Quantum Time-Space Tradeoff for Finding Multiple Collision Pairs

Yassine Hamoudi and Frédéric Magniez
Université de Paris, IRIF, CNRS, F-75006 Paris, France.
{hamoudi,magniez}@irif.fr

Abstract

We study the problem of finding $K$ collision pairs in a random function $f : [N] \rightarrow [N]$ by using a quantum computer. We prove that the number of queries to the function in the quantum random oracle model must increase significantly when the size of the available memory is limited. Namely, we demonstrate that any algorithm using $S$ qubits of memory must perform a number $T$ of queries that satisfies the tradeoff $T^3 S \geq \Omega(K^3 N)$. Classically, the same question has only been settled recently by Dinur [Din20, Eurocrypt’20], who showed that the Parallel Collision Search algorithm of van Oorschot and Wiener [OW99] achieves the optimal time-space tradeoff of $T^2 S = \Theta(K^2 N)$. Our result limits the extent to which quantum computing may decrease this tradeoff. We further show that any improvement to our lower bound would imply a breakthrough for a related question about the Element Distinctness problem. Our method is based on a novel application of Zhandry’s recording query technique [Zha19, Crypto’19] for proving lower bounds in the exponentially small success probability regime.

1 Introduction

The efficiency of a cryptographic attack is a hard-to-define concept that must express the interplay between different computational resources [Wie04, Ber05, Ber09]. Arguably, the two most used criteria are the time complexity, measured for instance as the number of queries to a random oracle, and the space complexity, which is the memory size needed to perform the attack. Time-space tradeoffs aim at connecting these two quantities together by studying how much time increases when the available space decreases. Devising security proofs that are sensitive to memory constraints is a challenging program. Indeed, very few tools are available to quantify the extent to which space impacts the security level of a scheme. A recent line of work [TT18, JT19, GJT20] has made some progress for the case of classical attackers with bounded memory. The development of quantum computing asks the question of whether the access to quantum operations and quantum memories may lower the security levels. The answer is unclear when taking space into account. Indeed, many quantum “speed-ups” come at the cost of a dramatic increase in the space requirement [BHT98, Amb07, LZ19]. A central open question is whether a speed-up both in term of time and space complexities is achievable for such problems?

The focus of this work is to provide time-space tradeoff lower bounds for the problem of finding multiple collision pairs in a random function. The search for a single collision pair is one of the cornerstones of cryptanalysis. Classically, the birthday attack can be achieved by the mean of a memoryless (i.e. logarithmic-size memory) algorithm using the Pollard’s rho method [Pol75]. On the other hand, the quantum BHT algorithm [BHT98] requires less queries to the
random function, but the product of its time and space complexities is in fact higher than that of the classical attack! In this paper, we address the more general problem of finding multiple collision pairs in a random function. This task plays a central role in low-memory meet-in-the-middle attacks [OW99, Din20]. It has applications in many problems, such as double and triple encryption [OW99], subset sum [DDKS12, DEM19], k-sum [Wag02], 3-collision [JL09], etc. Recently, it has also been used to attack the post-quantum cryptography candidates NTRU [Vre16] and SIKE [ACC+18].

The celebrated classical Parallel Collision Search algorithm of van Oorschot and Wiener [OW99] can find as many collision pairs as desired in a time that depends on the available memory. The question of whether this algorithm achieves the optimal classical time-space tradeoff has been settled positively only recently by Chakrabarti and Chen [CC17] (for the case of 2-to-1 random functions) and Dinur [Din20] (for the case of uniformly random functions). In the quantum setting, no time-space tradeoff was known prior to our work. In other words, it could have been the case that a memoryless quantum attack outperforms the Parallel Collision Search algorithm with unlimited memory capacity.

We point out that time-space tradeoffs have been studied for a long time in the complexity community [BFK+81, Bea91, BFMadH+87, Yao94, BSSV03, Abz90, MNT93]. However, only few results are known in the quantum setting (Sorting [KŠW07], Boolean Matrix-Vector and Matrix-Matrix Multiplication [KŚW07], Evaluating Solutions to Systems of Linear Inequalities [ÅS¥W09]). Apart from our work, all existing quantum tradeoffs are based on the hardness of finding multiple preimages of a given value in a random function. We use the machinery developed in our paper to give a new simpler proof of the latter result.

1.1 Our Results

The Collision Pairs Finding problem asks to find a certain number $K$ of disjoint collision pairs in a random function $f : [M] \rightarrow [N]$ where $M \geq N$. A collision pair (or simply collision) is a pair of values $x_1 \neq x_2$ such that $f(x_1) = f(x_2)$. Two collisions $(x_1, x_2)$ and $(x_3, x_4)$ are disjoint if $x_1, \ldots, x_4$ are all different. We measure the time $T$ of an algorithm solving this problem as the number of query access to $f$, and the space $S$ as the amount of memory used. We assume that the output is produced in an online fashion, meaning that a collision can be output as soon as it is discovered. The length of the output is not counted toward the space bound and a same collision may be output several times (but it contributes only once to the total count). The requirement for the collisions to be disjoint is made to simplify our proofs later on. We note that a random function $f : [N] \rightarrow [N]$ contains $(1 - 2/e)N$ disjoint collisions on average [FO89].

Classically, the Parallel Collision Search algorithm [OW99] achieves an optimal [Din20] time-space tradeoff of $T^2S = \tilde{\Theta}(K^2N)$ for any amount of space $S$ between $\tilde{\Theta}(\log N)$ and $\tilde{O}(K)$. In the quantum setting, the BHT algorithm [BHT98] can find a single collision in time $T = \tilde{O}(N^{1/3})$ and space $S = \tilde{O}(N^{1/3})$. In Appendix A, we adapt it for finding an arbitrary number $K$ of collisions at cost $T^2S \leq \tilde{O}(K^2N)$. For the sake of simplicity, we do not require these collisions to be disjoint. This is the same tradeoff as classically, except that the space parameter $S$ can hold larger values up to $\tilde{O}(K^{2/3}N^{1/3})$, hence the existence of a quantum speed-up when there is no memory constraint.

**Proposition A.1.** For any $1 \leq K \leq O(N)$ and $\tilde{\Theta}(\log N) \leq S \leq \tilde{O}(K^{2/3}N^{1/3})$, there exists a bounded-error quantum algorithm that can find $K$ collisions in a random function $f : [N] \rightarrow [N]$ by making $T = \tilde{O}(K/\sqrt{N/S})$ queries and using $S$ qubits of memory.

The BHT algorithm achieves the optimal time complexity for finding a single collision [AS04, Zha15]. Our first main result is to provide a similar lower bound for the problem of finding $K$ disjoint collisions. We demonstrate that the optimal time complexity must satisfy $T \geq \Omega(K^{2/3}N^{1/3})$. This bound is optimal, as shown by Proposition A.1 when $S = \Theta(K^{2/3}N^{1/3})$.

---

1The notation $\tilde{}$ is used to denote the presence of hidden polynomial factors in $\log(N)$ or $1/\log(N)$. 

---

2
Theorem 3.6. The success probability of finding a tradeoff next below this bound. This specific property is of crucial importance for proving our time-space tradeoff next.

More precisely, we prove that the optimal success probability decreases at an exponential rate in $K$ below this bound. This specific property is of crucial importance for proving our time-space tradeoff next.

Theorem 4.1. Every quantum algorithm for finding $K$ disjoint collisions in a random function $f : [M] → [N]$ with success probability $2/3$ must satisfy a time-space tradeoff of $T^3S ≥ Ω(K^3N)$, where $1 ≤ K ≤ N/8$.

Our second main result is the next quantum time-space tradeoff for the same problem of finding $K$ disjoint collisions in a random function. We summarize all the tradeoffs known for this problem in Table 1.

Corollary 4.2. Every quantum algorithm for finding $N/8$ disjoint collisions in a random function $f : [M] → [N]$ with success probability $2/3$ must satisfy a time-space tradeoff of $TS^{1/3} ≥ Ω(N^{4/3})$.

We further show that any improvement to this lower bound would imply a breakthrough for the *Element Distinctness* problem, which consists in finding a single collision in a random function $f : [N] → [N^2]$ (or, more generally, deciding if a function contains a collision). It is a long-standing open question to prove a time-space lower bound for this problem. Although there is some progress in the classical case [BFM97, BSSV03], no result is known in the quantum setting. We give a reduction that converts any tradeoff for finding multiple collisions into a tradeoff for Element Distinctness. We state a particular case of our reduction below.

Corollary 4.6. Suppose that there exists $ε ∈ (0, 1)$ such that every quantum algorithm for finding $Ω(N)$ disjoint collisions in a random function $f : [10N] → [N]$ must satisfy a time-space tradeoff of $TS^{1/3} ≥ Ω(N^{4/3+ε})$. Then, every quantum algorithm for solving Element Distinctness on domain size $N$ must satisfy a time-space tradeoff of $TS^{1/3} ≥ Ω(N^{2/3+2ε})$.

We point out that $TS^{1/3} ≥ Ω(N^{2/3})$ can already be deduced from the query complexity of Element Distinctness [AS04] and $S ≥ 1$. We conjecture that our current tradeoff for finding $K$ collisions can be improved to $T^2S ≥ Ω(K^2N)$, which would imply $T^2S ≥ Ω(N^2)$ for Element Distinctness (Corollary 4.7). This result would be optimal by Ambainis’ algorithm [Amb07].
Finally, we adapt the machinery developed in our paper to study the $K$-Search problem, which consists in finding $K$ preimages of one in a random function $f : [M] \rightarrow \{0, 1\}$ where $f(x) = 1$ with probability $K/N$ independently for each $x \in [M]$. Several variants of this problem have been considered in the literature before [KŠW07, Amb10, Špa08], where it was shown that the success probability must be exponentially small in $K$ when the number of quantum queries is smaller than $O(\sqrt{KN})$. Our proof is the first one to consider this particular input distribution, and it is arguably simpler and more intuitive than previous work. It can also be used to reprove all previous quantum time-space tradeoffs [KŠW07, AŠW09].

**Theorem 5.1.** The success probability of finding $K$ preimages of 1 in a random function $f : [M] \rightarrow \{0, 1\}$ where $f(x) = 1$ with probability $K/N$ for each $x \in [M]$ is at most $O(T^2/(KN))^{K/2} + 2^{-K}$ for any algorithm making $T$ quantum queries to $f$ and any $1 \leq K \leq N/8$.

