Synthesis of Reversible Logic Circuits

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Abstract

Reversible or information-lossless circuits have applications in digital signal processing, communication, computer graphics and cryptography. They are also a fundamental requirement in the emerging field of quantum computation. We investigate the synthesis of reversible circuits that employ a minimum number of gates and contain no redundant input-output line-pairs (temporary storage channels). We prove constructively that every even permutation can be implemented without temporary storage using NOT, CNOT and TOFFOLI gates. We describe an algorithm for the synthesis of optimal circuits and study the reversible functions on three wires, reporting the distribution of circuit sizes. Finally, in an application important to quantum computing, we synthesize oracle circuits for Grover’s search algorithm, and show a significant improvement over a previously proposed synthesis algorithm.

This work was partially supported by the Undergraduate Summer Research Program at the University of Michigan and by the DARPA QuIST program. The views and conclusions contained herein are those of the authors and should not be interpreted as necessarily representing official policies or endorsements, either expressed or implied, of the Defense Advanced Research Projects Agency (DARPA) or the U.S. Government.
1 Introduction

In most computing tasks, the number of output bits is relatively small compared to the number of input bits. For example, in a decision problem, the output is only one bit (yes or no) and the input can be as large as desired. However, computational tasks in digital signal processing, communication, computer graphics, and cryptography require that all of the information encoded in the input be preserved in the output. Some of those tasks are important enough to justify adding new microprocessor instructions to the HP PA-RISC (MAX and MAX-2), Sun SPARC (VIS), PowerPC (AltiVec), IA-32 and IA-64 (MMX) instruction sets. In particular, new bit-permutation instructions were shown to vastly improve performance of several standard algorithms, including matrix transposition and DES, as well as two recent cryptographic algorithms Twofish and Serpent. Bit permutations are a special case of reversible functions, that is, functions that permute the set of possible input values. For example, the butterfly operation \((x, y) \mapsto (x + y, x y)\) is reversible but is not a bit permutation. It is a key element of Fast Fourier Transform algorithms and has been used in application-specific Xtensa processors from Tensilica. One might expect to get further speed-ups by adding instructions to allow computation of an arbitrary reversible function. The problem of chaining such instructions together provides one motivation for studying reversible computation and reversible logic circuits, that is, logic circuits composed of gates computing reversible functions.

Reversible circuits are also interesting because the loss of information associated with irreversibility implies energy loss. Younis and Knight showed that some reversible circuits can be made asymptotically energy-lossless as their delay is allowed to grow arbitrarily large. Currently, energy losses due to irreversibility are dwarfed by the overall power dissipation, but this may change if power dissipation improves. In particular, reversibility is important for nanotechnologies where switching devices with gain are difficult to build.

Finally, reversible circuits can be viewed as a special case of quantum circuits because quantum evolution must be reversible. Classical (non-quantum) reversible gates are subject to the same “circuit rules,” whether they operate on classical bits or quantum states. In fact, popular universal gate libraries for quantum computation often contain as subsets universal gate libraries for classical reversible computation. While the speed-ups which make quantum computing attractive are not available without purely quantum gates, logic synthesis for classical reversible circuits is a first step toward synthesis of quantum circuits. Moreover, algorithms for quantum communications and cryptography often do not have classical counterparts because they act on quantum states, even if their action in a given computational basis corresponds to classical reversible functions on bit-strings. Another connection between classical and quantum computing comes from Grover’s quantum search algorithm. Circuits for Grover’s algorithm contain large
parts consisting of NOT, CNOT and TOFFOLI gates only \[14\].

We review existing work on classical reversible circuits. Toffoli \[20\] gives constructions for an arbitrary reversible or irreversible function in terms of a certain gate library. However, his method makes use of a large number of temporary storage channels, i.e. input-output wire-pairs other than those on which the function is computed (also known as ancilla bits). Sasao and Kinoshita show that any conservative function \((f(x)\) is conservative if \(x\) and \(f(x)\) always contain the same number of 1s in their binary expansions) has an implementation with only three temporary storage channels using a certain fixed library of conservative gates, although no explicit construction is given \[16\]. Kerntopf uses exhaustive search methods to examine small-scale synthesis problems and related theoretical questions about reversible circuit synthesis \[9\]. There has also been much recent work on synthesizing reversible circuits that implement non-reversible Boolean functions on some of their outputs, with the goal of providing the quantum phase shift operators needed by Grover’s quantum search algorithm \[8, 12, 21\]. Some work on local optimization of such circuits via equivalences has also been done \[12, 8\]. In a different direction, group theory has recently been employed as a tool to analyze reversible logic gates \[19\] and investigate generators of the group of reversible gates \[5\].

Our work pursues synthesis of optimal reversible circuits which can be implemented without temporary storage channels. In Section 3 we show by explicit construction that any reversible function which performs an even permutation on the input values can be synthesized using the CNTS (CNOT, NOT, TOFFOLI, and SWAP) gate library and no temporary storage. An arbitrary (possibly odd) permutation requires at most one channel of temporary storage for implementation. By examining circuit equivalences among generalized CNOT gates, we derive a canonical form for CNT-circuits. In Section 4 we present synthesis algorithms for implementing any reversible function by an optimal circuit with gates from an arbitrary gate library. Besides branch-and-bound, we use a dynamic programming technique that exploits reversibility. While we use gate count as our cost function throughout, this method allows for many different cost functions to be used. Applications to quantum computing are examined in Section 5.

2 Background

In conventional (irreversible) circuit synthesis, one typically starts with a universal gate library and some specification of a Boolean function. The goal is to find a logic circuit that implements the Boolean function and minimizes a given cost metric, e.g., the number of gates or the circuit depth. At a high level, reversible circuit synthesis is just a special case in which no fanout is allowed and all gates must be reversible.
2.1 Reversible Gates and Circuits

**Definition 1** A gate is reversible if the (Boolean) function it computes is bijective.

If arbitrary signals are allowed on the inputs, a necessary condition for reversibility is that the gate have the same number of input and output wires. If it has $k$ input and output wires, it is called a $k$ gate, or a gate on $k$ wires. We will think of the $m$th input wire and the $m$th output wire as really being the same wire. Many gates satisfying these conditions have been examined in the literature [15]. We will consider a specific set defined by Toffoli [20].

**Definition 2** A $k$-CNOT is a $(k + 1) \times (k + 1)$ gate. It leaves the first $k$ inputs unchanged, and inverts the last iff all others are 1. The unchanged lines are referred to as control lines.

Clearly the $k$-CNOT gates are all reversible. The first three of these have special names. The 0-CNOT is just an inverter or NOT gate, and is denoted by N. It performs the operation $(x) \rightarrow \overline{x}$, where $\overline{x}$ denotes XOR. The 1-CNOT, which performs the operation $(y; x) \rightarrow (y; \overline{x})$, is referred to as a Controlled-NOT [7], or CNOT (C). The 2-CNOT is normally called a TOFFOLI (T) gate, and performs the operation $(z; y; x) \rightarrow (z; y; \overline{x})$. We will also be using another reversible gate, called the SWAP (S) gate. It is a 2 gate which exchanges the inputs; that is, $(x; y) \rightarrow (y; x)$. One reason for choosing these particular gates is that they appear often in the quantum computing context, where no physical “wires” exist, and swapping two values requires non-trivial effort. [14]. We will be working with circuits from a given, limited-gate library. Usually, this will be the CNTS gate library, consisting of the CNOT, NOT, and TOFFOLI, and SWAP gates.

**Definition 3** A well-formed reversible logic circuit is an acyclic combinational logic circuit in which all gates are reversible, and are interconnected without fanout.

As with reversible gates, a reversible circuit has the same number of input and output wires; again we will call a reversible circuit with $n$ inputs an $n \times n$ circuit, or a circuit on $n$ wires. We draw reversible circuits as arrays of horizontal lines representing wires. Gates are represented by vertically-oriented symbols. For example, in Figure 1 we see a reversible circuit drawn in the notation introduced by Feynman [7]. The symbols represent inverters and the symbols

```
|         |         |
|--------|--------|
| x      | x'     |
| y      | y'     |
| z      | z'     |
```

Figure 1: 3 reversible circuit with two T gates and two N gates.
Figure 2: Truth table for the circuit in Figure 1.

represent controls. A vertical line connecting a control to an inverter means that the inverter is only applied if the wire on which the control is set carries a 1 signal. Thus, the gates used are, from left to right, TOFFOLI, NOT, TOFFOLI, and NOT.

