Time and Space Bounds for Reversible Simulation*

(Extended Abstract)

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Abstract. We prove a general upper bound on the tradeoff between time and space that suffices for the reversible simulation of irreversible computation. Previously, only simulations using exponential time or quadratic space were known. The tradeoff shows for the first time that we can simultaneously achieve subexponential time and subquadratic space. The boundary values are the exponential time with hardly any extra space required by the Lange-McKenzie-Tapp method and the \((\log 3)^3\)th power time with square space required by the Bennett method. We also give the first general lower bound on the extra storage space required by general reversible simulation. This lower bound is optimal in that it is achieved by some reversible simulations.

1 Introduction

Computer power has roughly doubled every 18 months for the last half-century (Moore’s law). This increase in power is due primarily to the continuing miniaturization of the elements of which computers are made, resulting in more and more elementary gates per unit area with higher and higher clock frequency, accompanied by less and less energy dissipation per elementary computing event. Roughly, a linear increase in clock speed is accompanied by a square increase in elements per unit area—so if all elements compute all of the time, then the dissipated energy per time unit rises cubicly (linear times square) in absence of energy decrease per elementary event. The continuing dramatic decrease in dissipated energy per elementary event is what has made Moore’s law possible. But there is a foreseeable end to this: There is a minimum quantum of energy dissipation associated with elementary events. This puts a fundamental limit on how far we can go with miniaturization, or does it?

Reversible Computation: R. Landauer \(8\) has demonstrated that it is only the ‘logically irreversible’ operations in a physical computer that necessarily dissipate energy by generating a corresponding amount of entropy for every

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bit of information that gets irreversibly erased; the logically reversible operations can in principle be performed dissipation-free. Currently, computations are commonly irreversible, even though the physical devices that execute them are fundamentally reversible. At the basic level, however, matter is governed by classical mechanics and quantum mechanics, which are reversible. This contrast is only possible at the cost of efficiency loss by generating thermal entropy into the environment. With computational device technology rapidly approaching the elementary particle level it has been argued many times that this effect gains in significance to the extent that efficient operation (or operation at all) of future computers requires them to be reversible (for example, in [8, 1, 2, 4, 7, 11, 5]). The mismatch of computing organization and reality will express itself in friction: computers will dissipate a lot of heat unless their mode of operation becomes reversible, possibly quantum mechanical. Since 1940 the dissipated energy per bit operation in a computing device has—with remarkable regularity—decreased at the inverse rate of Moore’s law [7] (making Moore’s law possible). Extrapolation of current trends shows that the energy dissipation per binary logic operation needs to be reduced below $kT$ (thermal noise) within 20 years. Here $k$ is Boltzmann’s constant and $T$ the absolute temperature in degrees Kelvin, so that $kT \approx 3 \times 10^{-21}$ Joule at room temperature. Even at $kT$ level, a future device containing 1 trillion ($10^{12}$) gates operating at 1 terahertz ($10^{12}$) switching all gates all of the time dissipates about 3000 watts. Consequently, in contemporary computer and chip architecture design the issue of power consumption has moved from a background worry to a major problem. For current research towards implementation of reversible computing on silicon see MIT’s Pendulum Project and linked web pages (http://www.ai.mit.edu/~cvieri/reversible.html).

On a more futuristic note, quantum computing [15, 14] is reversible. Despite its importance, theoretical advances in reversible computing are scarce and far between; all serious ones are listed in the references.

**Related Work:** Currently, almost no algorithms and other programs are designed according to reversible principles (and in fact, most tasks like computing Boolean functions are inherently irreversible). To write reversible programs by hand is unnatural and difficult. The natural way is to compile irreversible programs to reversible ones. This raises the question about efficiency of general reversible simulation of irreversible computation. Suppose the irreversible computation to be simulated uses $T$ time and $S$ space. A first efficient method was proposed by Bennett [3], but it is space hungry and uses $ST^{\log 3}$ time and space $S \log T$. If $T$ is maximal, that is, exponential in $S$, then the space use is $S^2$. This method can be modelled by a reversible pebble game. Reference [12] demonstrated that Bennett’s method is optimal for reversible pebble games and that simulation space can be traded off against limited erasing. In [3] it was shown that using a method by Sipser [16] one can reversibly simulate using only $O(S)$ extra space but at the cost of using exponential time. In [3] the authors provide

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1 By judicious choosing of simulation parameters this method can be tweaked to run in $ST^{1+\epsilon}$ time for every $\epsilon > 0$ at the cost of introducing a multiplicative constant depending on $1/\epsilon$. The complexity analysis of [3] was completed in [10].
an oracle construction (essentially based on [12]) that separates reversible and irreversible space-time complexity classes.

