Symmetries, graph properties, and quantum speedups

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Abstract

Aaronson and Ambainis (2009) and Chailloux (2018) showed that fully symmetric (partial) functions do not admit exponential quantum query speedups. This raises a natural question: how symmetric must a function be before it cannot exhibit a large quantum speedup?

In this work, we prove that hypergraph symmetries in the adjacency matrix model allow at most a polynomial separation between randomized and quantum query complexities. We also show that, remarkably, permutation groups constructed out of these symmetries are essentially the only permutation groups that prevent super-polynomial quantum speedups. We prove this by fully characterizing the primitive permutation groups that allow super-polynomial quantum speedups.

In contrast, in the adjacency list model for bounded-degree graphs—where graph symmetry is manifested differently—we exhibit a property testing problem that shows an exponential quantum speedup. These results resolve open questions posed by Ambainis, Childs, and Liu (2010) and Montanaro and de Wolf (2013).
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1 Introduction

One of the most fundamental problems in the field of quantum computing is the question of when quantum algorithms substantially outperform classical ones. While polynomial quantum speedups are known in many settings, super-polynomial quantum speedups are known (or even merely conjectured) for only a few select problems. Crucially, exponential quantum speedups only occur for certain “structured” problems such as period-finding (used in Shor’s factoring algorithm [Sho94]) and Simon’s problem [Sim97], in which the input is known in advance to have a highly restricted form. In contrast, for “unstructured” problems such as black-box search or NP-complete problems, only polynomial speedups are known (and in some models, it can be formally shown that only polynomial speedups are possible).

In this work, we are interested in formalizing and characterizing the structure necessary for fast quantum algorithms. In particular, we study the types of symmetries a function can have while still exhibiting super-polynomial quantum speedups.

1.1 Prior work

Despite the strong intuition in the field that structure is necessary for super-polynomial quantum speedups, only a handful of works have attempted to formalize this and characterize the required structure. All of them study the problem in the query complexity (black-box) model of quantum computation, which is a natural framework in which both period-finding and Simon’s problem can be formally shown to give exponential quantum speedups (see [BdW02] for a survey of query complexity, or [Cle04] for a formalization of period-finding specifically).

In the query complexity model, the goal is to compute a Boolean function \( f : \Sigma^n \rightarrow \{0, 1\} \) using as few queries to the bits of the input \( x \in \Sigma^n \) as possible, where \( \Sigma \) is some finite alphabet. Each query specifies an index \( i \in [n] := \{1, 2, \ldots, n\} \) and receives the response \( x_i \in \Sigma \). A query algorithm, which may depend on \( f \) but not on \( x \), must output \( f(x) \) (with bounded error in the worst-case) after as few queries as possible. Quantum query algorithms are allowed to make queries in superposition, and we are interested in how much advantage this gives them over randomized classical algorithms (for formal definitions of these notions, see [BdW02]).

Beals, Buhrman, Cleve, Mosca, and de Wolf [BBC+01] showed that all total Boolean functions \( f : \Sigma^n \rightarrow \{0, 1\} \) have a polynomial relationship between their classical and quantum query complexities (which we denote \( R(f) \) and \( Q(f) \), respectively). This means that super-polynomial speedups are not possible in query complexity unless we impose a promise on the input: that is, unless we define \( f : P \rightarrow \{0, 1\} \) with \( P \subset \Sigma^n \), and allow an algorithm computing \( f \) to behave arbitrarily on inputs outside of the promise set \( P \). For such promise problems (also called partial functions), provable exponential quantum speedups are known. This is the setting in which Simon’s problem and period-finding reside.

The question, then, is what we can say about the structure necessary for a partial Boolean function \( f \) to exhibit a super-polynomial quantum speedup. Towards this end, Aaronson and Ambainis [AA14] showed that symmetric functions do not allow super-polynomial quantum speedups, even with a promise. Chailloux [Cha18] improved this result by reducing the degree of the polynomial relationship between randomized and quantum algorithms for symmetric functions, and removing a technical requirement on the symmetry of those functions.\(^1\)

Other work attempted to characterize the structure necessary for quantum speedups in alternative ways. Ben-David [Ben16] showed that certain types of symmetric promises do not admit any

\(^1\) Aaronson and Ambainis required the function to be symmetric both under permuting the \( n \) bits of the input, and under permuting the alphabet symbols in \( \Sigma \); Chailloux showed that the latter is not necessary.
function with a super-polynomial quantum speedup, a generalization of \[\text{BBC}+01\] (who showed this when the promise set is \(\Sigma^n\)). Aaronson and Ben-David \[\text{AB16}\] showed that small promise sets, which contain only poly\((n)\) inputs out of \(|\Sigma|^n\), also do not admit functions separating quantum and classical algorithms by more than a polynomial factor.

A class of problems with significant symmetry, though much less than full permutation symmetry, is the class of graph properties. For such problems, the input describes a graph, and the output depends only on the isomorphism class of that graph. Thus the vertices can be permuted arbitrarily, but such a permutation induces a structured permutation on the edges, about which queries provide information.

The setting of graph property testing provides a natural class of partial graph properties. Here we are promised that the input graph either has a property, or is \(\epsilon\)-far from having the property, meaning that we must change at least an \(\epsilon\) fraction of the edges to make the property hold. Graph property testing has been extensively studied since its introduction by Goldreich, Goldwasser, and Ron \[\text{GGR98}\].

The behavior of classical graph property testers can differ substantially depending on the model in which the input graph is specified. For example, in the adjacency matrix model, Alon and Krivelevich \[\text{AK02}\] proved that bipartiteness can be tested in \(\tilde{O}(1/\epsilon^2)\) queries, which is surprisingly independent of the size of the input graph. In contrast, in the adjacency list model for bounded-degree graphs, Goldreich and Ron \[\text{GR97}\] proved that \(\Omega(\sqrt{n})\) queries are needed to test bipartiteness of \(n\)-vertex graphs.

Quantum algorithms for testing properties of bounded-degree graphs in the adjacency list model were studied by Ambainis, Childs, and Liu \[\text{ACL11}\]. They gave upper bounds of \(\tilde{O}(n^{1/3})\) for testing bipartiteness and expansion, demonstrating polynomial quantum speedups. Furthermore, they showed that at least \(\Omega(n^{1/4})\) quantum queries are required to test expansion, ruling out the possibility of an exponential quantum speedup. This work naturally raises the question (also highlighted by Montanaro and de Wolf \[\text{MdW16}\]) of whether there can ever be exponential quantum speedup for graph property testing, or for graph properties more generally.

1.2 Our contributions

In this work, we extend the results of Aaronson-Ambainis and Chailloux to other symmetry groups. We characterize general symmetries of functions as follows.

**Definition 1.** Let \(f:\mathcal{P}\to\{0,1\}\) be a function with \(\mathcal{P}\subseteq\Sigma^n\), where \(\Sigma\) is a finite alphabet and \(n\in\mathbb{N}\). We say that \(f\) is symmetric with respect to a permutation group \(G\) acting on domain \([n]\) if for all \(x\in\mathcal{P}\) and all \(\pi\in G\), the string \(x\circ\pi\) defined by \((x\circ\pi)_i:=x_{\pi(i)}\) satisfies \(x\circ\pi\in\mathcal{P}\) and \(f(x\circ\pi)=f(x)\).

The case where \(G=S_n\), the fully-symmetric permutation group, is the scenario handled in Chailloux’s work \[\text{Cha18}\]: he showed that \(R(f)=O(Q(f)^3)\) if \(f\) is symmetric under \(S_n\). Aaronson and Ambainis \[\text{AA14}\] required an even stronger symmetry property. We note that when \(\Sigma\) is large, say \(|\Sigma|=n\) or larger, the class of functions symmetric under \(S_n\) is already highly nontrivial: among others, it includes functions such as COLLISION, an important function whose quantum query complexity was established in \[\text{AS04}\]; \(k\)-SUM, whose quantum query complexity required the negative-weight adversary to establish \[\text{BS13}\]; and \(k\)-DISTINCTNESS, whose quantum query complexity is still open \[\text{BKT18}\]. Additionally, computational geometry functions such as CLOSESTPAIR (a function studied in recent work by Aaronson, Chia, Lin, Wang, and Zhang \[\text{ACL}^+19\]) are typically symmetric under \(S_n\), as the points are usually represented as alphabet symbols. If we round
the alphabet to be finite, the $G = S_n$ case already shows that no computational geometry problem of this form can have super-cubic quantum speedups.

In this work, we examine what happens when we relax the full symmetry $S_n$ to smaller symmetry groups $G$. We introduce some tools for showing that particular classes of permutation groups $G$ do not allow super-polynomial quantum speedups; that is, we provide tools for showing that every $f$ symmetric with respect to $G$ satisfies $Q(f) = R(f)^{\Omega(1)}$. Our first main result is the following theorem, in which $G$ is the graph symmetry: the permutation group acting on strings of length $\binom{n}{2}$ (which represent the possible edges of a graph), which includes all permutations of the edges which are induced by one of the $n!$ relabelings of the $n$ vertices. Functions that take the adjacency matrix of a graph as input, and whose output depends only on the isomorphism class of the graph (not on the labeling of its vertices), are symmetric with respect to the graph symmetry $G$.

**Theorem 2** (Informal version of Corollary 38). Any Boolean function $f$ defined on the adjacency matrix of a graph (and symmetric with respect to renaming the vertices of the graph) has $R(f) = O(Q(f)^6)$. This holds even if $f$ is a partial function.

This theorem holds even when the alphabet of $f$ is non-Boolean. We note that this is a strict generalization of the result of Aaronson and Ambainis [AA14], since any fully-symmetric function is necessarily symmetric under the graph symmetry as well. It is also a generalization of [Cha18], except that our polynomial degree (power 6) is larger than the power 3 of Chailloux. Our results also extend to other types of graph symmetries, including hypergraph symmetries and bipartite graph symmetries.

Next, we extend **Theorem 2** to give a more general characterization of which classes of symmetries are inconsistent with super-polynomial quantum speedups. Our main result in this regard is a dichotomy theorem for primitive permutation groups. Primitive permutation groups are sometimes described as the “building blocks” of permutation groups, because arbitrary permutation groups can always be decomposed into primitive groups (in a certain formal sense). We give a complete classification of which primitive groups $G$ allow super-polynomial quantum speedups and which do not, in terms of the minimal base size $b(G)$. The minimal base size is a key quantity in computational group theory that roughly captures the size of a permutation group $G$ relative to the number of points it acts on. Thus, the following theorem amounts to showing that symmetries of “large” primitive permutation groups do not allow large quantum speedups, while symmetries of “small” primitive permutation groups do.

**Theorem 3** (Informal version of Corollary 49). Let $G$ be a primitive permutation group acting on $[n]$, and let $b(G)$ denote the size of a minimal base for $G$. If $b(G) = n^{\Omega(1)}$, then $R(f) = Q(f)^{O(1)}$ for every $f$ that is symmetric under $G$. Otherwise, if $b(G) = n^{o(1)}$, then there exists a partial function $f$ that is symmetric under $G$ such that $R(f) = Q(f)^{o(1)}$.

By decomposing an arbitrary permutation group into primitive groups, we can extend the above theorem to show that if a permutation group $G$ does not allow super-polynomial speedups, then it must be constructed out of a constant number of primitive groups $H$ that satisfy $b(H) = n^{\Omega(1)}$, where $H$ acts on $n$ points. In the other direction, if any of the primitive factors $H$ of $G$ satisfy $b(H) = n^{o(1)}$, or if $G$ contains more than a constant number of primitive factors, then we exhibit a function that is symmetric under $G$ with a super-polynomial quantum speedup.

Remarkably, the proof of **Theorem 3** also includes a strong characterization of what the primitive permutation groups $G$ that satisfy $b(G) = n^{\Omega(1)}$ look like. Indeed, we show that as a consequence of the classification of finite simple groups, all such groups essentially look like hypergraph symmetries or minor extensions thereof. Thus, in some sense, permutation groups built out of hypergraph
symmetries are the only permutation groups that are inconsistent with super-polynomial speedups. We consider this one of the most surprising consequences of our work; a priori, one could expect there to be many different kinds of symmetries that disallow super-polynomial speedups.

Finally, we return to the topic of graph properties. While the aforementioned results show that no exponential speedup is possible for graph properties in the adjacency matrix model, they do not resolve the original open question of Ambainis, Childs, and Liu [ACL11], which specifically addressed graph property testing in the adjacency list model. Graph symmetries in this model manifest themselves in a different way that is not captured by Definition 1, so Theorem 2 does not apply. In this model, we show the following,

**Theorem 4** (Informal version of Theorem 56). There exists a graph property testing problem in the adjacency list model for which there is an exponential quantum speedup.

This is in stark contrast to the situation in the adjacency matrix case. Together, Theorem 2 and Theorem 4 fully settle the open question about the possibility of exponential quantum speedup for graph properties, showing that its answer is highly model-dependent: exponential speedup is impossible in the adjacency matrix model, but is possible in the adjacency list model, even in the restrictive setting of graph property testing.

### 1.3 Techniques

Here we briefly discuss the techniques used to prove the main results.

#### 1.3.1 Well-shuffling symmetries do not allow speedups

Our analysis begins with a basic observation of Chailloux [Cha18]. Suppose that $f$ is symmetric under the full symmetric group $S_n$ acting on $[n]$. Zhandry [Zha15] showed that distinguishing a random element of $S_n$ from a random small-range function $\alpha: [n] \rightarrow [n]$ with $|\alpha([n])| = r$ requires $\Omega(r^{1/3})$ quantum queries.\(^3\) Now, if $Q$ is a quantum algorithm solving $f$ using $T$ queries, then $Q$ also outputs $f(x)$ on input $x \circ \pi$ (the input $x$ with bits shuffled according to $\pi$) for any $\pi \in S_n$, since $f$ is symmetric under $S_n$. In particular, $Q(x \circ \pi)$ for a random $\pi \in S_n$ still outputs $f(x)$. However, the $T$-query algorithm $Q$ cannot distinguish a random $\pi \in S_n$ from a random function $\alpha: [n] \rightarrow [n]$ with range $r \approx T^3$. Therefore $Q$ must output $f(x)$ with at most constant error even when run on $x \circ \alpha$ for a random small-range function $\alpha$. This property can be used to simulate $Q$ classically: a classical algorithm $R$ can simply sample a small-range function $\alpha$, explicitly query the entire string $x \circ \alpha$ (using only $O(T^3)$ queries since $\alpha$ has range $O(T^3)$), and then simulate $Q$ on the string $x \circ \alpha$. This is an $O(T^3)$-query classical algorithm for computing $f$, created from a $T$-query quantum algorithm for $f$.

The above trick can be generalized from the fully-symmetric group $S_n$ to other permutation groups $G$, provided we can show that it is hard for a $T$-query quantum algorithm to distinguish $G$ from a function with range poly($T$). The question of whether there exists an arbitrary symmetric function $f$ with a quantum speedup is therefore reduced to the question of whether the concrete task of distinguishing $G$ from the set of all small-range functions can be done quickly using a quantum algorithm. That is, if $D_{n,r}$ is the set of all strings in $[n]^n$ that use only $r$ unique symbols, then

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\(^2\)Given a graph $x: [n] \times [d] \rightarrow [n] \cup \{\ast\}$ in the adjacency list model with $n$ vertices and bounded degree $d$, its isomorphism class essentially consists of graphs of the form $\pi^{-1} \circ x \circ (\pi \times \text{id}_d)$ where $\pi \in S_n$ is a relabeling of vertices and $\pi^{-1}$ is defined as the inverse of $\pi$ but also maps $\ast$ to $\ast$. A function $f$ is a graph property in this model if and only if its domain $P$ is a union of such isomorphism classes and $f$ is constant on each isomorphism class.

\(^3\)Actually, we show that a version of Zhandry’s result that is sufficient for our purposes follows easily from the collision lower bound, so his techniques are not necessary for our results.
we care about the quantum query cost of distinguishing $D_{n,r}$ from $G$; if this cost is $r^{\Omega(1)}$, then no function that is symmetric under $G$ can exhibit a super-polynomial quantum speedup. In this case, we call $G$ well-shuffling.

We show that the well-shuffling property is preserved under various operations one might perform on a permutation group, which allows us to prove that many permutation groups are well-shuffling simply by reduction to $S_n$. As an example, for undirected graph properties, the relevant transformation is from a permutation group $G$ acting on a set $[n]$ to the induced action $H$ on the $\binom{n}{2}$ unordered pairs of $[n]$. We then show that if $G$ is well-shuffling, then $H$ is also well-shuffling, because the problem of distinguishing $G$ from a small range function has a reduction to the problem of distinguishing $H$ from a small range function. More generally, this reduction works for ordered pairs or for tuples of any constant length.

1.3.2 Classification of symmetries with and without speedups

To characterize which permutation groups allow super-polynomial quantum speedups, we first show that large quantum speedups exist for any sufficiently “small” permutation group. In particular, if the minimal base size $b(G)$ of a permutation group $G$ acting on $n$ points satisfies $b(G) = n^{\omega(1)}$, then we exhibit a partial function that is symmetric under $G$ with a super-polynomial quantum speedup. We construct such a function by starting with an arbitrary promise problem that has a sufficiently large separation between randomized and quantum query complexities (e.g., Simon’s problem [Sim97] or Forrelation [AA15]). We then modify it to be symmetric under $G$ in a way that preserves the super-polynomial separation between randomized and quantum query complexities.

One might wonder whether a sort of converse holds, i.e., whether $b(G) = n^{\Omega(1)}$ implies that $G$ is well-shuffling, and therefore that all partial functions symmetric under $G$ admit at most a polynomial quantum speedup. We show this for primitive permutation groups $G$, essentially because the structure of “large” primitive groups is very well-understood. We base our proof on a theorem due to Liebeck [Lie84] that characterizes minimal base sizes of primitive groups via the classification of finite simple groups. Liebeck’s theorem allows us to show that primitive groups $G$ with $b(G) = n^{\Omega(1)}$ can all be constructed out of the symmetric group via transformations that preserve the well-shuffling property. As a result, the quantity $b(G)$ completely characterizes whether a primitive permutation group admits a super-polynomial speedup. Moreover, the transformations involved are the same transformations we use to show that hypergraph symmetries are well-shuffling, so primitive groups that satisfy $b(G) = n^{\Omega(1)}$ all essentially look like hypergraph symmetries.