Since we will be dealing only with bijective functions, i.e., permutations, we represent them using the cycle notation where a permutation is represented by disjoint cycles of variables. For example, the truth table in Figure 2 is represented by \( (2;3) (6;7) \) because the corresponding function swaps 010 (2) and 011 (3), and 110 (6) and 111 (7). The set of all permutations of \( n \) indices is denoted \( S_n \), so the set of bijective functions with \( n \) binary inputs is \( S_{2^n} \). We will call \( (2;3) (6;7) \) CNT-constructible since it can be computed by a circuit with gates from the CNT gate library. More generally:

**Definition 4** Let \( L \) be a (reversible) gate library. An \( L \)-circuit is a circuit composed only of gates from \( L \). A permutation \( \pi \) in \( S_{2^n} \) is \( L \)-constructible if it can be computed by an \( n \) \( L \)-circuit.

Figure 3a indicates that the circuit in Figure 1 is equivalent to one consisting of a single C gate. Pairs of circuits computing the same function are very useful, since we can substitute one

\[
\begin{array}{c|c}
\text{x} & \text{y} & \text{z} \\
0 & 0 & 0 \\
0 & 0 & 1 \\
0 & 1 & 0 \\
0 & 1 & 1 \\
1 & 0 & 0 \\
1 & 0 & 1 \\
1 & 1 & 0 \\
1 & 1 & 1 \\
\end{array}
\]

**Figure 2:** Truth table for the circuit in Figure 1.
for the other. On the right, we see similarly that three C gates can be used to replace the S gate appearing in the middle circuit of Figure 3. If allowed by the physical implementation, the S gate may itself be replaced with a wire swap. This, however, is not possible in some forms of quantum computation [14]. Figure 3 therefore shows us that the C and S gates in the CNTS gate library can be removed without losing computational power. We will still use the CNTS gate library in synthesis to reduce gate counts and potentially speed up synthesis. This is motivated by Figure 3, which shows how to replace four gates with one C gate, and thus up to 12 gates with one S gate.

Figure 4 illustrates the meaning of “temporary storage” [20]. The top \( n \times k \) wires transfer \( n \times k \) signals, collectively designated \( Y \), to the corresponding wires on the other side of the circuit. The signals \( Y \) are arbitrary, in the sense that the circuit \( K \) must assume nothing about them to make its computation. Therefore, the output on the bottom \( k \) wires must be only a function of their input values \( X \) and not of the “ancilla” bits \( Y \), hence the bottom output is denoted \( f(X) \). While the signals \( Y \) must leave the circuit holding the same values they entered it with, their values may be changed during the computation as long as they are restored by the end. These wires usually serve as an essential workspace for computing \( f(X) \). An example of this can be found in Figure 3: the C gate on the right needs two wires, but if we simulate it with two N gates and two T gates, we need a third wire. The signal applied to the top wire emerges unaltered.

**Definition 5** Let \( L \) be a reversible gate library. Then \( L \) is universal if for all \( k \) and all permutations \( \pi \in S_{2^k} \), there exists some \( l \) such that some \( L \)-constructible circuit computes \( \pi \) using \( l \) wires of temporary storage.

The concept of universality differs in the reversible and irreversible cases in two important ways. First, we do not allow ourselves access to constant signals during the computation, and second, we synthesize whole permutations rather than just functions with one output bit.

### 2.2 Prior Work

It is a result of Toffoli’s that the CNT gate library is universal; he also showed that one can bound the amount of temporary storage required to compute a permutation in \( S_{2^n} \) by \( n \times 3 \). Indeed,
much of the reversible and quantum circuit literature allows the presence of polynomially many temporary storage bits for circuit synthesis. Given that qubits are a severely limited resource in current implementation technologies, this may not be a realistic assumption. We are therefore interested in trying to synthesize permutations using no extra storage. To illustrate the limitations this puts on the set of computable permutations, suppose we restrict ourselves to the C gate library. The following results are well-known in the quantum circuits literature [15, 3]. We provide proofs both for completeness, and to accustom the reader to techniques we will require later.

**Definition 6** A function \( f : \{0,1\}^n \rightarrow \{0,1\}^m \) is linear iff \( f(x \oplus y) = f(x) \oplus f(y) \), where \( \oplus \) denotes bitwise XOR.

This is just the usual definition of linearity where we think of \( \{0,1\}^n \) as a vector space over the two-element field \( \mathbb{F}_2 \). In our work \( n = m \) because of reversibility. Thus, \( f \) can be thought of as a square matrix over \( \mathbb{F}_2 \). The composition of two linear functions is a linear function.

**Lemma 7** [3] Every C-constructible permutation computes an invertible linear transformation. Moreover, every invertible linear transformation is computable by a C-constructible circuit. No C-circuit requires more than \( n^2 \) gates.

**Proof:** To show that all C-circuits are linear, it suffices to prove that each C gate computes a linear transformation. Indeed, \( C(x_1 \ y_1; x_2 \ y_2) = (x_1 \ y_1; x_1 \ y_2; x_2 \ y_1) \), \( (x_2 \ y_2) = C(x_1 \ y_1) \). In the basis 10 : : : 0, 01 : : : 0, : : : 0, a C gate with the control on the \( i \)-th wire and the inverter on the \( j \)-th applied to an arbitrary vector will add the \( i \)-th entry to the \( j \)-th. Thus, the matrices corresponding to individual C gates account for all the elementary row-addition matrices. Any invertible matrix in \( GL(\mathbb{F}_2) \) can be written as a product of these. Thus, any invertible linear transformation can be computed by a C-circuit. Finally, any matrix over \( \mathbb{F}_2 \) may be row-reduced to the identity using fewer than \( n^2 \) row operations. \( \square \)

One might ask how inefficient the row reduction algorithm is in synthesizing C-circuits. A counting argument can be used to find asymptotic lower bounds on the longest circuits [17].

**Lemma 8** Let \( L \) be a gate library; let \( K_n \) be the set of L-constructible permutations on \( n \) wires, and let \( k_i \) be the cardinality of \( K_i \). Then the longest gate-minimal L-circuit on \( n \) wires has more than \( \log k_n = \log b \) gates, where \( b \) is the number of one-gate circuits on \( n \) wires. \( b = \text{poly}(n) \), so for large \( n \), worst-case circuits have length \( \Omega(\log k_n = \log n) \).

**Proof:** Suppose the longest gate-minimal L-circuit has \( x = 1 \) gates. Then every permutation in \( K_n \) is computed by an L-circuit of at most \( x = 1 \) gates. The number of such circuits is \( \sum_{i=1}^{x} b^i = < b^x \). Therefore, \( k_n < b^x \), and it follows that \( x > \log k_n = \log b \).
Finally, let $G$ be a gate in $L$ with the largest number of inputs, say $p$. Then, on $n$ wires, there are at most $n(p + 1) < n^p$ ways to make a 1-gate circuit using $G$. If $L$ has $q$ gates in total, then $b = qn^p = \text{poly}(n)$. Hence, $x > \log k_n = (p \log n + \log q) = \Omega(\log n)$. 

We now need to count the number of $C$-constructible permutations. On two wires, there are six, corresponding to the six circuits in Figure 5.

**Corollary 9** \( [17] \) $S_{2^n}$ has $\prod_{i=0}^{2^n} 2^n$ $C$-constructible permutations. Therefore, worst-case $C$-circuits require $\Omega(n^2 = \log n)$ gates.

**Proof:** A linear mapping is fully defined by its values on basis vectors. There are $2^n$ 1 ways of mapping the $2^n$-bit string $10: \ldots : 0$. Once we have fixed its image, there are $2^n$ 2 ways of mapping $010: \ldots : 0$, and so on. Each basis bit-string cannot map to the subspace spanned by the previous bit-strings. There are $2^n 2^i$ choices for the $i$-th basis bit-string. Once all basis bit-strings are mapped, the mapping of the rest is specified by linearity. The number of $C$-constructible permutations on $n$ wires is greater than $2^{n^2} = 2$. By Lemma 8 worst-case $C$-circuits require $\Omega(n^2 = \log n)$ gates.

Let us return to CNT-constructible permutations. A result similar to Lemma 7 requires:

**Definition 10** A permutation is called even if it can be written as the product of an even number of transpositions. The set of even permutations in $S_n$ is denoted $A_n$.

It is well-known that if a permutation can be written as the product of an even number of transpositions, then it may not be written as the product of an odd number of transpositions. Moreover, half the permutations in $S_n$ are even for $n > 1$.

**Lemma 11** \( [20] \) Any $n$- circuit with no $n$ gates computes an even permutation.

**Proof:** It suffices to prove this for a circuit consisting of only one gate, as the product of even permutations is even. Let $G$ be a gate in an $n$- circuit. By hypothesis, $G$ is not $n$- $n$, so there must be at least one wire which is unaffected by $G$. Without loss of generality, let this be the high-order wire. Then $2^{n-1} G(k) = G(2^{n-1} k)$, and $k < 2^{n-1}$ implies $G(k) < 2^{n-1}$. Thus every cycle in the cycle decomposition of $G$ appears in duplicate: once with numbers less than $2^{n-1}$, and once with the corresponding numbers with their high order bits set to one. But these cycles have the same length, and so their product is an even permutation. Therefore, $G$ is the product of even permutations, and hence is even.
To illustrate this result, consider the following example. A 2×2 circuit consisting of a single S gate performs the permutation (1,2), as the inputs 01 and 10 are interchanged, and the inputs 00 and 11 remain fixed. This permutation consists of one transposition, and is therefore odd. On the other hand, in a 3×3 circuit, one can check that a swap gate on the bottom two wires performs the permutation (1,2) (5,6), which is even.