### 1.2 Our Techniques

**Recording Query Technique.** We use the recording query framework of Zhandry [Zha19] to upper bound the success probability of a query-bounded algorithm in finding $K$ collision pairs. This method intends to reproduce the classical strategy where the queries made by an algorithm (the *attacker*) are recorded and answered with on-the-fly simulation of the oracle. Zhandry brought this technique to the quantum random oracle model by showing that, for the uniform input distribution, one can record *in superposition* the queries made by a quantum algorithm. Our first technical contribution (Section 2.2) is to simplify the analysis of Zhandry’s technique and, as a byproduct, to generalize it to any product distribution on the input. We notice that there has been other independent work on extending Zhandry’s recording technique [HI19, CMSZ19, CMS19]. In particular, [CMSZ19] attempted to generalize it to any arbitrary input distribution. However, the main result of that paper relies on [CMSZ19, Equation 14] which is generally not true for non-product distributions. Our approach is simpler and is based on defining a “recording query operator” that is specific to the input distribution under consideration. This operator can replace the standard quantum query operator without impacting the success probability of the algorithm, but with the effect of “recording” the quantum queries in an additional register. In this paper, we describe explicitly the recording query operators that correspond to the uniform distribution (Lemma 3.2) and to the product of Bernoulli distributions (Lemma 5.3).

**Finding collisions with time-bounded algorithms.** Our application of the recording technique to the Collision Pairs Finding problem has two stages. We first bound the probability that the algorithm has forced the recording of sufficiently many collisions after $T$ queries. Namely, we show that the norm of the quantum state representing the recording of a new disjoint collision at the $t$-th query is of order $\sqrt{t/N}$ (Proposition 3.4). This is related to the probability that a new random value collides with one of the at most $t$ previously recorded queries. The reason why the collisions have to be disjoint is to avoid the recording of more than one new collision in one query. By solving a simple recurrence relation, one gets that the amplitude of the basis states that have recorded at least $K/2$ disjoint collisions after $T$ queries is at most $O(T^{3/2}/(K\sqrt{N}))^{K/2}$. We notice that Liu and Zhandry [LZ19, Theorem 5] had to carry out a similar analysis for the multi-collision finding problem. For constant $K$, they obtained a similar bound of $O(T^{3/2}/\sqrt{N})^{K/2}$ for recording $K$ collision pairs. The second stage of our proof consists in relating the probability of having recorded many collisions to the actual success probability of the algorithm. If we used previous approaches (notably [Zha19, Lemma 5]), this step would degrade the upper bound on the success probability by adding a term that is polynomial in $K/N$. We manage to preserve the exponentially small dependence on $K$ by doing a more careful analysis of the relation between the recording and the standard query models (Proposition 3.5). We adopt a similar approach for analyzing the $K$-Search problem in Section 5.
Finding collisions with time-space bounded algorithms. We convert the above time-only bound into a time-space tradeoff by using the time-segmentation method [BFK+81, KŠW07]. Given a quantum circuit that solves the Collision Pairs Finding problem in time $T$ and space $S$, we slice it into $T/(S^{2/3}N^{1/3})$ consecutive subcircuits, each of them performing $O(S^{2/3}N^{1/3})$ queries. If no slice can output more than $S$ collisions with high probability then there must be at least $K/S$ slices in total, thus implying the desired tradeoff. Our time-only lower bound states that it is impossible to find more than $S$ collisions with probability larger than $2^{-\Omega(S)}$ in time less than $O(S^{2/3}N^{1/3})$. However, we must also take into account that the initial memory at the beginning of each slice carries out information from previous stages. As in previous work [Aar05, KŠW07], we can “eliminate” this memory by replacing it with the completely mixed state while increasing the success probability by a factor of $2^S$. This increase is counterbalanced by our exponentially small success probability lower bound proved before.

Element Distinctness. We connect the Collision Pairs Finding and Element Distinctness problems together by showing how to transform a low-space algorithm for the latter into one for the former (Proposition 4.4). If there exists a time-$T$ space-$S$ algorithm for Element Distinctness on domain size $\sqrt{N}$ then we can find $\tilde{\Omega}(N)$ collisions in a random function $f : [N] \to [N]$ by repeatedly sampling a subset $R \subset [N]$ of size $\sqrt{N}$ and using that algorithm on the function $f$ restricted to the domain $R$. Among other things, we must ensure that a same collision will not occur many times and that storing the sets $R$ does not use too much memory (it turns out that 4-wise independence is sufficient for our purpose). We end up with an algorithm with time $T = O(NT)$ and space $S = O(S)$ for finding $\tilde{\Omega}(N)$ collisions. Consequently, if the Element Distinctness problem on domain size $\sqrt{N}$ can be solved with a time-space tradeoff of $TS^{1/3} \leq O(N^{1/3+\epsilon})$, then there exists an algorithm for finding $\tilde{\Omega}(N)$ collisions that satisfies a time-space tradeoff of $TS^{1/3} \leq O(N^{4/3+\epsilon})$.

2 Models of Computation

We first present the standard model of quantum query complexity in Section 2.1 and its recording variant in Section 2.2. These models are used for investigating the time complexity of the Collision Pairs Finding problem in Section 3 and of the $K$-Search problem in Section 5. Then, we describe the more general circuit model that also captures the space complexity in Section 2.3. It is used in Section 4 for studying time-space tradeoffs.

2.1 Query Model

The (standard) model of quantum query complexity [BW02] measures the number of quantum queries an algorithm (also called an “attacker”) needs to make on an input function $f : [M] \to [N]$ to find an output $z$ satisfying some predetermined relation $R(f, z)$. We present this model in more details below.

**Quantum Query Algorithm.** A $T$-query quantum algorithm is specified by a sequence $U_0, \ldots, U_T$ of unitary transformations acting on the algorithm’s memory. The state $|\psi\rangle$ of the algorithm is made of three registers $Q$, $P$, $W$ where the query register $Q$ holds $x \in [M]$, the phase register $P$ holds $p \in [N]$ and the working register $W$ holds some value $w$. We represent a basis state in the corresponding Hilbert space as $|x, p, w\rangle_{QPW}$. We may drop the subscript $QPW$ when it is clear from the context. The state $|\psi^t_f\rangle$ of the algorithm after $t \leq T$ queries to some input function $f : [M] \to [N]$ is

$$|\psi^t_f\rangle = U_t O_f U_{t-1} \cdots U_1 O_f U_0 |0\rangle$$

where the oracle $O_f$ acts only on register $|x, p\rangle$ and is defined by

$$O_f |x, p\rangle = \omega_N^{pf(x)} |x, p\rangle \quad \text{and} \quad \omega_N = e^{\frac{2\pi}{N}}.$$
The output of the algorithm is written into a substring $z$ of the working register $w$. The success probability $\sigma_f$ of the quantum algorithm on $f$ is the probability that the output value $z$ obtained by measuring the working register of $\|\psi_T^f\|$ satisfies the relation $R(f, z)$. In other words, if we let $\Pi_{\text{success}}^f$ be the projector whose support consists of all basis states $|x, p, w\rangle$ such that the output substring $z$ of $w$ satisfies $R(f, z)$, then

$$\sigma_f = \left\|\Pi_{\text{success}}^f\psi_T^f\right\|^2.$$ 

**Oracle’s Register.** Here, we describe the variant used in the adversary method [Amb02] and in Zhandry’s work [Zha19]. It is represented as an interaction between an algorithm that aims at finding a correct output $z$, and a superposition of oracle’s inputs that respond to the queries from the algorithm.

The memory of the oracle is made of an input register $\mathcal{F}$ holding the description of a function $f : [M] \rightarrow [N]$. This register can be divided into $M$ subregisters $\mathcal{F}_1, \ldots, \mathcal{F}_M$ where $\mathcal{F}_x$ holds $f(x) \in [N]$ for each $x \in [M]$. The basis states in the corresponding Hilbert space are $|f\rangle_{\mathcal{F}} = \otimes_{x \in [M]} |f(x)\rangle_{\mathcal{F}_x}$. Given an input distribution $D$ on the set of functions $[N]^M$, the oracle’s initial state is the state $|\text{init}\rangle_{\mathcal{F}} = \sum_{f \in [N]^M} \sqrt{\Pr[f \leftarrow D]} |f\rangle$.

The query operator $O$ is a unitary transformation acting on the memory of the algorithm and the oracle. Its action is defined on each basis state by

$$O|x, p, w\rangle|f\rangle = (O_f|x, p, w\rangle)|f\rangle.$$ 

The joint state $|\psi_t\rangle$ of the algorithm and the oracle after $t$ queries is $|\psi_t\rangle = U_tO_{U_{t-1}} \cdots U_1 O_U(|0\rangle|\text{init}\rangle) = \sum_{f \in [N]^M} \sqrt{\Pr[f \leftarrow D]} |\psi_T^f\rangle|f\rangle$, where the unitaries $U_i$ have been extended to act as the identity on $\mathcal{F}$. The success probability $\sigma$ of a quantum algorithm on an input distribution $D$ is the probability that the output value $z$ and the input $f$ obtained by measuring the working and input registers of $|\psi_T\rangle$ satisfy the relation $R(f, z)$. In other words, if we let $\Pi_{\text{success}}$ be the projector whose support consists of all basis states $|x, p, w\rangle$ such that the output substring $z$ of $w$ satisfies $R(f, z)$, then

$$\sigma = \left\|\Pi_{\text{success}}|\psi_T\rangle\right\|^2.$$ 

### 2.2 Recording Model

The quantum recording query model is a modification of the standard query model defined in the previous section that is unnoticeable by the algorithm, but that will allow us to track more easily the progress made toward solving the problem under consideration. The original recording model was formulated by Zhandry in [Zha19]. Here, we propose a simplified and more general version of this framework that only requires the initial oracle’s state $|\text{init}\rangle_{\mathcal{F}}$ to be a product state $\otimes_{x \in [M]} |\text{init}_x\rangle_{\mathcal{F}_x}$ (instead of the uniform distribution over all basis states as in [Zha19]).

**Construction.** The range $[N]$ is augmented with a new symbol $\bot$. The input register $\mathcal{F}$ of the oracle can now contain $f : [M] \rightarrow [N] \cup \{\bot\}$, where $f(x) = \bot$ will represent the absence of knowledge from the algorithm about the image of $x$. Unlike in the standard query model, the oracle’s initial state is independent from the relational problem $R$ to be solved and is fixed to be $|\bot^M\rangle_{\mathcal{F}}$ (which represents the fact that the algorithm knows nothing about the input initially). We extend the query operator $O$ defined in the standard query model by setting

$$O|x, p, w\rangle|f\rangle = |x, p, w\rangle|f\rangle \quad \text{when } f(x) = \bot.$$ 

Given a product input distribution $D = D_1 \otimes \cdots \otimes D_M$ on $[N]^M$, the oracle’s initial state in the standard query model can be decomposed as $|\text{init}\rangle_{\mathcal{F}} = \otimes_{x \in [M]} |\text{init}_x\rangle_{\mathcal{F}_x}$ where $|\text{init}_x\rangle_{\mathcal{F}_x} := \sum_y |y\rangle \Pr[y \leftarrow D_x]|y\rangle_{\mathcal{F}_x}$. We define the corresponding “recording query operator” as follows.