To go beyond primitive groups, we use the fact that an arbitrary permutation group can be decomposed into transitive groups (by looking at its orbits), which can then be decomposed into primitive groups (by looking at its nontrivial block systems). We show that for a group $G$, if any of the primitive factors in such a decomposition of $G$ are consistent with super-polynomial speedups, or if the number of primitive factors is $\omega(1)$, then there exists a partial function that is symmetric under $G$ with a super-polynomial quantum speedup. These functions are generally easy to construct, though in one case that involves tree-like symmetries, we make essential use of the 1-Fault DIRECT TREES problem of Zhan, Kimmel, and Hassidim [ZKH12]: a partial function that has a super-polynomial quantum speedup, constructed by adding a promise to a Boolean evaluation tree in a way that preserves the symmetries of the tree. Interestingly, these results also show that the quantity $b(G)$ does not determine the well-shuffling property in general, because there exist permutation groups $G$ acting on $n$ points with $b(G) = \Omega(n)$ that are consistent with exponential quantum speedups. Nevertheless, we can at least say that if a permutation group $G$ is inconsistent with super-polynomial quantum speedups, then it must be decomposable into $O(1)$ well-shuffling primitive groups.
Finally, we show that the possibility of exponential quantum speedups for graph properties depends strongly on the input model: while the adjacency matrix model allows at most polynomial speedups, we give an example of an exponential speedup in the adjacency list model for bounded-degree graphs. In particular, we show that such a separation holds even in the more restrictive setting of property testing, where “no” instances are promised to be far from “yes” instances.

The main idea for this separation comes from work by Childs et al. [CCD+03], demonstrating an exponential algorithmic speedup by quantum walk. More specifically, they design a “welded trees” graph that consists of two depth-$k$ binary trees welded together by an alternating cycle on the leaves of the two trees, with the two degree-2 vertices (roots) referred to as “entrance” and “exit”. The main result of [CCD+03] is that a quantum computer, given the entrance, can find the exit in time $\text{poly}(k)$, whereas a classical computer needs exponential time $2^{\Omega(k)}$ to find the exit. However, it is not immediately clear how to lift this separation to the realm of graph property testing: it is not even a graph property (since the entrance is given and the exit can be labeled differently in isomorphic graphs), and it is still more challenging to find a related classically-hard property testing problem.

Our approach is to leverage the quantum advantage for finding entrance-exit pairs to test whether the given graph has a certain form derived from welded trees. The difficulty is that it is unclear how finding entrance-exit pairs helps a quantum computer test the structure of a welded trees. One of our main observations is that once the vertices in the weld of the binary trees are marked, it becomes easy to test the entire structure: we can separately test the two binary trees, and then test the weld itself. However, marking the weld vertices also enables a classical computer to find entrance-exit pairs, so it seems that this approach has no quantum advantage.

Our resolution is to use many copies of the welded trees structure and randomly assign a bit to every root (degree-2) vertex. Then we mark vertices by connecting every non-root vertex $v$ to a root $r$ via an “advice edge”, such that

1. if $v$ is in the weld, then the sum of the bits assigned to $r$ and its pair is odd, and
2. if $v$ is not in the weld, then the sum of the bits assigned to $r$ and its pair is even.

To keep the degree bounded, instead of directly connecting advice edges to the roots, we attach a binary tree “antenna” to the roots and connect the advice edges there, as shown in Figs. 1 and 2.

Still, it might seem that this construction has a fatal chicken-or-egg problem: we need to mark the weld vertices to test the structure, but then marking the weld vertices enables also a classical computer to find the root pairs, leaving no quantum advantage. However, this is a feature, not a bug. The chicken-or-egg problem has two fixed points, with neither a chicken nor an egg (corresponding to classical strategies for our problem) or with both a chicken and an egg (corresponding to the quantum case). The key point is that the marking of the weld vertices can only be read efficiently by a quantum computer.

To show that this property is hard to test classically, we show that it is exponentially difficult to distinguish a randomly selected graph of the form described above, from a graph that looks like the above, except the binary trees are self-welded, i.e., the leaves of the binary trees are connected with a random cycle only containing vertices from the same binary tree (this then disconnects the root pairs, so the advice edges no longer have the original role and are chosen arbitrarily). The intuition behind our classical lower bound proof is the same as that behind the lower bound proof of [CCD+03]: given a root of a (self)-welded trees graph, the graph appears to be an exponentially deep binary tree for a (random) classical walker who can only explore the graph locally. Finally, we show that a random welded trees graph is $\Omega(1)$-far from a random self-welded trees graph by...
observing that the welded trees graph is bipartite, whereas the self-welded trees graph is $\Omega(1)$-far from being bipartite with very high probability.

1.4 Organization

The remainder of the paper is organized as follows.

In Section 2 we introduce some basic notions from query complexity, the theory of permutation groups, and symmetric functions. Section 3 introduces the notion of well-shuffling permutation groups and shows that they do not allow super-polynomial quantum speedups. In Section 4 we present techniques for showing that permutation groups are well-shuffling, establishing in particular that hypergraph symmetries cannot have super-polynomial quantum speedups. In Section 5 we classify well-shuffling permutation groups, showing that hypergraph symmetries are essentially the only groups that are inconsistent with super-polynomial speedups. In Section 6 we present an exponential separation between classical and quantum property testing in the adjacency list model for bounded-degree graphs. Finally, in Section 7 we briefly discuss some open problems. The appendices present two technical results: Appendix A gives a proof of the quantum minimax lemma, and Appendix B establishes a super-polynomial quantum speedup for deep wreath product symmetries.

2 Preliminaries

2.1 Query complexity

We begin by introducing some standard notation from query complexity. A Boolean function is a \{0, 1\}-valued function $f$ on strings of length $n \in \mathbb{N}$. We use $\text{Dom}(f)$ to denote the domain of $f$, and we always have $\text{Dom}(f) \subseteq \Sigma^n$ where $\Sigma$ is a finite alphabet. The function $f$ is called total if $\text{Dom}(f) = \Sigma^n$; otherwise it is called partial.

For a (possibly partial) Boolean function $f$, we use $R_\epsilon(f)$ to denote its randomized query complexity with error $\epsilon$, as defined in [BdW02]. This is the minimum number of queries required in the worst case by a randomized algorithm that computes $f$ with worst-case error $\epsilon$. We use $Q_\epsilon(f)$ to denote the quantum query complexity with error $\epsilon$ of $f$, also defined in [BdW02]. This is the minimum number of queries required in the worst case by a randomized algorithm that computes $f$ with worst-case error $\epsilon$. When $\epsilon = 1/3$, we omit it and simply write $R(f)$ and $Q(f)$.

An important tool for lower bounding query complexity is the minimax theorem, the original version of which was given by Yao for zero-error (Las Vegas) randomized algorithms [Yao77]. Here we use a bounded-error, quantum version of the minimax theorem. Bounded-error versions of the minimax theorem can be shown using linear programming duality (see also [Ver98] who proved a minimax theorem in the setting where both the error and the expected query complexity are measured against the same hard distribution). A similar technique works for quantum query complexity; this result is folklore, and we prove it in Appendix A.

Lemma 5 (Minimax for bounded-error quantum algorithms). Let $f$ be a (possibly partial) Boolean function with $\text{Dom}(f) \subseteq \Sigma^n$, and let $\epsilon \in (0, 1/2)$. Then there is a distribution $\mu$ supported on $\text{Dom}(f)$ which is hard for $f$ in the following sense: any quantum algorithm using fewer than $Q_\epsilon(f)$ quantum queries for computing $f$ must have average error more than $\epsilon$ on inputs sampled from $\mu$.

Note that achieving average error $\epsilon$ against a known distribution $\mu$ is always easier than achieving worst-case error $\epsilon$; the minimax theorem says that there is a hard distribution against which achieving average error $\epsilon$ is just as hard as achieving worst-case error $\epsilon$.
2.2 Permutation groups

We review some basic definitions about permutation groups, following Hulpke [Hul10].

Definition 6 (Permutation group). A permutation group is a pair $(D, G)$ where $D$ is a set and $G$ is a set of bijections $\pi : D \to D$, such that $G$ forms a group under composition (i.e., $G$ contains the identity function and is closed under composition and inverse of the bijections). In this case, $G$ is said to act on $D$. We will often denote a permutation group simply by $G$, with the domain $D$ being implicit.

In other words, a permutation group is simply a set of permutations of a domain $D$ which is closed under composition and inverse. In this work we will generally take $D = [n]$, where $[n]$ denotes the set $\{1, 2, \ldots, n\}$ for $n \in \mathbb{N}$. The set $[n]$ will represent the indices of an input string, or equivalently, the queries an algorithm is allowed to make.

We define orbits, transitivity, and stabilizers of permutation groups, all of which are standard.

Definition 7 (Orbit). Let $G$ be a permutation group on domain $D$, and let $i \in D$. Then the orbit of $i$ is the set $\{\pi(i) : \pi \in G\}$. A subset of $D$ is an orbit of $G$ if it is the orbit of some $i \in D$ with respect to $G$.

Definition 8 (Transitivity). We say that a permutation group $G$ on domain $D$ is $k$-transitive if for all distinct $i_1, i_2, \ldots, i_k \in D$ and distinct $j_1, j_2, \ldots, j_k \in D$, there exists some $\pi \in G$ such that $\pi(i_t) = j_t$ for all $t = 1, 2, \ldots, k$. A group that is $1$-transitive is called simply transitive.

Definition 9 (Stabilizer). Let $G$ be a permutation group on domain $D$. The pointwise stabilizer of a set $S \subseteq D$ is the subgroup $\text{Stab}_G(S) := \{\pi \in G : \pi(i) = i \ \forall i \in S\}$. The setwise stabilizer of a set $S \subseteq D$ is the subgroup $\text{Stab}_G(S) := \{\pi \in G : \pi(i) \in S \ \forall i \in S\}$.

We use some facts about the structure of permutation groups. For this, we need some additional definitions, starting with the notions of primitive groups and wreath products.

Definition 10 (Block system). Let $G$ be a permutation group on domain $D$. A partition $\mathcal{B} = \{B_1, B_2, \ldots, B_k\}$ of $D$ is called a block system for $G$ if $\mathcal{B}$ is $G$-invariant. Formally, this means that $\pi(B_i) \in \mathcal{B}$ for every $\pi \in G$ and $i \in [k]$, where $\pi(B_i) := \{\pi(x) : x \in B_i\}$. The sets $B_i$ are called blocks for $G$.

As an example, consider the action of the group of linear transformations on the nonzero elements of a vector space. The set of lines through the origin (i.e., nonzero scalar multiples of a nonzero vector) form a block system.

It follows from this definition that if $\mathcal{B}$ is a block system for $G$, then for every block $B_i$ and permutation $\pi \in G$, either $\pi(B_i) = B_i$ or $\pi(B_i) \cap B_i = \emptyset$. Moreover, if $G$ acts transitively, then all blocks have the same size.

Definition 11 (Primitive group). Let $G$ be a transitive permutation group on domain $D$. $G$ is called primitive if the only block systems for $G$ are $\{D\}$ and the partition of $D$ into singletons. A permutation group that is not primitive is called imprimitive.

Primitive groups are sometimes described as the “building blocks” of permutation groups, for reasons we will see later.

Definition 12. Let $G$ be an abstract group. The direct power of $G$ with exponent $m$, denoted $G^{\times m}$, is the direct product of $m$ copies of $G$:

$$G^{\times m} := G \times \cdots \times G,$$

$m$ times.
If $G$ is a permutation group acting on $[n]$, there are two natural ways to construct $G^{\times m}$ as a permutation group. Let $\pi = (\pi_1, \pi_2, \ldots, \pi_m) \in G^{\times m}$. The first action is on $[m] \times [n]$ where we have $\pi(i, j) = \pi_i(j)$. The second action is on $[n]^m$ where we have $\pi(i_1, i_2, \ldots, i_m) = (\pi_1(i_1), \pi_2(i_2), \ldots, \pi_m(i_m))$. These two actions of $G^{\times m}$ give rise to two actions of the wreath product of two groups:

**Definition 13** (Wreath product). Let $G$ and $H$ be permutation groups acting on $[m]$ and $[n]$, respectively. The wreath product of $G$ and $H$, denoted $G \wr H$, is one of two permutation groups:

(i) The imprimitive action is the action on $[m] \times [n]$ generated by the natural action of $H^{\times m}$, along with permutations of the form $(i, j) \rightarrow (\sigma(i), j)$ for every $\sigma \in G$.

(ii) The product action is the action on $[n]^m$ generated by the natural action of $H^{\times m}$, along with permutations of the form $(i_1, i_2, \ldots, i_m) \rightarrow (i_{\sigma(1)}, i_{\sigma(2)}, \ldots, i_{\sigma(m)})$ for every $\sigma \in G$.

**Fact 14.** Let $G$ and $H$ be permutation groups acting on $[m]$ and $[n]$, respectively. If $n > 1$, then the imprimitive action and the primitive action of $G \wr H$ are isomorphic as groups, and both have order $|G| \cdot |H|^m$.

If it is clear from context which of the two actions we are referring to, then we may sometimes simply write $G \wr H$ without specifying the type of action.

This definition differs from the standard definition of the wreath product, which is an abstract group that can be defined even if $H$ is an abstract group. However, the definitions agree when $H$ is a permutation group, which is the only case we consider.

The wreath product imprimitive action carries a useful intuitive interpretation: if $g$ and $h$ are functions that are symmetric under permutation groups $G$ and $H$, respectively, then $G \wr H$ is the (visible)\(^4\) group of symmetries of the block-composed function $g \circ h$. Thus, the wreath product imprimitive action is associative.

Note that the wreath product is not commutative. This can be confusing because some authors instead denote this wreath product as $H \wr G$. We prefer the notation used here because of the correspondence to block composition of functions.

**Definition 15** (Base). Let $G$ be a permutation group on domain $D$. A base for $G$ is a set $S \subseteq D$ such that the pointwise stabilizer $\text{Stab}_G(S)$ is the trivial group. The minimal base size for $G$, denoted $b(G)$, is the size of the smallest base for $G$.

Bases of permutation groups are important in computational group theory for the following reason: if $S$ is a base for $G$, then a permutation $\pi \in G$, viewed as a function $D \rightarrow D$, is uniquely determined by its restriction to $S$. This can be useful particularly when $S$ is small compared to $D$.

In many applications, the minimal base size is a useful proxy for the size of a permutation group. This is quantified by the following well-known lemma, which we prove for completeness:

**Lemma 16.** Let $G$ be a permutation group on $[n]$. Then the order of the group $|G|$ and the minimal base size $b(G)$ satisfy $2^{b(G)} \leq |G| \leq n^{b(G)}$. Equivalently, $\log_2(|G|) \leq b(G) \leq \log_2(n^{b(G)})$.

\(^4\)In rare cases, depending on $g$ and $h$, there can be more symmetries. For example, let $g = h = \text{OR}_n$, which both have symmetry group $S_n$. The wreath product $S_n \wr S_n$ is not the same as $S_n \wr 2$, which is the group of symmetries of $g \circ h = \text{OR}_n$. However, we can say that $g \circ h$ is guaranteed to be at least symmetric under $G \wr H$ for any functions $g$ and $h$. 

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Proof. Without loss of generality, identify a minimal base for $G$ with $[k]$ where $k = b(G)$. Let $H_i = \text{Stab}_G([i])$ be the pointwise stabilizer of the first $i$ points of the base, with the understanding that $H_0 = G$. We have the inclusion

$$1 = H_k \subseteq H_{k-1} \subseteq \cdots \subseteq H_1 \subseteq H_0 = G.$$ 

Let $O_i \subseteq [n]$ denote the orbit of $i$ in $H_{i-1}$. By the orbit-stabilizer theorem, we have $|H_{i-1}| = |H_i| \cdot |O_i|$. Thus, we have $|G| = \prod_{i=1}^k |O_i|$. We also know that $2 \leq |O_i| \leq n$, where the upper bound is trivial, and the lower bound holds because $[k]$ was assumed to be a minimal base. We get the desired bound on $|G|$ by bounding each term in the product. \hfill \square

### 2.3 Symmetric functions

We introduce some notation that is used throughout the paper to talk about symmetric functions.

**Definition 17** (Permuting strings). Let $\pi$ be a permutation on $[n]$, and let $x \in \{0, 1\}^n$. We write $x \circ \pi$ to denote the string whose characters have been permuted by $\pi$; that is, $(x \circ \pi)_i := x_{\pi(i)}$. More generally, $x \circ \pi$ is similarly defined when $\pi$ is merely a function $[n] \rightarrow [n]$ rather than a permutation.

Note that if we view a string $x \in \Sigma^n$ as a function $[n] \rightarrow \Sigma$ with $x(i) := \alpha_i$, then $x \circ \pi$ is simply the usual function composition of $x$ and $\pi$. This notation allows us to easily define symmetric functions.

**Definition 18** (Symmetric function). Let $G$ be a permutation group on $[n]$, and let $f$ be a (possibly partial) Boolean function with $\text{Dom}(f) \subseteq \Sigma^n$. We say $f$ is symmetric under $G$ if for all $x \in \text{Dom}(f)$ and all $\pi \in G$ we have $x \circ \pi \in \text{Dom}(f)$ and $f(x \circ \pi) = f(x)$.

For asymptotic bounds such as $Q(f) = R(f)^{\Omega(1)}$ to be well-defined, we need to talk about classes of functions rather than individual functions. To do that, we need to talk about classes of permutation groups. We introduce the following definition, which defines, for a class of permutation groups $\mathcal{G}$, the set of all functions symmetric under some group in $\mathcal{G}$. We denote this set by $F(\mathcal{G})$.

**Definition 19** (Class of symmetric functions). Let $\mathcal{G} = \{G_i\}_{i \in I}$ be a (possibly infinite) set of finite permutation groups, with $G_i$ acting on $[n_i]$ for each $i \in I$. Here $I$ is an arbitrary index set and $n_i \in \mathbb{N}$ for all $i \in I$. Then define $F(\mathcal{G})$ to be the set of all (possibly partial) Boolean functions that are symmetric under some $G_i$. That is, we have $f \in F(\mathcal{G})$ if and only if $f : \text{Dom}(f) \rightarrow \{0, 1\}$ is a function with $\text{Dom}(f) \subseteq [m]^n$ for some $n, m \in \mathbb{N}$, and $f$ is symmetric under $G_i$ for some $i \in I$ such that $n_i = n$. (Here $[m]$ represents the alphabet $\Sigma$.)