3 Theoretical Results

Since the CNTS gate library contains no gates of size greater than three, Lemma 11 implies that every CNTS-constructible (without temporary storage) permutation is even for $n = 4$. The main result of this section is that the converse is also true.

Theorem 12 Every even permutation is CNT-constructible.

Before beginning the proof, we offer the following two corollaries. These give a way to synthesize circuits computing odd permutations using temporary storage, and also extend Theorem 12 to an arbitrary universal gate library.

Corollary 13 Every permutation, even or odd, may be computed in a CNT-circuit with at most one wire of temporary storage.

Proof: Suppose we have an $n$-gate G computing $\pi \in S_2^n$, and we place it on the bottom $n$ wires of an $(n+1)$-wire reversible circuit; let $\tilde{\pi}$ be the permutation computed by this new circuit. Then by Lemma 11, $\tilde{\pi}$ is even. By Theorem 12, $\tilde{\pi}$ is CNT-constructible. Let C be a CNT-circuit computing $\tilde{\pi}$. C computes $\pi$ with one line of temporary storage.

Corollary 14 For any universal gate library L and sufficiently large n, permutations in $A_2^n$ are L-constructible, and those in $S_2^n$ are realizable with at most one wire of temporary storage.

Proof: Since L is universal, there is some number $k$ such that we can compute the permutations corresponding to the NOT, CNOT, and TOFFOLI gates using a total of $k$ wires. Let $n > k$, and let $\pi \in A_2^n$. By Theorem 12 we can find a CNT-circuit C computing $\pi$, and can replace every N, C, or T gate with a circuit computing it. The second claim follows similarly from Theorem 12 and Corollary 13.

To prove Theorem 12 we begin by asking which permutations are C-constructible, N-constructible, and T-constructible. The first of these questions was answered in Section 2. We now summarize the properties of N-constructible permutations. In what follows, $\oplus$ denotes bitwise XOR.

Definition 15 Given an integer i, we denote by $N^i$ the circuit formed by placing an N gate on every wire corresponding to a 1 in the binary expansion of i.
We will use $N^i$ to signify both the circuit described above, and the permutation which this circuit computes. Technically, the latter is not uniquely determined by the $N^i$ notation, but also depends on the number $n$ of wires in the circuit; however, $n$ will always be clear from context. The $N^i$ notation is illustrated for the case of three wires in Figure 6.

**Lemma 16** Let $\pi \in S_{2^n}$ be $N$-constructible. There exists an $i$ such that $\pi(x) = x_i$. Moreover, the gate-minimal circuit for $\pi$ is $N^i$. There are $2^n$ $N$-constructible permutations in $S_{2^n}$.

**Proof:** Clearly, $N^i$ computes the permutation $\pi(x) = x_i$. It now suffices to show that an arbitrary $N$-circuit may be reduced to one of the $N^i$ circuits. Any pair of consecutive $N$ gates on the same wire may be removed without changing the permutation computed by the circuit. Applying this transformation until no more gates can be removed must leave a circuit with at most one $N$ gate per wire; that is, a circuit of the form $N^i$. $\square$

### 3.1 T-Constructible Permutations

Characterizing the T-constructible permutations is more difficult. We will begin by extending the $N^i$ notation defined above.

**Definition 17** Let $N^h$ be an $N$-circuit as defined above. Let $k$ be an integer such that the bitwise Boolean product $hk = 0$. Let there be $p$ 1s in the binary expansion of $h$, and $q$ in the binary expansion of $k$. Define $N^h_k$ to be the reversible circuit composed of $p \cdot q$-CNOT gates, with control bits on the wires specified by the binary expansion of $k$, and inverters as specified by the binary expansion of $h$. $N^h_k$ performs $N^h$ iff the wires specified by $k$ have the value 1.

In a 3-input circuit, there are 3 possible T gates, namely $N^1_6, N^2_5,$ and $N^4_3$. They compute the permutations $(6;7)$; $(5;7)$; $(3;7)$ respectively. By composing these three transpositions in all possible ways, we may form all 24 permutations of $3;5;6;7$. These are precisely the non-negative integers less than 8 which are not of the form 0 or 2$^i$. Clearly, no T gate can affect an input with fewer than two 1s in its binary expansion.
Lemma 18  Every T-circuit fixes 0 and 2^i for all i.

For \( k \) T-circuits, \( k > 3 \), there is an added restriction. As T gates are 3 3, there can be no \( k \) \( k \) gates in the circuit, so by Lemma 11 the circuit must compute an even permutation. On the other hand, we will show that these are the only restrictions on T-constructible permutations. We will do this by choosing an arbitrary even permutation, and then giving an explicit construction of a circuit which computes it using no temporary storage. The first step is to decompose the permutation into a product of pairs of disjoint transpositions.

Lemma 19  For \( n > 4 \), any even permutation in \( S_n \) may be written as the product of pairs of disjoint transpositions. If a permutation \( \pi \) moves \( k \) indices, it may be decomposed into no more than \( \frac{k+1}{2} \) pairs of transpositions.

Proof:  By a pair of disjoint transpositions, we mean something of the form \( (a; b) \cdot (c; d) \) where \( a; b; c; d \) are distinct. For \( k = 3 \), \( (x_0 x_1 x_2 \cdots x_k) = (x_0 x_1)(x_1 x_2)(x_0 x_2 \cdots x_k) \). Now \( (x_0 x_1)(x_1 x_2) \) are disjoint, iteratively applying this decomposition process will convert an arbitrary cycle into a product of pairs of disjoint transpositions possibly followed by a single transposition, a 3-cycle or both.

Consider an arbitrary permutation \( \pi = c_0 c_1 \cdots c_k \), where \( c_0 \cdots c_k \) are the disjoint cycles in its cycle decomposition. As shown above, we may rewrite this as \( \pi = \kappa_1 \cdots \kappa_m \tau_1 \cdots \tau_p \sigma_1 \cdots \sigma_q \), where the \( \kappa_i \) are pairs of disjoint transpositions, the \( \tau_i \) are transpositions, and the \( \sigma_i \) are 3-cycles. As the \( \tau_i \) come from pairwise disjoint cycles, they must in turn be pairwise disjoint. Moreover, there must be an even number of them as \( \pi \) was assumed to be even, and the \( \kappa_i \) and \( \sigma_i \) are all even. Pairing up the \( \tau_i \) arbitrarily leaves an expression of the form \( \kappa_1 \cdots \kappa_m \cdot \tau_1 \cdots \tau_p \cdot \sigma_1 \cdots \sigma_q \cdot \pi \). Again, the \( \sigma_i \) are pairwise disjoint. Note that \( (a; b; c) \cdot (d; e; f) = [(a; b) \cdot (d; e)][(a; c) \cdot (d; f)] \), we may therefore rewrite any pair of disjoint 3-cycles as two pairs of disjoint transpositions. Iterating this process leaves at most one 3-cycle, \( (x; y; z) \). Since we are working in \( A_n \) for \( n > 4 \), there are at least two other indices, \( v, w \). Using these, we have \( (x; y; z) = [(x; y) \cdot (v; w)][(v; w) \cdot (x; z)] \).

A careful count of transposition pairs gives the bound \( \frac{k+1}{2} \) in the statement of the lemma. This bound is tight in the case of a permutation consisting of a single 4n + 3 cycle.

By Lemma 19, it suffices to show that we may construct a circuit for an arbitrary disjoint transposition pair. We begin with an important special case. On \( n \) wires, a \( N_{2^4}^{1} \) gate computes the permutation \( \kappa_0 = 0^n 4; 2^n 3) 0^n 2; 2^n 1) \), which may be implemented by \( 8 \\ 5 \\ T \) gates [1 Corollary 7.4].

Lemma 20  On \( n \) wires, the permutation \( \kappa_0 = 0^n 4; 2^n 3) 0^n 2; 2^n 1) \) is T-constructible.

Consider now an arbitrary disjoint transposition pair, \( \kappa = (a; b) \cdot (c; d) \). Given a permutation \( \pi \) with the property \( \pi (a) = 2^n \ 4, \pi (b) = 2^n \ 3, \pi (c) = 2^n \ 2, \pi (d) = 2^n \ 1, \) we have \( \kappa = \pi \kappa_0 \pi^{-1} \).
where $\kappa_0$ is the permutation in Lemma 20. We have a circuit which computes $\kappa_0$. Given a circuit that computes $\pi$, we may obtain a circuit computing $\pi^\dagger$ by reversing it. We now construct a circuit computing $\pi$.