**Results:** Previous results seem to suggest that a reversible simulation is stuck with either quadratic space use or exponential time use. This impression turns out to be false:

Here we prove a tradeoff between time and space which has the exponential time simulation and the quadratic space simulation as extremes and for the first time gives a range of simulations using simultaneously subexponential (i.e., \(2^{o(n)}\) is subexponential if \(f(n) = o(n)\)) time and subquadratic space. The idea is to use Bennett’s pebbling game where the pebble steps are intervals of the simulated computation that are bridged by using the exponential simulation method. (It should be noted that embedding Bennett’s pebbling game in the exponential method gives no gain, and neither does any other iteration of embeddings of simulation methods.) Careful analysis shows that the simulation using \(k\) pebbles takes \(T' = S3^{k2^{O(T/2^k)}}\) time and \(S' = O(kS)\) space, and in some cases the upper bounds are tight. For \(k = 0\) we have the exponential time simulation method and for \(k = \log T\) we have Bennett’s method. Interesting values arise for say

(a) \(k = \log \log T\): \(T' = S(\log T)^{\log 32^{O(T/\log T)}}\) and \(S' = S \log \log T \leq S \log S\);

(b) \(k = \sqrt{\log T}\): \(S' = S\sqrt{\log T} \leq S\sqrt{S}\) and \(T' = S3^{\sqrt{\log T}2^{O(T/2\sqrt{\log T})}}\).

(c) Let \(T, S, T', S'\) be as above. Eliminating the unknown \(k\) shows the tradeoff between simulation time \(T'\) and extra simulation space \(S': T' = S3^{\frac{S}{S'}}2^{O(T/2^{\frac{S}{S'}})}\).

(d) Let \(T, S, T', S'\) be as above and let the irreversible computation be halting and compute a function from inputs of \(n\) bits to outputs. For general reversible simulation by a reversible Turing machine using a binary tape alphabet and a single tape, \(S' \geq n + \log T + O(1)\) and \(T' \geq T\). This lower bound is optimal in the sense that it can be achieved by simulations at the cost of using time exponential in \(S\).

**Main open problem:** The ultimate question is whether one can do better, and obtain improved upper and lower bounds on the tradeoff between time and space of reversible simulation, and in particular whether one can have almost linear time and almost linear space simultaneously.

## 2 Reversible Turing Machines

In the standard model of a Turing machine the elementary operations are rules in quadruple format \((p, s, a, q)\) meaning that if the finite control is in state \(p\) and the machine scans tape symbol \(s\), then the machine performs action \(a\) and subsequently the finite control enters state \(q\). Such an action \(a\) consists of either printing a symbol \(s'\) in the tape square scanned, or moving the scanning head one tape square left or right.

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\(^2\) The work reported in this paper dates from 1998; Dieter van Melkebeek has drawn our attention to the unpublished [17] with similar, independent but later, research.
Quadruples are said to overlap in domain if they cause the machine in the same state and scanning the same symbol to perform different actions. A deterministic Turing machine is defined as a Turing machine with quadruples no two of which overlap in domain.

Now consider the special format (deterministic) Turing machines using quadruples of two types: read/write quadruples and move quadruples. A read/write quadruple \((p, a, b, q)\) causes the machine in state \(p\) scanning tape symbol \(a\) to write symbol \(b\) and enter state \(q\). A move quadruple \((p, *, \sigma, q)\) causes the machine in state \(p\) to move its tape head by \(\sigma \in \{-1, +1\}\) squares and enter state \(q\), oblivious to the particular symbol in the currently scanned tape square. (Here ‘−1’ means ‘one square left’, and ‘+1’ means ‘one square right’.) Quadruples are said to overlap in range if they cause the machine to enter the same state and either both write the same symbol or (at least) one of them moves the head. Said differently, quadruples that enter the same state overlap in range unless they write different symbols. A reversible Turing machine is a deterministic Turing machine with quadruples no two of which overlap in range. A \(k\)-tape reversible Turing machine uses \((2k + 2)\) tuples which, for every tape separately, select a read/write or move on that tape. Moreover, any two tuples can be restricted to some single tape where they don’t overlap in range.