**Definition 2.1.** Let $S_1, \ldots, S_M$ acting on $\mathcal{F}_1, \ldots, \mathcal{F}_M$ respectively and satisfying $S_x|\bot\rangle_{\mathcal{F}_x} = |\text{init}_x\rangle_{\mathcal{F}_x}$. Then the recording query operator $R$ with respect to $(S_x, |\text{init}_x\rangle_{\mathcal{F}_x})_{x \in [M]}$ is the operator
and \( R \) obtained by composing \( O \) with the unitary \( S \) defined as follows

\[
S = \sum_{x \in [M]} |x\rangle \langle x| \otimes S_x \quad \text{and} \quad R = S^\dagger \cdot O \cdot S
\]

where \( S \) act as the identity on the unspecified registers.

Later in this paper, we describe two recording query operators that correspond to the uniform distribution (Lemma 3.2) and to the product of Bernoulli distributions (Lemma 5.3).

**Indistinguishability.** The joint state of the algorithm and the oracle after \( t \) queries in the recording query model is \( |\phi_t\rangle = U_tRU_{t-1} \cdots U_1RU_0(|0\rangle |\perp^M) \). Notice that the query operator \( R \) can only change the value of \( f(x') \) when it is applied on a state \( |x,p,w\rangle |f\rangle \) where \( x = x' \). As a consequence, we have the following simple fact.

**Fact 2.2.** The state \( |\phi_t\rangle \) is a linear combination of basis states \( |x,p,w\rangle |f\rangle \) where \( f \) contains at most \( t \) entries different from \( \perp \).

The entries of \( f \) that are different from \( \perp \) represent what the oracle has learnt (or “recorded”) from the algorithm’s queries so far. In the next lemma, we show that \( |\phi_t\rangle \) is related to the state \( |\psi_t\rangle \) (defined in Section 2.1) by \( |\psi_t\rangle = (\otimes_{x \in [M]} S_x) |\phi_t\rangle \). In particular, the states \( |\psi_t\rangle \) and \( |\phi_t\rangle \) cannot be distinguished by the algorithm since \( \otimes_{x \in [M]} S_x \) acts as the identity on the algorithm’s memory.

**Theorem 2.3.** Let \((U_0, \ldots, U_T)\) be a \( T \)-query quantum algorithm. Consider an oracle’s initial state \( |\text{init}\rangle = \otimes_{x \in [M]} |\text{init}_x\rangle \) in the standard query model and choose the \( M \) unitaries \( S_1, \ldots, S_M \) in the recording query model to be any transformations satisfying

\[
S_x |\perp\rangle \otimes S_x = |\text{init}_x\rangle \otimes S_x \quad \text{for all} \quad x \in [M].
\]

Then, the states

\[
\begin{align*}
|\psi_t\rangle &= U_tOU_{t-1} \cdots U_1OU_0(|0\rangle |\text{init}) \\
|\phi_t\rangle &= U_tRU_{t-1} \cdots U_1RU_0(|0\rangle |\perp^M)
\end{align*}
\]

obtained after \( t \) queries in the standard and recording query models respectively are related by

\[
|\psi_t\rangle = \mathcal{T} |\phi_t\rangle \quad \text{where} \quad \mathcal{T} = \bigotimes_{x \in [M]} S_x.
\]

**Proof.** We start by introducing the intermediate operator \( \bar{R} = \mathcal{T}^\dagger \cdot O \cdot \mathcal{T} \). We first claim that \( \bar{R} = R \). Indeed, the query operator \( O \) acts as the identity on all the registers of a basis state \( |x,p,w\rangle |f\rangle \), except \( |x\rangle |p\rangle |f(x)\rangle \). Thus, the actions of \( S_{x'} \) and \( S_x \) for \( x' \neq x \) cancel out in \( \bar{R} \) and \( R \). Since \( \mathcal{T} \) and \( S \) act the same way on registers \( QPF_x \), we obtain that \( \bar{R}(|x,p,w\rangle |f\rangle) = R(|x,p,w\rangle |f\rangle) \).

We further observe that \( U_i \) and \( \mathcal{T} \) commute for all \( i \) since they act as non-identities on disjoint registers. Consequently, we have that

\[
|\psi_t\rangle = U_tOU_{t-1}O \cdots U_1OU_0 \cdot \mathcal{T}(|0\rangle |\perp^M)
\]

since \( \mathcal{T}(|0\rangle |\perp^M) = |0\rangle |\text{init} \)

\[
= \mathcal{T}^\dagger U_tO \cdot \mathcal{T}^\dagger U_{t-1}O \cdots \mathcal{T}^\dagger U_1O \cdot \mathcal{T}^\dagger U_0 \cdot \mathcal{T}(|0\rangle |\perp^M)
\]

since \( \mathcal{T}^\dagger = I \)

\[
= \mathcal{T} U_tU_t^\dagger \cdot O \cdot U_{t-1}U_t^\dagger \cdot O \cdots \mathcal{T} U_1U_1^\dagger \cdot O \cdot U_0(|0\rangle |\perp^M)
\]

by commutation

\[
= \mathcal{T} U_tRU_{t-1} \cdots U_1RU_0(|0\rangle |\perp^M)
\]

by definition of \( \bar{R} \)

\[
= \mathcal{T} |\phi_t\rangle
\]

by definition of \( |\phi_t\rangle \)

\( \square \)
2.3 Space-Bounded Model

Our model of space-bounded computation is the quantum circuit model augmented with the oracle gates of the query model defined in the previous sections. The time complexity, denoted by $T$, is the number of gates in the circuit. In practice, we lower bound it by the number of oracle gates only. The space complexity, denoted by $S$, is the number of qubits on which the circuit is operating. In this paper, we assume that the output of the computation is written on a classical tape where the read/write head can only move to the right. Furthermore, this tape must be updated at predefined gates in the circuit.

3 Time Lower Bound for Collision Pairs Finding

In this section, we upper bound the success probability of finding $K$ disjoint collisions in the query-bounded model of Section 2.1. The proof uses the recording query model of Section 2.2. We first describe in Section 3.1 the recording query framework associated with this problem. In Section 3.2, we study the probability that an algorithm has recorded at least $k$ collisions after $t$ queries for any $k$ and $t$. We prove by induction on $t$ and $k$ that this quantity is exponentially small in $k$ when $t \leq O(k^{2/3}N^{1/3})$ (Proposition 3.4). Finally, in Section 3.3, we relate this progress measure to the actual success probability (Proposition 3.5), and we conclude that the latter is exponentially small in $K$ after $T \leq O(K^{2/3}N^{1/3})$ queries (Theorem 3.6).

3.1 Recording Query Operator

We describe the recording operator that corresponds to the uniform distribution on the set of functions $f : [M] \rightarrow [N]$. In the standard query model, the oracle’s initial state is $|\text{init}\rangle_T = \bigotimes_{x \in [M]} (\frac{1}{\sqrt{N}} \sum_{y \in [N]} |y\rangle_F x)$. Consequently, in the recording query model, we choose the unitary transformations $S_1, \ldots, S_M$ to be defined as follows.

**Definition 3.1.** For any $x \in [M]$, we define the unitary $S_x$ acting on the register $F_x$ to be

$$S_x : \left\{ \begin{array}{ll}
|\perp\rangle_{F_x} & \mapsto \frac{1}{\sqrt{N}} \sum_{y \in [N]} |y\rangle_{F_x} \\
\frac{1}{\sqrt{N}} \sum_{y \in [N]} |y\rangle_{F_x} & \mapsto |\perp\rangle_{F_x} \\
\frac{1}{\sqrt{N}} \sum_{y \in [N]} \omega_{N}^{py} |y\rangle_{F_x} & \mapsto \frac{1}{\sqrt{N}} \sum_{y \in [N]} \omega_{N}^{py} |y\rangle_{F_x} \text{ for } p = 1, \ldots, N - 1
\end{array} \right.$$ 

These unitaries verify $T |\perp M\rangle = |\text{init}\rangle$ where $T = \bigotimes_{x \in [M]} S_x$, as required by Theorem 2.3. The recording query operator is $R = S \cdot O \cdot S$ since $S^\dagger = S$. The next lemma shows how $R$ is acting on the basis states.

**Lemma 3.2.** If the recording query operator $R$ associated to Definition 3.1 is applied on a basis state $|x,p,w\rangle_f$ where $p \neq 0$ then the register $|f(x)\rangle_{F_x}$ is mapped to

$$\left\{ \begin{array}{ll}
\sum_{y \in [N]} \frac{\omega_{N}^{py}}{\sqrt{N}} |y\rangle & \text{if } f(x) = \perp \\
\frac{\omega_{N}^{pf(x)}}{\sqrt{N}} |\perp\rangle + \frac{1 + \omega_{N}^{pf(x)}(N-2)}{N} |f(x)\rangle + \sum_{y \in [N] \setminus \{f(x)\}} \frac{1 - \omega_{N}^{py} - \omega_{N}^{pf(x)}}{N} |y\rangle & \text{otherwise}
\end{array} \right.$$ 

and the other registers are unchanged. If $p = 0$ then none of the registers are changed.

**Proof.** By definition, the unitary $S_x$ maps $|\perp\rangle_{F_x} \mapsto \frac{1}{\sqrt{N}} \sum_{y \in [N]} |y\rangle$ and $|y\rangle_{F_x} \mapsto \frac{1}{\sqrt{N}} |\perp\rangle + \frac{1}{\sqrt{N}} \sum_{y' \in [N] \setminus \{y\}} \omega_{N}^{py'} |y'\rangle$ where $y \in [N]$ and $|\overline{y}\rangle := \frac{1}{\sqrt{N}} \sum_{y \in [N]} \omega_{N}^{py} |y\rangle$. Thus, the action on the register $F_x$ is:

- If $f(x) = \perp$ then $|f(x)\rangle_{F_x} \xrightarrow{S} \frac{1}{\sqrt{N}} \sum_{y \in [N]} |y\rangle \xrightarrow{O} \frac{1}{\sqrt{N}} \sum_{y \in [N]} \omega_{N}^{py} |y\rangle \xrightarrow{S} \frac{1}{\sqrt{N}} \sum_{y \in [N]} \omega_{N}^{py} |y\rangle$. 