**Lemma 21** Suppose $n > 3$, and $0 < a;b;c;d < 2^n$. Further suppose that none of $a;b;c;d$ is 0, or of the form $2^i$. Then there exists a $T$-constructible permutation $\pi$ with the property $\pi(a) = 2^n$ 1, $\pi(\phi) = 2^n$ 2, $\pi(\psi) = 2^n$ 3, $\pi(\emptyset) = 2^n$ 4, computable by a circuit of no more than $5n$ 2 $T$ gates.

**Proof:** To simplify notation, set $M = 2^{n+1}$ and $m = n + 1$. Now, we construct $\pi$ in five stages. First, we build a permutation $\pi_a$ such that $\pi_a(a) = M + 4$. Then, we build $\pi_b$ such that $\pi_b(\phi) = M + 1$, and $\pi_b(M + 4) = M + 4$. Similarly, $\pi_c$ will fix $M + 1$ and $M + 4$, while $\pi_d(\psi) = M + 2$, and $\pi_d$ will fix $M + 1, M + 2, M + 4$ while $\pi_d(\emptyset) = M + 7$. Finally, we build a circuit that maps $M + 4 \mapsto 2M + 4, M + 1 \mapsto 2M + 3, M + 2 \mapsto 2M + 2, M + 7 \mapsto 2M + 1$.

By hypothesis, $a$ is not 0 or of the form $2^i$. This means that $a$ has at least two 1s in its binary expansion, say in positions $h_a$ and $k_a$. Apply $T$ gates with controls on positions $h_a$ and $k_a$ to set the second and $m$th bits. More precisely, let $z_a = 2^{h_a} + 2^{k_a}$, apply a $N_{\emptyset}^{M_a}$ iff $a$ has a 0 in the $(n + 1)$st bit and $N_{\emptyset}^{h_a}$ iff $a$ has a 0 in the 2nd bit. Now, apply $T$ gates with the controls on the $m$th and 2nd bits to set the remaining bits to 0. Let $\pi_a$ be the permutation computed by the circuit given above.

$\pi_a(\phi)$ must again have two nonzero bits in its binary expansion; since $b \not\equiv a$ implies $\pi_a(\phi) \not\equiv \pi_a(\emptyset)$, some nonzero bit of $\pi_a(\phi)$ lies on neither the $m$th nor the 2nd wire. Controlling by this and another bit, use the techniques of the previous paragraph to build a circuit taking $\pi_a(\phi) \mapsto M + 1$. By construction, this fixes $M + 4$; let the permutation computed by this circuit be $\pi_a$.

Consider now the nonzero bits of $c^0 = \pi_b(\psi)$. Again, since $a;b \equiv c$, we have $M + 4 \equiv M + 1 \equiv c^0$. Therefore, there must be at least one bit in which $c^0$ differs from $M + 4$. This bit could be the $m$th or the second bit, and $c^0$ could have a zero in this position. However, as $c^0$ is guaranteed to have at least 2 non-zero bits, there must be some other bit which is 1 in $c^0$ and 0 in $M + 4$. Similarly, there must be some other bit which is 1 in $c^0$ and 0 in $M + 1$. Controlling by these two bits (or, if they are the same bit, by this bit and any other bit which is 1 in $c^0$), we may use the above method to set $c^0 \mapsto M + 2$.

Next, consider the nonzero bits of $d^0 = \pi_c(\emptyset)$. First, suppose there are two which are not on the $m$th wire. Controlling by these can take $d^0 \mapsto M + 7$ without affecting any of the other values, as none of $M + 1;M + 2;M + 4$ have 1s in both these positions. If there are no two 1s in the binary expansion of $d^0$ which both lie off the $m$th wire, there can be at most two 1s in the binary expansion, one of which lies on the $m$th wire. Since $a;b;c \not\equiv d$, the second must lie on some wire which is not the 0th, 1st, or 2nd; in this case we may again control by these two bits to take $d^0 \mapsto M + 7$ without affecting other values.
Finally, apply $N^4_{M+1}$ and $N^4_{M+2}$ gates, and then a $N^8_{M+4}$ circuit. The reader may verify that this completes stage 5. Each of the first 4 stages takes at most $n$ T gates, as we flip at most $n$ bits in each. The final stage uses exactly $n^2$ T gates.

We now have a key result to prove.

**Theorem 22** Every T-constructible permutation in $S_{2^n}$ fixes 0 and $2^i$ for all $i$, and is even if $n > 3$. Conversely, every permutation of this form is T-constructible. A T-constructible permutation which moves $s$ indices requires at most $3(s + 1) (3n - 7)$ T gates. There are $\frac{1}{2} \mathcal{Q}^n \ n \ (n - 1)!$ T-constructible permutations in $S_{2^n}$.

**Proof:** We have already dealt with the case $n = 3$; hence suppose $n > 3$. The first statement follows directly from Lemmas 11 and 18. Now let $\pi \in S_{2^n}$ be an arbitrary even permutation fixing 0, $2^i$. Use the method of Lemma 19 to decompose $\pi$ into pairs of disjoint transpositions which fix 0, $2^i$. We are justified in using Lemma 19 because, for $n > 3$, there are at least five numbers between 0 and $2^n - 1$ which are not of the form 0 or $2^i$. Finally, using the circuits implied by Lemmas 20 and 21 we may construct circuits for each of these transposition pairs. Chaining these circuits together gives a circuit for the permutation $\pi$. Collecting the length bounds of the various lemmas cited gives the length bound in the theorem. The final claim then follows.

### 3.2 Circuit Equivalences

Given a (possibly long) reversible circuit to perform a specified task, one approach to reducing the circuit size is to perform local optimizations using circuit equivalences. The idea is to find subcircuits amenable to reduction. This direction is pursued in a paper by Iwama et al. [8], which examines circuit transformation rules for generalized-CNOT circuits which only alter one bit of the circuit. In their scenario, other bits may be altered during computation, so long as they are returned to their initial state by the end of the computation. We present a more general framework for deriving equivalences, from which many of the equivalences from [8] follow as special cases.

First, let us introduce notation to better deal with control bits.

**Definition 23** Let $G^i$ be a reversible gate that only affects wires corresponding to the 1s in the binary expansion of $i$ (as in an $N_i$ gate). Let the bitwise Boolean product $i \cdot j = 0$. Then define $V_j (G^i)$ as the gate which computes $G^i$ iff the wires specified by $j$ all carry a 1.

In particular, $V_j (N^i) = N^i_j$, and $V_k V_j (G^i) = V_{k+j} (G^i)$. Addition, multiplication, etc., of lower indices will always be taken to be bitwise Boolean, with $+$, $\cdot$, $\oplus$ representing OR, AND, and XOR respectively. We denote the bitwise complement of $x$ as $\overline{x}$. 

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Lemma 24  Let $K$ be an $n$ reversible circuit such that $K(0x_1:::x_{n-1}) = (0x_1:::x_{n-1})$, and let $f : B^n \to B^n$ be the function defined by $K((f(x_1:::x_{n-1})) = (f(x_1:::x_{n-1}))$. Then $f$ is a well-defined permutation in $S_{2^n}$, and if $F$ is a circuit computing $f$, then $V_1(F) = K$.

Proof: $K$, by hypothesis, permutes the inputs with a leading 0 amongst themselves. By reversibility, it must permute inputs with a leading 1 amongst themselves as well.

Definition 25 The commutator of permutations $P$ and $Q$, denoted $[P;Q]$, is $PQP^{-1}Q^{-1}$.

Moreover, it has reasonable properties with respect to control bits as the following result indicates.

Corollary 26 $V_{h+j}(G_i;H^j) = V_{h+j}(V_{h+j+k_j}(G_i;H^j))$

Proof: The corollary provides a circuit equivalent to the commutator of two given gates with arbitrary control bits. Namely, such a circuit can be constructed in two steps. First, identify wires which act as control for one gate but are not touched by the other gate. Second, connect the latter gate to every such wire so that the wire controls the gate.

By induction, it suffices to show that this procedure can be done to one such wire. Without loss of generality, suppose control bits and only control bits appear on the first wire. Then the input to this wire goes through the circuit unchanged. At least one of the two gates whose commutator is being computed must, by hypothesis, be controlled by the first wire. Therefore, on an input of zero to the first wire, this gate (and therefore its inverse) leaves all signals unchanged. Since the other gate appears along with its inverse, the whole circuit leaves the input unchanged. Our result now follows from Lemma 24.