To show that every partial recursive function can be computed by a reversible Turing machine one can proceed as follows [1]. Take the standard irreversible Turing machine computing that function. We modify it by adding an auxiliary storage tape called the ‘history tape’. The quadruple rules are extended to 6-tuples to additionally manipulate the history tape. To be able to reversibly undo (retrace) the computation deterministically, the new 6-tuple rules have the effect that the machine keeps a record on the auxiliary history tape consisting of the sequence of quadruples executed on the original tape. Reversibly undoing a computation entails also erasing the record of its execution from the history tape. This notion of reversible computation means that only 1 : 1 recursive functions can be computed. To reversibly simulate an irreversible computation from \(x\) to \(f(x)\) one reversibly computes from input \(x\) to output \(\langle x, f(x) \rangle\).

Reversible Turing machines or other reversible computers will require special reversible programs. One feature of such programs is that they should be executable when read from bottom to top as well as when read from top to bottom. Examples are the programs \(F(\cdot)\) and \(A(\cdot)\) in [12]. In general, writing reversible programs will be difficult. However, given a general reversible simulation of irreversible computation, one can simply write an oldfashioned irreversible program in an irreversible programming language, and subsequently simulate it reversibly. This leads to the following:

**Definition 1.** An irreversible-to-reversible compiler receives an irreversible program as input and compiles it to a reversible program.

Note that there is a decisive difference between reversible circuits and reversible special purpose computers on the one hand, and reversible universal computers on the other hand [13]. While one can design a special-purpose reversible version for every particular irreversible circuit using reversible universal gates, such
A method does not yield an irreversible-to-reversible compiler that can execute any irreversible program on a fixed universal reversible computer architecture as we are interested in here.

3 Time Parsimonious Simulation

3.1 Background

We keep the discussion at an intuitive informal level; the cited references contain the formal details and rigorous constructions. An irreversible deterministic Turing machine has an infinite graph of all configurations where every configuration has outdegree at most one. In a reversible deterministic Turing machine every configuration also has indegree at most one. The problem of reversing an irreversible computation from its output is to revisit the input configurations starting from the output configuration by a process of reversibly traversing the graph.

The reversible Bennett strategy essentially reversibly visits only the linear graph of configurations visited by the irreversible deterministic Turing machine in its computation from input to output, and no other configurations in the graph. It does so by a recursive procedure of establishing and undoing intermediate checkpoints that are kept simultaneously in memory. It turns out that this can be done using limited time $T^{\log 3}$ and space $S\log T$.

3.2 Reversible Pebbling

Let $G$ be a linear list of nodes $\{1, 2, \ldots, T_G\}$. We define a pebble game on $G$ as follows. The game proceeds in a discrete sequence of steps of a single player. There are $n$ pebbles which can be put on nodes of $G$. At any time the set of pebbles is divided in pebbles on nodes of $G$ and the remaining pebbles which are called free pebbles. At every step either an existing free pebble can be put on a node of $G$ (and is thus removed from the free pebble pool) or be removed from a node of $G$ (and is added to the free pebble pool). Initially $G$ is unpebbled and there is a pool of free pebbles. The game is played according to the following rule:

**Reversible Pebble Rule:** If node $i$ is occupied by a pebble, then one may either place a free pebble on node $i + 1$ (if it was not occupied before), or remove the pebble from node $i + 1$.

We assume an extra initial node 0 permanently occupied by an extra, fixed pebble, so that node 1 may be (un)pebbled at will. This pebble game is inspired by the method of simulating irreversible Turing Machines on reversible ones in a space efficient manner. The placement of a pebble corresponds to checkpointing the next state of the irreversible computation, while the removal of a pebble corresponds to reversibly erasing a checkpoint. Our main interest is in determining the number of pebbles $k$ needed to pebble a given node $i$. 
The maximum number $n$ of pebbles which are simultaneously on $G$ at any one time in the game gives the space complexity $nS$ of the simulation. If one deletes a pebble not following the above rules, then this means a block of bits of size $S$ is erased irreversibly.