• If \( f(x) \in [N] \) then \( |f(x)\rangle_{F_x} = \frac{1}{\sqrt{N}} \sum_{p^0} \omega^{-pf(x)}_N |\bar{p}\rangle = \frac{1}{\sqrt{N}} |\bot\rangle + \frac{1}{\sqrt{N}} \sum_{p^0} \omega^{-pf(x)}_N |0\rangle \)

\( |\hat{p}\rangle \xrightarrow{O} \frac{1}{\sqrt{N}} |\bot\rangle + \frac{1}{\sqrt{N}} \sum_{p^0} \omega^{-pf(x)}_N |p+p\rangle = \frac{1}{\sqrt{N}} |\bot\rangle + \frac{1}{\sqrt{N}} \sum_{p^0} \omega^{-pf(x)}_N |\bar{p}\rangle \)

\[ S \sum_{y \in [N]} |y\rangle + \frac{\omega^{pf(x)}_N}{\sqrt{N}} |\bot\rangle + \frac{\omega^{pf(x)}_N}{\sqrt{N}} \sum_{y \in [N]} \omega^{pf(x)}_N |\phi\rangle = \frac{\omega^{pf(x)}_N}{N} |\bot\rangle + \frac{1+\omega^{pf(x)}_N (N-2)}{N} |\phi\rangle \]

\[ |f(x)\rangle + \sum_{y \in [N]} (f(x)) \frac{1-\omega^{pf(x)}_N}{N} |y\rangle. \]

\[ \blacksquare \]

We observe that the recording operator \( R \) is close to the mapping \( |\bot\rangle_{F_x} \mapsto \sum_{y \in [N]} \frac{\omega^{pf(x)}_N}{\sqrt{N}} |y\rangle \) and \( |f(x)\rangle_{F_x} \mapsto \omega^{pf(x)}_N |f(x)\rangle \) (when \( f(x) \neq \bot \)) up to lower order terms of amplitude \( O(1/N) \). This is analogous to a "lazy" classical oracle that would choose the value of \( f(x) \) uniformly at random the first time it is queried.

### 3.2 Analysis of the Recording Progress

We define a measure of progress based on the number of disjoint collisions contained in the oracle's register of the recording query model. We first give some projectors related to this quantity.

**Definition 3.3.** We define the following projectors by giving the basis states on which they project:

- \( \Pi_{\leq k}, \Pi_{\geq k} \) and \( \Pi_{\leq k} \): all basis states \( |x,p,w\rangle |f\rangle \) such that \( f \) contains respectively at most, exactly or at least \( k \) disjoint collisions (the entries with \( \bot \) are not considered as collisions).

- \( \Pi_{\leq k} \) and \( \Pi_{\geq k} \) for \( y \in [N] \): all basis states \( |x,p,w\rangle |f\rangle \) such that \( f \) contains exactly \( k \) disjoint collisions, \( f(x) = \bot \) or \( f(x) = y \) respectively.

We can now define the measure of progress \( q_{t,k} \) for \( t \) queries and \( k \) collisions as

\[ q_{t,k} = \| \Pi_{\geq k} |\phi_t\rangle \| \]

where \( |\phi_t\rangle \) is the state after \( t \) queries in the recording query model. The main result of this section is the following bound on the growth of \( q_{t,k} \).

**Proposition 3.4.** For all \( t \) and \( k \), we have that \( q_{t,k} \leq \binom{t}{k} \left( \frac{4\sqrt{t}}{N} \right)^k \).

**Proof.** First, \( q_{0,0} = 1 \) and \( q_{0,k} = 0 \) for all \( k \geq 1 \) since the initial state is \( |0\rangle |\bot\rangle^M \). Then, we prove that \( q_{t,k} \) satisfies the following recurrence relation

\[ q_{t+1,k+1} \leq q_{t,k+1} + 4 \sqrt{\frac{k}{N}} q_{t,k} \] (1)

From this result, it is trivial to conclude that \( q_{t,k} \leq \binom{t}{k} \left( \frac{4\sqrt{t}}{N} \right)^k \). In order to prove Equation 1, we first observe that \( q_{t+1,k+1} = \| \Pi_{\geq k+1} U_{t+1} R |\phi_t\rangle \| = \| \Pi_{\geq k+1} R |\phi_t\rangle \| \) since the unitary \( U_{t+1} \) applied by the algorithm at time \( t+1 \) does not act on the oracle’s memory. Then, on any basis state \( |x,p,w\rangle |f\rangle \), the recording query operator \( R \) acts as the identity on the registers \( F_x \) for \( x' \neq x \). Consequently, the basis states \( |x,p,w\rangle |f\rangle \) in \( |\phi_t\rangle \) that may contribute to \( q_{t+1,k+1} \) must have either already \( k+1 \) disjoint collisions in \( f \), or exactly \( k \) disjoint collisions in \( f \) and \( p \neq 0 \).

This implies that

\[ q_{t+1,k+1} \leq q_{t,k+1} + \| \Pi_{\geq k+1} R \Pi_{\leq k,\bot} |\phi_t\rangle \| + \sum_{y \in [N]} \| \Pi_{\geq k+1} R \Pi_{\leq k,y} |\phi_t\rangle \|. \]
We first bound the term \( \|\Pi_{\geq k+1}R\Pi_{=k,\perp}\phi_t\| \). Consider any basis state \(|x, p, w\rangle f\) in the support of \( \Pi_{=k,\perp} \). We further assume that \( f \) contains at most \( t \) entries different from \( \perp \) by Fact 2.2. By Lemma 3.2, we have \( R|x, p, w\rangle f = \sum_{y \in [N]} \frac{\omega_{xy}^N}{\sqrt{N}} |x, p, w\rangle y \Pi_{x \neq x'} |f(x')\rangle_f \). Since there are at most \( t \) entries in \( f \) that can collide with the value contained in the register \( F_x \), we have \( \|\Pi_{\geq k+1}R|x, p, w\rangle f\| \leq \sqrt{t/N} \). Finally, since any two basis states in the support of \( \Pi_{=k,\perp} \) remain orthogonal after \( \Pi_{\geq k+1}R \) is applied, we obtain that \( \|\Pi_{\geq k+1}R\Pi_{=k,\perp}\phi_t\| \leq \sqrt{t/N}\|\Pi_{=k,\perp}\phi_t\| \leq \sqrt{t/N}|\Pi_{=k,\perp}\phi_t\| \leq \sqrt{t/N}\|\Pi_{=k,\perp}\phi_t\| \leq \sqrt{t/N}\|\Pi_{=k,\perp}\phi_t\| \).

We now consider the term \( \|\Pi_{\geq k+1}R\Pi_{=k,y}\phi_t\| \) for any \( y \in [N] \). Again, we consider any basis state \(|x, p, w\rangle f\) in the support of \( \Pi_{=k,y} \) with at most \( t \) entries different from \( \perp \). Using Lemma 3.2, we have \( R|x, p, w\rangle f = \sum_{y' \neq f(x)} \frac{1-\omega_{xy}^N}{N} |x, p, w\rangle y' \Pi_{x \neq x'} |f(x')\rangle_f \). As before, there are at most \( t \) terms in this sum can be in the support of \( \Pi_{\geq k+1} \). Consequently, \( \|\Pi_{\geq k+1}R|x, p, w\rangle f\| \leq 3\sqrt{t/N} \) and \( \|\Pi_{\geq k+1}R\Pi_{=k,y}\phi_t\| \leq 3\sqrt{t/N}\|\Pi_{=k,y}\phi_t\| \).

We conclude that \( q_{t+1,k+1} \leq q_{t,k+1} + \sqrt{t/N}q_{t,k} + \sum_{y \in [N]} 3\sqrt{t/N}\|\Pi_{=k,y}\phi_t\| \leq q_{t,k+1} + \sqrt{t/N}q_{t,k} + 3\sqrt{t/N}q_{t,k} \), where the second step is by Cauchy-Schwarz' inequality.

\[ \|

3.3 From the Recording Progress to the Success Probability

We connect the success probability \( \sigma = \|\Pi_{\text{success}}|\psi_T\rangle\|^2 \) in the standard query model to the final progress \( q_{T,k} \) in the recording query model after \( T \) queries. We show that if the algorithm has made no significant progress for recording \( k \geq K/2 \) collisions then it needs to “guess” the positions of \( K-k \) other collisions. Classically, the probability to find the values of \( K-k \) collisions that have not been queried would be at most \( (1/N^2)^{K-k} \). Here, we show similarly that if a unit state contains at most \( k \) collisions in the quantum recording model, then after mapping it to the standard query model (by applying the operator \( T \) of Theorem 2.3) the probability that the output register contains the correct positions of \( K \) collisions is at most \( N^2(4K^2/N^2)^{K-k} \).

Proposition 3.5. For any state \(|\phi\rangle\), we have \( \|\Pi_{\text{success}}T\Pi_{\leq k}|\phi\rangle\| \leq N(2K^2/N^2)^{K-k}\|\Pi_{\leq k}|\phi\rangle\| \).

Proof. We assume that the output of the algorithm also contains the image of each collision pair under \( f \). Namely, the output \( z \) is represented as a list of \( K \) triples \((x_1, x_2, C_1), \ldots, (x_{2-K}, x_{2K}, C_K)\) \( \in [M]^2 \times [N] \). It is correct if the input function \( f : [M] \rightarrow [N] \) (in the standard query model) satisfies \( f(x_{2\ell-1}) = f(x_{2\ell}) = C_\ell \) for all \( 1 \leq \ell \leq K \), and the values \( x_1, x_2, \ldots, x_{2K} \) are all different. By definition, the support of \( \Pi_{\text{success}} \) consists of all basis states \(|x, p, w\rangle f\) such that the output substring \( z \) of \( w \) satisfies these conditions.

We define a new family of projectors \( \Pi_{a,b} \), where \( 0 \leq a + b \leq 2K \), whose supports consist of all basis states \(|x, p, w\rangle f\) satisfying the following conditions:

(A) The output substring \( z \) is made of \( K \) triples \((x_1, x_2, C_1), \ldots, (x_{2-K}, x_{2K}, C_K)\) where the \( x_i \)’s are all different.

(B) There are exactly \( a \) indices \( i \in [2K] \) such that \( f(x_i) = \perp \).

(C) There are exactly \( b \) indices \( i \in [2K] \) such that \( f(x_i) \neq \perp \) and \( f(x_i) \neq C_{(i/2)} \).