If we are computing the commutator of generalized CNOT gates, then we may pick $G_i;H^j$ to be single inverters $N^i;N^j$ with $i,j$ having only a single 1 apiece in their binary expansions. Then we must have $h = j = 0$ or $j$, and $k = i = 0$ or $i$. The four cases are accounted for as follows:

Lemma 27 Let $i,j$ have only a single 1 apiece in their binary expansions. Then $N^i;N^j = N^j$, $N^i;N^j = N^i$, $N^i;N^j = 1$, and $N^i;N^j = N^j$.

Proof: As these equivalences all involve only 2-bit circuits, we may check them for $i = 0, j = 1$ by evaluating both sides of each equivalence on each of 4 inputs.

3.3 $CTN$ and $CTT$ Constructible Permutations

While an arbitrary CNT-circuit may have the C, N, and T gates interspersed arbitrarily, we first consider circuits in which these gates are segregated by type.

Definition 28 For any gate libraries $L_1:::L_k$, a $L_1:::L_k$-circuit is an $L_1$-circuit followed by an $L_2$-circuit, . . . , followed by an $L_k$-circuit. A permutation computed by an $L_1:::L_k$-circuit is $L_1:::L_k$-constructible.
A CNT-circuit with all N gates appearing at the right end is called a CT N circuit.

**Theorem 29** Let π be CNT-constructible. Then π is also CT N-constructible. Moreover, π uniquely determines the permutations π_{CT} and π_N computed by the CT and N sub-circuits, respectively.

**Proof:** We move all the N gates toward the outputs of the circuit. Each box in Figure 7 indicates a way of replacing an N \( \otimes \) CT circuit with a CT N circuit. The equivalences in this figure come from Corollary 26. Moreover, every possible way for an N gate to appear to the immediate left of a C or a T is accounted for, up to permuting the input and output wires. Now, number the non-N gates in the circuit in a reverse topological order starting from the outputs. In particular, if two gates appear at the same level in a circuit diagram, they must be independent, and one can order them arbitrarily. Let \( d \) be the number of the highest-numbered gate with an N gate to its immediate left. All N gates past the \( d \)-th gate \( G \) can be reordered with the \( G \) gate without introducing new N gates on the other side of \( G \), and without introducing new gates between the N gates and the outputs. In any event, as there are no remaining N gates to the left of \( G \), \( d \) decreases. This process terminates with all the N gates are clustered together at the circuit outputs. If we always cancel redundant pairs of N gates, then no more than two new gates will be introduced for each non-inverter originally in the circuit; additionally, there will be at most \( n \) N gates when the process is complete. Thus if the original circuit had \( l \) gates, then the new circuit has at most \( 3 (\emptyset 1) + n \) gates. Note that C and T gates (and hence CT-circuits) fix 0. Thus \( \pi(\emptyset) = \pi_N(\emptyset) \), so \( \pi_N = N^{\pi(\emptyset)} \), and \( \pi_{CT} = \pi_{N^{\pi(\emptyset)}} \).

Thus, if we want a CNT-circuit computing a permutation \( \pi \), we can quickly compute \( \pi_N \) and then simplify the problem to that of finding a CT-circuit for \( \pi \pi_N \). By Theorem 29, we know that a minimal-gate circuit of this form has roughly three times as many gates as the gate-minimal circuit computing \( \pi \).

The next natural question is whether an arbitrary CT-circuit is equivalent to some T \( \otimes \) C circuit. The equivalences in Figure 7 suggest that the answer is yes. However, the proof of Theorem 29 requires that many N gates be able to simultaneously move past a C or T gate, while Figure 7 only shows how to move a single C gate past a single T gate.
Lemma 30  The permutation $\pi$ computed by a $T\mathcal{C}$-circuit determines the permutations $\pi_T$ and $\pi_C$ computed by the sub-circuits. An even permutation is TC-constructible iff it fixes 0 and the images of inputs of the form $2^i$ are linearly independent over $\mathbb{F}_2$.

Proof:  Let $\pi$ be an arbitrary permutation. If $\pi$ is $T\mathcal{C}$-constructible, then images of the inputs $2^i$ are unaffected by the $T$ subcircuit; by Lemma 7 they must be mapped to linearly independent values by the $C$ subcircuit. This mapping of basis vectors completely specifies the permutation $\pi_C$ computed by the $C$ subcircuit, and therefore also the permutation $\pi_T = \pi \pi_C^{-1}$ computed by the $T$ subcircuit. Conversely, suppose $\pi$ is even and fixes 0, and the images of $2^i$ are linearly independent. Then there is some $C$-circuit taking the values $2^i$ to their images under $\pi$. Let it compute the permutation $\pi_C$; then $\pi \pi_C^{-1}$ fixes the values 0 and $2^i$ by construction. Theorem 22 therefore guarantees that $\pi \pi_C^{-1}$ is $T$-constructible. 

We will later use this result to show the existence of CT-constructible permutations which are not $T\mathcal{C}$ constructible.

3.4 $T\mathcal{C}\mathcal{J}\mathcal{N}$-Constructible Permutations

We are now ready to prove Theorem 12. According to Lemma 30, zero-fixing even permutations are $T\mathcal{C}$-constructible if they map inputs of the form $2^i$ in a certain way. This suggests that $T\mathcal{C}$-circuits account for a relatively large fraction of such permutations.

Theorem 31  Every zero-fixing permutation in $S_2$ and every zero-fixing even permutation in $S_2^n$ for $n > 4$ is $T\mathcal{C}\mathcal{J}\mathcal{N}$-constructible, and hence is CT-constructible. None requires more than $n^2 C$ gates and $3 Q^n + n + 1$ $\beta n - 7) T$ gates.

Proof:  Let $\pi$ be any zero-fixing permutation. Note that if the images of $2^i$ under $\pi$ were linearly independent, Lemma 30 would imply that $\pi$ was $T\mathcal{C}$ constructible. So, we will build a permutation $\pi_T$ with the property that the images of $2^i$ under $\pi_T$ are linearly independent, ensuring that $\pi \pi_T$ is $T\mathcal{C}$-constructible. Given a $T\mathcal{C}$-circuit for $\pi \pi_T$ and a T-circuit for $\pi_T$, we can reverse the circuit for $\pi_T$ and append it to the end of the $T\mathcal{C}$-circuit for $\pi \pi_T$ to give at $T\mathcal{C}\mathcal{J}\mathcal{N}$-circuit for $\pi$. All that remains is to show we can build one such $\pi_T$.

The basis vectors $2^i$ must be mapped either to themselves, to other basis vectors, or to vectors with at least two 1s. Let $i_1 : : : i_k$ be the indices of basis vectors which are not the images of other basis vectors, and let $j_1 : : : j_k$ be the indices of basis vectors whose images have at least two 1s. Let $\tilde{i}_1 : : : \tilde{i}_{n-k}$ and $\tilde{j}_1 : : : \tilde{j}_{n-k}$ be the indices which are not in the $i_m$ and $j_m$ respectively. Consider the matrix $M_\pi$ in which the $i$th column is the binary expansion of $\pi (2^i)$. We take the entries of $M_\pi$ to be elements of $\mathbb{F}_2$. Our indexing system divides $M_\pi$ into four submatrices: $M_\pi (\tilde{i} \tilde{j})$, $M_\pi (i \tilde{j})$, $M_\pi (\tilde{i} j)$, and $M_\pi (i j)$. By construction, $M_\pi (i j)$ and $M_\pi (\tilde{i} \tilde{j})$ are square, $M_\pi (\tilde{i} j)$ is a permutation matrix, and $M_\pi (i \tilde{j})$ is a zero matrix. Therefore, $\det M_\pi = \det (M_\pi (\tilde{i} \tilde{j}))$, and $M_\pi$ is
invertible iff $M_\pi \psi_{ij}$ is. Moreover, there is an invertible linear transformation, computable by column-reduction, which zeroes out the matrix $M_\pi \psi_{ij}$ without affecting $M_\pi \psi_{ij}$ or $M_\pi \psi_{ij}$. As this transformation $L$ is invertible, it corresponds to a permutation $\pi_x$, and the matrix $ML$ is the matrix of images of $2^1 \... \under$ the permutation $\pi_x$. In particular, the columns of $(ML)_i$ must all be different, which implies that the columns of $M_\pi \psi_{ij}$ must all be different. Moreover, $\pi_x$ is linear, and therefore zero-fixing; hence $M_\pi \psi_{ij}$ can have no zero columns. Taken together, these facts imply that for $k = 1, 2, M_\pi \psi_{ij}$ is invertible, hence so is $M_\pi$, thus $\pi$ is $T_\mathcal{C}$-constructible.

Suppose $k = 3$, and consider the family of matrices $A(p)$ defined as follows. $A(p)$ is a $p \times p$ matrix with 1s on the diagonal, 1s in the first row, and 1s in the first column, except possibly in the $(1; 1)$ entry, which is 1 iff $p$ is odd. Row-reducing the $A_i$ to lower triangular matrices quickly shows that the $A_i$ are invertible for all $i$. Moreover, for $i < 3$, there is at least two 1s in every column. Therefore, there is a $T$-constructible permutation $\pi_T$ such that $M_{\pi_T \psi_{ij}} = A_k$. Thus $\pi_{\pi_T}$ is $T_\mathcal{C}$-constructible, and $\pi$ is $T_\mathcal{C}F$-constructible.