### 3 Well-shuffling permutation groups

In this section we first define the notion of a well-shuffling class of permutation groups, which is a class $\mathcal{G}$ of permutation groups $G$ that are hard to distinguish from the set of small-range functions via a quantum query algorithm. We then show that a well-shuffling class of permutation groups does not allow super-polynomial quantum speedups. This result (Theorem 23) converts the task of showing permutation groups do not allow quantum speedups into the task of showing those permutation groups are well-shuffling, a much simpler objective.

We start by defining the set of small-range strings $D_{n,r}$.

**Definition 20** (Small-range strings). For $n, r \in \mathbb{N}$, let $D_{n,r}$ be the set of all strings $\alpha$ in $[n]^n$ for which the number of unique alphabet symbols in $\alpha$ is at most $r$. 

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We identify a string $\alpha \in [n]^n$ with a function $[n] \to [n]$. Then $D_{n,r}$ is the set of all functions $[n] \to [n]$ with range size at most $r$. Next, we define $\text{cost}(G, r)$ as the quantum query complexity of distinguishing $G$ from $D_{n,r}$ (where $G$ is a permutation group acting on $[n]$).

**Definition 21** (Cost). Identify a permutation on $[n]$ with a string in $[n]^n$ in which each alphabet symbol occurs exactly once. Then a permutation group $G$ on $[n]$ corresponds to a subset of $[n]^n$. For $r < n$, let $\text{cost}_r(G, r)$ be the minimum number of quantum queries needed to distinguish $G$ from $D_{n,r}$ to worst-case error $\epsilon$; that is, $\text{cost}_r(G, r) := Q_\epsilon(f)$, where $f$ has domain $G \cup D_{n,r} \subseteq [n]^n$ and is defined by $f(x) = 1$ if $x \in G$ and $f(x) = 0$ if $x \in D_{n,r}$. When $r \geq n$, we set $\text{cost}_r(G, r) := \infty$.

We note that since $\text{cost}_r(G, r)$ is defined as the worst-case quantum query complexity of a Boolean function, it satisfies amplification, meaning that the precise value of $\epsilon$ does not matter so long as it is a constant in $(0, 1/2)$ and so long as we do not care about constant factors.

We define a well-shuffling class of permutation groups as follows.

**Definition 22** (Well-shuffling permutation groups). Let $\mathcal{G}$ be a collection of permutation groups. We say $\mathcal{G}$ is well-shuffling if $\text{cost}(G, r) = r^{\Omega(1)}$ for $G \in \mathcal{G}$ and $r \in \mathbb{N}$. More explicitly, we say $\mathcal{G}$ is well-shuffling with power $a > 0$ if there exists $b > 0$ such that $\text{cost}(G, r) \geq r^{1/a}/b$ for all $G \in \mathcal{G}$ and all $r \in \mathbb{N}$.

We note that $\text{cost}(G, r) \geq r^{1/a}/b$ is always satisfied when $r$ is greater than or equal to the domain size of $G$, since in that case $\text{cost}(G, r) = \infty$. Hence to show well-shuffling we only need to worry about $r$ smaller than $n$, the domain size of the permutation group $G$.

The following theorem plays a central role in this work: it shows that a well-shuffling collection of permutation groups does not allow super-polynomial quantum speedups.

**Theorem 23.** Let $f : \text{Dom}(f) \to \{0, 1\}$ be a partial Boolean function on $n \in \mathbb{N}$ bits, with $\text{Dom}(f) \subseteq \Sigma^n$ (where $\Sigma$ is a finite alphabet). Let $G$ be a permutation group on $[n]$, and suppose that $f$ is symmetric under $G$. Then there is a universal constant $c \in \mathbb{N}$ such that

$$R(f) \leq \min\{r \in \mathbb{N} : \text{cost}(G, r) \geq cQ(f)\}.$$ 

Consequently, if $\mathcal{G}$ is a well-shuffling collection of permutation groups with power $a$, then for all $f \in F(\mathcal{G})$ we have $R(f) = O(Q(f)^a)$.

To prove this theorem, we use the following minimax theorem for the cost measure.

**Lemma 24** (Minimax for cost). Let $r, n \in \mathbb{N}$ satisfy $r < n$, let $\epsilon \in [0, 1/2)$, and let $G$ be a permutation group on $[n]$. Then there is a distribution $\mu$ on $D_{n,r}$ that is hard in the following sense. Let $\mu'$ be the uniform distribution on $G \subseteq [n]^n$. Then any quantum algorithm for distinguishing $G$ from $D_{n,r}$ which uses fewer than $Q_\epsilon(f) = \text{cost}_\epsilon(G, r)$ queries must either make error $> \epsilon$ on average against $\mu$, or else make error $> \epsilon$ against $\mu'$ (i.e., it fails to distinguish $\mu$ from the uniform distribution on $G$).

**Proof.** Let $f$ be the function that asks to distinguish $G$ from $D_{n,r}$ in the worst case. Then by the minimax theorem (Lemma 5), there is a hard distribution $\nu$ for $f$, such that any quantum algorithm using fewer than $Q_\epsilon(f) = \text{cost}_\epsilon(G, r)$ queries must make more than $\epsilon$ error against $\nu$. Let $\nu'$ be the distribution we get by applying a uniformly random permutation from $G$ to a sample from $\nu$. Then $\nu'$ is still a hard distribution for $f$. Indeed, if it were not a hard distribution, there would be some quantum algorithm $Q$ solving $f$ against $\nu'$ using too few queries; but in that case, we could
design an algorithm $Q'$ for solving $f$ against $\nu$ simply by taking the input $x$, implicitly applying a uniformly random $\pi$ from $G$ to permute the bits of $x$ (this can be done without querying $x$, simply by redirecting all future queries $i \in [n]$ through the permutation $\pi$), and then running $Q$ on the permuted string.

Now, note that composing a uniformly random permutation from $G$ with an arbitrary (fixed) permutation from $G$ gives a uniformly random permutation from $G$. This means that $\nu'$ is some mixture of the uniform distribution $\mu'$ on $G \subseteq [n]^n$ and another distribution $\mu$ on $D_{n,r}$. Then any algorithm that succeeds on both $\mu$ and $\mu'$ with error at most $\epsilon$ will also succeed on $\nu'$ with error at most $\epsilon$, from which the desired result follows.

Using this lemma, we now prove Theorem 23.

**Proof of Theorem 23.** Let $Q$ be a quantum algorithm for $f$ that uses $Q(f)$ queries. Amplify it to $Q'$ by repeating $3$ times and taking the majority vote; then it uses $3Q(f)$ queries and makes worst-case error $7/27$ instead of $1/3$. Using Lemma 24, let $\mu$ be the hard distribution on $D_{n,r}$ which is hard to distinguish from $G$ to error $\epsilon$, where we pick $r$ later and pick $\epsilon$ to be a constant close to $1/2$. Sample $\alpha$ from $\mu$, and consider the string $x \circ \alpha$ with $(x \circ \alpha)_i = x_{\alpha(i)}$.

Now, since $f$ is invariant under $G$, $Q'$ outputs $f(x)$ with error at most $7/27$ when run on $x \circ \pi$ for each $\pi \in G$. In particular, consider picking $\pi$ from $G$ uniformly at random, and running $Q'$ on $x \circ \pi$ where the string $x \in \text{Dom}(f)$ is fixed. Compare this to the behavior of $Q'$ on $x \circ \alpha$, where $\alpha$ is sampled from the hard distribution $\mu$ on $D_{n,r}$.

If $Q'$ did not output $f(x)$ on $x \circ \alpha$ with error at most $1/3$, then we could convert $Q'$ to an algorithm distinguishing $\pi$ from $\alpha$ with constant error. This is because $Q'$ outputs $f(x)$ with error at most $7/27 < 1/3$ on input $x \circ \pi$; hence $Q'$ behaves differently when run on $x \circ \alpha$ and on $x \circ \pi$. We can convert $Q'$ to an algorithm $Q''$ which hard codes the input $x$, and receives either a random $\pi$ from $G$ or a random $\alpha$ from $\mu$ as input. This algorithm makes only $3Q(f)$ queries to $\pi$ or $\alpha$, but its acceptance probability differs by a constant gap between the two distributions, which (using some standard re-balancing) we can use to distinguish $G$ from $\mu$ with at most constant error.

Now, assuming the distribution $\mu$ was picked to be hard enough (i.e., $\epsilon$ was chosen sufficiently close to $1/2$), this means that $3Q(f)$, the query cost of $Q''$, is at least $\text{cost}_x(G, r)$. Since $\text{cost}_x(G, r)$ is the worst-case quantum query complexity of a Boolean function, it can be amplified. We conclude that if $Q'$ failed to output $f(x)$ on input $x \circ \alpha$ (with $\alpha \leftarrow \mu$) with error at most $1/3$, then we have $Q(f) = \Omega(\text{cost}(G, r))$, that is, $Q(f) > \text{cost}(G, r)/c$ for some universal constant $c$ (from amplification).

Now assume that $Q(f) \leq \text{cost}(G, r)/c$. Then $Q'$ has error at most $1/3$ for computing $f(x)$ when run on $\alpha(x)$, with $\alpha$ chosen from $\mu$. Since $\alpha \in D_{n,r}$ uses at most $r$ alphabet symbols, a randomized algorithm can simulate $Q'$ simply by picking $\alpha$ from $\mu$ and querying all the $r$ bits of $x$ used in the string $x \circ \alpha$, fully determining that string. This algorithm $R$ uses $r$ queries, and makes at most $1/3$ error, so we conclude that $R(f) \leq r$.

By correctly picking $r$, we conclude that $R(f) \leq \min\{r \in \mathbb{N} : \text{cost}(G, r) \geq cQ(f)\}$, as desired. Finally, note that if $\text{cost}(G, r) \geq r^{1/a}/b$, then by picking $r = (bcQ(f))^a$ we get $\text{cost}(G, r) \geq cQ(f)$. From this it follows that $R(f) \leq (bcQ(f))^a = O(Q(f)^a)$, as desired.

The upshot of Theorem 23 is that we can show a class of permutation groups $G$ does not allow super-polynomial quantum speedups simply by showing that it is well-shuffling—that is, by showing that $G \in G$ is hard to distinguish from the set of small-range functions $D_{n,r}$ using a quantum query algorithm.
4 Showing permutation groups are well-shuffling

In this section, we introduce some tools for showing that a collection of permutation groups is well-shuffling. Due to Theorem 23, a well-shuffling collection of permutation groups does not allow any super-polynomial quantum speedups for the class of functions symmetric under it, so these tools can be directly used to show that certain symmetries are not consistent with large quantum speedups.

4.1 The symmetric group

The first fundamental result is that the class of full symmetric permutation groups $S_n$ is well-shuffling. This was shown by Zhandry [Zha15] in a different context, though we also provide a simpler proof by a reduction from the collision problem.

**Theorem 25.** There is a universal constant $C$ such that any quantum algorithm distinguishing a permutation in $S_n$ from a string in $D_{n,r}$ must make at least $r^{1/3}/C$ queries.

This theorem says that $S_n$ is hard to distinguish from $D_{n,r}$ (moreover, Zhandry [Zha15] showed that the hard distribution over $D_{n,r}$ is uniform, but we do not need this fact).

**Proof.** When $n$ is a multiple of $r$, then each $(n/r)$-to-1 function has range $r$ and each 1-to-1 function is a permutation; hence distinguishing $(n/r)$-to-1 from 1-to-1 functions is a sub-problem of distinguishing $D_{n,r}$ from $S_n$. This sub-problem is the collision problem, from which an $\Omega(r^{1/3})$ lower bound directly follows [AS04, Amb05, Kut05]. When $n$ is not a multiple of $r$ but $r \leq n/2$, we can just set $n' = r\lfloor n/r \rfloor$, and then distinguishing $(n'/r)$-to-1 from 1-to-1 functions with domain size $n'$ still reduces to distinguishing $D_{n,r}$ from $S_n$. \hfill $\Box$

From Theorem 25, the following two corollaries immediately follow (in light of Theorem 23).

**Corollary 26.** The set of symmetric groups $S = \{S_n\}_{n \in \mathbb{N}}$ is well-shuffling with power 3.

**Corollary 27.** All (possibly partial) Boolean functions $f$ that are symmetric under the full symmetric group $S_n$ satisfy $R(f) = O(Q(f)^3)$.

Apart from Theorem 25, the main tools we use to prove the well-shuffling property are transformations on permutation groups that approximately preserve $\text{cost}(G, r)$. We outline several such transformations and invariances. Since we prove Theorem 25 by a reduction from collision, and since our main tools from here on out are additional reductions, effectively all lower bounds in this paper work by reductions from collision.

4.2 Reduction between similar-looking permutation groups

We next show that similar-looking permutation groups have similar costs. Here, two groups $G$ and $H$ are “similar looking” if a random permutation from $G$ is almost indistinguishable from a random permutation from $H$ when queried in sufficiently few places. More formally, we have the following theorem.

**Theorem 28** (Similar-looking permutation groups have similar costs). Suppose $G$ and $H$ are permutation groups on $[n]$ and $k \leq n$ is a positive integer such that for each $i_1, i_2, \ldots, i_k, j_1, j_2, \ldots, j_k \in [n]$,

\[ \left| \Pr_{\pi \sim G} [\forall \ell \pi(i_\ell) = j_\ell] - \Pr_{\pi \sim H} [\forall \ell \pi(i_\ell) = j_\ell] \right| \leq n^{-10k}. \]

Then $\text{cost}(H, r) \geq \Omega(\min\{k, \text{cost}(G, r)\})$. In particular, if $k \geq n^{\Omega(1)}$ and $\text{cost}(G, r) \geq r^{\Omega(1)}$, then $\text{cost}(H, r) \geq r^{\Omega(1)}$.\hfill $\Box$
Theorem 28, setting $\sigma_i = \max \{0, x_i \}$ which can increase the sum of absolute values by a factor of at most $x$. Consider feeding into the quantum algorithm an input string where $Q$ of $m$ variables outside of $D$ consists of $d$ monomials. To see this, first note that there are at most $\binom{n^2}{d}$ monomials, so by Cauchy-Schwartz, the sum of absolute values of coefficients is at most $\sqrt{(\binom{n^2}{d})}$. But polynomials bounded in $[0,1]$ are Cauchy-Schwartz, the sum of absolute values of coefficients is at most $\sqrt{(\binom{n^2}{d})}$. From this we can conclude that $p_m$ is bounded in $[0,1]$ for all inputs it receives in $\{0,1\}^d$. But polynomials bounded in $[0,1]$ on the Boolean hypercube can have sum of coefficients at most $5^d$ (one way to see this is to recall that a bounded polynomial in the $\{-1,1\}$ basis has its sum of squares of coefficients equal to at most 1, and has at most $2^d$ coefficients, so by Cauchy-Schwartz, the sum of absolute values of coefficients is at most $2^{d/2}$; converting the $\{-1,1\}$ basis to the $\{0,1\}$ basis requires plugging $2z - 1$ terms into the variables, which can increase the sum of absolute values by a factor of at most $3^d$, for a total of at most $(3\sqrt{2})^d \leq 5^d$). Assuming $n \geq 5$, we get an upper bound of $n^{3d}$ on the sum of absolute values of coefficients of $p$.

We have $d \leq 2k$, so this sum is also at most $n^{6k}$. Now, on each input $x$, the expected output of $p(\pi(x))$ when $\pi$ is sampled uniformly from $H$ is a linear combination of the expectations of the monomials of $p$. For each monomial, this expectation is just the probability that the monomial is satisfied, which by the condition on $G$ and $H$ is within $n^{-10k}$ of the expectation under $\pi \leftarrow G$. It follows that the expectation of $p(\pi(x))$ when $\pi \leftarrow H$ is within $n^{6k}n^{-10k} = n^{-4k}$ of the expectation of $p(\pi(x))$ when $\pi \leftarrow G$. But this expectation is simply the acceptance probability of $Q$. Hence the acceptance probability of $Q$ on the uniform distribution on $H$ is within $n^{-4k}$ of the acceptance probability of $Q$ on the uniform distribution on $G$.

Since $Q$ distinguishes $H$ from $D_{n,r}$, it distinguishes the uniform distribution on $H$ from any string in $D_{n,r}$. Since it does not distinguish the uniform distribution on $H$ from the uniform distribution on $G$, $Q$ must also distinguish the uniform distribution on $G$ from any input in $D_{n,r}$ to error $1/3 + n^{-4k}$. By amplifying, we can get this down to error $1/3$, meaning that $\text{cost}(G,r) = O(\text{cost}(H,r))$, as desired.

An immediate consequence is that highly transitive permutation groups are well-shuffling:

Corollary 29. If $G$ is k-transitive, then $\text{cost}(G,r) = \Omega(\min\{k, r^{1/3}\})$, where the constant in the big-$\Omega$ is universal.

Proof. This follows directly from Theorem 28, setting $H$ to be the $k$-transitive permutation group we care about and setting $G = S_n$. To see this, observe that $k$-transitivity completely determines
Corollary 29 is less powerful than it seems: it is a consequence of the classification of finite simple groups that the only 6-transitive permutation groups are the symmetric group \( S_n \) for \( n \geq 6 \) and the alternating group \( A_n \) for \( n \geq 8 \), both in their natural action on \([n]\) [DM12]. Nevertheless, it will be important for us later that the alternating group \( A_n \), which is \((n-2)\)-transitive, satisfies \( \text{cost}(A_n, r) = \Omega(r^{1/3}) \).

### 4.3 Transformations for graph symmetries

Next, we introduce some additional transformations on permutation groups that approximately preserve the cost. The transformations in this section will allow us to show that graph property permutation groups (and several variants of them) are well-shuffling.

#### 4.3.1 Transformation for directed graphs

We start by defining an extension of a permutation group \( G \) on \([n]\) to a permutation group acting on \([n]^{\ell}\). The notation in the definition below comes from [Ker13].

**Definition 30.** Let \( G \) be a permutation group on domain \( D \), and let \( \ell \in \mathbb{N} \). Define \( G^{(\ell)} \) to be the permutation group that acts on domain \( D^\ell \) by \( \pi(i_1, i_2, \ldots, i_\ell) = (\pi(i_1), \pi(i_2), \ldots, \pi(i_\ell)) \) for each \( \pi \in G \) (so the number of permutations in \( G^{(\ell)} \) is the same as the number of permutations in \( G \)).

Define \( G^{(\ell)} \) to be the permutation group \( G^{(\ell)} \) with domain restricted to the subset \( D^{(\ell)} \subseteq D^\ell \) consisting of all distinct \( \ell \)-tuples of elements of \( D \).