For any state \(|x, p, w\rangle f\) in the support of \( \Pi_{a,b} \), we claim that

\[ \|\Pi_{\text{success}}T|x, p, w\rangle f\| \leq \left( \frac{1}{N} \right)^a \left( \frac{1}{N} \right)^b \]

Indeed, we have \( T = \otimes_{x' \in [M]} S_{x'} \) and by Definition 3.1 the action of \( S_{x_i} \) on the register \(|f(x_i)\rangle_{F_{x_i}} \) is \(|f(x_i)\rangle \mapsto \frac{1}{\sqrt{N}} \sum_{y \in [N]} |y\rangle \) if \( f(x_i) = \perp \), and \(|f(x_i)\rangle \mapsto \frac{1}{\sqrt{N}} |\perp\rangle + (1 - \frac{1}{N})|f(x_i)\rangle \)
Let us now consider any linear combination $|\varphi\rangle = \sum_{x,p,w,f} \alpha_{x,p,w,f} |x,p,w,f\rangle$ of basis states that are in the support of $\Pi_{a,b}$. We claim that

$$\|\Pi_{\text{success}} T|\varphi\rangle\| \leq \left(\frac{2K}{N}\right)^{a+b} \||\varphi\rangle\|$$

(3)

First, for any two basis states $|x,p,w,f\rangle$ and $|\bar{x},\bar{p},\bar{w},\bar{f}\rangle$ where $z = ((x_1, x_2, C_1), \ldots, (x_{2K-1}, x_{2K}, C_K))$ is the output substring of $w$, if the tuples $(x,p,w,(f(x'))_{x'\in\{x_1,\ldots,x_{2K}\}})$ and $(\bar{x},\bar{p},\bar{w},(\bar{f}(x'))_{x'\in\{x_1,\ldots,x_{2K}\}})$ are different then the state $\Pi_{\text{success}} T|x,p,w,f\rangle$ must be orthogonal to $\Pi_{\text{success}} T|\bar{x},\bar{p},\bar{w},\bar{f}\rangle$. Moreover, for any $z = ((x_1, x_2, C_1), \ldots, (x_{2K-1}, x_{2K}, C_K))$ that satisfies condition (A), there are $(2K)^a(2K-a)(N-1)^b \leq (2K)^{a+b} N^b$ different ways to choose $(f(x_i))_{i\in[2K]}$ that satisfy conditions (B) and (C). Let us write $\bar{w} = \{x_1, \ldots, x_{2K}\}$ when then output substring $z$ of $w$ contains $x_1, \ldots, x_{2K}$. Then, by using the Cauchy-Schwarz inequality and Equation 2, we get that

$$\|\Pi_{\text{success}} T|\varphi\rangle\|^2 \leq \sum_{x,p,w,(f(x'))_{x'\in\bar{w}}} \left(\sum_{(f'(x'))_{x'\in\bar{w}}} |\alpha_{x,p,w,f}\|^2 \right) \left(\sum_{(f(x'))_{x'\in\bar{w}}} \|\Pi_{\text{success}} T|x,p,w,f\rangle\|^2\right)$$

$$\leq \||\varphi\rangle\|^2 \cdot (2K)^{a+b} N^b \left(\frac{1}{N}\right)^a \left(\frac{1}{N^2}\right)^b$$

which proves Equation 3.

In order to conclude the proof, we first observe that the support of $\Pi_{\leq k}$ is contained into the union of the supports of $\Pi_{a,b}$ for $a + b \geq 2(K-k)$. Thus, by the triangle inequality, $\|\Pi_{\text{success}} T\Pi_{\leq k}|\varphi\rangle\| \leq \sum_{a+b \geq 2(K-k)} \|\Pi_{\text{success}} T\Pi_{a,b} \Pi_{\leq k}|\varphi\rangle\|$. This quantity is at most $\sum_{a+b \geq 2(K-k)} \left(\frac{2K}{N}\right)^{a+b} \|\Pi_{a,b} \Pi_{\leq k}|\varphi\rangle\|$ by Equation 3. Finally, by using the Cauchy-Schwarz inequality and the fact that the supports of the projectors $\Pi_{a,b}$ are disjoint with each other, we have

$$\|\Pi_{\text{success}} T\Pi_{\leq k}|\varphi\rangle\| \leq \sqrt{\sum_{a+b \geq 2(K-k)} \left(\frac{2K}{N}\right)^{a+b}} \cdot \sqrt{\sum_{a+b \geq 2(K-k)} \|\Pi_{a,b} \Pi_{\leq k}|\varphi\rangle\|^2} \leq N(\frac{2K}{N})^{K-k} \|\Pi_{\leq k}|\varphi\rangle\|$$. 

We can now conclude the proof of the main result of this section.

**Theorem 3.6.** The success probability of finding $K$ disjoint collisions in a random function $f : [M] \rightarrow [N]$ is at most $O(T^3/(K^2 N))^{K/2} + 2^{-K}$ for any algorithm making $T$ quantum queries to $f$ and any $1 \leq K \leq N/8$.

**Proof.** Let $|\psi_T\rangle$ (resp. $|\phi_T\rangle$) denote the state of the algorithm after $T$ queries in the standard (resp. recording) query model. We recall that $|\psi_T\rangle = T|\phi_T\rangle$ (Theorem 2.3). Thus, by the triangle inequality, the success probability $\sigma = \|\Pi_{\text{success}} T|\psi_T\rangle\|^2$ satisfies $\sqrt{\sigma} = \|\Pi_{\text{success}} T\Pi_{\geq K/2}|\phi_T\rangle\| + \|\Pi_{\text{success}} T\Pi_{\leq K/2}|\phi_T\rangle\| \leq \|\Pi_{\geq K/2}|\phi_T\rangle\| + \|\Pi_{\text{success}} T\Pi_{\leq K/2}|\phi_T\rangle\|$. Using Propositions 3.4 and 3.5, we have that $\sqrt{\sigma} \leq (\frac{T}{K/2}) (4\sqrt{T/N})^{K/2} + N(2K/N)^{K/2} \leq O(T^{3/2}/(K\sqrt{N}))^{K/2} + 2^{-K/2}$. Finally, the upper bound on $\sigma$ is derived from the standard inequality $(u + v)^2 \leq 2u^2 + 2v^2$. 

$\square$
4 Time-Space Tradeoffs for Collision Pairs Finding

We study the Collision Pairs Finding problem in the space-bounded model defined in Section 2.3. We prove a time-space tradeoff lower bound in Section 4.1 and we argue the hardness of improving this result in Section 4.2.

4.1 Time-Space Tradeoff

We use the lower bound from Section 3 to derive a time-space tradeoff for the problem of finding $K$ disjoint collisions in a random function $f : [M] \to [N]$.

Theorem 4.1. Every quantum algorithm for finding $K$ disjoint collisions in a random function $f : [M] \to [N]$ with success probability $2/3$ must satisfy a time-space tradeoff of $T^3S \geq \Omega(K^3N)$, where $1 \leq K \leq N/8$.

Proof. Our proof relies on the time-segmentation method for large-output problems [BFK+81, KSW07]. Choose any quantum circuit $C$ running in time $T$ and using $S > \Omega(\log N)$ qubits of memory. We can represent $C$ as a succession $C_1||C_2||\ldots||C_L$ of $L = T/T'$ sub-circuits each running in time $T' = S^{2/3}N^{1/3}$. Define $X_j$ to be the random variable that counts the number of (mutually) disjoint collisions that $C$ outputs between time $(j-1)T'$ and $jT'$ (i.e. in the sub-circuit $C_j$) when the input is a random function $f : [M] \to [N]$. We must have $\sum_{j=1}^L \mathbb{E}[X_j] \geq \Omega(K)$ for the algorithm to be correct.

We claim that $\mathbb{E}[X_j] \leq 3S$ for all $j$. Assume by contradiction that $\mathbb{E}[X_j] \geq 3S$ for some $j$. Since $X_j$ is bounded between 0 and $N$ we have $\Pr[X_j > 2S] \geq S/N$. Consequently, by running $C_j$ on a completely mixed state we obtain $2S$ disjoint collisions with probability at least $S/N \cdot 2^{-S}$ in time $T'$. However, by Theorem 3.6, no quantum algorithm can find more than $2S$ disjoint collisions in time $T' = S^{2/3}N^{1/3}$ with success probability larger than $4^{-S+1}$. This contradiction implies that $\mathbb{E}[X_j] \leq 3S$ for all $j$. Consequently, the number of sub-circuits must be $L \geq \Omega(K/S)$ in order to have $\sum_{j=1}^L \mathbb{E}[X_j] \geq \Omega(K)$. Since each sub-circuit runs in time $S^{2/3}N^{1/3}$ the running time of $C$ is $T \geq \Omega(L \cdot S^{2/3}N^{1/3}) \geq \Omega(KN^{1/3}/S^{1/3})$.

As an illustration of the above result, we obtain the following time-space tradeoff for finding $K = N/8$ disjoint collisions in a random function.

Corollary 4.2. Every quantum algorithm for finding $N/8$ disjoint collisions in a random function $f : [M] \to [N]$ with success probability $2/3$ must satisfy a time-space tradeoff of $TS^{1/3} \geq \Omega(N^{1/3})$.

4.2 Connection to Element Distinctness

We prove that any improvement of the lower bound given in Theorem 4.1 would imply a breakthrough for the Element Distinctness problem.

Definition 4.3. The Element Distinctness problem $\text{ED}_N$ on domain size $N$ consists in finding a collision in a random function $f : [N] \to [N^2]$.

It is well-known that the query complexity of Element Distinctness is $T = \Theta(N^{2/3})$ [AS04, Amb07]. However, it is a long-standing open problem to provide any quantum time-space lower bound for this problem (even classically the question is not completely settled yet [Yao94, BSSV03]). Here, we show that any improvement to Corollary 4.2 would imply a non-trivial time-space tradeoff for Element Distinctness. This result relies on a reduction presented in Algorithm 1 and analyzed in Proposition 4.4 (the constants $c_0$, $c_1$, $c_2$ will be chosen in the proof).
If Conjecture 1 is true, then any quantum algorithm solving the Element Distinctness problem with success probability $\epsilon$ must satisfy a time-space tradeoff of $T^\alpha S^\beta \geq \tilde{O}(N^{2(\gamma-\alpha)})$ for some constants $\alpha, \beta, \gamma$. Then, there exists a bounded-error quantum algorithm for finding $\tilde{O}(N/\log N)$ disjoint collisions in a random function $f : [10N] \rightarrow [N]$ that satisfies a time-space tradeoff of $T^\alpha S^\beta \leq \tilde{O}(N^{\gamma})$. 

Proof. We use the constants $c_0, c_1, c_2$ specified in the proof of Proposition 4.4. First, we note that a random function $f : [10N] \rightarrow [N]$ contains $c_0N$ collisions and no multi-collisions of size larger than $\log N$ with large probability [FO89]. Consequently, any set of $c_1N$ collisions must contain at least $c_1N/\log N$ mutually disjoint collisions with large probability. Assume now that there exists an algorithm solving $\mathsf{ED}_{N^{\frac{1}{2}}} \rightarrow S_{\frac{1}{2}}$ in time $c_2N^{\frac{1}{2}}$ and space $S_{\frac{1}{2}}$ such that $(c_2N^{\frac{1}{2}})^\alpha S_{\frac{1}{2}}^{\beta} \leq \tilde{O}(N^{\gamma-\alpha})$. Then, by plugging it into Algorithm 1, one can find $c_1N/\log N$ disjoint collisions in a random function $f : [10N] \rightarrow [N]$ in time $T = O(NT^{\frac{1}{2}})$ and space $S = O(S^{\frac{1}{2}})$. We derive from the above tradeoff that $T^\alpha S^\beta \leq \tilde{O}(N^{\gamma})$.