Finally, we know from Corollary 32 that no more than $n^2$ gates are necessary to compute $\pi_C$. At most $2n$ indices need be moved by $\pi_T$, and no more than $2^n - n$ can be moved by the $T$-constructible part of $\pi$. Thus by Theorem 33 we need no more than $3 \cdot (2n + 1) \cdot (3n \cdot 7)$ gates for $\pi_T$ and no more than $3 \cdot (2n \cdot n) \cdot (3n \cdot 7)$ gates for $\pi$. Adding these gives the gate-count estimate above.

**Corollary 32** There exist $T_\mathcal{C}F$-constructible permutations which are not $T_\mathcal{C}$-constructible.

**Proof:** The permutation $\pi = (2; 6) (4; 7)$ fixes 0 and is even, hence is $T_\mathcal{C}F$-constructible in $S_{2^n}$ for all $n \geq 3$ by Theorem 33. However, $\pi (1) \pi (2) = 1 \pi (1) \pi (2) = 1$, and $6 = 7 = \pi (4)$, hence by Lemma 30, $\pi$ is not $T_\mathcal{C}$-constructible.

**Theorem 33** Every permutation in $S_{2^n}$ for $n = 1; 2; 3$ and every even permutation in $S_{2^n}$ for $n > 3$ is $T_\mathcal{C}F\mathcal{N}$-constructible, and hence CNT-constructible. None requires more than $n^2 C$ gates, $n N$ gates, and $3 \cdot (2^n + n + 1) \cdot (3n \cdot 7)$ $T$ gates.

**Proof:** Let $\pi$ be any permutation; then $\pi^0 = \pi N(\pi^0)$ fixes 0. For $n = 1$, $\pi^0$ must be the identity; for $n = 2$, $\pi^0$ permutes $1; 2; 3$, any such permutation is linear, hence $\pi^0$ is C-constructible. For $n = 3$, $\pi^0$ is $T_\mathcal{C}F$-constructible; for $n > 3$, $\pi^0$ is $T_\mathcal{C}F$-constructible iff it is even, which happens iff $\pi$ is even. Thus in all cases there is a $T_\mathcal{C}F$-circuit, $\Pi^0$ computing $\pi^0$, then $\Pi^0 N(\pi^0)$ is a $T_\mathcal{C}F\mathcal{N}$-circuit computing $\pi$.

We note that the size of a truth table for a circuit with $n$ inputs and $n$ outputs is $n2^n$ bits. The synthesis procedure used in the theorems above clearly runs in time proportional to the number of
gates in the final circuit. This is $O(n^2)$, hence the synthesis procedure detailed in the theorems has linear runtime in the input size.

Just as in Corollary 9, we may ask how far from optimal the foregoing construction is for long circuits. There are $2^n! = 2$ even permutations in $S_{2^n}$, and these are all CNT-constructible. Using Stirling’s approximation, $\log(k!) = k \log k$, and Lemma 8 gives:

**Corollary 34**  Worst case CNT-circuits on $n$ wires require $\Omega(n^{2^n} = \log n)$ gates.

So, for long CNT-circuits, the algorithm implied by Theorem 33 is asymptotically suboptimal by, at worst, a logarithmic factor, as it produces circuits of length $O(n^{2^n})$. This is remarkably similar to the result of Corollary 9 in which we found that using row reduction to build C-circuits is asymptotically suboptimal by a logarithmic factor in the case of long C-circuits. However, even a constant improvement in size is very desirable, and circuits for practical applications are almost never of the worst-case type considered in Corollaries 9 and 34.
4 Optimal Synthesis

We will now switch focus, and seek optimal realizations for permutations we know to be CNT-constructible. A circuit is optimal if no equivalent circuit has smaller cost; in our case, the cost function will be the number of gates in the circuit.

Lemma 35 (Property of Optimality) If $B$ is a sub-circuit of an optimal circuit $A$, then $B$ is optimal.

**Proof:** Suppose not. Then let $B^0$ be a circuit with fewer gates than $B$, but computing the same function. If we replace $B$ by $B^0$, we get another circuit $A^0$ which computes the same function as $A$. But since we have only modified $B$, $A^0$ must be as much smaller than $A$ as $B^0$ is smaller than $B$. $A$ was assumed to be optimal, hence this is a contradiction. (Note that equivalent, optimal circuits can have the same number of gates.) □

The algorithm detailed in this section relies entirely on the property of optimality for its correctness. Therefore, any cost function for which this property holds may, in principle, be used instead of gate count.

Lemma 35 allows us to build a library of small optimal circuits by dynamic programming because the first $m$ gates of an optimal $(m+1)$-gate circuit form an optimal subcircuit. Therefore, to examine all optimal $(m+1)$-gate circuits, we iterate through optimal $m$-gate circuits and add single gates at the end in all possible ways. We then check the resulting circuits against the library, and eliminate any which are equivalent to a smaller circuit. In fact, instead of storing a library of all optimal circuits, we store one optimal circuit per synthesized permutation and also store optimal circuits of a given size together.

One way to find an optimal circuit for a given permutation $\pi$ is to generate all optimal $k$-gate circuits for increasing values of $k$ until a circuit computing $\pi$ is found. This procedure requires $\Theta(n^2)$ memory in the worst case ($n$ is the number of wires) and may require more memory than is available. Therefore, we stop growing the circuit library at $m$-gate circuits, when hardware limitations become an issue. The second stage of the algorithm uses the computed library of optimal circuits and, in our implementation, starts by reading the library from a file. Since little additional memory is available, we trade off runtime for memory.

We use a technique known as depth-first search with iterative deepening (DFID) [10]. After a given permutation is checked against the circuit library, we seek circuits with $j = m + 1$ gates that implement this permutation. If none are found, we seek circuits with $j = m + 2$ gates, etc. This algorithm, in general, needs an additional termination condition to prevent infinite looping for inputs which cannot be synthesized with a given gate library. For each $j$, we consider all permutations optimally synthesizable in $m$ gates. For each such permutation $\rho$, we multiply $\pi$
CIRCUIT find_circ(COST, PERM)
// assumes circuit library stored in LIB

if (COST  k)

// If PERM can be computed by a circuit with  k gates,
// such a circuit must be in the library
return LIB[DEPTH].find(PERM)

else

// Try building the goal circuit from  k-gate circuits
for each C in LIB[k]

// Divide PERM by permutation computed by C
PERM2  PERM * INVERSE(C.perm)

// and try to synthesize the result
TEMP_CCT  find_circ(depth-k,PERM2)
if (TEMP_CCT != NIL) return TEMP_CCT * C

// Finally, if no circuit of the desired depth can be found
return NIL

Figure 8: Finding a circuit of cost  COST that computes permutation PERM
(NIL returned if no such circuit exists). TEMP_CCT and records in LIB represent
circuits, and include a field “perm” storing the permutation computed. The * character means both multiplication of permutations and concatenation of circuits, and
NIL*<anything>=NIL.

by ρ^1 and recursively try to synthesize the result using  j  m gates. When  j  m  m, this
can be done by checking against the existing library. Otherwise, the recursion depth increases.
Pseudocode for this stage of our algorithm is given in Figure 8.
Table 1: Number of permutations computable in an optimal $L$-circuit using a given number of gates. $L$ \textit{CNTS}. Runtimes are in seconds for a 2GHz Pentium-4 Xeon CPU.

In addition to being more memory-efficient than straightforward dynamic programming, our algorithm is faster than branching over all possible circuits. To quantify these improvements, consider a library of circuits of size $m$ or less, containing $l_m$ circuits of size $m$. We analyze the efficiency of the algorithms discussed by simulating them on an input permutation of cost $k$. Our algorithm requires $l_m^{b(k-1)=mc}$ references to the circuit library. Simple branching is no better than our algorithm with $m = 1$, and thus takes at least $l_1^k$ steps, which is $l_1^{b(k-1)=mc}$ times more than our algorithm. A speed-up can be expected because $l_m^{m^{mc}}$ but specific numerical values of that expression depend on the numbers of suboptimal and redundant optimal circuits of length $m$. Indeed, Table lists values of $l_m$ for various subsets of the CNTS gate library and $m = 3$. For example, for the NT gate library, $k = 12$, $b(k-1)=mc = 3$, $l_1 = 6$ and $l_m = 88$. Therefore the performance ratio is $l_1^{b(k-1)=mc} = 6^{12}=88^3 = 3194.2$. Yet, this comparison is incomplete because it does not account for time spent building circuit libraries. We point out that this charge is amortized over multiple synthesis operations. In our experiments, generating a circuit library on three wires of up to three gates ($m = 3$) from the CNTS gate library takes less than a minute
on a 2-GHz Pentium-4 Xeon. Using such libraries, all of Table 1 can be generated in minutes, but it cannot be generated even in several hours using branching.