We show that these transformations both preserve the cost, at least when \( \ell \) is constant. We start with \( G^{(\ell)} \).

**Theorem 31.** Let \( G \) be a permutation group on \([n]\), and let \( H \) be the permutation group \( G^{(\ell)} \). Then \( \text{cost}(H, r') \geq \text{cost}(G, r)/\ell \).

**Proof.** Let \( Q \) be an algorithm distinguishing \( H \) from \( D_{n^\ell, r, \ell} \). Let \( \mu \) be the hard distribution for \( G \), such that no algorithm using fewer than \( \text{cost}(G, r) \) queries can distinguish \( \mu \) from the uniform distribution on \( G \). Then \( \mu \) is a distribution on \( D_{n, r} \). Let \( \mu' \) be the distribution on \( D_{n^\ell, r, \ell} \) that we get by sampling \( \alpha \leftarrow \mu \) and returning \( \alpha' \) defined by \( \alpha'(z) = (\alpha(z_1), \alpha(z_2), \ldots, \alpha(z_\ell)) \) for each \( z \in [n]^\ell \) (here we identify \([n]^{\ell}\) with \([n]^\ell\)). Note that if \( \alpha \) has range \( r \), then \( \alpha' \) has range at most \( r'^{\ell} \).

Then \( Q \) distinguishes \( \mu' \) from the uniform distribution on \( H \). The latter distribution is the same as obtained when sampling \( \pi \) uniformly from \( G \) and returning \( \pi' \) defined by \( \pi'(z) = (\pi(z_1), \pi(z_2), \ldots, \pi(z_\ell)) \) for \( z \) in the domain of \( H \). This means that \( Q \) can be used to distinguish \( \mu \) from the uniform distribution on \( G \): all we need is to simulate every query of \( Q \) using \( \ell \) queries to the input \( \alpha \). The desired result follows.

To handle \( G^{(\ell)} \), we first observe that restricting the domain of a permutation group to some union of its orbits does not decrease its cost.

**Lemma 32.** Let \( G \) be a permutation group on \([n]\), and let \( S \subseteq [n] \) be a union of orbits of \( G \). Let \( G' \) be the permutation group \( G \) acting only on \( S \). Then \( \text{cost}(G', r) \geq \text{cost}(G, r) \).
Proof. We identify $S$ with $|S|$ without loss of generality. If $Q$ distinguishes $G'$ from $D_{n,r}$, then we can turn it into $Q'$ distinguishing $G$ from $D_{n,r}$ by having $Q'$ run $Q$ and make queries only from $|S|$.

The fact that $G^{(t)}$ does not decrease the cost of $G$ too much then follows as a corollary of Theorem 31 and Lemma 32.

**Corollary 33.** Let $G$ be a permutation group on $[n]$, and let $H$ be the permutation group $G^{(t)}$. Then \( \text{cost}(H, r^t) \geq \text{cost}(G, r)/t \).

Proof. It suffices to note that $G^{(t)}$ is the permutation group $G^{(t)}$ with domain restricted to $[n]^{(t)}$, which is a union of orbits because $\pi \in G^{(t)}$ always sends a tuple with unique entries to another tuple with unique entries (since a permutation on $[n]$ is applied to each entry). The desired result then follows from Theorem 31 and Lemma 32.

We now observe that the transformation $G^{(t)}$ immediately allows us to show that directed graph symmetries are well-shuffling.

**Corollary 34 (Directed graph symmetries).** The set $G = \{G_k\}_{k \in \mathbb{N}}$ of all directed graph symmetries is well-shuffling with power 6. Here the permutation group $G_k$ acts on a domain of size $n = k(k-1)$ representing the possible arcs of a $k$-vertex directed graph, and $G_k$ consists of all $k!$ permutations on these arcs that act by relabeling the vertices.

Proof. This immediately follows by observing that $G_k = S_k^{(2)}$. To see this, note that the domain of $G_k$ is the set of all ordered pairs $(x, y) \in [k]$ with $x \neq y$, which is precisely $[k]^{(2)}$, and the permutations in $G_k$ are just those in $S_k$ applied to both coordinates, which is precisely relabeling the vertices. Corollary 33 then gives \( \text{cost}(G_k, r) \geq \text{cost}(S_k, \sqrt{r})/2 \), which is at least $\Omega(r^{1/6})$ by Theorem 25.

The collection of directed hypergraph symmetries is similarly well-shuffling.

**Corollary 35 (Directed hypergraph symmetries).** The set $G_p = \{G_k\}_{k \in \mathbb{N}}$ consisting of all $p$-uniform directed hypergraph symmetries is well-shuffling with power $3p$.

Proof. This follows from the same argument as Corollary 34.

### 4.3.2 Transformation for undirected graphs

To handle undirected graphs, we introduce yet another operation on permutation groups which approximately preserves the cost.

**Theorem 36.** Let $G$ be a permutation group, and let $B = \{B_1, B_2, \ldots, B_k\}$ be a block system for $G$, where the blocks have equal size. Let $H$ be the permutation group on $[k]$ induced by the action of $G$ on the blocks. Then \( \text{cost}(H, r) \geq \text{cost}(G, r) \).

Proof. Let $Q$ be a quantum algorithm distinguishing $H$ from $D_{k,r}$ using \( \text{cost}(H, r) \) queries. We construct a quantum algorithm $Q'$ for distinguishing $G$ from $D_{n,r}$. Let $\mu$ be the distribution on $D_{n,r}$ that is hard to distinguish from the uniform distribution on $G$. The algorithm $Q'$ fixes a unique $i_t \in B_t$ for each $t = 1, 2, \ldots, k$. On input $\alpha$ from $G \cup D_{n,r}$, the algorithm $Q'$ runs $Q$ as follows: each query $t \in [k]$ that $Q$ makes is turned into the query $i_t \in [n]$ for $\alpha$, and the output $\alpha(i_t)$ is converted into the symbol $t'$ such that $\alpha(i_t) \in B_{t'}$ and returned to $Q$. In this way, the algorithm $Q'$ effectively runs $Q$ on the mapped string $\phi(\alpha) \in [k]^k$, where $\phi(\alpha)_t$ is the symbol $t'$ such that $\alpha(i_t) \in B_{t'}$.
Now, if $\alpha \in D_{n,r}$, then $\phi(\alpha) \in D_{k,r}$, while if $\alpha \in G$, we have $\phi(\alpha) \in H$. Since $Q$ distinguishes $H$ from $D_{k,r}$, it follows that $Q'$ distinguishes $G$ from $D_{n,r}$ using the same number of queries, as desired. \hfill \qed

We are now finally ready to prove the formal version of Theorem 2, showing that the collection of (undirected) graph symmetries is well-shuffling.

**Definition 37 (Graph Symmetries).** The collection of graph symmetries is the set $\mathcal{G} = \{G_k\}_{k \in \mathbb{N}}$ of permutation groups with $G_k$ acting on $[n]$ with $n = k(k-1)/2$, such that the domain $[n]$ represents the set of all possible edges in a $k$-vertex graph, and $G_k$ acts on these edges and permutes them in a way that corresponds to relabeling the vertices of the underlying graph.

**Corollary 38.** The set of all graph symmetries is well-shuffling with power 6. Hence $R(f) = O(Q(f)^6)$ for functions $f$ symmetric under a graph symmetry.

**Proof.** Let $G$ be a directed graph symmetry on domain size $k(k-1)$, and partition this domain into $k(k-1)/2$ sets of size 2 of the form $\{(x,y), (y,x)\}$ for $x, y \in [k]$. Then the induced permutation group $H$ on these sets (from Theorem 36) is precisely the undirected graph symmetry on graphs of size $k$. Since $\text{cost}(H, r) \geq \text{cost}(G, r)$, and since the directed graph symmetries are well-shuffling with power 6, it follows that the undirected graph symmetries are also well-shuffling with power 6. \hfill \qed

Using similar arguments, we can show a similar result for hypergraphs.

**Corollary 39.** For every constant $p \in \mathbb{N}$, the collection of all $p$-uniform hypergraph symmetries is well-shuffling with power $3p$.

### 4.3.3 Transformations for bipartite graphs

We introduce yet more operations on permutation groups for the case of bipartite graph symmetries.

**Definition 40 (Product of permutation groups).** Let $G_1$ and $G_2$ be two permutation groups acting on $[n_1]$ and $[n_2]$, respectively. Then the product permutation group $G_1 \times G_2$ is a permutation group acting on $[n_1 n_2]$ such that for any $(\pi_1, \pi_2) \in G_1 \times G_2$, and any $k \in [n_1]$ and $\ell \in [n_2]$, we have $(\pi_1, \pi_2)(k, \ell) = (\pi_1(k), \pi_2(\ell))$. (Here we identify $[n_1] \times [n_2]$ with $[n_1 n_2]$.)

**Theorem 41.** For all $G_1$ and $G_2$ acting on $[n_1]$ and $[n_2]$, respectively, and for all $r$, $\text{cost}(G_1 \times G_2, r^2) \geq \min\{\text{cost}(G_1, r), \text{cost}(G_2, r)\}$.

**Proof.** Let $H = G_1 \times G_2$ and $m = n_1 n_2$. Let $Q$ be an algorithm distinguishing $H$ from $D_{m,r^2}$. Let $\mu_1$ be the hard distribution for $G_1$ (on $D_{n_1,r}$), and let $\mu_2$ be the hard distribution for $G_2$ (on $D_{n_2,r}$). Let $\mu'$ be the distribution on $D_{m,r}$ that we get by sampling $\alpha_1 \leftarrow \mu_1$, and $\alpha_2 \leftarrow \mu_2$ independently, and returning $\alpha' = (\alpha(z_1), \alpha(z_2))$. Note that if $\alpha_1$ and $\alpha_2$ have range $r$, then $\alpha'$ has range at most $r^2$. Now, since $Q$ distinguishes $D_{m,r^2}$ from $G_1 \times G_2$, it must also distinguish $\mu'$ from the uniform distribution over $G_1 \times G_2$, which itself is the product of the uniform distribution on $G_1$ and the uniform distribution on $G_2$. Let $\nu_1$ be the uniform distribution on $G_1$, and let $\nu_2$ be the uniform distribution on $G_2$. Consider the behavior of $Q$ on $\mu_1 \times \nu_2$. It must either distinguish this distribution from $\mu_1 \times \nu_2$, or else from $\nu_1 \times \nu_2$ (since it distinguishes $\mu_1 \times \mu_2$ and $\nu_1 \times \nu_2$ from each other). In the first case, we can construct $Q'$ which artificially generates the sample from $\mu_1$ and uses $Q$ to distinguish $\mu_2$ from $\nu_2$. In the second case, we can construct $Q'$ which artificially generates the sample from $\nu_2$ and uses $Q$ to distinguish $\mu_1$ from $\nu_1$. Hence $\text{cost}(G_1 \times G_2, r^2) \geq \min\{\text{cost}(G_1, r), \text{cost}(G_2, r)\}$, as desired. \hfill \qed
Corollary 42. The collection $G$ of all bipartite graph symmetries with equal parts is well-shuffling.

Proof. This immediately follow by observing that bipartite graph symmetries are the symmetries $S_k \times S_k$. Then Theorem 41 and Theorem 25 give the desired result. \qed

4.3.4 Other transformations

We introduce one final transformation, which merges two permutation groups into one. This transformation also does not decrease the cost.

Lemma 43 (Merger). Let $G$ and $H$ be two permutation groups on $[n]$, and let $F = (G, H)$ be the permutation group on $[n]$ which is the closure of $G \cup H$ under composition. Then $\text{cost}(F, r) \geq \text{cost}(G, r)$.

Proof. Since $G$ is a subset of $F$, distinguishing $G$ from $D_{n, r}$ is strictly easier than distinguishing $F$ from $D_{n, r}$. \qed

5 Classification of well-shuffling permutation groups

In this section, we show that the results from the previous section are qualitatively optimal, in the sense that permutation groups constructed out of hypergraph symmetries are essentially the only groups that are not consistent with super-polynomial quantum speedups. Our starting point is an observation that super-polynomial quantum speedups can be constructed out of any permutation group with sufficiently small minimal base size.

Proposition 44. Let $G$ be a permutation group on $[n]$, and let $f$ be a (possibly partial) Boolean function with $\text{Dom}(f) \subseteq \Sigma^n$ for some alphabet $\Sigma$. Then there exists a partial Boolean function $g$ that is symmetric under $G$ such that $Q(g) \leq Q(f) + b(G)$ and $R(g) \geq R(f)$.

Proof. Define a promise problem $g$ as follows. A string $x \in (\Sigma \times [n])^n$ is in the promise if there exists a permutation $\pi \in G$ such that for all $i$, $x_{\pi(i)} = (y_i, i)$ where $y \in \Sigma^n$ is in the promise of $f$. In other words, the promise for $g$ contains all reorderings of strings of the form $(y_i, i)$ by permutations that are in $G$. Naturally, in this case we define $g(x) = f(y)$. Then clearly $R(g) \geq R(f)$, because any randomized algorithm that solves $g$ also solves $g$ restricted to the promise that $\pi$ is the identity permutation, in which case $g$ is equivalent to $f$. Moreover, this problem is clearly symmetric under $G$. On the other hand, $Q(g) \leq Q(f) + b(G)$: by querying a minimal base for $G$, the algorithm can learn the unique permutation $\pi$ such that $x_{\pi(i)} = (y_i, i)$. Now, the algorithm just permutes $x$ according to $\pi$ and runs the query algorithm for $f$. \qed

Put another way, if $b(G) = n^{o(1)}$, then by choosing an arbitrary function $f$ with $Q(f) = n^{o(1)}$ and $R(f) = n^\Omega(1)$ (e.g., Simon’s problem [Sim97] or Forrelation [AA15]), then one can construct a function $g$ symmetric under $G$ with $Q(g) = n^{o(1)}$ and $R(g) = n^\Omega(1)$. Thus, permutation groups that do not allow super-polynomial speedups must have large minimal base size.

Naturally, this raises the question of whether a sort of converse holds, i.e., whether $b(G) = n^{\Omega(1)}$ implies that $G$ is well-shuffling. We will show that this is true for primitive permutation groups (Theorem 47), and moreover that primitive groups with $b(G) = n^\Omega(1)$ all look roughly like symmetries of $p$-uniform hypergraphs (Corollary 48). Thus, we completely classify which primitive permutation groups are consistent with super-polynomial quantum speedups, and which are not.

For imprimitive groups, we will see that there exist permutation groups $G$ with $b(G) = n^\Omega(1)$ that are consistent with large quantum speedups. So, $b(G) = n^{\Omega(1)}$ is not equivalent to the well-shuffling
property in general. Nevertheless, we can show that primitive permutation groups with \( b(G) = n^{\Omega(1)} \) are the “building blocks” of well-shuffling group actions. Indeed, we show in Corollary 55 that if a collection of imprimitive permutation groups is inconsistent with super-polynomial speedups, then it must be built out of a constant number of well-shuffling primitive groups. Thus, in some sense, all permutation groups that do not allow large quantum speedups must involve symmetries of \( p \)-uniform hypergraphs.

5.1 Primitive groups

For primitive permutation groups, the following theorem due to Liebeck [Lie84] shows that there is a tight connection between minimal base size and the structure of the group. Its proof relies on the classification of finite simple groups.

**Theorem 45.** Let \( G \) be a primitive permutation group on \([n]\). Then one of the following holds:

(i) \( G \) is a subgroup of \( S_\ell \wr S_m \) containing \( A_n \times \ell \), where the action of \( S_m \) is on \( p \)-element subsets of \([m]\) and the wreath product has the product action of degree \( n = \binom{m}{p}^\ell \), or

(ii) \( b(G) < 9 \log_2 n \).

Here, \( S_n \) denotes the symmetric group on \([n]\) and \( A_n \) denotes the alternating group on \([n]\).

Observe that in case (i), when \( \ell = 1 \), this corresponds precisely to either the set of \( p \)-uniform undirected hypergraph symmetries, or the index-2 subgroup of even permutations of the vertices. When \( p \) is a constant, we know that these groups are well-shuffling. More generally, as long as both \( \ell \) and \( p \) are constant, then these groups are well-shuffling:

**Corollary 46.** For all constant \( \ell, p \), the collection of primitive permutation groups that satisfy case (i) of Theorem 45 is well-shuffling with power \( 3\ell p \).

**Proof.** Suppose \( G, \ell, p, \) and \( m \) are as in case (i) of Theorem 45. Let \( A \) denote the action of the alternating group \( A_m \) on \([m]\), and let \( H \) denote the action of \( A_m \) on size-\( p \) subsets of \([m]\). Formally, we have the following chain of inequalities:

\[
\begin{align*}
\text{cost}(G, r) & \geq \text{cost}(H^{\times \ell}, r) & \text{Lemma 43} \\
& \geq \text{cost}(H^{(\ell)}, r) & \text{Lemma 43} \\
& \geq \text{cost}(H, r^{1/\ell})/\ell & \text{Theorem 31} \\
& \geq \text{cost}(A^{(p)}, r^{1/\ell})/\ell & \text{Theorem 36} \\
& \geq \text{cost}(A, r^{1/\ell p})/\ell p & \text{Corollary 33} \\
& \geq \Omega(r^{1/3\ell p}). & \text{Corollary 29}
\end{align*}
\]

In words, in the first two lines, it suffices to show that a subgroup of \( G \) is well-shuffling, and thus that a subgroup of a subgroup is well-shuffling. \( H^{(\ell)} \) is the so-called “diagonal subgroup” of \( H^{\times \ell} \), consisting of the permutations \((\pi_1, \pi_2, \ldots, \pi_\ell)\) such that \( \pi_i = \pi_j \) for all \( i, j \); this is consistent with the notation from Definition 30. The third inequality appeals to the cost lower bound for power actions. Then, we observe that \( H \) is equivalent to the action of \( A^{(p)} \) on blocks, where \( A^{(p)} \) is the power action of \( A \) restricted to tuples with no repeats, and the blocks are sets of tuples that are reorderings of each other. The next inequality uses the cost lower bound for this restriction of the power action. Finally, we appeal to the fact that the alternating group on \( m \) points is \((m - 2)\)-transitive, and thus well-shuffling. \( \square \)
In fact, the following theorem shows that a collection of primitive permutation groups is well-shuffling if and only if the permutation groups are as in case (i) of Theorem 45, with \( \ell \) and \( p \) both constant:

**Theorem 47.** For all constant \( \epsilon > 0 \), there exists \( c \in \mathbb{N} \) such that the collection of primitive permutation groups \( G \) with \( b(G) \geq n^\epsilon \) is well-shuffling with power \( c \). Moreover, for all sufficiently large \( n \), all such \( G \) are as in case (i) of Theorem 45 with \( \ell p \leq c/3 \).