As an application of Proposition 4.5, we obtain the following result regarding the hardness of improving Corollary 4.2.

**Corollary 4.6.** Suppose that there exists $\epsilon \in (0, 1)$ such that every quantum algorithm for finding $\tilde{O}(N)$ disjoint collisions in a random function $f : [10N] \rightarrow [N]$ must satisfy a time-space tradeoff of $TS^{1/3} \geq \tilde{O}(N^{4/3+\epsilon})$. Then, every quantum algorithm for solving Element Distinctness on domain size $N$ must satisfy a time-space tradeoff of $TS^{1/3} \geq \tilde{O}(N^{2/3+2\epsilon})$.

We conjecture that the optimal tradeoff for finding $K$ collisions is $T^2S = \Theta(K^2N)$, which would imply an optimal time-space tradeoff of $T^2S \geq \tilde{O}(N^2)$ for Element Distinctness.

**Conjecture 1.** Every quantum algorithm for finding $K$ disjoint collisions in a random function $f : [M] \rightarrow [N]$ with success probability $2/3$ must satisfy a time-space tradeoff of $T^2S \geq \Omega(K^2N)$.

**Corollary 4.7.** If Conjecture 1 is true, then any quantum algorithm solving the Element Distinctness problem with success probability $2/3$ must satisfy a time-space tradeoff of $T^2S \geq \Omega(N^2)$. 

Algorithm 1: Finding collisions by using $\mathsf{ED}_{N^{\frac{1}{2}}}$. 

**Proposition 4.4.** If there exists an algorithm solving $\mathsf{ED}_N$ in time $T_N$ and space $S_N$ then Algorithm 1 runs in time $O(N^\epsilon S_N)$ and space $O(S_N)$, and it finds $c_1N$ collisions in any function $f : [N] \rightarrow [N]$ containing at least $c_0N$ collisions.

The proof of Proposition 4.4 is deferred to Appendix B. We now use the above reduction to transform any low-space algorithm for Element Distinctness into one for finding $\tilde{O}(N/\log N)$ disjoint collisions in a random function. Observe that Algorithm 1 does not necessarily output collisions that are mutually disjoint. Nevertheless, there is a small probability that a random function $f : [M] \rightarrow [N]$ contains multi-collisions of size larger than $\log N$ when $M \approx N$ [FO89]. As a consequence, there is only a log $N$ loss in the analysis.

**Proposition 4.5.** Suppose that there exists a bounded-error quantum algorithm for solving Element Distinctness on domain size $N$ that satisfies a time-space tradeoff of $T^\alpha S^\beta \leq \tilde{O}(N^{2(\gamma-\alpha)})$ for some constants $\alpha, \beta, \gamma$. Then, there exists a bounded-error quantum algorithm for finding $\tilde{O}(N/\log N)$ disjoint collisions in a random function $f : [10N] \rightarrow [N]$ that satisfies a time-space tradeoff of $T^\alpha S^\beta \leq \tilde{O}(N^{\gamma})$.

Proof. We use the constants $c_0, c_1, c_2$ specified in the proof of Proposition 4.4. First, we note that a random function $f : [10N] \rightarrow [N]$ contains $c_0N$ collisions and no multi-collisions of size larger than $\log N$ with large probability [FO89]. Consequently, any set of $c_1N$ collisions must contain at least $c_1N/\log N$ mutually disjoint collisions with large probability. Assume now that there exists an algorithm solving $\mathsf{ED}_{N^{\frac{1}{2}}} \rightarrow S_{\frac{1}{2}}$ in time $c_2N^{\frac{1}{2}}$ and space $S_{\frac{1}{2}}$ such that $(c_2N^{\frac{1}{2}})^\alpha S_{\frac{1}{2}}^{\beta} \leq \tilde{O}(N^{\gamma-\alpha})$. Then, by plugging it into Algorithm 1, one can find $c_1N/\log N$ disjoint collisions in a random function $f : [10N] \rightarrow [N]$ in time $T = O(NT^{\frac{1}{2}})$ and space $S = O(S^{\frac{1}{2}})$. We derive from the above tradeoff that $T^\alpha S^\beta \leq \tilde{O}(N^{\gamma})$. 

As an application of Proposition 4.5, we obtain the following result regarding the hardness of improving Corollary 4.2.

**Corollary 4.6.** Suppose that there exists $\epsilon \in (0, 1)$ such that every quantum algorithm for finding $\tilde{O}(N)$ disjoint collisions in a random function $f : [10N] \rightarrow [N]$ must satisfy a time-space tradeoff of $TS^{1/3} \geq \tilde{O}(N^{4/3+\epsilon})$. Then, every quantum algorithm for solving Element Distinctness on domain size $N$ must satisfy a time-space tradeoff of $TS^{1/3} \geq \tilde{O}(N^{2/3+2\epsilon})$.

We conjecture that the optimal tradeoff for finding $K$ collisions is $T^2S = \Theta(K^2N)$, which would imply an optimal time-space tradeoff of $T^2S \geq \tilde{O}(N^2)$ for Element Distinctness.

**Conjecture 1.** Every quantum algorithm for finding $K$ disjoint collisions in a random function $f : [M] \rightarrow [N]$ with success probability $2/3$ must satisfy a time-space tradeoff of $T^2S \geq \Omega(K^2N)$.

**Corollary 4.7.** If Conjecture 1 is true, then any quantum algorithm solving the Element Distinctness problem with success probability $2/3$ must satisfy a time-space tradeoff of $T^2S \geq \Omega(N^2)$.
5 Lower Bound for \( K \)-Search using the Recording Technique

In this section, we illustrate the use of the recording query model to upper bound the success probability of a query-bounded algorithm on a non-uniform input distribution. We consider the problem of finding \( K \) preimages of 1 in a random function \( f: [M] \rightarrow \{0, 1\} \) where \( f(x) = 1 \) with probability \( K/N \) independently for each \( x \in [M] \).

**Theorem 5.1.** The success probability of finding \( K \) preimages of 1 in a random function \( f: [M] \rightarrow \{0, 1\} \) where \( f(x) = 1 \) with probability \( K/N \) for each \( x \in [M] \) is at most \( O(T^2/(KN))^{K/2} + 2^{-K} \) for any algorithm making \( T \) quantum queries to \( f \) and any \( 1 \leq K \leq N/8 \).

We show that, similarly to the classical setting where a query can reveal one with probability \( K/N \), the amplitude on the basis states that record a new one increases by a factor of \( \sqrt{K/N} \) after each query (Proposition 5.5). Thus, the amplitude of the basis states that have recorded at least \( K/2 \) ones after \( T \) queries is at most \( O(T/\sqrt{KN})^{K/2} \). This implies that any algorithm with \( T < O(\sqrt{KN}) \) queries must likely output at least \( K/2 \) ones at positions that have not been recorded. These outputs can only be correct with probability \( O(K/N)^{K/2} \) (Proposition 5.6).

### 5.1 Recording Query Operator

We describe the recording operator that corresponds to the distribution on the set of functions \( f: [M] \rightarrow [N] \) where \( f(x) = 1 \) with probability \( K/N \) independently for each \( x \in [M] \). In the standard query model, the oracle’s initial state is \( \text{init} = \otimes_{i\in[M]}(\sqrt{1-K/N}|0\rangle_{F_x} + \sqrt{K/N}|1\rangle_{F_x}) \).

Consequently, in the recording query model, we choose the unitary transformations \( S_1, \cdots, S_M \) to be defined as follows.

**Definition 5.2.** For any \( x \in [M] \), we define the unitary \( S_x \) acting on the register \( F_x \) to be

\[
S_x|\perp\rangle_{F_x} = |+\rangle_{F_x}, \quad S_x|+\rangle_{F_x} = |\perp\rangle_{F_x}, \quad S_x|\rangle_{F_x} = |-\rangle_{F_x},
\]

where \( \alpha = \sqrt{1-K/N}, \beta = \sqrt{K/N} \) and \( |+\rangle_{F_x} = \alpha|0\rangle_{F_x} + \beta|1\rangle_{F_x}, |-\rangle_{F_x} = \beta|0\rangle_{F_x} - \alpha|1\rangle_{F_x} \).

These unitaries verify \( T|\perp\rangle^M = \text{init} \) where \( T = \otimes_{i\in[M]}S_i \), as required by Theorem 2.3. The recording query operator is \( R = S \cdot O \cdot S \) since \( S^\dagger = S \). The next lemma shows how \( R \) is acting on the basis states.

**Lemma 5.3.** If the recording query operator \( R \) associated to Definition 5.2 is applied on a basis state \( |x,p,w\rangle f \) where \( p = 1 \) then the register \( |f(x)\rangle_{F_x} \) is mapped to

\[
\left\{ \begin{array}{l}
(1 - 2\beta^2)|\perp\rangle + 2\alpha\beta^2|0\rangle - 2\alpha^2\beta|1\rangle \quad \text{if } f(x) = \perp \\
2\alpha\beta^2|\perp\rangle + (1 - 2\alpha^2\beta^2)|0\rangle + 2\alpha^3\beta|1\rangle \quad \text{if } f(x) = 0 \\
-2\alpha^2\beta|\perp\rangle + 2\alpha^3\beta|0\rangle + (1 - 2\alpha^4)|1\rangle \quad \text{if } f(x) = 1
\end{array} \right.
\]

and the other registers are unchanged. If \( p = 0 \) then none of the registers are changed.