Let us now see what additional information we can glean from Table 1. Adding the C gate to the NT library appears to significantly reduce circuit size, but further adding the S gate does not help as much. To illustrate this, we show sample worst-case circuits on three wires for the NT, CNT, and CNTS gate libraries in Figure 9.

The totals in Table 1 can be independently determined by the following arguments. Every reversible function on three wires can be synthesized using the CNT gate library [20], and there are $8! = 40;320$ of these. All can be synthesized with the NT library because the C gate is redundant in the CNT library; see Figure 3 a. On the other hand, adding the S gate to the library cannot decrease the number of synthesizable functions. Therefore, the totals in the NT and CNTS columns must be 40;320 as well. On the other side of the table, the number of possible N circuits is just $2^3 = 8$ since there are three wires, and there can be at most one N gate per wire in an optimal circuit (else we can cancel redundant pairs.) By Theorem 29, the number of CN-constructible permutations should be the product of the number of N-constructible permutations and the number of C constructible permutations, since any CN-constructible permutation can be written uniquely as a product of an N-constructible and a C-constructible permutation. So the total in the CN column should be the product of the totals in the C and N columns, which it is. Similarly, the total in the CNT column should be the product of the totals in the CT and N columns; this allows one to deduce the total number of CT-constructible permutations from values we know. Finally, we showed that there were 24 T-constructible permutations on 3 wires in Section 5 and Corollary 9 states that the number of permutations implementable on n wires with C gates is $\prod_{i=0}^{n-1} 2^i \cdot 2^i$. For $n = 3$ this yields 168 and agrees with Table 1.

We can also add to the discussion of T-C constructible circuits we began in Section 3. By Lemma 30, the number of T-C-constructible permutations can be computed as the product of the numbers of T-constructible and C-constructible permutations. Table 1 mentions 24 T-circuits and 168 C-circuits on three wires. The product, 4032, is less than 5040, the number of CT constructible permutations on three wires, as we would expect from Corollary 32.

Finally, the longest C-circuits we observed on 3, 4 and 5 wires merely permute the wires. Such wire-permutations on n wires never require more than 3 n 1) gates. However, from Corollary 9 we know that for large n, worst-case C-circuits require $\Omega (n^2 = log(n))$ gates. Identifying specific worst-case circuits and describing families with worst-case asymptotics remains a challenge.

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1Although complete statistics for all 16! 4-wire functions are beyond our reach, average synthesis times are less than one second when the input function can be implemented with eight gates or fewer. Functions requiring nine or more gates tend to take more than 1.5 hours to synthesize. In this case, memory constraints limit our circuit library to 4-gate circuits, and the large jump in runtime after the 8-gate mark is due to an extra level of recursion.
Finally, we note that while the exact runtime complexity of this algorithm is dependant on characteristics of the gate library chosen, for a complete gate library it is obviously exponential in the number of input wires to the circuit (this is guaranteed by Corollary 34), and in fact must be at least doubly-exponential in the number of input wires (that is, exponential in the size of the truth table). Scalability issues, therefore, restrict this approach to small problems. On the other hand, given that the state of the art in quantum computing is largely limited by ten qubits, such small circuits are of interest to physicists building quantum computing devices.

5 Quantum Search Applications

Quantum computation is necessarily reversible, and quantum circuits generalize their reversible counterparts in the classical domain [14]. Instead of wires, information is stored on qubits, whose states we write as $|0\rangle$ and $|1\rangle$ instead of 0 and 1. There is an added complexity — a qubit can be in a superposition state that combines $|0\rangle$ and $|1\rangle$. Specifically, $|0\rangle$ and $|1\rangle$ are thought of as vectors of the computational basis, and the value of a qubit can be any unit vector in the space they span. The scenario is similar when considering many qubits at once: the possible configurations of the corresponding classical system (bit-strings) are now the computational basis, and any unit vector in the linear space they span is a valid configuration of the quantum system. Just as the classical configurations of the circuit persist as basis vectors of the space of quantum configurations, so too classical reversible gates persist in the quantum context. Non-classical gates are allowed, in fact, any (invertible) norm-preserving linear operator is allowed as a quantum gate. However, quantum gate libraries often have very few non-classical gates [14]. An important example of a non-classical gate (and the only one used in this paper) is the Hadamard gate $H$. It operates on one qubit, and is defined as follows: $H|0\rangle = \frac{1}{\sqrt{2}}(|0\rangle + |1\rangle)$, and $H|1\rangle = \frac{1}{\sqrt{2}}(|0\rangle - |1\rangle)$. Note that because $H$ is linear, giving the images of the computational basis elements defines it completely.

During the course of a computation, the quantum state can be any unit vector in the linear space spanned by the computational basis. However, a serious limitation is imposed by quantum measurement, performed after a quantum circuit is executed. A measurement non-deterministically collapses the state onto some vector in a basis corresponding to the measurement being performed. The probabilities of outcomes depend on the measured state — basis vectors [nearly] orthogonal
Initialization: Form balanced superposition of search index states

Oracle: Is \( f(x) = 1? \)

Grover Operator: Iteratively transform the index states to increase the amplitude of the target states

Measurement: Detect and output the indexes of the target states

Figure 10: A high-level schematic of Grover’s search algorithm.

to the measured state are least likely to appear as outcomes of measurement. If \( H |j \rangle \) were measured in the computational basis, it would be seen as \( |j \rangle \) half the time, and \( |j \rangle \) the other half.

Despite this limitation, quantum circuits have significantly more computational power than classical circuits. In this work, we consider Grover’s search algorithm, which is faster than any known non-quantum algorithm for the same problem [6]. Figure 10 outlines a possible implementation of Grover’s algorithm. It begins by creating a balanced superposition of \( 2^n \) n-qubit states which correspond to the indexes of the items being searched. These index states are then repeatedly transformed using a Grover operator circuit, which incorporates the search criteria in the form of a search-specific predicate \( f(x) \). This circuit systematically amplifies the search indexes that satisfy \( f(x) = 1 \) until a final measurement identifies them with high probability.

A key component of the Grover operator is a so-called “oracle” circuit that implements a search-specific predicate \( f(x) \). This circuit transforms an arbitrary basis state \( |j \rangle \) to the state \( ( \frac{1}{\sqrt{2}} |f \rangle + \frac{i}{\sqrt{2}} |f \rangle ) |j \rangle \). The oracle is followed by (i) several Hadamard gates, (ii) a subcircuit which flips the sign on all computational basis states other than \( |j \rangle \), and (iii) more Hadamard gates. A sample Grover-operator circuit for a search on 2 qubits is shown in Figure 11 and uses one qubit of temporary storage [14]. The search space here is \( \{0,1,2,3\} \), and the desired indices are 0 and 3. The oracle circuit is highlighted by a dashed line. While the portion following the oracle is fixed, the oracle may vary depending on the search criterion. Unfortunately, most works on Grover’s algorithm do not address the synthesis of oracle circuits and their complexity. According to Bettelli et al. [4], this is a major obstacle for automatic compilation of high-level quantum programs, and little help is available.

**Lemma 36** [17] With one temporary storage qubit, the problem of synthesizing a quantum circuit
that transforms computational basis states $j \psi \rangle$ to $(1)^{f(\psi)} j \psi \rangle$ can be reduced to a problem in the synthesis of classical reversible circuits.

**Proof:** Define the permutation $\pi_f$ by $\pi_f(x; y) = (x; y \cdot f(x))$, and define a unitary operator $U_f$ by letting it permute the states of the computational basis according to $\pi_f$. The additional qubit is initialized to $j i = H j 1 i$ so that $U_f j \psi \rangle; \ i = (1)^{f(\psi)} j \psi \rangle; \ i$. If we now ignore the value of the last qubit, the system is in the state $(1)^{f(\psi)} j \psi \rangle$, which is exactly the state needed for Grover’s algorithm. Since a quantum operator is completely determined by its behavior on a given computational basis, any circuit implementing $\pi_f$ implements $U_f$. As reversible gates may be implemented with quantum technology, we can synthesize $U_f$ as a reversible logic circuit.

Quantum computers implemented so far are severely limited by the number of simultaneously available qubits. While $n$ qubits are necessary for Grover’s algorithm, one should try to minimize the number of additional temporary storage qubits. One such qubit is required by Lemma 36 to allow classical reversible circuits to alter the phase of quantum states.

**Corollary 37** For permutations $\pi_f (x; y) = (x; y \cdot f(x))$, such that $\forall x : f(x) = 1 g$ has even cardinality, no more temporary storage is necessary. For the remaining $\pi_f$, we need an additional qubit of temporary storage.