### 3.3 Algorithm

We describe the idea of Bennett’s simulation. This simulation is optimal among all reversible pebble games. The total computation of $T$ steps is broken into $2^k$ segments of length $m = T2^{-k}$. Every $m$th point of the computation is a node in the pebbling game; node $i$ corresponding to $im$ steps of computation.

For each pebble a section of tape is reserved long enough to store the whole configuration of the simulated machine. By enlarging the tape alphabet, each pebble will require space only $S + O(1)$.

Both the pebbling and unpebbling of a pebble $t$ on some node, given that the previous node has a pebble $s$ on it, will be achieved by a single reversible procedure $\text{bridge}(s, t)$. This looks up the configuration at section $s$, simulates $m$ steps of computation in a manner described in section and exclusive-or’s the result into section $t$. If $t$ was a free pebble, meaning that its tape section is all zeroes, the result is that pebble $t$ occupies the next node. If $t$ already pebbled that node then it will be zeroed as a result.

The essence of Bennett’s simulation is a recursive subdivision of a computation path into 2 halves, which are traversed in 3 stages; the first stage gets the midpoint pebbled, the second gets the endpoint pebbled, and the 3rd recovers the midpoint pebble. The following recursive procedure implements this scheme; $\text{Pebble}(s, t, n)$ uses free pebbles $0, \ldots, n - 1$ to compute the $2^n$th node after the one pebbled by $s$, and exclusive-or’s that node with pebble $t$ (either putting $t$ on the node or taking it off). Its correctness follows by straightforward induction. Note that it is its own reverse; executing it twice will produce no net change. The pebble parameters $s$ and $t$ are simply numbers in the range $-1, 0, 1, \ldots, k$. Pebble -1 is permanently on node 0, pebble $k$ gets to pebble the final node, and pebble $i$, for $0 \leq i < k$ pebbles nodes that are odd multiples of $2^i$. The entire simulation is carried out with a call $\text{pebble}(-1, k, k)$.

```plaintext
pebble(s, t, n)
{
    if (n = 0)
        bridge(s, t);
    fi
    if (n > 0)
        let $r = n - 1$
        pebble(s, r, n - 1);
        pebble(r, t, n - 1);
        pebble(s, r, n - 1)
    fi
}
```
As noted by Bennett, both branches and merges must be labeled with mutually exclusive conditions to ensure reversibility. Recursion can be easily implemented reversibly by introducing an extra stack tape, which will hold at most \( n \) stack frames of size \( O(\log n) \) each, for a total of \( O(n \log n) \).

This pebbling method is optimal in that no more than \( 2^n + 1 \) steps can be bridged with \( n \) pebbles \[12\]. A call \( \text{pebble}(s, t, n) \) results in \( 3^n \) calls to \( \text{bridge}(...) \). Bennett chose the number of pebbles large enough \( (n = \Omega(\log T)) \) so that \( m \) becomes small, on the order of the space \( S \) used by the simulated machine. In that case \( \text{bridge}(s, t) \) is easily implemented with the help of an additional \textit{history} tape of size \( m \) which records the sequence of transitions. Instead, we allow an arbitrary choice of \( n \) and resort to the space efficient simulation of \[9\] to bridge the pebbled checkpoints.

4 Space Parsimonious Simulation

Lange, McKenzie and Tapp, \[9\], devised a reversible simulation, \textit{LMT-simulation} for short, that doesn’t use extra space, at the cost of using exponential time. Their main idea of reversibly simulating a machine without using more space is by reversibly cycling through the configuration tree of the machine (more precisely the connected component containing the input configuration). This configuration tree is a tree whose nodes are the machine configurations and where two nodes are connected by an edge if the machine moves in one step from one configuration to the other. We consider each edge to consist of two \textit{half-edges}, each adjacent to one configuration.

The configuration tree can be traversed by alternating two permutations on half-edges: a swapping permutation which swaps the two half-edges constituting each edge, and a rotation permutation whose orbits are all the half-edges adjacent to one configuration. Both permutations can be implemented in a constant number of steps. For simplicity one assumes the simulated machine strictly alternates moving and writing transitions. To prevent the simulation from exceeding the available space \( S \), each pebble section is marked with special left and right markers \( \dagger, \ddagger \), which we assume the simulated machine not to cross. Since this only prevents crossings in the forward simulation, we furthermore, with the head on the left (right) marker, only consider previous moving transitions from the right (left).