**Proof.** By definition, the unitary \( S_x \) maps \( |\perp\rangle_{F_x} \mapsto |+\rangle, |0\rangle_{F_x} \mapsto \alpha|\perp\rangle + \beta|-\rangle, |1\rangle_{F_x} \mapsto \beta|\perp\rangle - \alpha|\rangle \). Thus, the action on the register \( F_x \) is

- If \( f(x) = \perp \) then \( |f(x)\rangle_{F_x} \overset{S_x}{\mapsto} |+\rangle \overset{O}{\mapsto} \alpha|0\rangle - \beta|1\rangle \overset{S_x}{\mapsto} (\alpha^2 - \beta^2)|\perp\rangle + 2\alpha\beta|-\rangle \).
- If \( f(x) = 0 \) then \( |f(x)\rangle_{F_x} \overset{S_x}{\mapsto} \alpha|\perp\rangle + \beta|-\rangle \overset{O}{\mapsto} \alpha|\perp\rangle + \beta(|0\rangle + |1\rangle) \overset{S_x}{\mapsto} 2\alpha\beta^2|\perp\rangle + (1 - 2\alpha^2\beta^2)|0\rangle + 2\alpha^3\beta|1\rangle \).
- If \( f(x) = 1 \) then \( |f(x)\rangle_{F_x} \overset{S_x}{\mapsto} \beta|\perp\rangle - \alpha|-\rangle \overset{O}{\mapsto} \beta|\perp\rangle - \beta(|0\rangle + |1\rangle) \overset{S_x}{\mapsto} -2\alpha^2\beta|\perp\rangle + 2\alpha^3\beta|0\rangle + (1 - 2\alpha^4)|1\rangle \).

\[\square\]

If \( \alpha \gg \beta \), the above lemma shows that \( R \) is close to the mapping \( |\perp\rangle_{F_x} \mapsto |\perp\rangle - 2\beta|1\rangle, |0\rangle_{F_x} \mapsto |0\rangle + 2\beta|1\rangle, |1\rangle_{F_x} \mapsto -|1\rangle + 2\beta(|0\rangle - |\perp\rangle) \) up to lower order terms of amplitude \( O(\beta^2) \).
5.2 Analysis of the Recording Progress

The measure of progress is based on the number of ones contained in the oracle’s register. We first give some projectors related to this quantity.

**Definition 5.4.** We define the following projectors by giving the basis states on which they project:

- \(\Pi_{\leq k}, \Pi_{=k}\) and \(\Pi_{\geq k}\): all basis states \(|x, p, w\rangle|f\rangle\) such that \(f\) contains respectively at most, exactly or at least \(k\) coordinates equal to \(1\).
- \(\Pi_{=k, \perp}\) and \(\Pi_{=k, 0}\): all basis states \(|x, p, w\rangle|f\rangle\) such that (1) \(f\) contains exactly \(k\) coordinates equal to one, (2) the phase multiplier is \(p = 1\) and (3) \(f(x) = \perp\) or \(f(x) = 0\) respectively.

We can now define the measure of progress \(q_{t,k}\) for \(t\) queries and \(k\) ones as

\[ q_{t,k} = \|\Pi_{\geq k}|\phi_t\| \]

where \(|\phi_t\rangle\) is the state after \(t\) queries in the recording query model. The main result of this section is the following bound on the growth of \(q_{t,k}\).

**Proposition 5.5.** For all \(t\) and \(k\), we have that \(q_{t,k} \leq \binom{4}{k} \left(\frac{4\sqrt{K}}{N}\right)^k\).

**Proof.** First, \(q_{0,0} = 1\) and \(q_{0,k} = 0\) for all \(k \geq 1\) since the initial state is \(|0\rangle_{\perp M}\). Then, we prove that \(q_{t,k}\) satisfies the following recurrence relation

\[ q_{t+1,k+1} \leq q_{t,k+1} + 4\sqrt{\frac{K}{N}} q_{t,k} \]  \hspace{1cm} (4)

From this result, it is trivial to conclude that \(q_{t,k} \leq \binom{4}{k} \left(\frac{4\sqrt{K}}{N}\right)^k\). In order to prove Equation 4, we first observe that \(q_{t+1,k+1} = \|\Pi_{\geq k+1} U_{t+1} R|\phi_t\|\) and \(U_{t+1}\) is the unitary applied by the algorithm at time \(t + 1\). Then, on a basis state \(|x, p, w\rangle|f\rangle\) the recording query operator \(R\) acts as the identity on the registers \(F_x\) for \(x' \neq x\). Consequently, the basis states \(|x, p, w\rangle|f\rangle\) in \(|\phi_t\rangle\) that may contribute to \(q_{t+1,k+1}\) must have either already \(k + 1\) ones in \(f\), or exactly \(k\) ones in \(f\) and \(f(x) \neq 1\), \(p = 1\). This implies that

\[ q_{t+1,k+1} \leq q_{t,k+1} + \|\Pi_{\geq k+1} R \Pi_{=k, \perp} |\phi_t\|\| \end{aligned} + \|\Pi_{\geq k+1} R \Pi_{=k, 0} |\phi_t\|\| \end{aligned}\]

We first bound the term \(\|\Pi_{\geq k+1} R \Pi_{=k, \perp} |\phi_t\|\|\). Consider any basis state \(|x, p, w\rangle|f\rangle\) in the support of \(\Pi_{=k, \perp}\). By Lemma 5.3, we have \(\Pi_{\geq k+1} R |x, p, w\rangle|f\rangle = -2\alpha^2 \beta |x, p, w\rangle|1\rangle_{F_x} \otimes_{x' \neq x} |f(x')\rangle_{F_x'}\). Since any two basis states in the support of \(\Pi_{=k, \perp}\) remain orthogonal after \(\Pi_{\geq k+1} R\) is applied, we obtain that \(\|\Pi_{\geq k+1} R \Pi_{=k, \perp} |\phi_t\|\| = 2\alpha^2 \beta \|\Pi_{=k, \perp} |\phi_t\|\| \leq 2\sqrt{K/N(1 - K/N)} q_{t,k}\).

Similarly, for any state \(|x, p, w\rangle|f\rangle\) in the support of \(\Pi_{=k, 0}\), we have \(\|\Pi_{\geq k+1} R |x, p, w\rangle|f\rangle\| = 2\alpha^3 \beta \|\Pi_{=k, 0} |\phi_t\|\| \leq 2\sqrt{K/N(1 - K/N)} 3/2 q_{t,k}\). We can now conclude the proof,

\[ q_{t+1,k+1} \leq q_{t,k+1} + 2\sqrt{\frac{K}{N}} \left(1 - \frac{K}{N}\right) q_{t,k} + 2\sqrt{\frac{K}{N}} \left(1 - \frac{K}{N}\right) 3/2 q_{t,k} \leq q_{t,k+1} + 4\sqrt{\frac{K}{N}} q_{t,k} \]

\[ \square \]
5.3 From the Recording Progress to the Success Probability

We connect the success probability \( \sigma = ||\Pi_{\text{success}}|\psi_T\rangle||^2 \) in the standard query model to the final progress \( q_{T,k} \) in the recording query model after \( T \) queries. We show that if the algorithm has made no significant progress for \( k \geq K/2 \) then it needs to “guess” that \( f(x) = 1 \) for about \( K-k \) positions where the \( F_x \) register does not contain 1. Classically, the probability to find \( K-k \) preimages of one at positions that have not been queried would be \( (K/N)^{K-k} \). Here, we show similarly that if a unit state contains at most \( K \) ones in the quantum recording model, then after mapping it to the standard query model (by applying the operator \( T \) of Theorem 2.3) the probability that the output register contains the correct positions of \( K \) preimages of one is at most \( 3K (K/N)^{K-k} \).

**Proposition 5.6.** For any state \( |\phi\rangle \), we have \( \|\Pi_{\text{success}} T \Pi_{\leq k} |\phi\rangle\| \leq 3K/2 \left( \frac{K}{N} \right)^{K-k} \|\Pi_{\leq k} |\phi\rangle\| \).

**Proof.** Let \( |x, p, w\rangle f \) be any basis state in the support of \( \Pi_{\leq k} \). The output value \( z \) is a substring of \( w \) made of \( K \) distinct values \( x_1, \ldots, x_K \in [M] \) indicating positions where the input \( f \) is supposed to contain ones. By definition of \( \Pi_{\leq k} \), we have \( f(x_i) \neq 1 \) for at least \( K-k \) indices \( i \in [K] \). For each such index \( i \), after applying \( T = \otimes_{x' \in [M]} S_{x'} \), the amplitude of \( |1\rangle_{F_x} \) is \( \sqrt{K/N} \) (if \( f(x_i) = 1 \)) or \( \sqrt{K/N} (1 - \sqrt{K/N}) \) (if \( f(x_i) = 0 \)) by Definition 5.2. Consequently,

\[
\|\Pi_{\text{success}} T |x, p, w\rangle f\| \leq \left( \frac{K}{N} \right)^{K-k} \tag{5}
\]

Let us now consider any state \( |\phi\rangle \) and denote \( |\varphi\rangle = \Pi_{\leq k} |\phi\rangle = \sum_{x,p,w,f} \alpha_{x,p,w,f} |x, p, w\rangle f \). For any two basis states \( |x, p, w\rangle f \) and \( |\tilde{x}, \tilde{p}, \tilde{w}\rangle f' \) where \( z = (x_1, \ldots, x_K) \) is the output in \( w \), if \( (x, p, w, (f(x'))_{x' \not\in \{x_1, \ldots, x_K\}}) \neq (\tilde{x}, \tilde{p}, \tilde{w}, (f(x'))_{x' \not\in \{x_1, \ldots, x_K\}}) \) then \( \Pi_{\text{success}} T |x, p, w\rangle f \) must be orthogonal to \( \Pi_{\text{success}} T |\tilde{x}, \tilde{p}, \tilde{w}\rangle f' \). There are \( 3^K \) choices for \( |x, p, w\rangle f \) once we set the value of \( (x, p, w, (f(x'))_{x' \not\in \{x_1, \ldots, x_K\}}) \) since it remains to choose \( f(x') \in \{\bot, 0, 1\} \) for \( x' \not\in \{x_1, \ldots, x_K\} \). Let us write \( w_x = \{x_1, \ldots, x_K\} \) when then output substring \( z \) of \( w \) contains \( x_1, \ldots, x_K \). By using the Cauchy-Schwarz inequality and Equation 5, we get that

\[
\|\Pi_{\text{success}} T |\varphi\rangle\|^2 = \sum_{x,p,w,(f(x'))_{x' \not\in w_x}} \left\| \sum_{x',p',w'} \alpha_{x,p,w,f} \Pi_{\text{success}} T |x, p, w\rangle f \right\|^2 \\
\leq \sum_{x,p,w,(f(x'))_{x' \not\in w_x}} \left( \sum_{x',p',w'} |\alpha_{x,p,w,f}|^2 \right) \left( \sum_{(f(x'))_{x' \not\in w_x}} \|\Pi_{\text{success}} T |x, p, w\rangle f\| \right)^2 \\
\leq |||\varphi|||^2 \cdot 3^K \left( \frac{K}{N} \right)^{K-k} .
\]

We can now conclude the proof of the main result.