**Proof:** The permutation $\pi_f$ swaps $(x; y)$ with $(x; y \cdot f(x))$, and therefore performs one transposition for each element of $\forall x : f(x) = 1 g$. It is therefore even exactly when this set has even cardinality. The lemma follows from Corollary 13.

Given $\pi_f$, we can use the algorithm of Section 4 to construct an optimal circuit for it. Table 2 gives the optimal circuit sizes of functions $\pi_f$ corresponding to 3-input 1-output functions $f$ (“3+1 oracles”) which can be synthesized on four wires. These circuits are significantly smaller than many optimal circuits on four wires. This is not surprising, as they perform less computation.

In Grover oracle circuits, the main input lines preserve their input values and only the temporary storage lines can change their values. Therefore, Travaglione et al. [21] studied circuits where some lines cannot be changed even at intermediate stages of computation. In their terminology, a circuit with $k$ lines that we are allowed to modify and an arbitrary number of read-only
By induction, we can get as many uncomplemented literals in this product as we like. Given a circuit \( W \), and denote this operation by \( \text{w} \). For generality, let the \( k \) be the number of literals appearing (without complementation). If \( k > 0 \) then 3 \( k+1 \) 2 gates are required.

**Proof:** Assume we are given an XOR sum-of-products decomposition of \( f \). Then it suffices to know how to transform \( (\pi; a; b) = (\pi; a; b \ p) \) for an arbitrary product of uncomplemented literals \( p \), because then we can add the terms in an XOR decomposition term by term. So, without loss of generality, let \( p = x_1 :: x_m \). Denote by \( T(a; b; c) \) a T gate with controls on \( a; b \) and inverter on \( c \). Similarly, denote by \( C(a; b) \) a C gate with control on \( a \) and inverter on \( b \). Number the ROM wires \( 1 :: k \), and the non-ROM wires \( k+1 \) and \( k+2 \). Let us first suppose that there is at least one uncomplemented literal, and put a C(1; \( k+2 \)) on the circuit; note that C(1; \( k+2 \)) applied to the input \( (\pi; a; b) \) gives \( (\pi; a; b \ x_1) \). We will write this as \( C(1; k+2) = (\pi; a; b) \) ! (\( \pi; a; b \ x_1 \)), and denote this operation by \( W_1 \). Then, we define the circuit \( W_2 \) as the sequence of gates \( T(2; k+2; k+1)W_0 T(2; k+2; k+1)W_0 \), and one can check that \( W_2 = (\pi; a; b) \text{! (\( \pi; a \ x_1 x_2 \))} \). We define \( W_2 \) by exchanging the wires \( k+1 \) and \( k+2 \); clearly \( W_2 = (\pi; a; b) \text{! (\( \pi; a; b \ x_1 x_2 \))} \). In general, given a circuit \( W_i = (\pi; a; b \ x_1 :: x_i) \) ! (\( \pi; a \ x_1 :: x_i \)), we define \( W_{i+1} = T(\emptyset; k+2; k+1)W_i \); one can check that \( W_{i+1} = (\pi; a; b) \text{! (\( \pi; a \ x_1 x_{i+1} \))} \). Define \( W_{i+1} \) by exchanging the wires \( k+1 \) and \( k+2 \); then clearly \( W_{i+1} = (\pi; a; b) \text{! (\( \pi; a; b \ x_1 :: x_{i+1} \))} \). By induction, we can get as many uncomplemented literals in this product as we like. \( \square \)

The heuristic presented above has the property that none of its gates has more than one control bit on a ROM bit. Indeed, Travaglione et al. [21] had restricted their attention to circuits with precisely this property. However, they note [21] that their results do not depend on this restriction.

We applied the construction of Lemma [21] to all 256 functions implementable in 2-bit ROM-based circuits with 3 bits of ROM. The circuit size distribution is given in the line labeled XOR in Table 3. In comparing with circuits lengths resulting from our synthesis algorithm of Section

| Circuit Size | 0 | 1 | 2 | 3 | 4 | 5 | 6 | 7 | Total |
|--------------|---|---|---|---|---|---|---|---|-------|
| No. of circuits | 1 | 7 | 21 | 35 | 35 | 24 | 4 | 1 | 128 |

Table 2: Optimal 3+1 oracle circuits for Grover’s search.
Table 3: Circuit size distribution of 3+2 ROM-based circuits synthesized using various algorithms.

| Size | 0 | 1 | 2 | 3 | 4 | 5 | 6 | 7 | 8 | 9 | 10 | 11 | 12 | 13 |
|------|---|---|---|---|---|---|---|---|---|---|----|----|----|----|
| XOR  | 1 | 4 | 6 | 4 | 4 | 12| 18| 12| 6 | 12 | 19 | 16 | 10 | 8  |
| OPT T| 1 | 4 | 6 | 4 | 4 | 12| 21| 24| 29| 33| 44 | 46 | 22 | 5  |
| OPT  | 1 | 7 | 21| 35| 36| 28| 28| 35| 21| 7 | 1  | 0  | 0  | 0  |

We consider two cases. First, in the OPT T line, we only look at circuits satisfying the restriction mentioned above. Then, in the OPT line, we relax this restriction and give the circuit size distribution for optimal circuits.

Most functions computable by a 2-bit ROM-based circuit actually require two writeable bits. Whether or not a given function can be computed by a 1-bit ROM-based CNT-circuit, can be determined by the following constructive procedure. Observe that gates in 1-bit ROM circuits can be reordered arbitrarily, as no gate affects the control bits of any other gate. Thus, whether or not a C or T gate flips the controlled bit, depends only on the circuit inputs. Furthermore, multiple copies of the same gate on the same wires cancel out, and we can assume that at most one is present in an optimal circuit. A synthesis procedure can then check which gates are present by applying the permutation on every possible input combination with zero, one, or two 1s in its binary expansion. (Again, we have relaxed the restriction that only 1 control may be on a ROM wire). If the value of the function is 1, the circuit needs an N, C or T gate controlled by those bits.

Observe that adding the S gate to the gate library during \( k + 1 \) ROM synthesis will never decrease circuit sizes — no two wires can be swapped since at least one of them is a ROM wire. In the case of \( k + 2 \) ROM synthesis, only the two non-ROM wires can be swapped, and one of them must be returned to its initial value by the end of the computation. We ran an experiment comparing circuit lengths in the 3+2 ROM-based case and found no improvement in circuit sizes upon adding the S gate, but we have been unable to prove this in the general case.

\(^2\)Using a circuit library with 6 gates (191Mb file, 1.5 min to generate), the OPT line takes 5 min to generate. The use of a 5-gate library improved the runtimes by at least 2x if we do not synthesize the only circuit of size 11. For the OPT T line, we first find the 250 optimal circuits of size 12 (15 min) using a 6-gate library (61Mb, 5min). The remaining 6 functions were synthesized in 5 min with a 7-gate library (376Mb, 10 min). This required more than 1Gb of RAM.
6 Conclusions

We have explored a number of promising techniques for synthesizing optimal and near-optimal reversible circuits that require little or no temporary storage. In particular, we have proven that every even permutation function can be synthesized without temporary storage using the CNT gate library. Similarly, any permutation, even or odd, can be synthesized with up to one bit of temporary storage. We have recently discovered that A. DeVos has independently demonstrated this result, however, his proof relies on non-trivial group-theoretic notions and resorts to a computer algebra package for a special case. [5] We give a much more elementary analysis, and moreover our proof techniques are sufficiently constructive to be interpreted as a synthesis heuristic. We have also derived various equivalences among CNT-circuits that are useful for synthesis purposes, and given a decomposition of a CNT-circuit into a $T \mathcal{C}^j \mathcal{N}$-circuit.

To further investigate the structure of reversible circuits, we developed a method for synthesizing optimal reversible circuits. While this algorithm scales better than its counterparts for irreversible computation [11], its runtime is still exponential. Nonetheless, it can be used to study small problems in detail, which may be of interest to physicists building quantum computing devices because the current state of the art is largely limited by 10 qubits. One might think that an exhaustive search procedure would suffice for small problems, but in fact, even for three-input circuits, an exhaustive search is nowhere near finished after 15 hours; our procedure terminates in minutes. Our experimental data about all optimal reversible circuits on three wires using various subsets of the CNTS library reveal some interesting characteristics of optimal reversible circuits. Such statistics, extrapolated to larger circuits, can be used in the future to guide heuristics, and may suggest new theorems about reversible circuits.

Finally, we have applied our optimal synthesis tool to the design of oracle circuits for a key quantum computing application, Grover’s search algorithm, and obtained much smaller circuits than previous methods. Ultimately, we aim to extend the proposed methods to handle larger and more general circuits, with the eventual goal of synthesizing quantum circuits containing dozens of qubits.
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