5 The Tradeoff Simulation

To adapt the LMT simulation to our needs, we equip our simulating machine with one extra tape to hold the simulated configuration and another extra tape counting the difference between forward and backward steps simulated. \( m = 2^n \) steps of computation can be bridged with a \( \log m \) bits binary counter, incremented with each simulated forward step, and decremented with each simulated backward step—incurring an extra \( O(\log m) \) factor slowdown in simulation speed. Having obtained the configuration \( m \) steps beyond that of pebble
s, it is exclusive-or’d into section t and then the LMT simulation is reversed to end up with a zero counter and a copy of section s, which is blanked by an exclusive-or from the original.

bridge(s, t)
{
    copy section s onto (blanked) simulation tape
    setup: goto enter;
    loop1: come from endloop1;
    simulate step with swap&rotate and adjust counter
    if (counter=0)
        rotate back;
        if (simulation tape = section s)
            enter: come from start;
        fi (simulation tape = section s)
    fi (counter=0)
    endloop1: if (counter!=m) goto loop1;
    exclusive-or simulation tape into section t
    if (counter!=m)
        loop2: come from endloop2;
        reverse-simulate step with anti-rotate&swap and adjust counter
        if (counter=0)
            rotate back;
            if (simulation tape = section s) goto exit;
        fi (counter=0)
        endloop2: goto loop2;
    exit: clear simulation tape using section s
}

5.1 Complexity Analysis

Let us analyze the time and space used by this simulation.

**Theorem 1.** An irreversible computation using time T and space S can be simulated reversibly in time \( T' = 3^k 2^{O(T/2^k)} S \) and space \( S' = S(1 + O(k)) \), where \( k \) is a parameter that can be chosen freely \( 0 \leq k \leq \log T \) to obtain the required tradeoff between reversible time \( T' \) and space \( S' \).

**Proof.** (Sketch) Every invocation of the bridge() procedure takes time \( O(2^{O(m)} S) \). That is, every configuration has at most \( O(1) \) predecessor configurations where it can have come from (constant number of states, constant alphabet size and choice of direction). Hence there are \( \leq 2^{O(m)} \) configurations to be searched and about as many potential start configurations leading in \( m \) moves to the goal configuration, and every tape section comparison takes time \( O(S) \). The pebbling game over \( 2^k \) nodes takes \( 3^k \) (un)pebbling steps each of which is an invocation of bridge(). Filling in \( m = T/2^k \) gives the claimed time bound. Each of the \( k + O(1) \) pebbles takes space \( O(S) \), as does the simulation tape and the counter, giving the claimed total space. \( \square \)
It is easy to verify that for some simulations the upper bound is tight. The boundary cases, \( k = 0 \) gives the LMT-simulation using exponential time and no extra space, and \( k = \log T \) gives Bennett’s simulation using at most square space and subquadratic time. Taking intermediate values of \( k \) we can choose to reduce time at the cost of an increase of space use and vice versa. In particular, special values \( k = \log \log T \) and \( k = \sqrt{T} \) give the results using simultaneously subexponential time and subquadratic space exhibited in the introduction. Eliminating \( k \) we obtain:

**Corollary 1.** Let \( T, S, T', S' \) be as above. Then there is a reversible simulation that has the following tradeoff between simulation time \( T' \) and extra simulation space \( S' \):

\[
T' = S^{3^{\frac{k}{3}}} 2^{O(T/2^{\frac{k}{3}})}.
\]

### 5.2 Local Irreversible Actions

Suppose we have an otherwise reversible computation containing local irreversible actions. Then we need to reversibly simulate only the subsequence of irreversible steps, leaving the connecting reversible computation segments unchanged. That is, an irreversibility parsimonious computation is much cheaper to reversibly simulate than an irreversibility hungry one.