**Proof.** For sufficiently large \( n \), we cannot be in case (ii) of Theorem 45. This is because \( b(G) = O(\log n) \), so by Proposition 44, we can construct a function symmetric under \( G \) with a super-polynomial gap between quantum and randomized query complexity.

Therefore, suppose we are in case (i) of Theorem 45. If \( (\binom{m}{p}) = 1 \), then by the definition of the wreath product (Definition 13), \( G \) is the trivial group. Otherwise, we suppose \( (\binom{m}{p}) \geq 2 \) and therefore \( m \geq 2 \). The minimal base size of \( G \) is upper bounded by

\[
\begin{align*}
b(G) &\leq \log_2(|G|) \\
&\leq \log_2(|S_\ell \wr S_m|) \\
&= \log_2(\ell! \cdot m!^\ell) \\
&\leq \ell \log_2 \ell + \ell m \log_2 m \\
&\leq 2\ell^2 m^2.
\end{align*}
\]

Thus we have

\[
\left(\frac{m}{p}\right)^{\ell p} \leq \left(\frac{m}{p}\right)^{\ell} = n^\epsilon \leq b(G) \leq 2\ell^2 m^2,
\]

so

\[
\ell p (\log_2 m - \log_2 p) \leq 1 + 2 \log_2 \ell + 2 \log_2 m.
\]

Consider two cases. First, suppose \( m \leq p^2 \). Without loss of generality, we may always assume \( p \leq m/2 \), because the action of \( S_m \) on \( p \)-element subsets is equivalent to the action on \( (m - p) \)-element subsets, by identifying each \( p \)-element subset with its complement in \( [m] \). So we have

\[
\ell p \leq \ell p (\log_2 m - \log_2 p) \leq 1 + 2 \log_2 \ell + 2 \log_2 m \leq 1 + 2 \log_2 \ell + 4 \log_2 p.
\]

But clearly \( \ell p \leq 1 + 2 \log_2 \ell + 4 \log_2 p \) is possible only for bounded values of \( \ell p \).

Otherwise, suppose \( m \geq p^2 \). Then we have

\[
\ell p \leq \frac{1 + 2 \log_2 \ell + 2 \log_2 m}{\log_2 m - \log_2 p} \leq \frac{2 + 2 \log_2 \ell + 2 \log_2 m}{\log_2 m} = 4 + \frac{2 + 4 \log_2 \ell}{\log_2 m} \leq 6 + 4 \log_2 \ell.
\]

Likewise, \( \ell p \leq 6 + 4 \log_2 \ell \) is possible only for bounded values of \( \ell p \). So we may suppose that \( \ell p \leq c/3 \) for some constant \( c \). Then, the theorem follows from Corollary 46.

**Corollary 48.** Let \( \mathcal{G} \) be a collection of primitive permutation groups. The following are equivalent:

(i) \( \mathcal{G} \) is well-shuffling.

(ii) \( b(G) = n^{\Omega(1)} \) for \( G \in \mathcal{G} \) acting on \([n]\).

(iii) All but finitely many \( G \in \mathcal{G} \) satisfy case (i) of Theorem 45 with \( \ell p = O(1) \).
Proposition 44, and (iii) implies (i) by Corollary 49. Theorem 47.}

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is to understand which \textit{transitive} permutation groups (i.e., groups with a single orbit) allow super-polynomial quantum speedups. The following well-known embedding theorem shows that imprimitive transitive groups can be expressed as subgroups of a wreath product constructed from a nontrivial block system. See Theorem II.49 of [Hul10] for a proof.

**Lemma 52** (Embedding theorem). Let $G$ be a transitive imprimitive permutation group on domain $D$ with $B = \{B_1, B_2, \ldots, B_k\}$ a nontrivial block system for $G$. Let $H$ denote the permutation group acting on $[k]$ induced by the action of $G$ on blocks $B$. Let $\text{Stab}_G(\{B_i\})|_{B_i}$ denote the permutation group induced by the action of $\text{Stab}_G(\{B_i\})$ restricted to the orbit $B_i$. Then the action of $G$ is permutation isomorphic to a subgroup of $H \varwr \text{Stab}_G(\{B_i\})|_{B_i}$ in imprimitive action, meaning that $G$ can be viewed as a subgroup under an appropriate relabeling that identifies $D$ with $[k] \times B_i$.

Applying Lemma 52 recursively, one can always decompose a transitive imprimitive permutation group as a subgroup of an iterated wreath product of primitive groups. Thus, primitive groups are sometimes described as the “building blocks” of permutation groups. Note that such a decomposition is not unique in general, because a permutation group can have many nontrivial block systems.

The next proposition shows that if any terms in such a wreath product decomposition are consistent with super-polynomial speedups, then so is the group as a whole.

**Proposition 53.** Let $G$ be a permutation group on $[n]$ that is a subgroup of an iterated wreath product $G_1 \varwr G_2 \varwr \cdots \varwr G_d$ in imprimitive action. Suppose $f$ is a (possibly partial) function symmetric under $G_i$ for some $i$. Then there exists a partial function $g$ symmetric under $G$ with $Q(g) = Q(f)$ and $R(g) = R(f)$.

Proof. Suppose $G_i$ acts on $n_i$ points, so that $n = \prod_{i=1}^d n_i$. It suffices to exhibit such a function when $G$ is the full wreath product, and each $G_j$ is the symmetric group $S_{n_j}$ for every $j \neq i$. Define the function $\text{Triv}_m$ to be a promise problem that only takes two inputs, $0^m$ and $1^m$, outputting 0 on $0^m$ and 1 on $1^m$. The block composition $\text{Triv}_{n_1 n_2 \cdots n_{i-1}} \circ f \circ \text{Triv}_{n_{i+1} n_{i+2} \cdots n_d}$ is symmetric under $G$, and has quantum and randomized query complexities identical to those of $f$. \hfill \Box

Less trivially, if a wreath product decomposition of a group is sufficiently deep, then the group is also consistent with large speedups. The following construction is based on the so-called “1-FAULT DIRECT TREES” problem of [ZKH12, Kim12]:

**Theorem 54.** Let $G$ be a permutation group on $[n]$ that is a subgroup of an iterated wreath product of nontrivial permutation groups $G_1 \varwr G_2 \varwr \cdots \varwr G_d$ in imprimitive action. Then there exists a function $f$ symmetric under $G$ such that $Q(f) = O(1)$ but $R(f) = \Omega(\log d)$.

Proof. Suppose $G_i$ acts on $n_i \geq 2$ points, so that $n = \prod_{i=1}^d n_i$. It suffices to show that there exists a super-polynomial speedup when $G$ is the full wreath product and each $G_i$ is the symmetric group $S_{n_i}$. In this case, $G$ can be described the group of symmetries of a depth-$d$ rooted tree, where all of the vertices at distance $i-1$ from the root have exactly $n_i$ children. $G$ is the action of this group of symmetries on the leaves of the tree. We construct a problem symmetric under $G$ by taking a Boolean evaluation tree and adding a promise that gives an exponential speedup.

If $n_i = 2$ for all $i$, such a problem is already known: the 1-FAULT DIRECT TREES problem is a promise problem symmetric under $G$ with quantum query complexity $O(1)$ [Kim12] and randomized query complexity $\Omega(\log d)$ [ZKH12]. In the simplest case, the problem is a restriction of the depth-$d$ binary NAND tree, but it can also be generalized to Boolean evaluation trees composed of other functions.

We take a slight generalization of the 1-FAULT DIRECT TREES problem where the Boolean functions can have different fan-in at each depth. Specifically, we take the tree to be the block
composition of functions \( f_1 \circ f_2 \circ \cdots \circ f_d \) where the function \( f_i \) at depth \( i \) is defined by \( f_i : S \to \{0, 1\} \) where \( S \subseteq \{0, 1\}^{n_i} \) consists of the \( n_i \)-bit strings that have Hamming weight in \( \{0, \lfloor n_i/2 \rfloor, \lceil n_i/2 \rceil, n_i\} \), and \( f_i(x) = 0 \) if and only if \( x = 1^{n_i} \). Note that \( f_i \) is essentially just a padded version of the binary NAND function. The proof that this problem has the desired query complexity bounds follows almost immediately from the proofs in [Kim12, ZKH12]; for completeness, we provide the details in Appendix B.

We remark that Theorem 54 shows the existence of super-polynomial speedups under permutation groups with large base. Indeed, if we define \( G = S_2 \wr \cdots \wr S_2 \) to be the depth-\( d \) iterated wreath product in imprimitive action of the symmetric group \( S_2 \), then \( b(G) = 2^{d-1} \) even though \( G \) acts on \( 2^d \) points.

Putting all of these together, we can say the following about collections of permutation groups that are inconsistent with super-polynomial speedups:

**Corollary 55.** Let \( \mathcal{G} \) be a collection of permutation groups. Suppose that for every \( f \in F(\mathcal{G}) \), we have \( R(f) = Q(f)^{O(1)} \). Then the following must hold:

(i) Permutation groups in \( \mathcal{G} \) have \( O(1) \) orbits.

(ii) The collection of restrictions of groups \( G \in \mathcal{G} \) to subsets of their orbits is also inconsistent with super-polynomial quantum speedups.

(iii) All wreath product decompositions of groups \( G \in \mathcal{G} \) restricted to an orbit have \( O(1) \) primitive factors.

(iv) The collection of primitive factors that appear in such wreath product decompositions is well-shuffling. Equivalently, by Corollary 48, those primitive factors \( G \) that act on \( [n] \) have \( b(G) = n^{\Omega(1)} \), and all but finitely many satisfy case (i) of Theorem 45 with \( \ell p = O(1) \).

Thus, we have formally shown that collections of permutation groups that do not allow super-polynomial quantum speedups must be constructed out of a constant number of well-shuffling primitive groups. As we have shown before, such primitive groups are essentially just extensions of permutation groups that correspond to \( p \)-uniform hypergraph symmetries.

### 6 Exponential speedup for adjacency-list graph property testing

Having shown that graph properties in the adjacency *matrix* model cannot have exponential quantum speedup, we now turn to the adjacency *list* model. Specifically, we describe a graph property testing problem in the adjacency list model for which there is an exponential quantum speedup. We work with graphs that are undirected and have degree at most 5, while for convenience\(^5\) we allow the graphs to have self-loops and parallel edges (we count these with multiplicity toward the degree bound).

Recall that in property testing, the goal is to distinguish between some set of yes instances and the set of no instances consisting of all graphs that are \( \varepsilon \)-far from the yes instances. More precisely we say that a graph \( G = (V, E) \) (with maximum degree at most 5) is \( \varepsilon \)-far from the property \( \mathcal{P} \), if

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\(^5\)This is simply for clarity of presentation; we use self-loops and double edges as markers, for nodes and edges respectively. However, one could also work with simple graphs only and get the same exponential separation. This can be achieved by using slightly more complicated marker structures, e.g., one could replace self-loops with an edge connected to a triangle, etc.
for any yes instance $G' = (V, E') \in \mathcal{P}$ with $n$ vertices, the symmetric difference of the edge sets has size $|E \triangle E'| \geq \varepsilon n$, where we count edges with multiplicity. Since we work with graph properties, we also require that $\mathcal{P}$ is invariant under permuting the vertices.

Figure 1: An illustration of a candy graph in the yes instance (top right) and our shorthand symbol for it (top left), and a double-bow-tie graph which appears in the particular no instances considered for our classical lower bound proof (bottom right) and our shorthand symbol for it (bottom left).

The graph property $\mathcal{P}_k$ that we want to test is the following. A graph has the property $\mathcal{P}_k$ if its single-edges form a graph that contains $2j(2^k - 1)$ disconnected instances of “candy” graphs for some $j \in \mathbb{N}$, as shown in Figure 1. A candy graph is obtained from 4 binary trees of depth $k$: two “antenna” trees $A_1, A_2$ and two “body” trees $B_1, B_2$. The candy graph is formed by first connecting the leaves of $B_1$ and $B_2$ by a collection of alternating cycles containing all leaves of the body trees, and then merging the roots of $A_1 - B_1$ and $A_2 - B_2$. Additionally, in the full graph, roughly half of the candy (sub)graphs get a self-loop on exactly one of the roots, and the other candy graphs have either no self-loops or both roots get a self-loop.7 There is also a double edge attached to every non-root vertex of the candy graphs, namely every body vertex is connected to a distinct antenna vertex by a double edge in the following way: if the body vertex is a “weld” vertex (i.e., it was a leaf of a body tree), then it gets attached to an antenna vertex inside a candy graph that has exactly

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6In the lower bound proof we assume that there is a single long cycle, but this is hard to test, so for the definition of the property we allow smaller cycles as well.

7Strictly speaking, a $2^k/(2^k+1 - 2)$ fraction of candy graphs should get exactly one self-loop and the remaining fraction should be split between getting two or zero self-loops, because there are $2 \cdot 2^k$ weld vertices while only $2(2^k - 2)$ interior vertices in each candy graph, cf. the following main text. With a slight modification to the construction this could be balanced out, for example by replacing the roots by a path of length 5, and treating the middle vertex of this length-5 path as the new root.
one root with self-loop, while every interior vertex (i.e., non-weld and non-root) gets attached to an antenna vertex in a candy graph where the parity of the number of self-loops is even.

Figure 2: Multiple candy (sub)graphs which together form a yes-instance graph with property $P_k$. In the particular no instances we consider, the candy graphs are replaced by double-bow-tie graphs. Here, we only show six of them and only five of the many “advice edges” (indicated by double lines) that connect each body vertex to a distinct antenna vertex. The circles in the figure represent self-loops at the roots of the candy graphs, which provide advice about whether a body vertex is in the interior or the weld. An even parity of circles indicates interior, while an odd parity indicates weld.

While we do not claim that testing the above graph property is particularly useful, we can prove that a quantum computer has an exponential advantage in testing this property. The intuition is the following: this graph property is exponentially difficult to test on a classical computer because the welded trees locally look like exponentially deep trees, unless the weld vertices can be distinguished. This is where quantum computers get their advantage: a quantum computer can “read” the advice hidden in the structure of the double edges very efficiently using the quantum walk algorithm of Childs et al. [CCD+03]. Once we can tell apart the weld vertices from the other vertices, testing the structure could be performed even on a classical computer.

We prove the following theorem in this section:

**Theorem 56.** In the adjacency list model, there exists a constant $\epsilon$ such that testing whether a bounded-degree graph has property $P_k$ or is $\epsilon$-far from having it can be done by a quantum algorithm using $Q(P_k) = \text{poly}(k)$ queries, with constant success probability and perfect completeness. On the other hand, any classical randomized algorithm that can test the property with bounded two-sided error needs $R(P_k) = \exp(\Omega(k))$ queries.

**Proof outline.** First we prove in Section 6.1 that the property $P_k$ can be tested with perfect completeness and constant success probability with $\text{poly}(k)$ queries, given the proper advice. Then in Section 6.2 we show that for graphs having property $P_k$ the advice can be computed on a quantum computer with probability 1. This completes the proof of our efficient quantum tester.

Finally, in Section 6.3 we show that there are two particular distributions of yes and no instances that are exponentially difficult to distinguish on a classical computer, where the considered no instances are constant-far from having property $P_k$ with high probability. This rules out any sub-exponential-time classical property tester of $P_k$ that succeeds with constant probability. 

\[\Box\]
6.1 Efficient classical testability with advice

In this subsection, we describe how to test the property \( P_k \) efficiently with advice. First, we rigorously define this notion, which has a similar flavor to the complexity classes \( \text{NP} \) and \( \text{MA} \). Then, we proceed by showing an efficient tester with advice.

The required advice is precisely what our quantum algorithm can provide as per the above subsection: marking the weld vertices. Formally, the advice is a bit associated with every vertex, and for yes instances the advice is supposed to mark weld vertices.

There are two major parts of our analysis. First, we show that the test always accepts yes instances if the advice is correctly provided. Then we show that if the test accepts a graph with high probability, then the graph must actually be close to a yes instance. The latter implies that no instances that are far from complying with the property must actually be rejected with high probability.

6.1.1 Property testing with advice

We say that a property \( P \) can be tested with advice\(^8\) in \( Q \) queries if there is a tester \( T \) that is

Complete. For every instance \( x \in P \) there is a witness \( w \in \{0,1\}^* \), such that \( T \) accepts \( (x, w) \) with probability at least \( 2/3 \) and makes at most \( Q \) queries to \( x \) and \( w \).

Sound. For every instance \( x \) that is at least \( \varepsilon \)-far from \( P \), and for any witness \( w \in \{0,1\}^* \), \( T \) accepts \( (x, w) \) with probability at most \( 1/3 \).

Additionally, if for every yes instance there is a witness that makes the tester accept with certainty, we say that the tester has perfect completeness.

6.1.2 Testing the binary trees, the weld, and the advice

In this subsection, we describe our classical tester using the advice. We build increasingly more complex tests, such that passing them enforces more and more structure on the graph. Our final tester ensures that any graph passing the test with high probability must be \( \varepsilon \)-close to a yes instance. The query and time complexity of our final tester is \( \text{poly}(k/\varepsilon) \).

As a first step we test the binary tree structures that are supposed to remain after removing every double edge, self-loop, and edge between marked (therefore supposedly weld) vertices. Therefore, while testing the binary tree structures—i.e., in Algorithm 1, Algorithm 2, and the consistency check below—we ignore those edges (whenever we say that we ignore some edges we also exclude them from degree counts). Also, if we encounter any vertex that has more than 4 single-edge neighbors (excluding self-loops), then we immediately reject the graph, so we can also assume without loss of generality that the modified graph has degree at most 4 in Algorithm 1 and Algorithm 2. Similarly, if Algorithm 1 or Algorithm 2 returns REJECT, we immediately reject the instance.

The basic primitive of our tester is a binary-tree tester described in Algorithm 1, based on non-backtracking walks. The main idea of the tester is that for any non-root vertex \( v \) at level \( \ell \) (i.e., distance \( \ell \) from the root) in a binary tree of depth \( k \), any non-backtracking walk starting towards the parent of \( v \) can continue for at least \( k - \ell + 2 \) steps, while any other non-backtracking walk must terminate after \( k - \ell \) steps. Based on this observation we can find the parent of \( v \) efficiently. Algorithm 1 summarizes this procedure for finding the parent vertex, and adds some consistency checks that are helpful for catching potential corruptions of the structure.