**Proof of Theorem 5.1.** Let \( |\psi_T\rangle \) (resp. \( |\phi_T\rangle \)) denote the state of the algorithm after \( T \) queries in the standard (resp. recording) query model. According to Theorem 2.3, we have \( |\psi_T\rangle = T|\phi_T\rangle \). Thus, by the triangle inequality, the success probability \( \sigma = ||\Pi_{\text{success}} |\psi_T\rangle||^2 \) satisfies

\[
\sqrt{\sigma} \leq \|\Pi_{\text{success}} T \Pi_{\geq K/2} |\phi_T\rangle\| + \|\Pi_{\text{success}} T \Pi_{\leq K/2} |\phi_T\rangle\| \leq \|\Pi_{\geq K/2} |\phi_T\rangle\| + \|\Pi_{\text{success}} T \Pi_{\leq K/2} |\phi_T\rangle\| .
\]

Using Propositions 5.5 and 5.6, we have that \( \sqrt{\sigma} \leq \left( \frac{T}{K/2} \right) \left( 4\sqrt{K/N} \right)^{K/2} + 3K/2 \left( \sqrt{K/N} \right)^{K/2} \leq O(T/\sqrt{K/N})^{K/2} + 2K/2 \). Finally, the upper bound on \( \sigma \) is derived from the standard inequality \((u + v)^2 \leq 2u^2 + 2v^2\). \( \square \)
Acknowledgements

The authors want to thank the anonymous referees for their valuable comments and suggestions which helped to improve this paper. This research was supported in part by the ERA-NET Cofund in Quantum Technologies project QuantAlgo and the French ANR Blanc project RDAM.

References


[Ber05] D. J. Bernstein. Understanding brute force, 2005. ECRYPT STVL Workshop on Symmetric Key Encryption. 1


A Quantum Algorithm for Collision Pairs Finding

| **Input** | a random function $f : [N] \rightarrow [N]$, two integers $1 \leq K \leq O(N)$ and $S$. |
| **Output** | at least $K$ collisions in $f$. |

1. Repeat $\tilde{O}(K/S)$ times:
   (a) Sample a subset $R \subset [N]$ of size $S$ uniformly at random.
   (b) Construct a table containing all pairs $(x, f(x))$ for $x \in R$. Sort the table according to the second entry of each pair.
   (c) Define the function $g : [N] \setminus R \rightarrow \{0, 1\}$ by $g(x) = 1$ iff there exists $x' \in R$ such that $f(x) = f(x')$. Run the Grover search algorithm [BBHT98] on $g$, using the table computed at step 1.(b), to find all the pairs $(x, x') \in R \times ([N] \setminus R)$ such that $f(x) = f(x')$. Output all of these pairs.

Algorithm 2: Finding $K$ collisions using a memory of size $S$.

**Proposition A.1.** For any $1 \leq K \leq O(N)$ and $\tilde{\Omega}(\log N) \leq S \leq \tilde{O}(K^{2/3}N^{1/3})$, there exists a bounded-error quantum algorithm that can find $K$ collisions in a random function $f : [N] \rightarrow [N]$ by making $T = \tilde{O}(K\sqrt{N/S})$ queries and using $S$ qubits of memory.

**Proof.** We prove that Algorithm 2 satisfies the statement of the proposition. For simplicity, we do not try to tune the hidden factors in the big O notations.

The probability that a fixed pair $(x, x')$ satisfies $(x, x') \in R \times ([N] \setminus R)$ for at least one iteration of step 1 is $\Omega(K/S\cdot S/N\cdot(1-S/N)) = \Omega(K/N)$. Since a random function $f : [N] \rightarrow [N]$ contains $\Omega(N)$ collisions with high probability, the algorithm will encounter $\Omega(K)$ collisions in total. Thus, if the Grover search algorithm never fails we obtain the desired number of collisions.

The expected number of pre-images of 1 under $g$ is $O(S)$. Consequently, the complexity of Grover’s search at step 1.(c) is $O(\sqrt{SN})$. The overall query complexity is $T = \tilde{O}(K/S\cdot \sqrt{SN}) = \tilde{O}(K\sqrt{N/S})$, and the space complexity is $\tilde{O}(S)$. \hfill \Box

B Analysis of Algorithm 1

In this section we prove the following statement about Algorithm 1.

**Proposition 4.4.** If there exists an algorithm solving ED$_N$ in time $T_N$ and space $S_N$ then Algorithm 1 runs in time $O(NT_N\sqrt{N})$ and space $O(S_N\sqrt{N})$, and it finds $c_1N$ collisions in any function $f : [N] \rightarrow [N]$ containing at least $c_0N$ collisions.

**Proof.** We choose $c_0 = 40$, $c_1 = 1/10^4$ and $c_2 = 8$. We study the probabilities of the following events to occur at a fixed round of Algorithm 1:

- **Event A:** The hash function $h$ is collision free (i.e. $h(i) \neq h(j)$ for all $i \neq j$).
- **Event B:** None of the collisions output during the previous rounds is present in the image of $h$.
- **Event C:** The function $f \circ h : [\sqrt{N}] \rightarrow [N]$ contains a collision.
- **Event D:** The algorithm for ED$_{\sqrt{N}}$ finds a collision at step 2.(b).

Algorithm 1 succeeds if and only if the event $A \land B \land C \land D$ occurs during at least $c_1N$ rounds. We now lower bound the probability of this event to happen.
For event A, let us consider the random variable \( X = \sum_{i \neq j \in [\sqrt{N}]} 1_{h(i) = h(j)} \). Using that \( h \) is pairwise independent, we have \( \mathbb{E}[X] = \left( \frac{\sqrt{N}}{2} \right)^2 \frac{1}{N} \leq \frac{1}{2} \). Thus, by Markov’s inequality, Pr[A] = 1 − Pr[X ≥ 1] ≥ \( \frac{1}{2} \).

For event B, let us assume that \( k < c_1 N \) collisions \((x_1, x_2), \ldots, (x_{2k-1}, x_{2k})\) have been output so far. For any \( i \in [k] \), the probability that \( x_{2i-1} \) and \( x_{2i} \) occur in \( \{h(1), \ldots, h(\sqrt{N})\} \) is at most \( \left( \frac{\sqrt{N}}{2} \right)^2 \frac{1}{N} \) since \( h \) is pairwise independent. By a union bound, \( \Pr[B] \geq 1 - \frac{k}{N} \geq 1 - c_1 \).

For event C, let us consider the binary random variables \( Y_{i,j} = 1_{f \circ h(i) = f \circ h(j)} \) for \( i \neq j \in [\sqrt{N}] \), and let \( Y = \sum_{i \neq j} Y_{i,j} \) be twice the number of collisions in \( f \circ h \). Note that we may have \( Y_{i,j} = 1 \) because \( h(i) = h(j) \) (this is taken care of in event A). For each \( y \in [N] \), let \( N_y = |\{x : f(x) = y\}| \) denote the number of elements that are mapped to \( y \) by \( f \). Using that \( h \) is 4-wise independent, for any \( i \neq j \neq k \neq \ell \) we have,

\[
\begin{align*}
\Pr[Y_{i,j} = 1] &= \frac{\sum_{y \in [N]} N_y^2}{N^2} \\
\Pr[Y_{i,j} = 1 \land Y_{k,\ell} = 1] &= \Pr[Y_{i,j} = 1] \cdot \Pr[Y_{k,\ell} = 1].
\end{align*}
\]

Consequently, \( \mathbb{E}[Y] = \left( \frac{\sqrt{N}}{2} \right)^2 \sum_{y \in [N]} N_y^2 / N^2 \) and

\[
\mathbb{E}[Y] = \sum_{\{i,j\}} \mathbb{E}[Y_{i,j}] + \sum_{\{i,j\} \neq \{i,k\}} \mathbb{Cov}[Y_{i,j}, Y_{i,k}] + \sum_{\{i,j\} \cap \{k,\ell\} = \emptyset} \mathbb{Cov}[Y_{i,j}, Y_{k,\ell}]
\]

\[
\leq \sum_{\{i,j\}} \mathbb{E}[Y_{i,j}^2] + \sum_{\{i,j\} \neq \{i,k\}} \mathbb{E}[Y_{i,j} Y_{i,k}]
\]

\[
= \left( \frac{\sqrt{N}}{2} \right)^2 \sum_{y \in [N]} N_y^2 / N^2 + 3 \left( \frac{\sqrt{N}}{3} \right) \sum_{y \in [N]} N_y^3 / N^3
\]

where we have used that \( Y_{i,j} \) and \( Y_{k,\ell} \) are independent when \( i \neq j \neq k \neq \ell \). The term \( \sum_{y \in [N]} N_y^2 \) is equal to the number of pairs \((x, x') \in [N]^2\) such that \( f(x) = f(x') \). Each collision in \( f \) gives two such pairs, and we must also count the pairs \((x, x)\). Thus, \( \sum_{y \in [N]} N_y^2 \geq (1 + 2c_0)N \). Moreover, \( \sum_{y \in [N]} N_y^3 \leq (\sum_{y \in [N]} N_y^2)^{3/2} \). Consequently,

\[
\mathbb{E}[Y] / \mathbb{E}[Y]^2 \leq 1 + \sqrt{\frac{\left( \sum_{y \in [N]} N_y^2 \right)^{1/2}}{\left( \frac{\sqrt{N}}{2} \right)^2 \sum_{y \in [N]} N_y^2 / N^2}} \leq 4 \left( 1 + \frac{1 + \sqrt{1 + 2c_0}}{1 + 2c_0} \right).
\]

Finally, according to Chebyshev’s inequality, \( \Pr[Y = 0] \leq \Pr[|Y - \mathbb{E}[Y]| \geq \mathbb{E}[Y]] \leq \frac{\mathbb{E}[Y]}{\mathbb{E}[Y]^2} \). Thus, \( \Pr[C] = 1 - \Pr[Y = 0] \geq 1 - \frac{4(1 + \sqrt{1 + 2c_0})}{1 + 2c_0} \).

For event D, we have \( \Pr[D | A \land B \land C] \geq 2/3 \) assuming the bounded-error algorithm for solving \( \text{ED}_{\sqrt{N}} \) succeeds with probability 2/3.

We can now lower bound the probability of the four events together.

\[
\Pr[A \land B \land C \land D] = \Pr[D | A \land B \land C] \cdot \Pr[A \land B \land C]
\]

\[
\geq \Pr[D | A \land B \land C] \cdot (\Pr[A] + \Pr[B] + \Pr[C] - 2)
\]

\[
\geq \frac{2}{3} \left( \frac{1}{2} - c_1 - \frac{4(1 + \sqrt{1 + 2c_0})}{1 + 2c_0} \right) \geq 1/250
\]

Let \( \tau \) be the number of rounds after which \( c_1 N \) collisions have been found (i.e. \( A \land B \land C \land D \) has occurred \( c_1 N \) times). We have \( \mathbb{E}[\tau] \leq 8c_1 N \), and by Markov’s inequality \( \Pr[\tau \geq c_2 N] \leq 250c_1 / c_2 \leq 1/3 \). Thus, with probability at least 2/3, Algorithm 1 outputs at least \( c_1 N \) collisions in \( f \).