### 5.3 Reversible Simulation of Unknown Computing Time

In the previous analysis we have tacitly assumed that the reversible simulator knows in advance the number of steps \( T \) taken by the irreversible computation to be simulated. In this context one can distinguish on-line computations and off-line computations to be simulated. On-line computations are computations which interact with the outside environment and in principle keep running forever. An example is the operating system of a computer. Off-line computations are computations which compute a definite function from an input (argument) to an output (value). For example, given as input a positive integer number, compute as output all its prime factors. For every input such an algorithm will have a definite running time.

There is a well-known simple device to remove this dependency for batch computations without increasing the simulation time (and space) too much. Suppose we want to simulate a computation with unknown computation time \( T \). Then we simulate \( t \) steps of the computation with \( t \) running through the sequence of values \( 2, 2^2, 2^3, \ldots \). For every value \( t \) takes on we reversibly simulate the first \( t \) steps of the irreversible computation. If \( T > t \) then the computation is not finished at the end of this simulation. Subsequently we reversibly undo the computation until the initial state is reached again, set \( t := 2t \) and reversibly simulate again. This way we continue until \( t \geq T \) at which bound the computation finishes. The total time spent in this simulation is

\[
T'' = 2 \sum_{i=1}^{[\log T]} S^{3^{\frac{i}{3}}} 2^{O(2^{i-\frac{3i}{3}})} \leq 2T'.
\]
6 Lower Bound on Reversible Simulation

It is not difficult to show a simple lower bound on the extra storage space required for general reversible simulation. We consider only irreversible computations that are halting computations performing a mapping from an input to an output. For convenience we assume that the Turing machine has a single binary work tape delimited by markers †, ‡ that are placed $S$ positions apart. Initially the binary input of length $n$ is written left adjusted on the work tape. At the end of the computation the output is written left adjusted on the work tape. The markers are never moved. Such a machine clearly can perform every computation as long as $S$ is large enough with respect to $n$. Assume that the reversible simulator is a similar model albeit reversible. The average number of steps in the computation is the uniform average over all equally likely inputs of $n$ bits.

**Theorem 2.** To generally simulate an irreversible halting computation of a Turing machine as above using storage space $S$ and $T$ steps on average, on inputs of length $n$, by a general reversible computation using storage space $S'$ and $T'$ steps on average, the reversible simulator Turing machine having $q'$ states, requires trivially $T' \geq T$ and $S' \geq n + \log T - O(1)$ up to a logarithmic additive term.

Proof. There are $2^n$ possible inputs to the irreversible computation, the computation on every input using on average $T$ steps. A general simulation of this machine cannot use the semantics of the function being simulated but must simulate every step of the simulated machine. Hence $T' \geq T$. The simulator being reversible requires different configurations for every step of everyone of the simulated computations that is, at least $2^n T$ configurations. The simulating machine has not more than $q' 2^{S'} 2^{S'}$ distinct configurations—$2^{S'}$ distinct values on the work tape, $q'$ states, and $S'$ head positions for the combination of input tape and work tape. Therefore, $q' 2^{S'} 2^{S'} \geq 2^n T$. That is, $q' S' 2^{S'} - n \geq T$ which shows that $S' - n - \log S' \geq \log T - \log q'$. □

For example, consider irreversible computations that don’t use extra space apart from the space to hold the input, that is, $S = n$. An example is the computation of $f(x) = 0$.

- If $T$ is polynomial in $n$ then $S' = n + \Omega(\log n)$.
- If $T$ is exponential in $n$ then $S' = n + \Omega(n)$.

Thus, in some cases the LMT-algorithm is required to use extra space if we deal with halting computations computing a function from input to output. In the final version of the paper the authors have added that their simulation uses some extra space for counting (essentially $O(S)$) in case we require halting computations from input to output, matching the lower bound above for $S = n$ since their simulation uses on average $T'$ steps exponential in $S$.

**Optimality and Tradeoffs:** The lower bound of Theorem 2 is optimal in the following sense. As one extreme, the LMT-algorithm of discussed above uses $S' = n + \log T$ space for simulating irreversible computations of total functions on inputs of $n$ bits, but at the cost of using $T' = \Omega(2^S)$ simulation time. As the other
extreme, Bennett’s simple algorithm in [1] uses $T' = O(T)$ reversible simulation time, but at the cost of using $S' = \Omega(T)$ additional storage space. This implies that improvements in determining the complexity of reversible simulation must consider time-space tradeoffs.

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