\(^8\)Note that this terminology differs from the standard use of “advice” in complexity theory, which typically refers to a string that can only depend on the input length.
Algorithm 2 recursively applies Algorithm 1 to find a path to the root. Although this recursive structure may not be the most efficient way of finding such a root-path, it allows us to argue that a root-path found for a vertex should match the root-path of its parent, even in slightly corrupted trees.

Algorithm 1 FindParent(v)

```plaintext
input: vertex v (in a graph with maximum degree at most 5)
output: parent vertex u (if the graph is a binary tree of depth k)
1: if deg(v) = 4 then return ROOT (* in a binary tree only the root has degree 4 *)
2: if deg(v) \not\in \{1, 3, 4\} then REJECT (* every vertex is expected to have 1, 3, or 4 neighbors *)
3: r \leftarrow \emptyset (* stores a root candidate that is any degree-4 vertex encountered *)
4: for each w \in adj(v) do (* if we get here we know that the degree of v is 1 or 3 *)
5: p_0^{(w)} \leftarrow v (* the starting vertex of the non-backtracking walk *)
6: p_1^{(w)} \leftarrow w (* the first vertex on the path in the direction towards w *)
7: \ell(w) \leftarrow 2k (* upper bound on the length of a non-backtracking walk *)
8: for each t = 1, \ldots, 2k - 1 do (* the steps of the non-backtracking walk *)
9: if deg(p_t^{(w)}) = 1 then (* a leaf is found *)
10: \ell(w) \leftarrow t (* store the length of the path *)
11: quit loop and goto line 17 (* quit the loop and end the walk *)
12: if deg(p_t^{(w)}) = 4 then (* a root candidate is found *)
13: if r = \emptyset then r \leftarrow p_t (* update root candidate *)
14: else REJECT (* there is a unique root *)
15: p_{t+1}^{(w)} \leftarrow \text{a uniformly random vertex in adj}(p_t^{(w)}) \setminus \{p_{t-1}^{(w)}\} (* make a random step *)
16: end for
17: if deg(p_{\ell(w)}^{(w)}) > 1 then REJECT (* verify that a leaf is found *)
18: end for
19: u \leftarrow \operatorname{argmax}_w (\ell(w)) (* the longest path must go through the parent *)
20: if \exists w_1, w_2 \in adj(v) \setminus \{u\} such that \ell(w_1) \neq \ell(w_2) then REJECT (* consistency check *)
21: if r \neq \emptyset \land r \notin \{p_1^{(u)}, p_2^{(u)}, \ldots, p_{\ell(u)}^{(u)}\} then REJECT (* consistency check *)
22: return u
```

Consistency test. During this test we ignore every double edge and self-loop, and also any edge between marked vertices. Given vertex v, reject if v is marked and has degree greater than 1. Run Algorithm 2 to find a path to the root (degree-4 vertex). If the root-path is longer than k, or any vertex (other than v) is marked on the root-path, then reject. Also reject if v is marked and the path is shorter than k. Repeat this procedure 100 times and reject unless always the same path is found. If this consistency check or any individual run fails, then reject, otherwise accept.

Definition 57. A vertex v is consistent if it passes the consistency test with probability at least 1/2. For a consistent vertex, there is a consistent root and a consistent root-path to the consistent root that Algorithm 2 finds with probability at least 0.99.

Lemma 58. For any edge e between two neighboring consistent vertices u and v, one of the consistent root-paths, say (u, p_1, \ldots, p_\ell), must be a concatenation of the edge e and the other consistent root-path, i.e., p_1 = v.
Algorithm 2 FindRootPath(v)

| input:  | vertex v (in a graph with maximum degree at most 5) |
| output: | a path to the root (if the graph is a binary tree of depth k) |
| 1:     | \( p_0 \leftarrow v \) (\( v \) is the first vertex on the path to the root \( \star \)) |
| 2:     | \( \ell \leftarrow k \) (\( \star \) the path to the root cannot be longer than \( k \) \( \star \)) |
| 3:     | for each \( t = 1, \ldots, k \) do (\( \star \) find a path to the root step-by-step \( \star \)) |
| 4:     | \( p_t \leftarrow \text{FindParent}(p_{t-1}) \) (\( \star \) find the parent of the current vertex \( \star \)) |
| 5:     | if \( p_t = \text{ROOT} \) then (\( \star \) the root is found \( \star \)) |
| 6:     | \( \ell \leftarrow t \) (\( \star \) store the length of the path \( \star \)) |
| 7:     | quit loop and goto 9 (\( \star \) store the length of the path and stop \( \star \)) |
| 8:     | end for |
| 9:     | if \( \deg(p_t) \neq 4 \) then REJECT (\( \star \) verify that a root is found \( \star \)) |
| 10:    | return \( (p_0, p_1, \ldots, p_t) \) |

Proof. First note that due to Line 2, any consistent vertex \( u, v \) must have degree 1, 3, or 4. If \( v \) has degree 4, then for any neighbor \( u \) the consistent root-path must be \( (u, v) \) due to Line 21. Also note that the statement is trivial if either \( u \) or \( v \) has degree 1.

Finally, we treat the case when both \( u \) and \( v \) have degree 3. Towards a contradiction let us assume that the statement of the lemma does not hold. Let \( v_c \neq u \) denote the neighbor of \( v \) which is not its immediate parent on the consistent root-path, and similarly let \( u_c \neq v \) denote the neighbor of \( u \) which is not its immediate parent on the consistent root-path. Due to Line 20, a non-backtracking walk from \( u \) in the direction \( u_c \) has typically (with probability at least 0.99) the same length as a non-backtracking walk in the direction \( v \). This also means that there is a fixed length that has high probability (\( > 0.98 \)), which we denote by \( \ell_u \). We denote by \( \ell_u \) the analoguous typical length starting from \( u \) towards either \( u_c \) or \( v \). When we start a non-backtracking walk in the direction of \( u \) we continue with a non-backtracking walk towards \( u_c \) with probability at least \( 1/3 \). This implies that \( \ell_v > \ell_u \). Since the situation is symmetric we can also deduce that \( \ell_v < \ell_u \) which is a contradiction.

It is easy to see that for any consistent vertex \( v \) all vertices on its consistent root-path are also consistent vertices. This is due to the recursive structure of Algorithm 2.

Consequently we can conclude that the graph \( G_C \) induced by consistent vertices is a (sub-)binary forest. More precisely, every connected component can be injectively mapped into a graph obtained by merging the roots of two binary trees of depth \( k \). The mapping is defined via the root-paths, and also ensures that marked consistent vertices get mapped to leaves. Moreover, for any \( \delta > 0 \), if the consistency test is passed with probability \( 1 - \delta \) for a uniformly random vertex \( v \in V \), then there are at least \( (1 - 2\delta)|V| \) consistent vertices, since by definition non-consistent vertices fail to pass the consistency check with probability at least \( 1/2 \).

Weld-consistency test. During this test we ignore every double edge and self-loop (additionally, when we call the consistency test as a subroutine, we also ignore edges between marked vertices). Run the consistency test on \( v \) and denote by \( r \) the root found. If the length of the root-path is less than \( k \) (i.e., \( v \) is not a “leaf”), then accept. Let \( e \) denote the last edge of the root-path from \( v \) to \( r \).

- If \( v \) is not marked, then go to the root \( r \) and perform 100 non-backtracking random walks of length \( k \) along \( e \). If any walk terminates before \( k \) steps or any marked vertex is found, then reject, otherwise accept.

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• If \( v \) is marked, then reject unless exactly 2 of its neighbors are also marked. Run the consistency test on both marked neighbors, and denote the roots \( r' \) and \( r'' \). Reject if \( r' \neq r'' \) or \( r = r' \). Perform a non-backtracking random walk from \( r \) along all 4 edges of length \( k \); if any walk stops before \( k \) steps or not exactly 2 out of the 4 end-points is marked, then reject. Also reject if the walk along \( e \) does not end in a marked vertex. Reject unless both marked end-points have exactly two marked neighbors; traverse to a random marked neighbor of both marked end-points, run a consistency check, and report its root. If the consistency check fails or either of the found roots does not match \( r' \), then reject. Repeat the process of this paragraph 100 times.

**Definition 59.** A vertex \( v \) is weld-consistent if it passes the above test with probability at least \( 1/2 \).

We already established that consistent vertices form a (sub-)binary forest, where some leaves at level \( k \) may be marked. The weld-consistency test might remove some of the marked vertices, but ensures that the (sub-)binary trees are paired up via marked-marked vertex edges.

**Lemma 60.** All weld-consistent marked vertices in a (sub-)binary tree \( t_1 \) are connected to at most 2 consistent marked vertices in another (sub-)binary tree \( t_2 \) of consistent vertices.

**Proof.** Indeed, for any weld-consistent vertex \( v \), its consistent root \( r \) has the property that a non-backtracking walk of length \( k \) along the 4 outgoing edges after traversing the 2 marked leaves finds the same root \( r' \) with high probability. We call \( r' \) the consistent-root-pair of \( r \). It is easy to see that all consistent marked neighbors of \( v \) must have the same consistent root as the consistent-root-pair of \( r \). \( \square \)

By Lemma 60 we have established that weld-consistent vertices form pairs of (sub-)binary trees, that are connected by edges between marked vertices at level \( k \), with each marked vertex having at most 2 marked neighbors. Moreover, these (sub-)binary candy graphs are isomorphic to an induced subgraph of an actual candy graph. Also at any root, at most two of the branches contain marked leaves. We call such a graph a (sub-)binary candy ensemble.

Similarly as before, for any \( \delta > 0 \), if the weld-consistency test is passed with probability \( 1 - \delta \) for a uniformly random vertex \( v \in V \), then there are at least \( (1 - 2\delta)|V| \) weld-consistent vertices. This way we can ensure that there are many weld-consistent vertices and that they form a (sub-)binary candy ensemble. By the following test we also ensure that a typical connected component is close to being a complete candy graph.

**Completeness test.** During this test we still ignore every double edge and self-loop. Given vertex \( v \), run the weld-consistency test, and then go to the root \( r \). From \( r \), along each edge perform a non-backtracking walk. Reject if any walk fails to find a leaf (degree-1 vertex) or a marked vertex (with 2 marked neighbors) after exactly \( k \) steps, or any non-degree-3 vertex is encountered other than a leaf. Also reject if there are not exactly 2 marked vertices among the 4 reached end-points. Traverse to a random marked neighbor of a randomly chosen marked end-point, find its root-path, and let \( r' \) denote the root that is found. Also run the weld-consistency check for every vertex on every sampled path and reject if any of the vertices fail the weld-consistency check. Go to \( r' \) and run the same test, except for traversing a marked-marked edge at the end. Repeat the whole process described in this paragraph \( \Theta(k/\varepsilon) \) times.

**Lemma 61.** If \( v \) is a weld-consistent vertex within a (sub-)binary candy graph induced by fewer than \( (1 - \varepsilon)(2^{k+3} - 6) \) weld-consistent vertices, then the completeness test rejects with high probability.
**Proof.** Let us consider the connected component of $v$ formed by weld-consistent vertices. We already established that this connected component can be embedded into a candy graph. If at least an $\varepsilon$-fraction of the vertices of the candy graph are missing from the connected component of $v$, then we must encounter a non-weld consistent vertex or a degree-deficient vertex (i.e., less than degree-3 interior or weld vertex) with probability $\Omega(\varepsilon/k)$ in each of the $\Theta(k/\varepsilon)$ runs (since by the pigeonhole principle there is a level set of the candy graph, where at least an $\Omega(\varepsilon/k)$-fraction of the vertices are missing). This implies the statement of the lemma.

If there are more than $\varepsilon|V|$ consistent vertices that are in a sub-binary tree of weld-consistent vertices of size at most $(1-\varepsilon)\left(2^{k+1}-1\right)$, then the completeness check fails with probability at least $\varepsilon$. Therefore, we can see that if the graph passes the completeness check with probability at least $1-\varepsilon$, then the graph is $O(\varepsilon)$-close to the desired structure of having a disjoint union of welded trees, when disregarding self-loops and double edges. (One can actually form the desired structure by removing non-consistent vertices, and then completing the potentially incomplete trees missing $O(\varepsilon)$ edges.)

**Advice test.** Given vertex $v$, run the completeness test, and reject unless it has precisely one double edge or has 4 single-edge neighbors, with the potential addition of a self-loop (root vertex). If $v$ has a double edge neighbor $u$, then run the completeness test on $u$. Reject unless precisely one of $u$ and $v$ appear to be in a branch with non-marked leaves in the completeness test. Let $a$ denote the one in the non-marked branch, and $b$ the other. Check the parity of the number of loops of $a$’s root $r$ and the root-pair $r'$ of $r$ found during the completeness test. Reject if the parity does not match the marking of $b$.

**Definition 62.** A vertex $v$ is advice-consistent if it passes the advice test with probability at least $1/2$.

We can see that if the graph passes the advice test with probability at least $1-\varepsilon$, then at most an $O(\varepsilon)$-fraction of the weld-consistent vertices have an incorrect advice edge. Thus, if the number of vertices is a multiple of $2\cdot(2^k-1)\cdot(2^{k+3}-6)$, then we can just take the induced subgraph of the advice-consistent vertices and complete it using $O(\varepsilon)$ edges to get to a yes instance with property $P_k$.

**The final test.** Verify that the number of vertices is a multiple of $2\cdot(2^k-1)\cdot(2^{k+3}-6)$, then run the advice test $O(1/\varepsilon)$ times. If we have a yes instance with the correct advice, then the test always accepts, so the tester has perfect completeness. On the other hand, if the test accepts with probability at least $1/3$, then we know that the graph must be $\varepsilon$-close to a yes instance. Therefore, any $\varepsilon$-far no instance will be rejected with probability at least $2/3$.

### 6.2 Marking the weld vertices using a quantum computer

In this subsection, we describe how to efficiently mark weld vertices in a graph $G \in P_k$. More precisely, given a vertex $v$ in $G$ the task is to tell whether it is a weld vertex or not, in query and time complexity poly($k$). If the graph $G$ does not have property $P_k$ we do not place any restriction on the output, so we will always assume that $G \in P_k$. The procedure that we describe below has success probability $1$.

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Note that it is possible, that one also needs to remove an $O(\varepsilon)$-fraction of the candy-components, if there is a surplus of candies with even/odd number of marker loops.
Let us denote by $W$ the set of weld vertices in a graph $G$ with property $\mathcal{P}_k$. Given a vertex $v$, we decide whether or not $v \in W$ as follows:

1.) Along every non-double edge adjacent to $v$, start a non-backtracking (random) walk, which only traverses single edges. If a degree-3 vertex (i.e., a leaf with a double edge) is encountered within the first $k$ steps of any of these walks (including $v$ itself), then conclude $v \notin W$.

2.) Now we know that $v$ is a non-root body vertex. We traverse the double edge to the antenna vertex $a$ in antenna $A$. From now on we completely ignore double edges (and even exclude them from degree counts), and use Algorithm 2 to find the root $r_a$ of the antenna $A$.

3.) From the root $r_a$ we run the quantum walk algorithm of Childs et al. [CCD+03] to find the other root $r_b$ in the candy graph. We conclude $v \in W$ if exactly one of $r_a, r_b$ has a self-loop, otherwise we conclude $v \notin W$.

Now we briefly prove the correctness of the above procedure. For the first step, note that the degree-3 vertices are precisely the leaves of the antennas. Moreover, if a non-backtracking walk is started from an antenna vertex towards the leaves, then it will always find a leaf within $k$ steps. On the other hand, any body vertex is at least $k + 1$ steps from an antenna leaf, if only single-edges can be traversed.

The second step relies on the correctness of Algorithm 2, which we already analyzed in the previous section.

The third step deserves a bit more explanation. We would like to use the quantum algorithm of Childs et al. [CCD+03] on the body part of the candy graph. As we discuss in the next paragraph, for this it suffices to construct an adjacency list “oracle” for a graph, where all vertices have bounded degree (say, 3) and the body welded tree structure containing $r_a$ is one of the connected components. We construct this “oracle” starting from the input oracle and removing double edges. Then we disconnect the antennae by treating the roots (vertices with at least 4 non-double edges) in a special way. From a root we start a non-backtracking (random) walk of length $k$ along each edge. If we find a leaf after $k$ steps, then we know that the initial edge leads to the antenna vertex and therefore we remove the corresponding edge adjacent to the root.\footnote{Strictly speaking this results in a corrupted adjacency list oracle, because we did not delete the edge from the neighbor list of the vertex in the antenna connected to the root. This could be easily fixed with a bit of caution, but we note that the edge also gets automatically deleted if we use the block-encoding framework [GSLW19].}

Once we have the adjacency list oracle for the modified graph, we can construct a block-encoding of the modified adjacency matrix $A$ divided by the maximum degree (= 3) using standard techniques (see for example [GSLW19]). Once we have a block-encoding we can also perform Hamiltonian simulation [BCK15, LC19] efficiently (or implement polynomials of $A$ [GSLW19]), which in turn enables implementing the algorithm of [CCD+03]. Since the quantum walk algorithm of [CCD+03] finds the other root in the tree with probability $\Omega(1/\text{poly}(k))$, and the success probability can also be computed, we can make the algorithm succeed with probability 1 using amplitude amplification.

Once we have found the other root in the candy graph we can easily compute the parity of the number of self-loops, which in turn correctly identifies the weld vertices, since $G \in \mathcal{P}_k$. As all steps of the above procedure succeed with certainty, we get the following.

**Lemma 63.** The above quantum algorithm given any vertex $v$ in a graph $G \in \mathcal{P}_k$ outputs a bit which is 1 iff $v$ is a weld vertex. Moreover, the algorithm has query and time complexity $\text{poly}(k)$.\footnote{Strictly speaking this results in a corrupted adjacency list oracle, because we did not delete the edge from the neighbor list of the vertex in the antenna connected to the root. This could be easily fixed with a bit of caution, but we note that the edge also gets automatically deleted if we use the block-encoding framework [GSLW19].}
6.3 Classical lower bound

We show that our problem is classically hard by showing that it is hard to distinguish between a graph chosen randomly from $G_1$—a particular “difficult” subset of yes instances in $P_k$—and a graph chosen randomly from $G_2$—a particular “difficult” subset of no instances. $G_1$ is the set of yes instances containing $2^k - 1$ candy subgraphs, all of which are welded along a single cycle, such that exactly half of the candy subgraphs with an even number of self-loops have no self-loops. $G_2$ contains the same graphs as $G_1$ except that each candy graph is modified to take on the shape of a “double-bow-tie”: the weld vertices within a “bow-tie” are connected by an arbitrary (single) cycle, and exactly half of the roots have a self-loop.

