{\ell=1}^{L} \sum_{S,T \in C_{\ell}} h_{S} h_{T} \hat{\Sigma}_{S,T}.}$$

(E10)

For simplicity, we have assumed that all cliques were sampled the same number of times $N$, as was the case in our experiment. This is the statistical uncertainty for the fidelity estimate reported in the main text. Note that there are no covariances between estimators of different cliques, since they correspond to separate, independently obtained samples.

### 2. SPAM-corrected estimates

To correct for SPAM errors in our device, we ran a series of calibrating experiments to characterize measurement-readout error rates. This corresponds to constructing a $2^{n} \times 2^{n}$ stochastic matrix $M$ such that, for the probability distribution $q = (q_{z} \mid z \in \{0,1\}^{n})^{T}$ encoded in the prepared quantum state, we instead observe (due to measurement errors solely) the distribution $p = Mq$. By computing the inverse matrix $M^{-1}$, we can recover the SPAM-error-free distribution $q = M^{-1}p$.

In practice, we only have access to the estimate $\hat{p}$ of $p$, constructed from the samples $b_{1}, \ldots, b_{N}$. Applying this inversion procedure to $\hat{p}$, we obtain

$$\hat{q}_{z} = \sum_{z' \in \{0,1\}^{n}} [M^{-1}]_{z,z'} \hat{p}_{z'},$$

$$= \sum_{z' \in \{0,1\}^{n}} \left[ [M^{-1}]_{z,z'} \frac{1}{N} \sum_{k=1}^{N} \delta_{z',b_{k}} \right],$$

$$= \frac{1}{N} \sum_{k=1}^{N} [M^{-1}]_{z,b_{k}}. \quad \text{(E11)}$$

All SPAM-corrected expectation values (and their associated uncertainties) are then evaluated by simply replacing $\hat{p}$ with $\hat{q}$. For instance, in terms of $M^{-1}$, the expectation-value estimates become

$$\hat{\mu}_{S} = \frac{1}{N} \sum_{k=1}^{N} \left[ \sum_{z \in \{0,1\}^{n}} [M^{-1}]_{z,b_{k}} \omega_{S}(-1)^{f^{(S)}_{z}} \right]. \quad \text{(E12)}$$

Similarly, the SPAM-corrected covariance matrix between clique elements is

$$\hat{\Sigma}_{S,T} = \frac{1}{N} \sum_{z,z' \in \{0,1\}^{n}} \left( \sum_{k=1}^{N} [M^{-1}]_{z,b_{k}} [M^{-1}]_{z',b_{k}} \right) \omega_{S} \omega_{T} (-1)^{f^{(S)}_{z} + f^{(T)}_{z'}} - \hat{\mu}_{S} \hat{\mu}_{T}, \quad \text{(E13)}$$

TABLE III: Expectation values for the 63 nontrivial stabilizers of $|C_{0}\rangle$ estimated from the CBF experiment. Uncertainties denote 1σ statistical uncertainties (see Appendix E). Horizontal rules indicate our partitioning of the stabilizers into 37 locally commuting sets, such that we estimate all stabilizers in that set from the same global Pauli measurement.

| Input  | Stabilizer       | Expectation value |
|--------|------------------|-------------------|
|        | Raw value        | SPAM-corrected    |
| 000001 | $+ZIZIX$         | 0.7164(10)        | 0.7616(10)        |
| 001000 | $+IZXZI$         | 0.6732(10)        | 0.7178(11)        |
| 001001 | $+ZZZZZ$         | 0.4952(12)        | 0.5027(14)        |
| 000100 | $+IIIZX$         | 0.7208(10)        | 0.7654(10)        |
| 010000 | $+ZXZII$         | 0.7320(10)        | 0.7787(10)        |
| 010010 | $+ZXZZZ$         | 0.5352(12)        | 0.6046(12)        |
| 000011 | $+IIIZY$         | 0.7656(9)         | 0.8302(10)        |
| 010011 | $+IXZYY$         | 0.6168(11)        | 0.6813(12)        |
| 000100 | $+IIZXZ$         | 0.6764(10)        | 0.7205(11)        |
| 100000 | $+XZIIIZ$        | 0.7200(10)        | 0.7646(10)        |
| 100100 | $+XXZZZ$         | 0.5324(12)        | 0.6019(14)        |
| 000101 | $+ITIXX$         | 0.5748(12)        | 0.6221(13)        |
| 010001 | $+IXZIX$         | 0.5584(12)        | 0.6034(13)        |
Table III. (Continued).

| Input  | Stabilizer | Expectation value | Raw value | SPAM-corrected |
|--------|------------|-------------------|-----------|----------------|
| 010100 | +ZXIXZI    | 0.5900(11)        | 0.6448(12)|                |
| 000110 | +IZZYZ    | 0.5360(12)        | 0.5803(14)|                |
| 100111 | +XZZYXI   | 0.4368(13)        | 0.4862(14)|                |
| 000111 | −IZIXXY   | 0.6112(11)        | 0.6765(12)|                |
| 010111 | −IXIXYX   | 0.5936(11)        | 0.6449(12)|                |
| 001010 | +IZIXZ    | 0.5536(12)        | 0.5984(13)|                |
| 100010 | +XZIZXI   | 0.5392(12)        | 0.5884(13)|                |
| 101000 | +XIXIZX   | 0.5832(11)        | 0.6311(12)|                |
| 001011 | +ZZZIYX   | 0.6084(11)        | 0.6718(12)|                |
| 001100 | +IZYZZ    | 0.7376(10)        | 0.8037(10)|                |
| 101100 | +XIIYZZ   | 0.6316(11)        | 0.6981(12)|                |
| 001101 | +ZZYYIX   | 0.6068(11)        | 0.6712(12)|                |
| 001110 | −IZYXZ    | 0.5924(11)        | 0.6565(13)|                |
| 101110 | −XYIXY    | 0.5308(12)        | 0.5784(13)|                |
| 011111 | +ZZZXXY   | 0.7056(10)        | 0.7970(11)|                |
| 010101 | +IXIXYX   | 0.5386(12)        | 0.5960(13)|                |
| 111110 | −XYXXXY   | 0.6228(11)        | 0.6925(12)|                |
| 010100 | +ZIXYZZ   | 0.4936(12)        | 0.5483(14)|                |
| 011000 | +ZZYYIZ   | 0.5768(12)        | 0.6275(13)|                |
| 011010 | +ZYZXIX   | 0.4676(13)        | 0.5171(14)|                |
| 011001 | +IYYZZX   | 0.4500(13)        | 0.4976(14)|                |
| 011101 | +IYYY    | 0.5200(12)        | 0.5628(13)|                |
| 101101 | +IYYIY    | 0.5584(12)        | 0.6037(13)|                |
| 110110 | +YYIIY    | 0.5016(12)        | 0.5487(13)|                |
| 011100 | −ZYYXZ    | 0.6444(11)        | 0.7192(12)|                |

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