We prove that classically at least $2^\Omega(k)$ queries are needed to distinguish between random instances from $G_1$ and $G_2$. With a slight abuse of notation, we also use $G_i$ to mean a random graph sampled from the set $G_i$. More precisely, we define the sampling procedures for $G_i$ as follows:

a. Call a graph a depth-$k$ root-joined binary tree if it is obtained by identifying the root vertices of two binary trees of depth $k$, one of which is called antenna and the other body. Create $4(2^k - 1)$ depth-$k$ root-joined binary trees, and arrange them in $2(2^k - 1)$ pairs.

b. For the first $2^k$ pairs mark one of the roots with a self-loop, and for the last $2^k - 1$ pairs mark both roots with a self-loop.

c. Pick a random bijection between body interior vertices and antenna vertices that are in a pair with exactly one self-loop, and pick another random bijection between body weld vertices and antenna vertices that are in a pair with an even number of self-loops. Connect vertices along the bijections with a double edge.

d. $G_1$: In each pair, join the body weld vertices in the two root-joined binary trees of the pair by a single random cycle that alternates between vertices in the two trees.

$G_2$: In each pair, join the body weld vertices within both root-joined binary trees by a single random cycle.

When an adjacency-list oracle is sampled for the above graphs, then vertex labels are randomly permuted, as well as the adjacency lists of every vertex.\footnote{Note that the above procedure samples from graph isomorphism classes non-uniformly; indeed, classes with richer automorphism groups tend to be sampled more often.}

First, we verify that a random no instance in $G_2$ is indeed far from the yes instances in expectation. Let us call a graph in $G_i$ reduced if all its advice edges and self-loops have been removed (i.e., in the above procedure we ignore steps b and c). Observe that reduced graphs in $G_1$ are bipartite (simply alternately color each layer in each welded tree). On the other hand, we prove that reduced graphs in $G_2$ are typically far from even being bipartite.

Second, we argue that welded and self-welded trees are hard to distinguish by a classical randomized algorithm that makes few queries. Then, it follows that graphs from $G_1$ and $G_2$ are themselves hard to distinguish since they are based on the welded and self-welded trees, respectively, but are otherwise the same. We argue this formally by defining a fixed simulator $G_s$ which we show behaves like both $G_1$ and $G_2$ when only sub-exponentially many queries are used.

6.3.1 A random graph in $G_2$ is typically far from any yes instance

We show that the reduced version of $G_2$ is typically constant-far from reduced graphs in $G_1$. It is easy to see that this also implies that (non-reduced) $G_2$ is typically constant-far from $G_1$.\footnote{Note that the above procedure samples from graph isomorphism classes non-uniformly; indeed, classes with richer automorphism groups tend to be sampled more often.}
Lemma 64. With probability at least $1 - \exp(-\Omega(2^k))$, we need to remove at least $2^k \cdot (2^k - 1)/16$ edges from a random reduced graph in $\mathcal{G}_2$ to make it bipartite.

Proof. Let us consider one self-welded component in a reduced graph $\mathcal{G}_2$, and in particular its induced subgraph $B$, spanned by the vertices in the weld layer and the layer adjacent to it, so that $m = 3 \cdot 2^{k-1}$ is the total number of vertices in $B$. We show the following:

Claim 65. With probability at least $1 - \exp(-\Omega(m))$, we need to remove at least $m/96$ edges from $B$ to make it bipartite.

This implies the statement of the lemma because there are $4 \cdot (2^k - 1)$ copies of $B$ in any graph in $\mathcal{G}_2$. If all of them are far from bipartite, in the sense of needing to remove at least $m/96$ edges, then the total number of edges that need to be removed is at least $2^k \cdot (2^k - 1)/16$. On the other hand, the probability of any of them not being far from bipartite is at most $4 \cdot (2^k - 1) \cdot \exp(-\Omega(2^k)) = \exp(-\Omega(2^k))$.

Proof. This can be argued following the strategy of [GR97, Lemma 7.4]. The sampling procedure in terms of $B$ can be described as follows: take $2^k - 1$ disjoint paths of length 2, enumerate all paths and give labels $1, 2, \ldots, 2^k$ to the endpoints of the paths, then sample a random permutation $\pi$ of $[2^k]$, and connect $\pi^{-1}(1)$ to $\pi^{-1}(2)$, $\pi^{-1}(2)$ to $\pi^{-1}(3)$, ..., and $\pi^{-1}(2^k)$ to $\pi^{-1}(1)$. Observe that $B$ can be alternatively sampled by fixing a cycle of length $2^k$, and randomly matching its vertices as follows: sample a random permutation $\pi$ of $[2^k]$, then connect with a length-2 path $\pi^{-1}(1)$ to $\pi^{-1}(2)$, $\pi^{-1}(2)$ to $\pi^{-1}(3)$, ..., and $\pi^{-1}(2^k)$ to $\pi^{-1}(1)$. It is easy to see that for a given $\pi$ the graphs sampled in the two different ways are isomorphic (via the permutation of the cycle vertices $\pi$).

It is more convenient to work with the second sampling procedure. Consider a particular partition $(U_1, U_2)$ of the cycle vertices in $B$ and any partition $(V_1, V_2)$ of all vertices in $B$ that extends $(U_1, U_2)$ in the sense that $U_1 \subset V_1$. We say an edge violates $(V_1, V_2)$ if it connects two vertices from the same $V_i$. Let $c := 2m/3$ denote the number of cycle vertices. We separate the analysis into two cases:

Case 1. There are at least $c/64$ violating cycle edges.

Case 2. There are fewer than $c/64$ violating cycle edges. Then assume without loss of generality that $|U_1| \leq |U_2|$. If $|U_1| \leq c/2 - c/128$, then there are at most $c - c/64$ non-violating cycle edges, so there are at least $c/64$ violating cycle edges, which contradicts our assumption. Hence $|U_1| \geq c/2 - c/128 \geq 3c/8$.

To randomly match vertices in $U_1$, we can take an arbitrary unmatched vertex $u$ in $U_1$ and connect it randomly to a cycle vertex $w$ by a path of length 2. Note that if $w$ is in $U_2$, then we must have one violating edge on the length-2 path with respect to any $(V_1, V_2)$ partition. We continue until all the vertices in $U_1$ are matched.

At each step, the probability of $w$ being in $U_2$ is at least $1/2$, so the probability that fewer than $c/64$ violating edges are created in a total of at least $3c/16$ steps is at most

$$
\sum_{i=0}^{c/64-1} \binom{3c/16}{i} \cdot 2^{i-3c/16} \leq 2^{-11c/64} \sum_{i=0}^{c/64} \binom{3c/16}{i} \leq 2^{35/96}H(16/3/64) - \frac{11c}{64} \leq 2^{-0.11c}, \quad (1)
$$

where $H(p) := -p \log(p) - (1-p) \log(1-p)$ is the binary entropy function, and the second inequality follows from [Juk11, Corollary 22.9] stating

$$
\forall 0 < h \leq n/2 : \sum_{j=0}^{h} \binom{n}{j} \leq 2^{n \cdot H(h/n)}. \quad (2)
$$
Note that this bounds the probability that any partition \((V_1, V_2)\) extending \((U_1, U_2)\) has fewer than \(c/64\) violating edges.

We call a partition \((V_1, V_2)\) of all vertices in \(B_1\) “bad” if it has fewer than \(c/64\) violating edges. Then by the union bound

\[
\Pr(\exists \text{ bad partition } (V_1, V_2)) \leq \sum_{(U_1, U_2)} \Pr(\exists \text{ bad partition extending } (U_1, U_2)) \leq \#\{(U_1, U_2) : \text{ Case 2}\} \times 2^{-0.11c},
\]

where \((U_1, U_2)\) denotes a partition of the cycle edges and \#\{(U_1, U_2) : \text{ Case 2}\} is the number of such partitions with fewer than \(c/64\) violating cycle edges. For each \(i < c/64\), each partition which has \(i\) violating cycle edges is determined by the choice of those edges. Therefore

\[
\#\{(U_1, U_2) : \text{ Case 2}\} = \sum_{i=0}^{c/64-1} \binom{c}{i} \leq 2^{0.09c}.
\]

Therefore \(\Pr(\exists \text{ bad partition } (V_1, V_2)) \leq 2^{-0.02c} = \exp(-\Omega(c)) = \exp(-\Omega(m))\) as desired. \(\square\)

This completes the proof of the lemma. \(\square\)

### 6.3.2 Hardness of distinguishing welded and self-welded trees

Let \(B\) be a classical randomized algorithm. We consider the difficulty of \(B\) winning four different games by querying an oracle that provides black-box access to either a self-welded or welded tree. To be clear, in this subsection, by a “self-welded tree” we mean two binary trees of depth \(k\) where the weld vertices in each are connected to themselves by a single cycle; by a “welded tree” we mean two binary trees of depth \(k\) where the weld vertices in each are connected to weld vertices in the other by a single cycle. A self-welded tree looks like a candy graph without the antenna and a welded tree looks like a double-bow-tie graph without the antenna (cf. Figure 1).

The difficulty of winning can be quantified by bounding the winning probability by a function of \(t\), the number of queries that \(B\) makes. Throughout, the winning probability is over three or four sources of randomness: the internal randomness of \(B\), the cycle forming the (self-)weld, the vertex labelings, and also the random choice of the starting vertex in two of the games. These four games are described in the four lemmas below. They relate to the difficulty of distinguishing welded from self-welded trees because, unless they are won, the welded and self-welded trees both look just like a large (effectively infinite) binary tree to \(B\).

We can, without loss of generality, assume that \(B\) does not query labels which have already been queried and that each query to a vertex label returns the labels of all neighbors of that vertex. For notational convenience, we shall use the concept of the “knowledge graph” of \(B\), as introduced in [GR97, Section 7]. The knowledge graph of \(B\) is the graph that it sees, which consists of all vertices whose labels it has either queried or has been returned by the oracle in answer to those queries, as well as the edges between them. The knowledge graph can change every time \(B\) makes a query. In the following, we make the further assumption that \(B\) can only query labels in its knowledge graph. This assumption shall later be justified by Lemma 70.

**Lemma 66.** Let Game A be that of finding a cycle in a self-welded tree given the label of root. If \(B\) uses \(t \leq 2^{k-1}\) queries, then its probability of winning Game A is at most \(O(t^2 \cdot 2^{-k/2})\).
Proof. As explained in [CCD+03], the winning probability can be expressed as the probability that a random embedding $\pi$ of a random rooted binary tree $T$, with $t$ vertices and root at $\text{ROOT}$, into a random self-welded tree $G$ contains a cycle, where the randomness in the embedding corresponds to the random vertex labelings, and the randomness in the rooted binary tree corresponds to the internal randomness of $\mathcal{B}$. The terms “random embedding” and “rooted binary tree” are defined in [CCD+03, Sec. IV].

Therefore, it suffices to show that for an arbitrary fixed $T$, the probability that its random embedding $\pi(T)$ in a random $G$ contains a cycle is at most $O(t^2 \cdot 2^{-k/2})$. We denote this probability by $E_G[P^G(T)]$ to match the notation of [CCD+03]. Our proof is very similar to that of [CCD+03, Lemma 8].

The probability that $\pi(T)$ contains a cycle is the same as the probability that there exist vertices $a \neq b$ in $T$ such that $\pi(a) = \pi(b)$.

First note that the probability that $\pi(T)$ ever goes to a depth smaller than $k/2$ after reaching a weld vertex is at most $t^2 \cdot 2^{-k/2-1}$, as $\pi$ must embed $T$ to the left $k/2 + 1$ times starting from some vertex in $T$ and there are at most $t$ paths from root to leaf in $T$, each containing at most $t$ vertices.

Now fix any pair of vertices $a \neq b$ in $T$. We shall bound the probability that $\pi(a) = \pi(b)$. Let $P$ be the path in $T$ from $a$ to $b$, and $c$ the vertex in $T$ closest to the root on $P$. Let $P_1$ and $P_2$ be the paths in $T$ from $c$ to $a$ and $b$ respectively. The vertices at depths $k/2 + 1, \ldots, k$ divide into $2^{k/2}$ complete binary trees of depth $k/2$ which we denote $S_1, \ldots, S_{2k/2}$.

We have already considered the case that $\pi(P_1)$ ever goes to depth smaller than $k/2$ after reaching a weld vertex, so we can assume it does not. In this case, $\pi(a) = \pi(b)$ must lie in one of the subtrees $S_j$ and one of $\pi(P_1)$ or $\pi(P_2)$ must go from a weld vertex to a vertex in $S_j$. The probability of the latter occurring is at most $2^{k/2}/(2^k - t) \leq 2 \cdot 2^{-k/2}$ for $t \leq 2^{k-1}$ due to the randomness of the self-weld. There are $\binom{k}{2}$ choices of $a, b$, so the probability of finding a cycle is at most $t^2 \cdot 2^{-k/2}$.

Overall, we have shown
$$E_G[P^G(T)] \leq t^2 \cdot 2^{-k/2} + t^2 \cdot 2^{-k/2-1} = O(t^2 \cdot 2^{-k/2})$$

as claimed. \hfill $\square$

Lemma 67. Let Game $B$ be that of finding $\text{ROOT}$ or a cycle in a self-welded tree given the label of a uniformly random starting vertex. If $B$ uses $t \leq 2^{k-4}$ queries, then its probability of winning is at most $O(t^2 \cdot 2^{-k/4})$.

Proof. Again, we can bound this probability by bounding the probability that a random embedding $\pi$ of an arbitrary fixed rooted binary tree $T$ contains $\text{ROOT}$ or a cycle. The only difference in setup is that the root is now at the random vertex in the self-welded tree and the probability is also over this randomness. First, the probability of the random vertex being at depth less than $3k/4$ is at most $2^{3k/4} \cdot 2^{-k} = 2^{-k/4}$. We may suppose that $B$ wins in this case.

Therefore, consider when the random vertex is at depth more than $3k/4$. Then, similarly to the proof of the Lemma 6, we can bound the probability that $\pi(T)$ ever goes to depth smaller than $k/2$ by $t^2 \cdot 2^{-k/4-1}$.

The probability of finding a cycle is again $t^2 \cdot 2^{-k/2}$ by the same argument as before. Therefore, the overall probability of $B$ finding the $\text{ROOT}$ or a cycle starting from a random vertex is $O(t^2 \cdot 2^{-k/4})$. \hfill $\square$

Lemma 68. Given a $\text{ROOT}$, let Game $C$ be that of finding the other $\text{ROOT}$ or a cycle in a welded tree. If $B$ uses $t \leq 2^{k-1}$ queries, then its probability of winning is at most $O(t^2 \cdot 2^{-k/2})$.

\textsuperscript{12}Strictly speaking, $T$ has $O(t)$ vertices as we are assuming the oracle answers each query to a vertex label with all labels of its neighbors. But, because the lemma is only up to big-$O$, this is unimportant.
Proof. The proof is given by [CCD+03, Lemma 8] and is very similar to that of Lemma 66.

Lemma 69. Let Game D be that of finding a root or a cycle in a welded tree given the label of a uniformly random starting vertex. If B uses \( t \leq 2^{k-1} \) queries, then its probability of winning is at most \( O(t^2 \cdot 2^{-k/4}) \).

Proof. The proof is essentially the same as that of Lemma 67 and Lemma 68.

6.3.3 Hardness of distinguishing \( G_1 \) and \( G_2 \)

Let \( A \) be a classical randomized algorithm that distinguishes \( G_1 \) from \( G_2 \). Our strategy for proving the classical lower bound is to consider how the “knowledge graph” of \( A \) changes as it interacts with oracles that provide black-box descriptions of a random graph in \( G_1 \) or \( G_2 \). Recall that the knowledge graph of \( A \) is the graph that it sees. In the following, we refer to vertices in the knowledge graph at each step as “known” and vertices outside it as “unknown”.

We shall show that, regardless of whether we are given a graph from \( G_1 \) or \( G_2 \), the knowledge graph of \( A \) looks the same at each step with high probability if only a small number of queries have been made. To be more precise, we shall define a graph simulator \( G_s \) such that the knowledge graph of \( A \) when it interacts with either \( G_1 \) or \( G_2 \) looks the same as when it interacts with \( G_s \) with all but exponentially small probability, if only a polynomial number of queries are made.

To the benefit of \( A \), we assume that \( A \) has knowledge of which labels are antenna, root, or body labels. We also suppose that all reached root labels have been queried for free so that they and their neighbors all belong to the knowledge graph of \( A \) at the outset. We can assume that \( A \) does not query labels that have already been queried and that each query to a vertex label returns the labels of all neighbors of that vertex. Then, at each step, \( A \) must choose to perform one of the following Actions:

I. Query an unknown body label.

II. Query an unknown antenna label.

III. Query a known body label.

IV. Query a known antenna label.

For ease of reference, we call the candy subgraphs of \( G_1 \) and double-bow-tie subgraphs of \( G_2 \) their “base graphs”. At each step of \( A \), we say that a base graph has been tapped if any one of its vertices, other than a root, is already in the knowledge graph of \( A \).

At each step of \( A \), we assume to its benefit that:

A1. \( A \) wins if, upon taking Action I or Action II, the label of a vertex in a tapped base graph is revealed.

A2. \( A \) wins if, upon taking Actions III or IV, the vertex whose label is queried is connected by an advice edge to a vertex in a tapped base graph.

A3. When \( A \) performs Action II, the entire half-antenna in which the queried antenna label resides is revealed (i.e., all vertex labels within the antenna as well as edges between them become known).

A4. When \( A \) performs Action IV, the entire half-antenna in which the queried antenna label resides is revealed.
The point of making these assumptions is simply to make it easier to analyze how the knowledge graph of \( \mathcal{A} \) changes. We consider A1 and A2 because when they do not occur, \( \mathcal{A} \) is essentially restricted to local traversal in each base graph. We consider A3 and A4 so that we do not need to unduly worry about how \( \mathcal{A} \) traverses the antenna, which is irrelevant for distinguishing \( \mathcal{G}_1 \) from \( \mathcal{G}_2 \).

**Lemma 70.** If \( \mathcal{A} \) uses \( t \) queries, then its probability of winning via either A1 or A2 above is at most \( O(t^2 \cdot 2^{-k}) \).

**Proof.** In \( t \) queries, at most \( 2t \) base graphs can be tapped. However, there are \( 2^k \) base graphs in \( \mathcal{G}_1 \) and \( \mathcal{G}_2 \). Therefore, the probability of winning via A1 at each step is at most \( O(t/2^k) \), so the probability of winning via A1 at any step is \( O(t^2/2^k) \). The same bound holds for winning via A2 due to the randomness in the connection of advice edges.

Let us consider a graph simulator \( \mathcal{G}_s \) that can respond to the different Actions of \( \mathcal{A} \). As mentioned previously, we shall argue that its responses are similar to the responses of a random graph from either \( \mathcal{G}_1 \) and \( \mathcal{G}_2 \) when the number of queries is not too large. The responses of \( \mathcal{G}_s \) depend on the Actions of \( \mathcal{A} \) as follows:

I. \( \mathcal{G}_s \) returns a label chosen uniformly at random from antenna labels not currently in the knowledge graph. \( \mathcal{G}_s \) also returns three labels chosen uniformly at random from body labels not currently in the knowledge graph.

II. \( \mathcal{G}_s \) returns an entire half-antenna with labels chosen uniformly at random from antenna labels not currently in the knowledge graph. \( \mathcal{G}_s \) also returns one body label chosen uniformly at random from body labels not currently in the knowledge graph.

III. \( \mathcal{G}_s \) does one of the following two case-dependent actions.

(a) If the queried body label is known because it is connected to a known antenna vertex, \( \mathcal{G}_s \) returns three labels chosen uniformly at random from body labels not currently in the knowledge graph.

(b) If the queried body label is known because it is connected to a known body vertex, \( \mathcal{G}_s \) returns two labels chosen uniformly at random from body labels not currently in the knowledge graph as well as a label chosen uniformly at random from antenna labels not currently in the knowledge graph.

IV. \( \mathcal{G}_s \) does one of the following two case-dependent actions.

(a) If the queried antenna label is known because it is connected to a known body vertex, \( \mathcal{G}_s \) returns an entire half-antenna with labels chosen uniformly at random from antenna labels not currently in the knowledge graph.

(b) If the queried antenna label is known because it is connected to a known antenna vertex, \( \mathcal{G}_s \) returns a label chosen uniformly at random from body labels not currently in the knowledge graph.

Note that in cases III and IV, \( \mathcal{G}_s \) also returns vertex labels that are consistent with its knowledge graph when querying a known label, but we omitted describing this for convenience. In particular, in case IV(b), \( \mathcal{G}_s \) returns antenna labels consistent with the relevant half-antenna which must be entirely known by assumptions A3 and A4.
Now, suppose that \( \mathcal{A} \) makes at most \( t \leq 2^{k-1} \) queries to either \( \mathcal{G}_1 \) or \( \mathcal{G}_2 \). Then Lemma 70 says that the probability of it winning, at any step, via A1 or A2 is at most \( O(t^2 \cdot 2^{-k}) \). We henceforth assume that neither A1 nor A2 ever occurs. In this case, we explain why \( \mathcal{G}_s \) responds like \( \mathcal{G}_1 \) and \( \mathcal{G}_2 \) when \( \mathcal{A} \) performs each of the following Actions:

I. When \( \mathcal{A} \) queries unknown body label, because A1 does not occur, the corresponding body vertex as well as its neighbors lie in two untapped base graphs. Therefore, the labels returned are outside the knowledge graph of \( \mathcal{A} \) and so are distributed at random among those not in its knowledge graph, just like those returned by \( \mathcal{G}_s \) in response to Action I.

II. When \( \mathcal{A} \) queries an unknown antenna label, because A1 does not occur, the corresponding entire half-antenna (cf. A3) and the neighboring body vertex lie in two untapped base graphs. Therefore, the labels returned are outside the knowledge graph of \( \mathcal{A} \) and so are distributed at random among those not in its knowledge graph, just like those returned by \( \mathcal{G}_s \) in response to Action II.

III. When \( \mathcal{A} \) queries a known body label, there are two cases, depending on how the body vertex is known.

(a) If it is known because it is connected to a known antenna vertex by an advice edge, then it must be in an untapped base graph because A2 does not occur. Hence \( \mathcal{G}_s \) behaves the same as \( \mathcal{G}_i \).

(b) If it is known because it is connected to a known body vertex, then Lemma 66 and Lemma 67 (respectively Lemma 68 and Lemma 69) imply that \( \mathcal{G}_s \) behaves the same as \( \mathcal{G}_2 \) (respectively \( \mathcal{G}_1 \)), except with probability at most \( O(t^2 \cdot 2^{-k/4}) \). Lemma 66 and Lemma 68 address the case when the queried body vertex is connected to the root of a base graph. Lemma 67 and Lemma 69 address the case when it is connected to a random vertex in the body of a base graph.

IV. When \( \mathcal{A} \) queries a known antenna label, there are two cases, depending on how the antenna vertex is known.

(a) If it is known because it is connected to a known body vertex by an advice edge, then it must be in an untapped base graph because A2 does not occur. Hence \( \mathcal{G}_s \) behaves the same as \( \mathcal{G}_i \) (cf. A4).

(b) If it is known because it is connected to a known antenna vertex, then the entire half-antenna must already be known (cf. A4), and the body vertex connected to it must be in an untapped base graph because A2 does not occur. Hence \( \mathcal{G}_s \) behaves the same as \( \mathcal{G}_i \).

Therefore, the probability of \( \mathcal{A} \) distinguishing \( \mathcal{G}_i \) from \( \mathcal{G}_s \) is at most \( O(t^2 \cdot 2^{-k/4}) \) for each \( i = 1, 2 \). Therefore, the probability of distinguishing \( \mathcal{G}_1 \) from \( \mathcal{G}_2 \) is also at most \( O(t^2 \cdot 2^{-k/4}) \). Now, Lemma 64 implies that there exists a constant \( \epsilon \) such that for all \( k \) sufficiently large, the probability that \( \mathcal{G}_2 \) is \( \epsilon \)-close to \( \mathcal{G}_1 \) is at most 0.1. Let \( \mathcal{G}_2' \) be a random graph from those in \( \mathcal{G}_2 \) that are \( \epsilon \)-far from \( \mathcal{G}_1 \). Then the probability of distinguishing \( \mathcal{G}_2' \) from \( \mathcal{G}_2 \) is at most 0.1. Therefore, the probability of distinguishing \( \mathcal{G}_2' \) from \( \mathcal{G}_1 \) is at most 0.1 + \( O(t^2 \cdot 2^{-k/4}) \). Hence, to distinguish \( \mathcal{G}_1 \) from \( \mathcal{G}_2' \) with probability at least 2/3 requires \( t = \exp(\Omega(k)) \) queries. This establishes the desired classical lower bound in Theorem 56.
7 Open problems

We conclude by discussing a few open problems.

We proved that $p$-uniform hypergraph properties have at most a power $3p$ separation between quantum and randomized query complexity. However, we do not know if this is tight.

**Open Problem 71.** What is the largest possible separation between $Q(f)$ and $R(f)$ for $p$-uniform hypergraph properties $f$? That is, what is the largest $k$ for which there exists such an $f$ with $R(f) = \Omega(Q(f)^k)$?

We remark that the problem is open even for the case $p = 1$ of fully symmetric functions, where the best upper bound is $k \leq 3$ due to Chailloux [Cha18], and the best lower bound is $k \geq 2$ for the OR function [Gro96]. For larger $p$, it is at least possible to exhibit functions with a power separation of $k \geq p/2$ by appealing to Proposition 44, choosing the function $f$ in Proposition 44 to be Forrelation [AA15], and using an upper bound of $b(G) = O(m)$ for $G$ the action of $S_m$ on $\binom{m}{p}$ points (Theorem 3.2 of [Hal12]). Thus there must be some dependence on $p$.

In Corollary 55, we showed that well-shuffling permutation groups must be constructed out of a constant number of primitive groups with sufficiently large minimal base size. We conjecture that a converse holds, which would imply a complete dichotomy regarding which permutation groups allow super-polynomial quantum speedups and which do not:

**Conjecture 72.** Let $\mathcal{G}$ be a collection of permutation groups that satisfies conditions (i)–(iv) of Corollary 55. Then $\mathcal{G}$ is well-shuffling, and thus $R(f) = Q(f)^{O(1)}$ for every $f \in F(\mathcal{G})$.

Conjecture 72 is based on our intuition that it is challenging to find an interesting family of permutation groups that satisfies these conditions, but that cannot be constructed out of well-shuffling primitive groups using the transformations that preserve the well-shuffling property. For example, even small subgroups of a wreath product of well-shuffling groups can remain well-shuffling: the product of two well-shuffling permutation groups $G_1 \times G_2$ (as defined in Definition 40) is well-shuffling, as shown in Theorem 41. But $G_1 \times G_2$ can be viewed as a subgroup of $G_1 \wr G_2$ in imprimitive action, and is potentially much smaller than the full wreath product. Based on this intuition, we expect that it is possible to prove Conjecture 72 essentially in the same way we classified the well-shuffling primitive groups, via reduction to the symmetric and alternating groups.

In this form, Conjecture 72 would also imply that the well-shuffling property completely characterizes whether a collection of permutation groups allows super-polynomial quantum speedups or not. It would be interesting to prove such a characterization directly, without appealing to the structure of such groups.

In Proposition 44, we showed how to construct super-polynomial quantum speedups out of any sufficiently small permutation group. However, the construction requires large alphabets (indeed, even larger than the number of input symbols). Might it be possible to construct such functions over a smaller alphabet, or even a Boolean alphabet? Note that some permutation groups cannot be realized as the group of symmetries of a function with Boolean inputs [Kis98].

Finally, while we constructed a separation between classical and quantum graph property testing in the adjacency list model, it remains unclear whether such a separation exists for graph properties of practical interest. For example, it remains open whether there could be an exponential quantum speedup for testing bipartiteness [ACL11]. Going beyond that particular example, it would be interesting to investigate the possibility of a super-polynomial quantum speedup for testing monotone graph properties (i.e., those that remain satisfied when adding edges to a yes instance).
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A Proof of the quantum minimax lemma

We prove Lemma 5, which we restate below.

Lemma 5 (Minimax for bounded-error quantum algorithms). Let $f$ be a (possibly partial) Boolean function with $\text{Dom}(f) \subseteq \Sigma^n$, and let $\epsilon \in (0, 1/2)$. Then there is a distribution $\mu$ supported on $\text{Dom}(f)$ which is hard for $f$ in the following sense: any quantum algorithm using fewer than $Q_\epsilon(f)$ quantum queries for computing $f$ must have average error more than $\epsilon$ on inputs sampled from $\mu$.

Proof. By [BSS03], there is a finite bound $B$ expressible in terms of $n$ and $|\Sigma|$ on the necessary size of the work space register for a quantum algorithm computing $f$ with error at most $\epsilon$. This means the quantum query algorithms we deal with can be assumed without loss of generality to have work space size $B$. A quantum algorithm making $T$ queries can be represented as a sequence of $T$ unitary matrices of size upper bounded by $B$; this can be arranged as a finite vector of complex numbers. It is not hard to see that the set of all such valid quantum algorithms is a compact set.

For a quantum algorithm $Q$, let $\text{err}(Q, x)$ denote the error $Q$ makes when run on input $x \in \text{Dom}(x)$; this is $\Pr[Q(x) \neq f(x)]$, where $Q(x)$ is the random variable for the measured output of $Q$ when run on $x$. We note that $\text{err}(Q, x)$ is a continuous function of $Q$. Let $v_Q$ be the vector in $\mathbb{R}^{|\text{Dom}(f)|}$ defined by $v_Q[x] := \text{err}(Q, x)$. Then $v_Q$ is a continuous function of $Q$. Further, let $V$ be the set of all such vectors $v_Q$ for valid quantum algorithms $Q$ which make at most $Q_\epsilon(f) - 1$ queries. Since the set of such valid quantum algorithms is compact and since $v_Q$ is continuous in $Q$, we conclude that $V$ is compact. Furthermore, we claim that $V$ is convex: this is because for any two quantum algorithms $Q$ and $Q'$, there is a quantum algorithm $Q''$ that behaves like their mixture (in terms of its error on each input $x$).

Next, let $\Delta \subseteq \mathbb{R}^{|\text{Dom}(f)|}$ be the set of all probability distributions over $\text{Dom}(f)$. Then $\Delta$ is also convex and compact. Finally, define $\alpha : V \times \Delta \rightarrow \mathbb{R}$ by $\alpha(v, \mu) := \mathbb{E}_{x \sim \mu} v[x] = \sum_{x \in \text{Dom}(f)} \mu[x] v[x]$. Then $\alpha$ is continuous in each coordinate, and is saddle: that is, $\alpha(\cdot, \mu)$ is convex for each $\mu \in \Delta$ (indeed, it is linear), and $\alpha(v, \cdot)$ is concave for each $v \in V$ (indeed, it is also linear). A standard minimax theorem (e.g., [Sio58]) then gives us

$$\min_{v \in V} \max_{\mu \in \Delta} \alpha(v, \mu) = \max_{\mu \in \Delta} \min_{v \in V} \alpha(v, \mu).$$

For the left-hand side, it is clear that the maximum over $\mu$ (once the vector $v$ has been chosen) is the same as the maximum over $x \in \text{Dom}(f)$ of $v[x]$. This makes the left-hand side the minimum
over $v \in V$ of $\|v\|_\infty$, or equivalently, the minimum worst-case error of quantum algorithms making at most $Q_\epsilon(f) - 1$ queries. By the definition of $Q_\epsilon(f)$, this minimum must be strictly greater than $\epsilon$ (or else $Q_\epsilon(f)$ would be smaller). Hence the left-hand side is strictly greater than $\epsilon$.

Looking at the right-hand side, we get a single distribution $\mu$ such that every quantum algorithm $Q$ making at most $Q_\epsilon(f) - 1$ queries must make error greater than $\epsilon$ against $\mu$, as desired. \qed

**B Quantum speedup for deep wreath product symmetries**

Here, we complete the proof of Theorem 54. We define 1-Fault Direct Trees almost exactly as in Definition 2 of [Kim12], with the addition that we allow the Boolean evaluation tree to be an arbitrary block composition $f_1 \circ f_2 \circ \cdots \circ f_d$ of (possibly different) direct Boolean functions $f_i$, where $f_i$ has $n_i$ inputs. As long as each $f_i$ is a symmetric function, then 1-Fault Direct Trees is symmetric under the iterated wreath product $S_{n_1} \wr S_{n_2} \wr \cdots \wr S_{n_d}$ of symmetric groups because for any vertex $v$, the quantity $\kappa(v)$ defined in Definition 2 of [Kim12] does not depend on the ordering of the children of $v$.

Recall that we choose $f_i: S \to \{0, 1\}$ where $S \subseteq \{0, 1\}^{n_i}$ consists of the $n_i$-bit strings that have Hamming weight in $\{0, [n_i/2], [n_i/2], n_i\}$, and $f_i(x) = 0$ if and only if $x = 1^{n_i}$. The function $f_i$ is defined to be trivial on $x$ if $x = 0^{n_i}$ or $1^{n_i}$; otherwise $f_i$ is defined to be fault on $x$. If $f_i(x)$ is trivial on $x$, then all input bits of $x$ are defined to be strong. Otherwise, if $f_i(x)$ is fault on $x$, the bits $j$ such that $x_j = 1$ are defined to be weak, and the bits $j$ such that $x_j = 0$ are defined to be strong.

Using the notation from Definition 2.1 of [ZKH12], we define a span program $P_i$ for $f$ by $v_j = \frac{1}{\sqrt{n_i}} \in \mathbb{R}^1$ and $\chi_j = \pi_j$ for all $j \in [n_i]$. We now verify that $P_i$ satisfies the key properties that are needed to prove the quantum query upper bound in [Kim12].

We have $r_0 = \frac{1}{\sqrt{n_i}}(1, 1, \ldots, 1) \in \mathbb{C}^{n_i}$ and $C = 1$ as in Definition 2.2 of [ZKH12], which gives the following:

- $\text{WSIZE}_1(P_i, x) = 1$ if input $x$ makes $f_i$ trivial. If $x = 0^n$, choosing $\vec{w} = r_0$ suffices to show this. Otherwise, if $x = 1^n$, then we are forced to choose $\vec{w} = r_0$.

- $\text{WSIZE}_1(P_i, x) \leq 3$ if input $x$ makes $f_i$ fault. Suppose $x = 0^{[n_i/2]}1^{[n_i/2]}$. Choosing $w_j = \frac{\sqrt{n_i}}{[n_i/2]}$ for $j \in [[n_i/2]]$ gives witness size $\frac{n_i}{[n_i/2]} \leq 3$ (tight for $n_i = 3$). Similarly, for $x = 0^{[n_i/2]}1^{[n_i/2]}$, we take $w_j = \frac{\sqrt{n_i}}{[n_i/2]}$ for $j \in [[n_i/2]]$ to get witness size $\frac{n_i}{[n_i/2]} \leq 2$.

- For $\text{WSIZE}_d(P_i, x)$, $s_j$ do not affect the witness size, where the $j$th input bit is weak. This is because the only weak input bits are those where $x_j = 1$ and $f_i$ is fault on $x$. Then $f_i(x) = 1$, and we have $\chi_j = 0$, so we are forced to take $w_j = 0$.

Now that these three conditions are satisfied, the proof of Theorem 2 of [Kim12] goes through without modifications, and shows that this version of 1-Fault Direct Trees has quantum query complexity $O(1)$.

It remains to lower bound the randomized query complexity of 1-Fault Direct Trees as we have defined it. Suppose we further promise that at every vertex in the tree corresponding to $f_i$, the child subtrees are partitioned into two sets, one of size $\lfloor n_i/2 \rfloor$ and one of size $\lceil n_i/2 \rceil$, such that all of the subtrees in one part have the same values at corresponding leaves. Then with this additional promise, the problem remains at least as hard as the 1-Fault NAND Trees problem: if we always take the partition to be the $\lfloor n_i/2 \rfloor$ left vertices and the $\lceil n_i/2 \rceil$ right vertices, then it is in fact equivalent to 1-Fault NAND Trees where some inputs are repeated. [ZKH12] proved that depth-$d$ 1-Fault NAND Trees have randomized query complexity at least $\Omega(\log d)$, which completes our proof.

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